

# Introduction to Modern Analysis

Shmuel Kantorovitz

*Bar Ilan University,  
Ramat Gan, Israel*

OXFORD  
UNIVERSITY PRESS

**Oxford Graduate Texts in Mathematics**

*Series Editors*

R. Cohen   S. K. Donaldson  
S. Hildebrandt   T. J. Lyons  
M. J. Taylor

1. Keith Hannabuss: *An Introduction to Quantum Theory*
2. Reinhold Meise and Dietmar Vogt: *Introduction to Functional Analysis*
3. James G. Oxley: *Matroid Theory*
4. N. J. Hitchin, G. B. Segal, and R. S. Ward: *Integrable Systems: Twistors, Loop Groups, and Riemann Surfaces*
5. Wulf Rossmann: *Lie Groups: An Introduction Through Linear Groups*
6. Q. Liu: *Algebraic Geometry and Arithmetic Curves*
7. Martin R. Bridson and Simon M. Salamon (eds): *Invitations to Geometry and Topology*
8. Shmuel Kantorovitz: *Introduction to Modern Analysis*
9. Terry Lawson: *Topology: A Geometric Approach*
10. Meinolf Geck: *An Introduction to Algebraic Geometry and Algebraic Groups*

# OXFORD

UNIVERSITY PRESS

Great Clarendon Street, Oxford OX2 6DP

Oxford University Press is a department of the University of Oxford.  
It furthers the University's objective of excellence in research, scholarship,  
and education by publishing worldwide in

Oxford New York

Auckland Cape Town Dar es Salaam Hong Kong Karachi  
Kuala Lumpur Madrid Melbourne Mexico City Nairobi  
New Delhi Shanghai Taipei Toronto

With offices in

Argentina Austria Brazil Chile Czech Republic France Greece  
Guatemala Hungary Italy Japan Poland Portugal Singapore  
South Korea Switzerland Thailand Turkey Ukraine Vietnam

Oxford is a registered trade mark of Oxford University Press  
in the UK and in certain other countries

Published in the United States  
by Oxford University Press Inc., New York

© Oxford University Press 2003

The moral rights of the author have been asserted  
Database right Oxford University Press (maker)

First published 2003

First published in paperback 2006

All rights reserved. No part of this publication may be reproduced,  
stored in a retrieval system, or transmitted, in any form or by any means,  
without the prior permission in writing of Oxford University Press,  
or as expressly permitted by law, or under terms agreed with the appropriate  
reprographics rights organization. Enquiries concerning reproduction  
outside the scope of the above should be sent to the Rights Department,  
Oxford University Press, at the address above

You must not circulate this book in any other binding or cover  
and you must impose the same condition on any acquirer

British Library Cataloguing in Publication Data  
Data available

Library of Congress Cataloging in Publication Data  
Data available

Typeset by Newgen Imaging Systems (P) Ltd., Chennai, India  
Printed in Great Britain  
on acid-free paper by  
Biddles Ltd., King's Lynn

ISBN 0-19-852656-3 978-0-19-852656-8  
ISBN 0-19-920315-6 (Pbk.) 978-0-19-920315-4 (Pbk.)

1 3 5 7 9 10 8 6 4 2

To Ita, Bracha, Pnina, Pinchas, and Ruth



# Preface

This book grew out of lectures given since 1964 at Yale University, the University of Illinois at Chicago, and Bar Ilan University. The material covers the usual topics of Measure Theory and Functional Analysis, with applications to Probability Theory and to the theory of linear partial differential equations. Some relatively advanced topics are included in each chapter (excluding the first two): the Riesz–Markov representation theorem and differentiability in Euclidean spaces (Chapter 3); Haar measure (Chapter 4); Marcinkiewicz’s interpolation theorem (Chapter 5); the Gelfand–Naimark–Segal representation theorem (Chapter 7); the Von Neumann double commutant theorem (Chapter 8); the spectral representation theorem for normal operators (Chapter 9); the extension theory for unbounded symmetric operators (Chapter 10); the Lyapounov Central Limit theorem and the Kolmogoroff ‘Three Series theorem’ (Application I); the Hormander–Malgrange theorem, fundamental solutions of linear partial differential equations with variable coefficients, and Hormander’s theory of convolution operators, with an application to integration of pure imaginary order (Application II). Some important complementary material is included in the ‘Exercises’ sections, with detailed hints leading step-by-step to the wanted results. Solutions to the end of chapter exercises may be found on the companion website for this text: <http://www.oup.co.uk/academic/companion/mathematics/kantorovitz>.

*Ramat Gan*  
*July 2002*

S. K.





# Contents

|          |  |           |
|----------|--|-----------|
| <b>1</b> | <b>Measures</b>                                  | <b>1</b>  |
| 1.1      | Measurable sets and functions                    | 1         |
| 1.2      | Positive measures                                | 7         |
| 1.3      | Integration of non-negative measurable functions | 9         |
| 1.4      | Integrable functions                             | 15        |
| 1.5      | $L^p$ -spaces                                    | 22        |
| 1.6      | Inner product                                    | 29        |
| 1.7      | Hilbert space: a first look                      | 32        |
| 1.8      | The Lebesgue–Radon–Nikodym theorem               | 34        |
| 1.9      | Complex measures                                 | 39        |
| 1.10     | Convergence                                      | 46        |
| 1.11     | Convergence on finite measure space              | 49        |
| 1.12     | Distribution function                            | 50        |
| 1.13     | Truncation                                       | 52        |
|          | Exercises  | 54        |
| <b>2</b> | <b>Construction of measures</b>                  | <b>57</b> |
| 2.1      | Semi-algebras                                    | 57        |
| 2.2      | Outer measures                                   | 59        |
| 2.3      | Extension of measures on algebras                | 62        |
| 2.4      | Structure of measurable sets                     | 63        |
| 2.5      | Construction of Lebesgue–Stieltjes measures      | 64        |
| 2.6      | Riemann versus Lebesgue                          | 67        |
| 2.7      | Product measure                                  | 69        |
|          | Exercises  | 73        |
| <b>3</b> | <b>Measure and topology</b>                      | <b>77</b> |
| 3.1      | Partition of unity                               | 77        |
| 3.2      | Positive linear functionals                      | 79        |
| 3.3      | The Riesz–Markov representation theorem          | 87        |
| 3.4      | Lusin’s theorem                                  | 89        |
| 3.5      | The support of a measure                         | 92        |
| 3.6      | Measures on $\mathbb{R}^k$ ; differentiability   | 93        |
|          | Exercises  | 97        |

|          |   |            |
|----------|---|------------|
| <b>4</b> | <b>Continuous linear functionals</b>  | <b>102</b> |
| 4.1      | Linear maps   | 102        |
| 4.2      | The conjugates of Lebesgue spaces   | 104        |
| 4.3      | The conjugate of $C_c(X)$   | 109        |
| 4.4      | The Riesz representation theorem  | 111        |
| 4.5      | Haar measure  | 113        |
|          | Exercises   | 121        |
| <b>5</b> | <b>Duality</b>  | <b>123</b> |
| 5.1      | The Hahn–Banach theorem   | 123        |
| 5.2      | Reflexivity   | 127        |
| 5.3      | Separation  | 130        |
| 5.4      | Topological vector spaces   | 133        |
| 5.5      | Weak topologies   | 135        |
| 5.6      | Extremal points   | 139        |
| 5.7      | The Stone–Weierstrass theorem   | 143        |
| 5.8      | Operators between Lebesgue spaces: Marcinkiewicz’s<br>interpolation theorem | 145        |
|          | Exercises   | 150        |
| <b>6</b> | <b>Bounded operators</b>  | <b>153</b> |
| 6.1      | Category  | 153        |
| 6.2      | The uniform boundedness theorem   | 154        |
| 6.3      | The open mapping theorem  | 156        |
| 6.4      | Graphs  | 159        |
| 6.5      | Quotient space  | 160        |
| 6.6      | Operator topologies   | 161        |
|          | Exercises   | 164        |
| <b>7</b> | <b>Banach algebras</b>  | <b>170</b> |
| 7.1      | Basics  | 170        |
| 7.2      | Commutative Banach algebras   | 178        |
| 7.3      | Involution  | 181        |
| 7.4      | Normal elements   | 183        |
| 7.5      | General $B^*$ -algebras   | 185        |
| 7.6      | The Gelfand–Naimark–Segal construction                                      | 190        |
|          | Exercises   | 195        |
| <b>8</b> | <b>Hilbert spaces</b>   | <b>203</b> |
| 8.1      | Orthonormal sets  | 203        |
| 8.2      | Projections   | 206        |
| 8.3      | Orthonormal bases   | 208        |
| 8.4      | Hilbert dimension   | 211        |
| 8.5      | Isomorphism of Hilbert spaces   | 212        |
| 8.6      | Canonical model   | 213        |
| 8.7      | Commutants  | 214        |
|          | Exercises   | 215        |

|           |   |            |
|-----------|---|------------|
| <b>9</b>  | <b>Integral representation</b>                                    | <b>223</b> |
| 9.1       | Spectral measure on a Banach subspace                             | 223        |
| 9.2       | Integration   | 224        |
| 9.3       | Case $Z = X$  | 226        |
| 9.4       | The spectral theorem for normal operators                         | 229        |
| 9.5       | Parts of the spectrum   | 231        |
| 9.6       | Spectral representation   | 233        |
| 9.7       | Renorming method  | 235        |
| 9.8       | Semi-simplicity space   | 237        |
| 9.9       | Resolution of the identity on $Z$                                 | 239        |
| 9.10      | Analytic operational calculus                                     | 243        |
| 9.11      | Isolated points of the spectrum                                   | 246        |
| 9.12      | Compact operators   | 248        |
|           | Exercises   | 252        |
| <b>10</b> | <b>Unbounded operators</b>  | <b>258</b> |
| 10.1      | Basics  | 258        |
| 10.2      | The Hilbert adjoint   | 261        |
| 10.3      | The spectral theorem for unbounded selfadjoint operators          | 264        |
| 10.4      | The operational calculus for unbounded selfadjoint operators      | 265        |
| 10.5      | The semi-simplicity space for unbounded operators in Banach space | 267        |
| 10.6      | Symmetric operators in Hilbert space                              | 271        |
|           | Exercises   | 275        |
|           | <b>Application I Probability</b>                                  | <b>283</b> |
| I.1       | Heuristics  | 283        |
| I.2       | Probability space   | 285        |
| I.3       | Probability distributions   | 298        |
| I.4       | Characteristic functions  | 307        |
| I.5       | Vector-valued random variables                                    | 315        |
| I.6       | Estimation and decision   | 324        |
| I.7       | Conditional probability   | 336        |
| I.8       | Series of $L^2$ random variables                                  | 349        |
| I.9       | Infinite divisibility   | 355        |
| I.10      | More on sequences of random variables                             | 359        |
|           | <b>Application II Distributions</b>                               | <b>364</b> |
| II.1      | Preliminaries   | 364        |
| II.2      | Distributions   | 366        |
| II.3      | Temperate distributions   | 376        |
| II.4      | Fundamental solutions   | 392        |
| II.5      | Solution in $\mathcal{E}'$  | 396        |
| II.6      | Regularity of solutions   | 398        |

|                     |                             |     |
|---------------------|-----------------------------|-----|
| II.7                | Variable coefficients       | 400 |
| II.8                | Convolution operators       | 404 |
| II.9                | Some holomorphic semigroups | 415 |
| <b>Bibliography</b> |                             | 421 |
| <b>Index</b>        |                             | 425 |

# 1

## Measures

### 1.1 Measurable sets and functions

The setting of abstract measure theory is a family  $\mathcal{A}$  of so-called *measurable* subsets of a given set  $X$ , and a function

$$\mu : \mathcal{A} \rightarrow [0, \infty],$$

so that the *measure*  $\mu(E)$  of the set  $E \in \mathcal{A}$  has some ‘intuitively desirable’ property, such as ‘countable additivity’:

$$\mu \left( \bigcup_{i=1}^{\infty} E_i \right) = \sum_{i=1}^{\infty} \mu(E_i),$$

for mutually disjoint sets  $E_i \in \mathcal{A}$ . In order to make sense, this setting has to deal with a family  $\mathcal{A}$  that is closed under countable unions. We then arrive to the concept of a *measurable space*.

**Definition 1.1.** Let  $X$  be a (non-empty) set. A  $\sigma$ -algebra of subsets of  $X$  (briefly, a  $\sigma$ -algebra *on*  $X$ ) is a subfamily  $\mathcal{A}$  of the family  $\mathbb{P}(X)$  of all subsets of  $X$ , with the following properties:

- (1)  $X \in \mathcal{A}$ ;
- (2) if  $E \in \mathcal{A}$ , then the complement  $E^c$  of  $E$  belongs to  $\mathcal{A}$ ;
- (3) if  $\{E_i\}$  is a *sequence* of sets in  $\mathcal{A}$ , then its *union* belongs to  $\mathcal{A}$ .

The ordered pair  $(X, \mathcal{A})$ , with  $\mathcal{A}$  a  $\sigma$ -algebra on  $X$ , is called a *measurable space*. The sets of the family  $\mathcal{A}$  are called *measurable sets* (or  $\mathcal{A}$ -measurable sets) in  $X$ .

Observe that by (1) and (2), the empty set  $\emptyset$  belongs to the  $\sigma$ -algebra  $\mathcal{A}$ . Taking then  $E_i = \emptyset$  for all  $i > n$  in (3), we see that  $\mathcal{A}$  is closed under *finite unions*; if this weaker condition replaces (3),  $\mathcal{A}$  is called an *algebra* of subsets of  $X$  (briefly, an algebra *on*  $X$ ).

By (2) and (3), and DeMorgan's Law,  $\mathcal{A}$  is closed under countable intersections (finite intersections, in the case of an algebra). In particular, any algebra on  $X$  is closed under differences  $E - F := E \cap F^c$ .

The intersection of an arbitrary family of  $\sigma$ -algebras on  $X$  is a  $\sigma$ -algebra on  $X$ . If all the  $\sigma$ -algebras in the family contain some fixed collection  $\mathcal{E} \subset \mathbb{P}(X)$ , the said intersection is the smallest  $\sigma$ -algebra on  $X$  (with respect to set inclusion) that contains  $\mathcal{E}$ ; it is called *the  $\sigma$ -algebra generated by  $\mathcal{E}$* , and is denoted by  $[\mathcal{E}]$ .

An important case comes up naturally when  $X$  is a topological space (for some topology  $\tau$ ). The  $\sigma$ -algebra  $[\tau]$  generated by the topology is called the *Borel* ( $\sigma$ )-algebra [denoted  $\mathcal{B}(X)$ ], and the sets in  $\mathcal{B}(X)$  are the *Borel sets* in  $X$ . For example, the countable intersection of  $\tau$ -open sets (a so-called  $G_\delta$ -set) and the countable union of  $\tau$ -closed sets (a so-called  $F_\sigma$ -set) are Borel sets.

**Definition 1.2.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be measurable spaces. A map  $f : X \rightarrow Y$  is *measurable* if for each  $B \in \mathcal{B}$ , the set

$$f^{-1}(B) := \{x \in X; f(x) \in B\} := [f \in B]$$

*belongs to  $\mathcal{A}$ .*

A constant map  $f(x) = p \in Y$  is trivially measurable, since  $[f \in B]$  is either  $\emptyset$  or  $X$  (when  $p \in B^c$  and  $p \in B$ , respectively), and so belongs to  $\mathcal{A}$ .

When  $Y$  is a topological space, we shall usually take  $\mathcal{B} = \mathcal{B}(Y)$ , the Borel algebra on  $Y$ . In particular, for  $Y = \mathbb{R}$  (the real line),  $Y = [-\infty, \infty]$  (the 'extended real line'), or  $Y = \mathbb{C}$  (the complex plane), with their usual topologies, we shall call the measurable map a *measurable function* (more precisely, an  $\mathcal{A}$ -measurable function). If  $X$  is a topological space, a  $\mathcal{B}(X)$ -measurable map (function) is called a Borel map (function).

Given a measurable space  $(X, \mathcal{A})$  and a map  $f : X \rightarrow Y$ , for an arbitrary set  $Y$ , the family

$$\mathcal{B}_f := \{F \in \mathbb{P}(Y); f^{-1}(F) \in \mathcal{A}\}$$

is a  $\sigma$ -algebra on  $Y$  (because the inverse image operation preserves the set theoretical operations:  $f^{-1}(\bigcup_\alpha F_\alpha) = \bigcup_\alpha f^{-1}(F_\alpha)$ , etc.), and it is the largest  $\sigma$ -algebra on  $Y$  for which  $f$  is measurable.

If  $Y$  is a topological space, and  $f^{-1}(V) \in \mathcal{A}$  for every *open*  $V$ , then  $\mathcal{B}_f$  contains the topology  $\tau$ , and so contains  $\mathcal{B}(Y)$ ; that is,  $f$  is measurable. Since  $\tau \subset \mathcal{B}(Y)$ , the converse is trivially true.

**Lemma 1.3.** *A map  $f$  from a measurable space  $(X, \mathcal{A})$  to a topological space  $Y$  is measurable if and only if  $f^{-1}(V) \in \mathcal{A}$  for every open  $V \subset Y$ .*

In particular, if  $X$  is also a topological space, and  $\mathcal{A} = \mathcal{B}(X)$ , it follows that every continuous map  $f : X \rightarrow Y$  is a Borel map.

**Lemma 1.4.** *A map  $f$  from a measurable space  $(X, \mathcal{A})$  to  $[-\infty, \infty]$  is measurable if and only if*

$$[f > c] \in \mathcal{A}$$

*for all real  $c$ .*

The non-trivial direction in the lemma follows from the fact that  $(c, \infty] \in \mathcal{B}_f$  by hypothesis for all real  $c$ ; therefore, the  $\sigma$ -algebra  $\mathcal{B}_f$  contains the sets

$$\bigcup_{n=1}^{\infty} (b - 1/n, \infty]^c = \bigcup_{n=1}^{\infty} [-\infty, b - 1/n] = [-\infty, b)$$

and  $(a, b) = [-\infty, b) \cap (a, \infty]$  for every real  $a < b$ , and so contains all countable unions of ‘segments’ of the above type, that is, all open subsets of  $[-\infty, \infty]$ .

The sets  $[f > c]$  in the condition of Lemma 1.4 can be replaced by any of the sets  $[f \geq c]$ ,  $[f < c]$ , or  $[f \leq c]$  (for all real  $c$ ), respectively. The proofs are analogous.

For  $f : X \rightarrow [-\infty, \infty]$  measurable and  $\alpha$  real, the function  $\alpha f$  (defined pointwise, with the usual arithmetics  $\alpha \cdot \infty = \infty$  for  $\alpha > 0$ ,  $= 0$  for  $\alpha = 0$ , and  $= -\infty$  for  $\alpha < 0$ , and similarly for  $-\infty$ ) is measurable, because for all real  $c$ ,  $[\alpha f > c] = [f > c/\alpha]$  for  $\alpha > 0$ ,  $= [f < c/\alpha]$  for  $\alpha < 0$ , and  $\alpha f$  is constant for  $\alpha = 0$ .

If  $\{a_n\} \subset [-\infty, \infty]$ , one denotes the superior (inferior) limit, that is, the ‘largest’ (‘smallest’) limit point, of the sequence by  $\limsup a_n$  ( $\liminf a_n$ , respectively).

Let  $b_n := \sup_{k \geq n} a_k$ . Then  $\{b_n\}$  is a decreasing sequence, and therefore

$$\exists \lim_n b_n = \inf_n b_n.$$

Let  $\alpha := \limsup a_n$  and  $\beta = \lim b_n$ . For any given  $n \in \mathbb{N}$ ,  $a_k \leq b_n$  for all  $k \geq n$ , and therefore  $\alpha \leq b_n$ . Hence  $\alpha \leq \beta$ .

On the other hand, for any  $t > \alpha$ ,  $a_k > t$  for at most *finitely many* indices  $k$ . Therefore, there exists  $n_0$  such that  $a_k \leq t$  for all  $k \geq n_0$ , hence  $b_{n_0} \leq t$ . But then  $b_n \leq t$  for all  $n \geq n_0$  (because  $\{b_n\}$  is decreasing), and so  $\beta \leq t$ . Since  $t > \alpha$  was arbitrary, it follows that  $\beta \leq \alpha$ , and the conclusion  $\alpha = \beta$  follows. We showed

$$\limsup a_n = \lim_n \left( \sup_{k \geq n} a_k \right) = \inf_{n \in \mathbb{N}} \left( \sup_{k \geq n} a_k \right). \quad (1)$$

Similarly

$$\liminf a_n = \lim_n \left( \inf_{k \geq n} a_k \right) = \sup_{n \in \mathbb{N}} \left( \inf_{k \geq n} a_k \right). \quad (2)$$

**Lemma 1.5.** *Let  $\{f_n\}$  be a sequence of measurable  $[-\infty, \infty]$ -valued functions on the measurable space  $(X, \mathcal{A})$ . Then the functions  $\sup f_n$ ,  $\inf f_n$ ,  $\limsup f_n$ ,  $\liminf f_n$ , and  $\lim f_n$  (when it exists), all defined pointwise, are measurable.*

**Proof.** Let  $h = \sup f_n$ . Then for all real  $c$ ,

$$[h > c] = \bigcup_n [f_n > c] \in \mathcal{A},$$

so that  $h$  is measurable by Lemma 1.4.

As remarked above,  $-f_n = (-1)f_n$  are measurable, and therefore  $\inf f_n = -\sup(-f_n)$  is measurable.

The proof is completed by the relations (1), (2), and

$$\lim f_n = \limsup f_n = \liminf f_n,$$

when the second equality holds (i.e. if and only if  $\lim f_n$  exists).  $\square$

In particular, taking a sequence with  $f_k = f_n$  for all  $k > n$ , we see that  $\max\{f_1, \dots, f_n\}$  and  $\min\{f_1, \dots, f_n\}$  are measurable, when  $f_1, \dots, f_n$  are measurable functions into  $[-\infty, \infty]$ . For example, the *positive* (*negative*) parts  $f^+ := \max\{f, 0\}$  ( $f^- := -\min\{f, 0\}$ ) of a measurable function  $f : X \rightarrow [-\infty, \infty]$  are (non-negative) measurable functions. Note the decompositions

$$f = f^+ - f^-; \quad |f| = f^+ + f^-.$$

**Lemma 1.6.** *Let  $g : Y \rightarrow Z$  be a continuous function from the topological space  $Y$  to the topological space  $Z$ . If  $f : X \rightarrow Y$  is measurable on the measurable space  $(X, \mathcal{A})$ , then the composite function  $h(x) = g(f(x))$  is measurable.*

Indeed, for every open subset  $V$  of  $Z$ ,  $g^{-1}(V)$  is open in  $Y$  (by continuity of  $g$ ), and therefore

$$[h \in V] = [f \in g^{-1}(V)] \in \mathcal{A},$$

by measurability of  $f$ .

If

$$Y = \prod_{k=1}^n Y_k$$

is the product space of topological spaces  $Y_k$ , the projections  $p_k : Y \rightarrow Y_k$  are continuous. Therefore, if  $f : X \rightarrow Y$  is measurable, so are the ‘component functions’  $f_k(x) := p_k(f(x)) : X \rightarrow Y_k$  ( $k = 1, \dots, n$ ), by Lemma 1.6. Conversely, if the topologies on  $Y_k$  have countable bases (for all  $k$ ), a countable base for the topology of  $Y$  consists of sets of the form  $V = \prod_{k=1}^n V_k$  with  $V_k$  varying in a countable base for the topology of  $Y_k$  (for each  $k$ ). Now,

$$[f \in V] = \bigcap_{k=1}^n [f_k \in V_k] \in \mathcal{A}$$

if all  $f_k$  are measurable. Since every open  $W \subset Y$  is a countable union of sets of the above type,  $[f \in W] \in \mathcal{A}$ , and  $f$  is measurable. We proved:

**Lemma 1.7.** *Let  $Y$  be the cartesian product of topological spaces  $Y_1, \dots, Y_n$  with countable bases to their topologies. Let  $(X, \mathcal{A})$  be a measurable space. Then  $f : X \rightarrow Y$  is measurable iff the components  $f_k$  are measurable for all  $k$ .*

For example, if  $f_k : X \rightarrow \mathbb{C}$  are measurable for  $k = 1, \dots, n$ , then  $f := (f_1, \dots, f_n) : X \rightarrow \mathbb{C}^n$  is measurable, and since  $g(z_1, \dots, z_n) := \sum \alpha_k z_k$  ( $\alpha_k \in \mathbb{C}$ ) and  $h(z_1, \dots, z_n) = z_1 \dots z_n$  are continuous from  $\mathbb{C}^n$  to  $\mathbb{C}$ , it follows from Lemma 1.6 that (finite) linear combinations and products of complex measurable functions are measurable. Thus, the complex measurable functions form



an algebra over the complex field (similarly, the real measurable functions form an algebra over the real field), for the usual pointwise operations.

If  $f$  has values in  $\mathbb{R}, [-\infty, \infty]$ , or  $\mathbb{C}$ , its measurability implies that of  $|f|$ , by Lemma 1.6.

By Lemma 1.7, a complex function is measurable iff its real part  $\Re f$  and imaginary part  $\Im f$  are both measurable.

If  $f, g$  are measurable with values in  $[0, \infty]$ , the functions  $f+g$  and  $fg$  are well-defined pointwise (with values in  $[0, \infty]$ ) and measurable, by the continuity of the functions  $(s, t) \rightarrow s+t$  and  $(s, t) \rightarrow st$  from  $[0, \infty]^2$  to  $[0, \infty]$  and Lemma 1.7.

The function  $f : X \rightarrow \mathbb{C}$  is *simple* if its range is a *finite* set  $\{c_1, \dots, c_n\} \subset \mathbb{C}$ . Let  $E_k := [f = c_k]$ ,  $k = 1, \dots, n$ . Then  $X$  is the disjoint union of the sets  $E_k$ , and

$$f = \sum_{k=1}^n c_k I_{E_k},$$

where  $I_E$  denotes the *indicator* of  $E$  (also called the *characteristic function* of  $E$  by non-probabilists, while probabilists reserve the later name to a different concept):

$$I_E(x) = 1 \quad \text{for } x \in E \quad \text{and} \quad = 0 \quad \text{for } x \in E^c.$$

Since a singleton  $\{c\} \subset \mathbb{C}$  is closed, it is a Borel set. Suppose now that the simple (complex) function  $f$  is defined on a measurable space  $(X, \mathcal{A})$ . If  $f$  is measurable, then  $E_k := [f = c_k]$  is measurable for all  $k = 1, \dots, n$ . Conversely, if all  $E_k$  are measurable, then for each open  $V \subset \mathbb{C}$ ,

$$[f \in V] = \bigcup_{\{k; c_k \in V\}} E_k \in \mathcal{A},$$

so that  $f$  is measurable. In particular, an indicator  $I_E$  is measurable iff  $E \in \mathcal{A}$ .

Let  $B(X, \mathcal{A})$  denote the complex algebra of all *bounded* complex  $\mathcal{A}$ -measurable functions on  $X$  (for the pointwise operations), and denote

$$\|f\| = \sup_X |f| \quad (f \in B(X, \mathcal{A})).$$

The map  $f \rightarrow \|f\|$  of  $B(X, \mathcal{A})$  into  $[0, \infty)$  has the following properties:

- (1)  $\|f\| = 0$  iff  $f = 0$  (the zero function);
- (2)  $\|\alpha f\| = |\alpha| \|f\|$  for all  $\alpha \in \mathbb{C}$  and  $f \in B(X, \mathcal{A})$ ;
- (3)  $\|f + g\| \leq \|f\| + \|g\|$  for all  $f, g \in B(X, \mathcal{A})$ ;
- (4)  $\|fg\| \leq \|f\| \|g\|$  for all  $f, g \in B(X, \mathcal{A})$ .

For example, (3) is verified by observing that for all  $x \in X$ ,

$$|f(x) + g(x)| \leq |f(x)| + |g(x)| \leq \sup_X |f| + \sup_X |g|.$$

A map  $\|\cdot\|$  from any (complex) vector space  $Z$  to  $[0, \infty)$  with Properties (1)–(3) is called a *norm* on  $Z$ . The above example is the *supremum norm* or *uniform*

norm on the vector space  $Z = B(X, \mathcal{A})$ . Property (1) is the *definiteness* of the norm; Property (2) is its *homogeneity*; Property (3) is the *triangle inequality*. A vector space with a specified norm is a *normed space*. If  $Z$  is an algebra, and the specified norm satisfies Property (4) also,  $Z$  is called a *normed algebra*. Thus,  $B(X, \mathcal{A})$  is a normed algebra with respect to the supremum norm. Any normed space  $Z$  is a metric space for the metric *induced by the norm*

$$d(u, v) := \|u - v\| \quad u, v \in Z.$$

Convergence in  $Z$  is convergence with respect to this metric (unless stated otherwise). Thus, convergence in the normed space  $B(X, \mathcal{A})$  is precisely *uniform convergence* on  $X$  (this explains the name ‘uniform norm’).

If  $x, y \in Z$ , the triangle inequality implies  $\|x\| = \|(x - y) + y\| \leq \|x - y\| + \|y\|$ , so that  $\|x\| - \|y\| \leq \|x - y\|$ . Since we may interchange  $x$  and  $y$ , we have

$$|\|x\| - \|y\|| \leq \|x - y\|.$$

In particular, *the norm function is continuous on  $Z$* .

The simple functions in  $B(X, \mathcal{A})$  form a subalgebra  $B_0(X, \mathcal{A})$ ; it is *dense* in  $B(X, \mathcal{A})$ :

**Theorem 1.8 (Approximation theorem).** *Let  $(X, \mathcal{A})$  be a measurable space. Then:*

- (1)  $B_0(X, \mathcal{A})$  is dense in  $B(X, \mathcal{A})$  (i.e. every bounded complex measurable function is the uniform limit of a sequence of simple measurable complex functions).
- (2) If  $f : X \rightarrow [0, \infty]$  is measurable, then there exists a sequence of measurable simple functions

$$0 \leq \phi_1 \leq \phi_2 \leq \cdots \leq f,$$

such that  $f = \lim \phi_n$ .

**Proof.** (1) Since any  $f \in B(X, \mathcal{A})$  can be written as

$$f = u^+ - u^- + iv^+ - iv^-$$

with  $u = \Re f$  and  $v = \Im f$ , it suffices to prove (1) for  $f$  with range in  $[0, \infty)$ . Let  $N$  be the first integer such that  $N > \sup f$ . For  $n = 1, 2, \dots$ , set

$$\phi_n := \sum_{k=1}^{N2^n} \frac{k-1}{2^n} I_{E_{n,k}},$$

where

$$E_{n,k} := f^{-1} \left( \left[ \frac{k-1}{2^n}, \frac{k}{2^n} \right) \right).$$

The simple functions  $\phi_n$  are measurable,

$$0 \leq \phi_1 \leq \phi_2 \leq \cdots \leq f,$$

and

$$0 \leq f - \phi_n < \frac{1}{2^n},$$

so that indeed  $\|f - \phi_n\| \leq (1/2^n)$ , as wanted.

If  $f$  has range in  $[0, \infty]$ , set

$$\phi_n := \sum_{k=1}^{n2^n} \frac{k-1}{2^n} I_{E_{n,k}} + n I_{F_n},$$

where  $F_n := [f \geq n]$ . Again  $\{\phi_n\}$  is a non-decreasing sequence of non-negative measurable simple functions  $\leq f$ . If  $f(x) = \infty$  for some  $x \in X$ , then  $x \in F_n$  for all  $n$ , and therefore  $\phi_n(x) = n$  for all  $n$ ; hence  $\lim_n \phi_n(x) = \infty = f(x)$ . If  $f(x) < \infty$  for some  $x$ , let  $n > f(x)$ . Then there exists a unique  $k, 1 \leq k \leq n2^n$ , such that  $x \in E_{n,k}$ . Then  $\phi_n(x) = ((k-1)/2^n)$  while  $((k-1)/2^n) \leq f(x) < (k/2^n)$ , so that

$$0 \leq f(x) - \phi_n(x) < 1/2^n \quad (n > f(x)).$$

Hence  $f(x) = \lim_n \phi_n(x)$  for all  $x \in X$ . □

## 1.2 Positive measures

**Definition 1.9.** Let  $(X, \mathcal{A})$  be a measurable space. A (*positive*) *measure* on  $\mathcal{A}$  is a function

$$\mu : \mathcal{A} \rightarrow [0, \infty]$$

such that  $\mu(\emptyset) = 0$  and

$$\mu\left(\bigcup_{k=1}^{\infty} E_k\right) = \sum_{k=1}^{\infty} \mu(E_k) \tag{1}$$

for any sequence of mutually disjoint sets  $E_k \in \mathcal{A}$ . Property (1) is called  *$\sigma$ -additivity* of the function  $\mu$ . The ordered triple  $(X, \mathcal{A}, \mu)$  will be called a (*positive*) *measure space*.

Taking in particular  $E_k = \emptyset$  for all  $k > n$ , it follows that

$$\mu\left(\bigcup_{k=1}^n E_k\right) = \sum_{k=1}^n \mu(E_k) \tag{2}$$

for any *finite* collection of mutually disjoint sets  $E_k \in \mathcal{A}, k = 1, \dots, n$ . We refer to Property (2) by saying that  $\mu$  is (*finitely*) *additive*.

Any finitely additive function  $\mu \geq 0$  on an algebra  $\mathcal{A}$  is necessarily *monotonic*, that is,  $\mu(E) \leq \mu(F)$  when  $E \subset F (E, F \in \mathcal{A})$ ; indeed

$$\mu(F) = \mu(E \cup (F - E)) = \mu(E) + \mu(F - E) \geq \mu(E).$$

If  $\mu(E) < \infty$ , we get

$$\mu(F - E) = \mu(F) - \mu(E).$$

**Lemma 1.10.** *Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let*

$$E_1 \subset E_2 \subset E_3 \subset \cdots$$

*be measurable sets with union  $E$ . Then*

$$\mu(E) = \lim_n \mu(E_n).$$

**Proof.** The sets  $E_n$  and  $E$  can be written as *disjoint* unions

$$\begin{aligned} E_n &= E_1 \cup (E_2 - E_1) \cup (E_3 - E_2) \cup \cdots \cup (E_n - E_{n-1}), \\ E &= E_1 \cup (E_2 - E_1) \cup (E_3 - E_2) \cup \cdots, \end{aligned}$$

where all differences belong to  $\mathcal{A}$ . Set  $E_0 = \emptyset$ . By  $\sigma$ -additivity,

$$\begin{aligned} \mu(E) &= \sum_{k=1}^{\infty} \mu(E_k - E_{k-1}) \\ &= \lim_n \sum_{k=1}^n \mu(E_k - E_{k-1}) = \lim_n \mu(E_n). \end{aligned}$$

□

In general, if  $E_j$  belong to an *algebra*  $\mathcal{A}$  of subsets of  $X$ , set  $A_0 = \emptyset$  and  $A_n = \bigcup_{j=1}^n E_j$ ,  $n = 1, 2, \dots$ . The sets  $A_j - A_{j-1}$ ,  $1 \leq j \leq n$ , are disjoint  $\mathcal{A}$ -measurable subsets of  $E_j$  with union  $A_n$ . If  $\mu$  is a non-negative *additive* set function on  $\mathcal{A}$ , then

$$\mu\left(\bigcup_{j=1}^n E_j\right) = \mu(A_n) = \sum_{j=1}^n \mu(A_j - A_{j-1}) \leq \sum_{j=1}^n \mu(E_j). \quad (*)$$

This is the *subadditivity property* of non-negative additive set functions (on algebras).

If  $\mathcal{A}$  is a  $\sigma$ -algebra and  $\mu$  is a positive measure on  $\mathcal{A}$ , then since  $A_1 \subset A_2 \subset \cdots$  and  $\bigcup_{n=1}^{\infty} A_n = \bigcup_{j=1}^{\infty} E_j$ , letting  $n \rightarrow \infty$  in (\*), it follows from Lemma 1.10 that

$$\mu\left(\bigcup_{j=1}^{\infty} E_j\right) \leq \sum_{j=1}^{\infty} \mu(E_j).$$

This property of positive measures is called  *$\sigma$ -subadditivity*.

For *decreasing* sequences of measurable sets, the ‘dual’ of Lemma 1.10 is false in general, unless we assume that the sets have *finite* measure:

**Lemma 1.11.** *Let  $\{E_k\} \subset \mathcal{A}$  be a decreasing sequence (with respect to set-inclusion) such that  $\mu(E_1) < \infty$ . Let  $E = \bigcap_k E_k$ . Then*

$$\mu(E) = \lim_n \mu(E_n).$$

**Proof.** The sequence  $\{E_1 - E_k\}$  is increasing, with union  $E_1 - E$ . By Lemma 1.10 and the *finiteness* of the measures of  $E$  and  $E_k$  (subsets of  $E_1$ !),

$$\begin{aligned} \mu(E_1) - \mu(E) &= \mu\left(\bigcup_k (E_1 - E_k)\right) \\ &= \lim \mu(E_1 - E_n) = \mu(E_1) - \lim \mu(E_n), \end{aligned}$$

and the result follows by cancelling the *finite* number  $\mu(E_1)$ .  $\square$

If  $\{E_k\}$  is an *arbitrary* sequence of subsets of  $X$ , set  $F_n = \bigcap_{k \geq n} E_k$  and  $G_n = \bigcup_{k \geq n} E_k$ . Then  $\{F_n\}$  ( $\{G_n\}$ ) is increasing (decreasing, respectively), and  $F_n \subset E_n \subset G_n$  for all  $n$ .

One defines

$$\liminf_n E_n := \bigcup_n F_n; \quad \limsup_n E_n := \bigcap_n G_n.$$

These sets belong to  $\mathcal{A}$  if  $E_k \in \mathcal{A}$  for all  $k$ . The set  $\liminf E_n$  consists of all  $x$  that *belong to  $E_n$  for all but finitely many  $n$* ; the set  $\limsup E_n$  consists of all  $x$  that *belong to  $E_n$  for infinitely many  $n$* . By Lemma 1.10,

$$\mu(\liminf E_n) = \lim_n \mu(F_n) \leq \liminf \mu(E_n). \quad (3)$$

If the measure of  $G_1$  is finite, we also have by Lemma 1.11

$$\mu(\limsup E_n) = \lim_n \mu(G_n) \geq \limsup \mu(E_n). \quad (4)$$

## 1.3 Integration of non-negative measurable functions

**Definition 1.12.** Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and  $\phi : X \rightarrow [0, \infty)$  a measurable simple function. The *integral over  $X$  of  $\phi$  with respect to  $\mu$* , denoted

$$\int_X \phi d\mu$$

or briefly

$$\int \phi d\mu,$$

is the finite sum

$$\sum_k c_k \mu(E_k) \in [0, \infty],$$

where

$$\phi = \sum_k c_k I_{E_k}, \quad E_k = [\phi = c_k],$$

and  $c_k$  are the distinct values of  $\phi$ .

Note that

$$\int I_E d\mu = \mu(E) \quad E \in \mathcal{A}$$

and

$$0 \leq \int \phi d\mu \leq \|\phi\| \mu([\phi \neq 0]). \quad (1)$$

For an *arbitrary* measurable function  $f : X \rightarrow [0, \infty]$ , consider the (non-empty) set  $S_f$  of measurable simple functions  $\phi$  such that  $0 \leq \phi \leq f$ , and define

$$\int f d\mu := \sup_{\phi \in S_f} \int \phi d\mu. \quad (2)$$

For any  $E \in \mathcal{A}$ , the integral *over*  $E$  of  $f$  is defined by

$$\int_E f d\mu := \int f I_E d\mu. \quad (3)$$

Let  $\phi, \psi$  be measurable simple functions; let  $c_k, d_j$  be the distinct values of  $\phi$  and  $\psi$ , taken on the (mutually disjoint) sets  $E_k$  and  $F_j$ , respectively. Denote  $Q := \{(k, j) \in \mathbb{N}^2; E_k \cap F_j \neq \emptyset\}$ .

If  $\phi \leq \psi$ , then  $c_k \leq d_j$  for  $(k, j) \in Q$ . Hence

$$\begin{aligned} \int \phi d\mu &= \sum_k c_k \mu(E_k) = \sum_{(k,j) \in Q} c_k \mu(E_k \cap F_j) \\ &\leq \sum_{(k,j) \in Q} d_j \mu(E_k \cap F_j) = \sum_j d_j \mu(F_j) = \int \psi d\mu. \end{aligned}$$

Thus, the integral is *monotonic on simple functions*.

If  $f$  is simple, then  $\int \phi d\mu \leq \int f d\mu$  for all  $\phi \in S_f$  (by monotonicity of the integral on simple functions), and therefore the supremum in (2) is less than or equal to the integral of  $f$  as a simple function; since  $f \in S_f$ , the reverse inequality is trivial, so that the two definitions of the integral of  $f$  coincide for  $f$  simple.

Since  $S_{cf} = cS_f := \{c\phi; \phi \in S_f\}$  for  $0 \leq c < \infty$ , we have (for  $f$  as above)

$$\int cf d\mu = c \int f d\mu \quad (0 \leq c < \infty). \quad (4)$$

If  $f \leq g$  ( $f, g$  as above),  $S_f \subset S_g$ , and therefore  $\int f d\mu \leq \int g d\mu$  (monotonicity of the integral with respect to the ‘integrand’).

In particular, if  $E \subset F$  (both measurable), then  $fI_E \leq fI_F$ , and therefore  $\int_E f d\mu \leq \int_F f d\mu$  (monotonicity of the integral with respect to the set of integration).

If  $\mu(E) = 0$ , then any  $\phi \in S_{fI_E}$  assumes its non-zero values  $c_k$  on the sets  $E_k \cap E$ , that have measure 0 (as measurable subsets of  $E$ ), and therefore  $\int \phi d\mu = 0$  for all such  $\phi$ , hence  $\int_E f d\mu = 0$ .

If  $f = 0$  on  $E$  (for some  $E \in \mathcal{A}$ ), then  $fI_E$  is the zero function, hence has zero integral (by definition of the integral of simple functions!); this means that  $\int_E f d\mu = 0$  when  $f = 0$  on  $E$ .

Consider now the set function

$$\nu(E) := \int_E \phi d\mu \quad E \in \mathcal{A}, \quad (5)$$

for a *fixed* simple measurable function  $\phi \geq 0$ . As a special case of the preceding remark,  $\nu(\emptyset) = 0$ . Write  $\phi = \sum c_k I_{E_k}$ , and let  $A_j \in \mathcal{A}$  be mutually disjoint ( $j = 1, 2, \dots$ ) with union  $A$ . Then

$$\phi I_A = \sum c_k I_{E_k \cap A},$$

so that, by the  $\sigma$ -additivity of  $\mu$  and the possibility of interchanging summation order when the summands are non-negative,

$$\begin{aligned} \nu(A) &:= \sum_k c_k \mu(E_k \cap A) = \sum_k c_k \sum_j \mu(E_k \cap A_j) \\ &= \sum_j \sum_k c_k \mu(E_k \cap A_j) = \sum_j \nu(A_j). \end{aligned}$$

Thus  $\nu$  is a positive measure. This is actually true for *any* measurable  $\phi \geq 0$  (not necessarily simple), but this will be proved later.

If  $\psi, \chi$  are *simple* functions as above (the distinct values of  $\psi$  and  $\chi$  being  $a_1, \dots, a_p$  and  $b_1, \dots, b_q$ , assumed on the measurable sets  $F_1, \dots, F_p$  and  $G_1, \dots, G_q$ , respectively), then the simple measurable function  $\phi := \psi + \chi$  assumes the constant value  $a_i + b_j$  on the set  $F_i \cap G_j$ , and therefore, defining the measure  $\nu$  as above, we have

$$\nu(F_i \cap G_j) = (a_i + b_j) \mu(F_i \cap G_j). \quad (6)$$

But  $a_i$  and  $b_j$  are the constant values of  $\psi$  and  $\chi$  on the set  $F_i \cap G_j$  (respectively), so that the right-hand side of (6) equals  $\nu'(F_i \cap G_j) + \nu''(F_i \cap G_j)$ , where  $\nu'$  and  $\nu''$  are the measures defined as  $\nu$ , with the integrands  $\psi$  and  $\chi$  instead of  $\phi$ . Summing over all  $i, j$ , since  $X$  is the disjoint union of the sets  $F_i \cap G_j$ , the additivity of the measures  $\nu, \nu'$ , and  $\nu''$  implies that  $\nu(X) = \nu'(X) + \nu''(X)$ , that is,

$$\int (\psi + \chi) d\mu = \int \psi d\mu + \int \chi d\mu \quad (7)$$

Property (7) is the additivity of the integral over non-negative measurable simple functions. This property too will be extended later to *arbitrary* non-negative measurable functions.

**Theorem 1.13.** *Let  $(X, \mathcal{A}, \mu)$  be a positive measure space. Let*

$$f_1 \leq f_2 \leq f_3 \leq \cdots : X \rightarrow [0, \infty]$$

*be measurable, and denote  $f = \lim f_n$  (defined pointwise). Then*

$$\int f \, d\mu = \lim \int f_n \, d\mu. \quad (8)$$

*This is the Monotone Convergence theorem of Lebesgue.*

**Proof.** By Lemma 1.5,  $f$  is measurable (with range in  $[0, \infty]$ ). The monotonicity of the integral (and the fact that  $f_n \leq f_{n+1} \leq f$ ) implies that

$$\int f_n \, d\mu \leq \int f_{n+1} \, d\mu \leq \int f \, d\mu,$$

and therefore the limit in (8) exists ( $:= c \in [0, \infty]$ ) and the *inequality  $\geq$*  holds in (8). It remains to show the inequality  $\leq$  in (8). Let  $0 < t < 1$ . Given  $\phi \in S_f$ , denote

$$A_n = [t\phi \leq f_n] = [f_n - t\phi \geq 0] \quad (n = 1, 2, \dots).$$

Then  $A_n \in \mathcal{A}$  and  $A_1 \subset A_2 \subset \cdots$  (because  $f_1 \leq f_2 \leq \cdots$ ). If  $x \in X$  is such that  $\phi(x) = 0$ , then  $x \in A_n$  (for all  $n$ ). If  $x \in X$  is such that  $\phi(x) > 0$ , then  $f(x) \geq \phi(x) > t\phi(x)$ , and there exists therefore  $n$  for which  $f_n(x) \geq t\phi(x)$ , that is,  $x \in A_n$  (for that  $n$ ). This shows that  $\bigcup_n A_n = X$ . Consider the measure  $\nu$  defined by (5) (for the simple function  $t\phi$ ). By Lemma 1.10,

$$t \int \phi \, d\mu = \nu(X) = \lim_n \nu(A_n) = \lim_n \int_{A_n} t\phi \, d\mu.$$

However  $t\phi \leq f_n$  on  $A_n$ , so the integrals on the right are  $\leq \int_{A_n} f_n \, d\mu \leq \int_X f_n \, d\mu$  (by the monotonicity property of integrals with respect to the set of integration). Therefore  $t \int \phi \, d\mu \leq c$ , and so  $\int \phi \, d\mu \leq c$  by the arbitrariness of  $t \in (0, 1)$ . Taking the supremum over all  $\phi \in S_f$ , we conclude that  $\int f \, d\mu \leq c$  as wanted.  $\square$

For *arbitrary* sequences of non-negative measurable functions we have the following *inequality*:

**Theorem 1.14 (Fatou's lemma).** *Let  $f_n : X \rightarrow [0, \infty]$ ,  $n = 1, 2, \dots$ , be measurable. Then*

$$\int \liminf_n f_n \, d\mu \leq \liminf_n \int f_n \, d\mu.$$

**Proof.** We have

$$\liminf_n f_n := \lim_n \left( \inf_{k \geq n} f_k \right).$$



Denote the infimum on the right by  $g_n$ . Then  $g_n, n = 1, 2, \dots$ , are measurable,  $g_n \leq f_n$ ,

$$0 \leq g_1 \leq g_2 \leq \dots,$$

and  $\lim_n g_n = \liminf_n f_n$ . By Theorem 1.13,

$$\int \liminf_n f_n d\mu = \int \lim g_n d\mu = \lim \int g_n d\mu.$$

But the integrals on the right are  $\leq \int f_n d\mu$ , therefore their limit is  $\leq \liminf \int f_n d\mu$ .  $\square$

Another consequence of Theorem 1.13 is the additivity of the integral of non-negative measurable functions.

**Theorem 1.15.** *Let  $f, g : X \rightarrow [0, \infty]$  be measurable. Then*

$$\int (f + g) d\mu = \int f d\mu + \int g d\mu.$$

**Proof.** By the Approximation theorem (Theorem 1.8), there exist simple measurable functions  $\phi_n, \psi_n$  such that

$$0 \leq \phi_1 \leq \phi_2 \leq \dots, \quad \lim \phi_n = f,$$

$$0 \leq \psi_1 \leq \psi_2 \leq \dots, \quad \lim \psi_n = g.$$

Then the measurable simple functions  $\chi_n = \phi_n + \psi_n$  satisfy

$$0 \leq \chi_1 \leq \chi_2 \leq \dots, \quad \lim \chi_n = f + g.$$

By Theorem 1.13 and the additivity of the integral of (non-negative measurable) simple functions (cf. (7)), we have

$$\begin{aligned} \int (f + g) d\mu &= \lim \int \chi_n d\mu = \lim \int (\phi_n + \psi_n) d\mu \\ &= \lim \int \phi_n d\mu + \lim \int \psi_n d\mu = \int f d\mu + \int g d\mu. \end{aligned}$$

$\square$

The additivity property of the integral is also true for *infinite* sums of non-negative measurable functions:

**Theorem 1.16 (Beppo Levi).** *Let  $f_n : X \rightarrow [0, \infty], n = 1, 2, \dots$ , be measurable. Then*

$$\int \sum_{n=1}^{\infty} f_n d\mu = \sum_{n=1}^{\infty} \int f_n d\mu.$$

**Proof.** Let

$$g_k = \sum_{n=1}^k f_n; \quad g = \sum_{n=1}^{\infty} f_n.$$

The measurable functions  $g_k$  satisfy

$$0 \leq g_1 \leq g_2 \leq \dots, \quad \lim g_k = g,$$

and by Theorem 1.15 (and induction)

$$\int g_k d\mu = \sum_{n=1}^k \int f_n d\mu.$$

Therefore, by Theorem 1.13

$$\int g d\mu = \lim_k \int g_k d\mu = \lim_k \sum_{n=1}^k \int f_n d\mu = \sum_{n=1}^{\infty} \int f_n d\mu.$$

□

We may extend now the measure property of  $\nu$ , defined earlier with a *simple* integrand, to the general case of a non-negative measurable integrand:

**Theorem 1.17.** *Let  $f : X \rightarrow [0, \infty]$  be measurable, and set*

$$\nu(E) := \int_E f d\mu, \quad E \in \mathcal{A}.$$

*Then  $\nu$  is a (positive) measure on  $\mathcal{A}$ , and for any measurable  $g : X \rightarrow [0, \infty]$ ,*

$$\int g d\nu = \int gf d\mu. \quad (*)$$

**Proof.** Let  $E_j \in \mathcal{A}, j = 1, 2, \dots$  be mutually disjoint, with union  $E$ . Then

$$f I_E = \sum_{j=1}^{\infty} f I_{E_j},$$

and therefore, by Theorem 1.16,

$$\nu(E) := \int f I_E d\mu = \sum_j \int f I_{E_j} d\mu = \sum_j \nu(E_j).$$

Thus  $\nu$  is a measure.

If  $g = I_E$  for some  $E \in \mathcal{A}$ , then

$$\int g d\nu = \nu(E) = \int I_E f d\mu = \int gf d\mu.$$

By (4) and Theorem 1.15 (for the measures  $\mu$  and  $\nu$ ), (\*) is valid for  $g$  *simple*. Finally, for general  $g$ , the Approximation theorem (Theorem 1.8) provides a sequence of simple measurable functions

$$0 \leq \phi_1 \leq \phi_2 \leq \dots; \quad \lim \phi_n = g.$$

Then the measurable functions  $\phi_n f$  satisfy

$$0 \leq \phi_1 f \leq \phi_2 f \leq \dots; \quad \lim \phi_n f = gf,$$

and Theorem 1.13 implies that

$$\int g \, d\nu = \lim_n \int \phi_n \, d\nu = \lim_n \int \phi_n f \, d\mu = \int g f \, d\mu.$$

□

Relation (\*) is conveniently abbreviated as

$$d\nu = f \, d\mu.$$

Observe that if  $f_1$  and  $f_2$  coincide *almost everywhere* (briefly, ‘a.e.’ or  $\mu$ -a.e., if the measure needs to be specified), that is, if they coincide except on a *null set*  $A \in \mathcal{A}$  (more precisely, a  $\mu$ -null set, that is, a measurable set  $A$  such that  $\mu(A) = 0$ ), then the corresponding measures  $\nu_i$  are equal, and in particular  $\int f_1 \, d\mu = \int f_2 \, d\mu$ . Indeed, for all  $E \in \mathcal{A}$ ,  $\mu(E \cap A) = 0$ , and therefore

$$\nu_i(E \cap A) = \int_{E \cap A} f_i \, d\mu = 0, \quad i = 1, 2$$

by one of the observations following Definition 1.12. Hence

$$\begin{aligned} \nu_1(E) &= \nu_1(E \cap A) + \nu_1(E \cap A^c) = \nu_1(E \cap A^c) \\ &= \nu_2(E \cap A^c) = \nu_2(E). \end{aligned}$$

## 1.4 Integrable functions

Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $f$  be a measurable function with range in  $[-\infty, \infty]$  or  $\overline{\mathbb{C}} := \mathbb{C} \cup \{\infty\}$  (the Riemann sphere). Then  $|f| : X \rightarrow [0, \infty]$  is measurable, and has therefore an integral ( $\in [0, \infty]$ ). In case this integral is *finite*, we shall say that  $f$  is *integrable*. In that case, the measurable set  $[|f| = \infty]$  has measure zero. Indeed, it is contained in  $[|f| > n]$  for all  $n = 1, 2, \dots$ , and

$$n\mu([|f| > n]) = \int_{[|f| > n]} n \, d\mu \leq \int_{[|f| > n]} |f| \, d\mu \leq \int |f| \, d\mu.$$

Hence for all  $n$

$$0 \leq \mu([|f| = \infty]) \leq \frac{1}{n} \int |f| \, d\mu,$$

and since the integral on the right is finite, we must have  $\mu([|f| = \infty]) = 0$ .

In other words, an integrable function is *finite a.e.*

We observed above that non-negative measurable functions that coincide a.e. have equal integrals. This property is desirable in the general case now considered. If  $f$  is measurable, and if we redefine it as the finite arbitrary constant  $c$  on a set  $A \in \mathcal{A}$  of measure zero, then the new function  $g$  is also measurable. Indeed, for any open set  $V$  in the range space,

$$[g \in V] = \{[g \in V] \cap A^c\} \cup \{[g \in V] \cap A\}.$$

The second set on the right is empty if  $c \in V^c$ , and is  $A$  if  $c \in V$ , thus belongs to  $\mathcal{A}$  in any case. The first set on the right is equal to  $[f \in V] \cap A^c \in \mathcal{A}$ , by the measurability of  $f$ . Thus  $[g \in V] \in \mathcal{A}$ .

If  $f$  is integrable, we can redefine it as an arbitrary *finite* constant on the set  $[|f| = \infty]$  (that has measure zero) and obtain a new *finite-valued* measurable function, whose integral should be the same as the integral of  $f$  (by the ‘desirable’ property mentioned before). This discussion shows that we may restrict ourselves to *complex* (or, as a special case, to *real*) valued measurable functions.

**Definition 1.18.** Let  $(X, \mathcal{A}, \mu)$  be a positive measure space. The function  $f : X \rightarrow \mathbb{C}$  is *integrable* if it is measurable and

$$\|f\|_1 := \int |f| d\mu < \infty.$$

The set of all (complex) *integrable* functions will be denoted by

$$L^1(X, \mathcal{A}, \mu),$$

or briefly by  $L^1(\mu)$  or  $L^1(X)$  or  $L^1$ , when the unmentioned ‘objects’ of the measure space are understood.

Defining the operations pointwise,  $L^1$  is a complex vector space, since the inequality

$$|\alpha f + \beta g| \leq |\alpha| |f| + |\beta| |g|$$

implies, by monotonicity, additivity, and homogeneity of the integral of non-negative measurable functions:

$$\|\alpha f + \beta g\|_1 \leq |\alpha| \|f\|_1 + |\beta| \|g\|_1 < \infty,$$

for all  $f, g \in L^1$  and  $\alpha, \beta \in \mathbb{C}$ .

In particular,  $\|\cdot\|_1$  satisfies the triangle inequality (take  $\alpha = \beta = 1$ ), and is trivially homogeneous.

Suppose  $\|f\|_1 = 0$ . For any  $n = 1, 2, \dots$ ,

$$\begin{aligned} 0 \leq \mu([|f| > 1/n]) &= \int_{[|f| > 1/n]} d\mu = n \int_{[|f| > 1/n]} (1/n) d\mu \\ &\leq n \int_{[|f| > 1/n]} |f| d\mu \leq n \|f\|_1 = 0, \end{aligned}$$

so  $\mu([|f| > 1/n]) = 0$ . Now the set where  $f$  is *not* zero is

$$[|f| > 0] = \bigcup_{n=1}^{\infty} [|f| > 1/n],$$

and by the  $\sigma$ -subadditivity property of positive measures, it follows that this set has measure zero. Thus, the vanishing of  $\|f\|_1$  implies that  $f = 0$  a.e. (the converse is trivially true). One verifies easily that the relation ‘ $f = g$  a.e.’ is

an equivalence relation for complex measurable functions (transitivity follows from the fact that the union of two sets of measure zero has measure zero, by subadditivity of positive measures). All the functions  $f$  in the same equivalence class have the same value of  $\|f\|_1$  (cf. discussion following Theorem 1.17).

We use the same notation  $L^1$  for the space of all *equivalence classes* of integrable functions, with operations performed as usual on representatives of the classes, and with the  $\|\cdot\|_1$ -norm of a class equal to the norm of any of its representatives;  $L^1$  is a normed space (for the norm  $\|\cdot\|_1$ ). It is customary, however, to think of the elements of  $L^1$  as functions (rather than equivalence classes of functions!).

If  $f \in L^1$ , then  $f = u + iv$  with  $u := \Re f$  and  $v := \Im f$  real measurable functions (cf. discussion following Lemma 1.7), and since  $|u|, |v| \leq |f|$ , we have  $\|u\|_1, \|v\|_1 \leq \|f\|_1 < \infty$ , that is,  $u, v$  are *real* elements of  $L^1$  (conversely, if  $u, v$  are real elements of  $L^1$ , then  $f = u + iv \in L^1$ , since  $L^1$  is a complex vector space).

Writing  $u = u^+ - u^-$  (and similarly for  $v$ ), we obtain four non-negative (finite) measurable functions (cf. remarks following Lemma 1.5), and since  $u^+ \leq |u| \leq |f|$  (and similarly for  $u^-$ , etc.), they have *finite integrals*. It makes sense therefore to *define*

$$\int u \, d\mu := \int u^+ \, d\mu - \int u^- \, d\mu$$

(on the right, one has the difference of two *finite* non-negative real numbers!).

Doing the same with  $v$ , we then let

$$\int f \, d\mu := \int u \, d\mu + i \int v \, d\mu.$$

Note that according to this definition,

$$\Re \int f \, d\mu = \int \Re f \, d\mu,$$

and similarly for the imaginary part.

**Theorem 1.19.** *The map  $f \rightarrow \int f \, d\mu \in \mathbb{C}$  is a continuous linear functional on the normed space  $L^1(\mu)$ .*

**Proof.** Consider first real-valued functions  $f, g \in L^1$ . Let  $h = f + g$ . Then

$$h^+ - h^- = (f^+ - f^-) + (g^+ - g^-),$$

and since all functions above have *finite* values,

$$h^+ + f^- + g^- = h^- + f^+ + g^+.$$

By Theorem 1.15,

$$\int h^+ \, d\mu + \int f^- \, d\mu + \int g^- \, d\mu = \int h^- \, d\mu + \int f^+ \, d\mu + \int g^+ \, d\mu.$$

All integrals above are finite, so we may subtract  $\int h^- + \int f^- + \int g^-$  from both sides of the equation. This yields:

$$\int h \, d\mu = \int f \, d\mu + \int g \, d\mu.$$

The additivity of the integral extends trivially to *complex* functions in  $L^1$ .

If  $f \in L^1$  is real and  $c \in [0, \infty)$ ,  $(cf)^+ = cf^+$  and similarly for  $f^-$ . Therefore, by (4) (following Definition 1.12),

$$\int cf \, d\mu = \int cf^+ \, d\mu - \int cf^- \, d\mu = c \int f^+ \, d\mu - c \int f^- \, d\mu = c \int f \, d\mu.$$

If  $c \in (-\infty, 0)$ ,  $(cf)^+ = -cf^-$  and  $(cf)^- = -cf^+$ , and a similar calculation shows again that  $\int(cf) = c \int f$ . For  $f \in L^1$  complex and  $c$  real, write  $f = u + iv$ . Then

$$\int cf = \int(cu + icv) := \int(cu) + i \int(cv) = c \left( \int u + i \int v \right) := c \int f.$$

Note next that

$$\int(if) = \int(-v + iu) = - \int v + i \int u = i \int f.$$

Finally, if  $c = a + ib$  ( $a, b$  real), then by additivity of the integral and the previous remarks,

$$\int(cf) = \int(af + ibf) = \int(af) + \int ibf = a \int f + ib \int f = c \int f.$$

Thus

$$\int(\alpha f + \beta g) \, d\mu = \alpha \int f \, d\mu + \beta \int g \, d\mu$$

for all  $f, g \in L^1$  and  $\alpha, \beta \in \mathbb{C}$ .

For  $f \in L^1$ , let  $\lambda := \int f \, d\mu \in \mathbb{C}$ . Then, since the left-hand side of the following equation is *real*,

$$\begin{aligned} |\lambda| &= e^{i\theta} \lambda = e^{i\theta} \int f \, d\mu = \int (e^{i\theta} f) \, d\mu = \Re \int (e^{i\theta} f) \, d\mu = \int \Re(e^{i\theta} f) \, d\mu \\ &\leq \int |e^{i\theta} f| \, d\mu = \int |f| \, d\mu. \end{aligned}$$

We thus obtained the important inequality

$$\left| \int f \, d\mu \right| \leq \int |f| \, d\mu. \quad (1)$$

If  $f, g \in L^1$ , it follows from the linearity of the integral and (1) that

$$\left| \int f \, d\mu - \int g \, d\mu \right| = \left| \int (f - g) \, d\mu \right| \leq \|f - g\|_1. \quad (2)$$

In particular, if  $f$  and  $g$  represent the same equivalence class, then  $\|f - g\|_1 = 0$ , and therefore  $\int f d\mu = \int g d\mu$ . This means that the functional  $f \rightarrow \int f d\mu$  is *well-defined* as a functional on the *normed space*  $L^1(\mu)$  (of equivalence classes!), and its continuity follows trivially from (2).  $\square$

In term of sequences, continuity of the integral on the normed space  $L^1$  means that if  $\{f_n\} \subset L^1$  converges to  $f$  in the  $L^1$ -metric, then

$$\int f_n d\mu \rightarrow \int f d\mu. \quad (3)$$

A useful sufficient condition for convergence in the  $L^1$ -metric, and therefore, for the validity of (3), is contained in the *Dominated Convergence theorem* of Lebesgue:

**Theorem 1.20.** *Let  $(X, \mathcal{A}, \mu)$  be a measure space. Let  $\{f_n\}$  be a sequence of complex measurable functions on  $X$  that converge pointwise to the function  $f$ . Suppose there exists  $g \in L^1(\mu)$  (with values in  $[0, \infty)$ ) such that*

$$|f_n| \leq g \quad (n = 1, 2, \dots). \quad (4)$$

*Then  $f, f_n \in L^1(\mu)$  for all  $n$ , and  $f_n \rightarrow f$  in the  $L^1(\mu)$ -metric.*

In particular, (3) is valid.

**Proof.** By Lemma 1.5,  $f$  is measurable. By (4) and monotonicity

$$\|f\|_1, \|f_n\|_1 \leq \|g\|_1 < \infty,$$

so that  $f, f_n \in L^1$ .

Since  $|f_n - f| \leq 2g$ , the measurable functions  $2g - |f_n - f|$  are non-negative. By Fatou's Lemma (Theorem 1.14),

$$\int \liminf_n (2g - |f_n - f|) d\mu \leq \liminf_n \int (2g - |f_n - f|) d\mu. \quad (5)$$

The left-hand side of (5) is  $\int 2g d\mu$ . The integral on the right-hand side is  $\int 2g d\mu + (-\|f_n - f\|_1)$ , and its  $\liminf$  is

$$= \int 2g d\mu + \liminf_n (-\|f_n - f\|_1) = \int 2g d\mu - \limsup_n \|f_n - f\|_1.$$

Subtracting the finite number  $\int 2g d\mu$  from both sides of the inequality, we obtain

$$\limsup_n \|f_n - f\|_1 \leq 0.$$

However if a non-negative sequence  $\{a_n\}$  satisfies  $\limsup a_n \leq 0$ , then it converges to 0 (because  $0 \leq \liminf a_n \leq \limsup a_n \leq 0$  implies  $\liminf a_n = \limsup a_n = 0$ ). Thus  $\|f_n - f\|_1 \rightarrow 0$ .  $\square$

Rather than assuming pointwise convergence of the sequence  $\{f_n\}$  at every point of  $X$ , we may assume that the sequence converges *almost everywhere*, that is,  $f_n \rightarrow f$  on a set  $E \in \mathcal{A}$  and  $\mu(E^c) = 0$ . The functions  $f_n$  could be defined only a.e., and we could include the countable union of all the sets where these functions are not defined (which is a set of measure zero, by the  $\sigma$ -subadditivity of measures) in the ‘exceptional’ set  $E^c$ . The limit function  $f$  is defined a.e., in any case. For such a function, measurability means that  $[f \in V] \cap E \in \mathcal{A}$  for each open set  $V$ .

If  $f_n$  (defined a.e.) converge pointwise a.e. to  $f$ , then with  $E$  as above, the restrictions  $f_n|_E$  are  $\mathcal{A}_E$ -measurable, where  $\mathcal{A}_E$  is the  $\sigma$ -algebra  $\mathcal{A} \cap E$ , because

$$[f_n|_E \in V] = [f_n \in V] \cap E \in \mathcal{A}_E.$$

By Lemma 1.5,  $f|_E := \lim f_n|_E$  is  $\mathcal{A}_E$ -measurable, and therefore the a.e.-defined function  $f$  is ‘measurable’ in the above sense. We may define  $f$  as an arbitrary constant  $c \in \mathbb{C}$  on  $E^c$ ; the function thus extended to  $X$  is  $\mathcal{A}$ -measurable, as seen by the argument preceding Definition 1.18.

Now  $f_n I_E$  are  $\mathcal{A}$ -measurable, converge pointwise everywhere to  $f I_E$ , and if  $|f_n| \leq g \in L^1$  for all  $n$  (wherever the functions are defined), then  $|f_n I_E| \leq g \in L^1$  (everywhere!). By Theorem 1.20,

$$\|f_n - f\|_1 = \|f_n I_E - f I_E\|_1 \rightarrow 0.$$

We then have the following *a.e. version* of the Lebesgue Dominated Convergence theorem:

**Theorem 1.21.** *Let  $\{f_n\}$  be a sequence of a.e.-defined measurable complex functions on  $X$ , converging a.e. to the function  $f$ . Let  $g \in L^1$  be such that  $|f_n| \leq g$  for all  $n$  (at all points where  $f_n$  is defined). Then  $f$  and  $f_n$  are in  $L^1$ , and  $f_n \rightarrow f$  in the  $L^1$ -metric (in particular,  $\int f_n d\mu \rightarrow \int f d\mu$ ).*

A useful ‘almost everywhere’ proposition is the following:

**Proposition 1.22.** *If  $f \in L^1(\mu)$  satisfies  $\int_E f d\mu = 0$  for every  $E \in \mathcal{A}$ , then  $f = 0$  a.e.*

**Proof.** Let  $E = [u := \Re f \geq 0]$ . Then  $E \in \mathcal{A}$ , so

$$\|u^+\|_1 = \int_E u d\mu = \Re \int_E f d\mu = 0,$$

and therefore  $u^+ = 0$  a.e. Similarly  $u^- = v^+ = v^- = 0$  a.e. (where  $v := \Im f$ ), so that  $f = 0$  a.e.  $\square$

We should remark that, in general, a measurable a.e.-defined function  $f$  can be extended as a measurable function on  $X$  only by defining it as *constant* on the exceptional null set  $E^c$ . Indeed, the null set  $E^c$  could have a *non-measurable* subset  $A$ . Suppose  $f : E \rightarrow \mathbb{C}$  is not onto, and let  $a \in f(E)^c$ . If we assign on  $A$  the (constant complex) value  $a$ , and any value  $b \in f(E)$  on  $E^c - A$ , then the extended function is not measurable, because  $[f = a] = A \notin \mathcal{A}$ .



In order to be able to extend  $f$  in an arbitrary fashion and always get a measurable function, it is sufficient that subsets of null sets should be measurable (recall that a ‘null set’ is measurable by definition!). A measure space with this property is called a *complete* measure space. Indeed, let  $f'$  be an arbitrary extension to  $X$  of an a.e.-defined measurable function  $f$ , defined on  $E \in \mathcal{A}$ , with  $E^c$  null. Then for any open  $V \subset \mathbb{C}$ ,

$$[f' \in V] = ([f' \in V] \cap E) \cup ([f' \in V] \cap E^c).$$

The first set in the union is in  $\mathcal{A}$ , by measurability of the a.e.-defined function  $f$ ; the second set is in  $\mathcal{A}$  as a subset of the null set  $E^c$  (by completeness of the measure space). Hence  $[f' \in V] \in \mathcal{A}$ , and  $f'$  is measurable.

We say that the measure space  $(X, \mathcal{M}, \nu)$  is an *extension* of the measure space  $(X, \mathcal{A}, \mu)$  (both on  $X$ !) if  $\mathcal{A} \subset \mathcal{M}$  and  $\nu = \mu$  on  $\mathcal{A}$ . It is important to know that any measure space  $(X, \mathcal{A}, \mu)$  has a (unique) ‘minimal’ complete extension  $(X, \mathcal{M}, \nu)$ , where minimality means that if  $(X, \mathcal{N}, \sigma)$  is any complete extension of  $(X, \mathcal{A}, \mu)$ , then it is an extension of  $(X, \mathcal{M}, \nu)$ . Uniqueness is of course trivial. The existence is proved below by a ‘canonical’ construction.

**Theorem 1.23.** *Any measure space  $(X, \mathcal{A}, \mu)$  has a unique minimal complete extension  $(X, \mathcal{M}, \nu)$  (called the completion of the given measure space).*

**Proof.** We let  $\mathcal{M}$  be the collection of all subsets  $E$  of  $X$  for which there exist  $A, B \in \mathcal{A}$  such that

$$A \subset E \subset B, \quad \mu(B - A) = 0. \quad (6)$$

If  $E \in \mathcal{A}$ , we may take  $A = B = E$  in (6), so  $\mathcal{A} \subset \mathcal{M}$ . In particular  $X \in \mathcal{M}$ .

If  $E \in \mathcal{M}$  and  $A, B$  are as in (6), then  $A^c, B^c \in \mathcal{A}$ ,

$$B^c \subset E^c \subset A^c$$

and  $\mu(A^c - B^c) = \mu(B - A) = 0$ , so that  $E^c \in \mathcal{M}$ .

If  $E_j \in \mathcal{M}, j = 1, 2, \dots$  and  $A_j, B_j$  are as in (6) (for  $E_j$ ), then if  $E, A, B$  are the respective unions of  $E_j, A_j, B_j$ , we have  $A, B \in \mathcal{A}, A \subset E \subset B$ , and

$$B - A = \bigcup_j (B_j - A) \subset \bigcup_j (B_j - A_j).$$

The union on the right is a null set (as a countable union of null sets, by  $\sigma$ -subadditivity of measures), and therefore  $B - A$  is a null set (by monotonicity of measures). This shows that  $E \in \mathcal{M}$ , and we conclude that  $\mathcal{M}$  is a  $\sigma$ -algebra.

For  $E \in \mathcal{M}$  and  $A, B$  as in (6), we let  $\nu(E) = \mu(A)$ . The function  $\nu$  is *well defined* on  $\mathcal{M}$ , that is, the above definition does not depend on the choice of  $A, B$  as in (6). Indeed, if  $A', B'$  satisfy (6) with  $E$ , then

$$A - A' \subset E - A' \subset B' - A',$$

so that  $A - A'$  is a null set. Hence by additivity of  $\mu$ ,  $\mu(A) = \mu(A \cap A') + \mu(A - A') = \mu(A \cap A')$ . Interchanging the roles of  $A$  and  $A'$ , we also have  $\mu(A') = \mu(A \cap A')$ , and therefore  $\mu(A) = \mu(A')$ , as wanted.

If  $E \in \mathcal{A}$ , we could choose  $A = B = E$ , and so  $\nu(E) = \mu(E)$ . In particular,  $\nu(\emptyset) = 0$ . If  $\{E_j\}$  is a sequence of mutually disjoint sets in  $\mathcal{M}$  with union  $E$ , and  $A_j, B_j$  are as in (6) (for  $E_j$ ), we observed above that we could choose  $A$  for  $E$  (for (6)) as the union of the sets  $A_j$ . Since  $A_j \subset E_j, j = 1, 2, \dots$  and  $E_j$  are mutually disjoint, so are the sets  $A_j$ . Hence

$$\nu(E) := \mu(A) = \sum_j \mu(A_j) := \sum_j \nu(E_j),$$

and we conclude that  $(X, \mathcal{M}, \nu)$  is a measure space extending  $(X, \mathcal{A}, \mu)$ . It is complete, because if  $E \in \mathcal{M}$  is  $\nu$ -null and  $A, B$  are as in (6), then for any  $F \subset E$ , we have

$$\emptyset \subset F \subset B,$$

and since  $\mu(B - A) = 0$ ,

$$\mu(B - \emptyset) = \mu(B) = \mu(A) := \nu(E) = 0,$$

so that  $F \in \mathcal{M}$ .

Finally, suppose  $(X, \mathcal{N}, \sigma)$  is any complete extension of  $(X, \mathcal{A}, \mu)$ , let  $E \in \mathcal{M}$ , and let  $A, B$  be as in (6). Write  $E = A \cup (E - A)$ . The set  $B - A \in \mathcal{A} \subset \mathcal{N}$  is  $\sigma$ -null ( $\sigma(B - A) = \mu(B - A) = 0$ ). By completeness of  $(X, \mathcal{N}, \sigma)$ , the subset  $E - A$  of  $B - A$  belongs to  $\mathcal{N}$  (and is of course  $\sigma$ -null). Since  $A \in \mathcal{A} \subset \mathcal{N}$ , we conclude that  $E \in \mathcal{N}$  and  $\mathcal{M} \subset \mathcal{N}$ . Also since  $\sigma = \mu$  on  $\mathcal{A}$ ,  $\sigma(E) = \sigma(A) + \sigma(E - A) = \mu(A) := \nu(E)$ , so that  $\sigma = \nu$  on  $\mathcal{M}$ .  $\square$

## 1.5 $L^p$ -spaces

Let  $(X, \mathcal{A}, \mu)$  be a (positive) measure space, and let  $p \in [1, \infty)$ . If  $f : X \rightarrow [0, \infty]$  is measurable, so is  $f^p$  by Lemma 1.6, and therefore  $\int f^p d\mu \in [0, \infty]$  is well defined. We denote

$$\|f\|_p := \left( \int f^p d\mu \right)^{1/p}.$$

**Theorem 1.24 (Holder's inequality).** *Let  $p, q \in (1, \infty)$  be conjugate exponents, that is,*

$$\frac{1}{p} + \frac{1}{q} = 1. \tag{1}$$

*Then for all measurable functions  $f, g : X \rightarrow [0, \infty]$ ,*

$$\int fg d\mu \leq \|f\|_p \|g\|_q. \tag{2}$$

**Proof.** If  $\|f\|_p = 0$ , then  $\|f^p\|_1 = 0$ , and therefore  $f = 0$  a.e.; hence  $fg = 0$  a.e., and the left-hand side of (2) vanishes (as well as the right-hand side). By symmetry, the same holds true if  $\|g\|_q = 0$ . So we may consider only the case where  $\|f\|_p$  and  $\|g\|_q$  are both *positive*. Now if one of these quantities is infinite,

the right-hand side of (2) is infinite, and (2) is trivially true. So we may assume that both quantities belong to  $(0, \infty)$  (*positive and finite*). Denote

$$u = f/\|f\|_p, \quad v = g/\|g\|_q. \quad (3)$$

Then

$$\|u\|_p = \|v\|_q = 1. \quad (4)$$

It suffices to prove that

$$\int uv \, d\mu \leq 1, \quad (5)$$

because (2) would follow by substituting (3) in (5).

The logarithmic function is concave  $((\log t)'' = -(1/t^2) < 0)$ . Therefore by (1)

$$\frac{1}{p} \log s + \frac{1}{q} \log t \leq \log \left( \frac{s}{p} + \frac{t}{q} \right)$$

for all  $s, t \in (0, \infty)$ . Equivalently,

$$s^{1/p} t^{1/q} \leq \frac{s}{p} + \frac{t}{q}, \quad s, t \in (0, \infty). \quad (6)$$

When  $x \in X$  is such that  $u(x), v(x) \in (0, \infty)$ , we substitute  $s = u(x)^p$  and  $t = v(x)^q$  in (6) and obtain

$$u(x)v(x) \leq \frac{u(x)^p}{p} + \frac{v(x)^q}{q}, \quad (7)$$

and this inequality is trivially true when  $u(x), v(x) \in \{0, \infty\}$ . Thus (7) is valid on  $X$ , and integrating the inequality over  $X$ , we obtain by (4) and (1)

$$\int uv \, d\mu \leq \frac{\|u\|_p^p}{p} + \frac{\|v\|_q^q}{q} = \frac{1}{p} + \frac{1}{q} = 1.$$

□

**Theorem 1.25 (Minkowski's inequality).** *For any measurable functions  $f, g : X \rightarrow [0, \infty]$ ,*

$$\|f + g\|_p \leq \|f\|_p + \|g\|_p \quad (1 \leq p < \infty). \quad (8)$$

**Proof.** Since (8) is trivial for  $p = 1$  (by the additivity of the integral of non-negative measurable functions, we get even an equality), we consider  $p \in (1, \infty)$ . The case  $\|f + g\|_p = 0$  is trivial. By convexity of the function  $t^p$  (for  $p > 1$ ),  $((s + t)/2)^p \leq (s^p + t^p)/2$  for  $s, t \in (0, \infty)$ . Therefore, if  $x \in X$  is such that  $f(x), g(x) \in (0, \infty)$ ,

$$(f(x) + g(x))^p \leq 2^{p-1}[f(x)^p + g(x)^p], \quad (9)$$

and (9) is trivially true if  $f(x), g(x) \in \{0, \infty\}$ , and holds therefore on  $X$ . Integrating, we obtain

$$\|f + g\|_p^p \leq 2^{p-1}[\|f\|_p^p + \|g\|_p^p]. \quad (10)$$

If  $\|f + g\|_p = \infty$ , it follows from (10) that at least one of the quantities  $\|f\|_p, \|g\|_p$  is infinite, and (8) is then valid (as the trivial equality  $\infty = \infty$ ). This discussion shows that we may restrict our attention to the case

$$0 < \|f + g\|_p < \infty. \quad (11)$$

We write

$$(f + g)^p = f(f + g)^{p-1} + g(f + g)^{p-1}. \quad (12)$$

By Holder's inequality,

$$\int f(f + g)^{p-1} d\mu \leq \|f\|_p \|(f + g)^{p-1}\|_q = \|f\|_p \|f + g\|_p^{p/q},$$

since  $(p - 1)q = p$  for conjugate exponents  $p, q$ . A similar estimate holds for the integral of the second summand on the right-hand side of (12). Adding these estimates, we obtain

$$\|f + g\|_p^p \leq (\|f\|_p + \|g\|_p) \|f + g\|_p^{p/q}.$$

By (11), we may divide this inequality by  $\|f + g\|_p^{p/q}$ , and (8) follows since  $p - p/q = 1$ .  $\square$

In a manner analogous to that used for  $L^1$ , if  $p \in [1, \infty)$ , we consider the set

$$L^p(X, \mathcal{A}, \mu)$$

(or briefly,  $L^p(\mu)$ , or  $L^p(X)$ , or  $L^p$ , when the unmentioned parameters are understood) of all (*equivalence classes*) of measurable complex functions  $f$  on  $X$ , with

$$\|f\|_p := \| |f| \|_p < \infty.$$

Since  $\|\cdot\|_p$  is trivially homogeneous, it follows from (8) that  $L^p$  is a normed space (over  $\mathbb{C}$ ) for the pointwise operations and the norm  $\|\cdot\|_p$ . We can restate Holder's inequality in the form:

**Theorem 1.26.** *Let  $p, q \in (1, \infty)$  be conjugate exponents. If  $f \in L^p$  and  $g \in L^q$ , then  $fg \in L^1$ , and*

$$\|fg\|_1 \leq \|f\|_p \|g\|_q.$$

A sufficient condition for convergence in the  $L^p$ -metric follows at once from Theorem 1.21:

**Proposition.** *Let  $\{f_n\}$  be a sequence of a.e.-defined measurable complex functions on  $X$ , converging a.e. to the function  $f$ . For some  $p \in [1, \infty)$ , suppose*

there exists  $g \in L^p$  such that  $|f_n| \leq g$  for all  $n$  (with the usual equivalence class ambiguity). Then  $f, f_n \in L^p$ , and  $f_n \rightarrow f$  in the  $L^p$ -metric.

**Proof.** The first statement follows from the inequalities  $|f|^p, |f_n|^p \leq g^p \in L^1$ . Since  $|f - f_n|^p \rightarrow 0$  a.e. and  $|f - f_n|^p \leq (2g)^p \in L^1$ , the second statement follows from Theorem 1.21.  $\square$

The positive measure space  $(X, \mathcal{A}, \mu)$  is said to be *finite* if  $\mu(X) < \infty$ . When this is the case, the Holder inequality implies that  $L^p(\mu) \subset L^r(\mu)$  topologically (i.e. the inclusion map is continuous) when  $1 \leq r < p < \infty$ . Indeed, if  $f \in L^p(\mu)$ , then by Holder's inequality with the conjugate exponents  $p/r$  and  $s := p/(p-r)$ ,

$$\begin{aligned} \|f\|_r^r &= \int |f|^r \cdot 1 \, d\mu \\ &\leq \left[ \int (|f|^r)^{p/r} \, d\mu \right]^{r/p} \left[ \int 1^s \, d\mu \right]^{1/s} = \mu(X)^{1/s} \|f\|_p^r. \end{aligned}$$

Since  $1/rs = (1/r) - (1/p)$ , we obtain

$$\|f\|_r \leq \mu(X)^{1/r-1/p} \|f\|_p. \quad (13)$$

Hence  $f \in L^r(\mu)$ , and (13) (with  $f - g$  replacing  $f$ ) shows the continuity of the inclusion map of  $L^p(\mu)$  into  $L^r(\mu)$ .

Taking in particular  $r = 1$ , we get that  $L^p(\mu) \subset L^1(\mu)$  (topologically) for all  $p \geq 1$ , and

$$\|f\|_1 \leq \mu(X)^{1/q} \|f\|_p, \quad (14)$$

where  $q$  is the conjugate exponent of  $p$ .

We formalize the above discussion for future reference.

**Proposition.** *Let  $(X, \mathcal{A}, \mu)$  be a finite positive measure space. Then  $L^p(\mu) \subset L^r(\mu)$  (topologically) for  $1 \leq r < p < \infty$ , and the norms inequality (13) holds.*

Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be measurable spaces, and let  $h : X \rightarrow Y$  be a measurable map (cf. Definition 1.2). If  $\mu$  is a measure on  $\mathcal{A}$ , the function  $\nu : \mathcal{B} \rightarrow [0, \infty]$  given by

$$\nu(E) = \mu(h^{-1}(E)), \quad E \in \mathcal{B} \quad (15)$$

is well-defined, and is clearly a measure on  $\mathcal{B}$ . Since  $I_{h^{-1}(E)} = I_E \circ h$ , we can write (15) in the form

$$\int_Y I_E \, d\nu = \int_X I_E \circ h \, d\mu.$$

By linearity of the integral, it follows that

$$\int_Y f \, d\nu = \int_X f \circ h \, d\mu \quad (16)$$

for every  $\mathcal{B}$ -measurable simple function  $f$  on  $Y$ . If  $f : Y \rightarrow [0, \infty]$  is  $\mathcal{B}$ -measurable, use the Approximation Theorem 1.8 to obtain a non-decreasing

sequence  $\{f_n\}$  of  $\mathcal{B}$ -measurable non-negative simple functions converging pointwise to  $f$ ; then  $\{f_n \circ h\}$  is a similar sequence converging to  $f \circ h$ , and the Monotone Convergence Theorem shows that (16) is true for all such  $f$ .

If  $f : Y \rightarrow \mathbb{C}$  is  $\mathcal{B}$ -measurable, then  $f \circ h$  is a (complex)  $\mathcal{A}$ -measurable function on  $X$ , and for any  $1 \leq p < \infty$ ,

$$\int_Y |f|^p d\nu = \int_X |f|^p \circ h d\mu = \int_X |f \circ h|^p d\mu.$$

Thus,  $f \in L^p(\nu)$  for some  $p \in [1, \infty)$  if and only if  $f \circ h \in L^p(\mu)$ , and

$$\|f\|_{L^p(\nu)} = \|f \circ h\|_{L^p(\mu)}.$$

In particular (case  $p = 1$ ),  $f$  is  $\nu$ -integrable on  $Y$  if and only if  $f \circ h$  is  $\mu$ -integrable on  $X$ . When this is the case, writing  $f$  as a linear combination of four non-negative  $\nu$ -integrable functions, we see that (16) is valid for all such  $f$ . Formally

**Proposition.** *Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be measurable spaces, and let  $h : X \rightarrow Y$  be a measurable map. For any (positive) measure  $\mu$  on  $\mathcal{A}$ , define  $\nu(E) := \mu(h^{-1}(E))$  for  $E \in \mathcal{B}$ . Then:*

- (1)  $\nu$  is a (positive) measure on  $\mathcal{B}$ ;
- (2) if  $f : Y \rightarrow [0, \infty]$  is  $\mathcal{B}$ -measurable, then  $f \circ h$  is  $\mathcal{A}$ -measurable and (16) is valid;
- (3) if  $f : Y \rightarrow \mathbb{C}$  is  $\mathcal{B}$ -measurable, then  $f \circ h$  is  $\mathcal{A}$ -measurable;  $f \in L^p(\nu)$  for some  $p \in [1, \infty)$  if and only if  $f \circ h \in L^p(\mu)$ , and in that case, the map  $f \rightarrow f \circ h$  is norm-preserving; in the special case  $p = 1$ , the map is integral preserving (i.e. (16) is valid).

If  $\phi$  is a simple complex measurable function with distinct non-zero values  $c_j$  assumed on  $E_j$ , then

$$\|\phi\|_p^p = \sum_j |c_j|^p \mu(E_j)$$

is finite if and only if  $\mu(E_j) < \infty$  for all  $j$ , that is, equivalently, if and only if

$$\mu(|\phi| > 0) < \infty.$$

Thus, the simple functions in  $L^p$  (for any  $p \in [1, \infty)$ ) are the (measurable) simple functions vanishing outside a measurable set of finite measure (depending on the function). These functions are dense in  $L^p$ . Indeed, if  $0 \leq f \in L^p$  (without loss of generality, we assume that  $f$  is everywhere defined!), the Approximation Theorem provides a sequence of simple measurable functions

$$0 \leq \phi_1 \leq \phi_2 \leq \cdots \leq f$$

such that  $\phi_n \rightarrow f$  pointwise. By the proposition following Theorem 1.26,  $\phi_n \rightarrow f$  in the  $L^p$ -metric.

For  $f \in L^p$  complex, we may write  $f = \sum_{k=0}^3 i^k g_k$  with  $0 \leq g_k \in L^p$  ( $g_0 := u^+$ , etc., where  $u = \Re f$ ). We then obtain four sequences  $\{\phi_{n,k}\}$  of simple

functions in  $L^p$  converging, respectively, to  $g_k, k = 0, \dots, 3$ , in the  $L^p$ -metric; if  $\phi_n := \sum_{k=0}^3 i^k \phi_{n,k}$ , then  $\phi_n$  are simple  $L^p$ -functions, and  $\phi_n \rightarrow f$  in the  $L^p$ -metric. We proved:

**Theorem 1.27.** *For any  $p \in [1, \infty)$ , the simple functions in  $L^p$  are dense in  $L^p$ .*

Actually,  $L^p$  is the *completion* of the normed space of all measurable simple functions vanishing outside a set of finite measure, with respect to the  $L^p$ -metric (induced by the  $L^p$ -norm). The meaning of this statement is made clear by the following definition.

**Definition 1.28.** Let  $Z$  be a metric space, with metric  $d$ . A *Cauchy sequence* in  $Z$  is a sequence  $\{z_n\} \subset Z$  such that  $d(z_n, z_m) \rightarrow 0$  when  $n, m \rightarrow \infty$ . The space  $Z$  is *complete* if every Cauchy sequence in  $Z$  converges in  $Z$ . If  $Y \subset Z$  is dense in  $Z$ , and  $Z$  is complete, we also say that  $Z$  is the *completion* of  $Y$  (for the metric  $d$ ). The completion of  $Y$  (for the metric  $d$ ) is *unique* in a suitable sense.

A *complete normed space* is called a *Banach space*.

In order to get the conclusion preceding Definition 1.28, we still have to prove that  $L^p$  is complete:

**Theorem 1.29.**  *$L^p$  is a Banach space for each  $p \in [1, \infty)$ .*

We first prove the following

**Lemma 1.30.** *Let  $\{f_n\}$  be a Cauchy sequence in  $L^p(\mu)$ . Then it has a subsequence converging pointwise  $\mu$ -a.e.*

**Proof of Lemma.** Since  $\{f_n\}$  is Cauchy, there exists  $m_k \in \mathbb{N}$  such that  $\|f_n - f_m\|_p < 1/2^k$  for all  $n > m > m_k$ . Set

$$n_k = k + \max(m_1, \dots, m_k).$$

Then  $n_{k+1} > n_k > m_k$ , and therefore  $\{f_{n_k}\}$  is a subsequence of  $\{f_n\}$  satisfying

$$\|f_{n_{k+1}} - f_{n_k}\|_p < 1/2^k \quad k = 1, 2, \dots \quad (17)$$

Consider the series

$$g = \sum_{k=1}^{\infty} |f_{n_{k+1}} - f_{n_k}|, \quad (18)$$

and its partial sums  $g_m$ . By Theorem 1.25 and (17),

$$\|g_m\|_p \leq \sum_{k=1}^m \|f_{n_{k+1}} - f_{n_k}\|_p < \sum_{k=1}^{\infty} 1/2^k = 1$$

for all  $m$ . By Fatou's lemma,

$$\int g^p d\mu \leq \liminf_m \int g_m^p d\mu = \liminf_m \|g_m\|_p^p \leq 1.$$

Therefore  $g < \infty$  a.e., that is, the series (18) converges a.e., that is, the series

$$f_{n_1} + \sum_{k=1}^{\infty} (f_{n_{k+1}} - f_{n_k}) \quad (19)$$

converges absolutely pointwise a.e. to its sum  $f$  (extended as 0 on the null set where the series does not converge). Since the partial sums of (19) are precisely  $f_{n_m}$ , the lemma is proved.  $\square$

**Proof of Theorem 1.29.** Let  $\{f_n\} \subset L^p$  be Cauchy. Thus for any  $\epsilon > 0$ , there exists  $n_\epsilon \in \mathbb{N}$  such that

$$\|f_n - f_m\|_p < \epsilon \quad (20)$$

for all  $n, m > n_\epsilon$ . By the lemma, let then  $\{f_{n_k}\}$  be a subsequence converging pointwise a.e. to the (measurable) complex function  $f$ . Applying Fatou's lemma to the non-negative measurable functions  $|f_{n_k} - f_m|$ , we obtain

$$\|f - f_m\|_p^p = \int \lim_k |f_{n_k} - f_m|^p d\mu \leq \liminf_k \|f_{n_k} - f_m\|_p^p \leq \epsilon^p \quad (21)$$

for all  $m > n_\epsilon$ . In particular,  $f - f_m \in L^p$ , and therefore  $f = (f - f_m) + f_m \in L^p$ , and (21) means that  $f_m \rightarrow f$  in the  $L^p$ -metric.  $\square$

**Definition 1.31.** Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $f : X \rightarrow \mathbb{C}$  be a measurable function. We say that  $M \in [0, \infty]$  is an a.e. upper bound for  $|f|$  if  $|f| \leq M$  a.e. The infimum of all the a.e. upper bounds for  $|f|$  is called the *essential supremum* of  $|f|$ , and is denoted  $\|f\|_\infty$ . The set of all (*equivalence classes of*) measurable complex functions  $f$  on  $X$  with  $\|f\|_\infty < \infty$  will be denoted by  $L^\infty(\mu)$  (or  $L^\infty(X)$ , or  $L^\infty(X, \mathcal{A}, \mu)$ , or  $L^\infty$ , depending on which ‘parameter’ we wish to stress, if at all).

By definition of the essential supremum, we have

$$|f| \leq \|f\|_\infty \quad \text{a.e.} \quad (22)$$

In particular,  $\|f\|_\infty = 0$  implies that  $f = 0$  a.e. (that is,  $f$  is the zero class).

If  $f, g \in L^\infty$ , then by (22),  $|f + g| \leq |f| + |g| \leq \|f\|_\infty + \|g\|_\infty$  a.e., and so  $\|f + g\|_\infty \leq \|f\|_\infty + \|g\|_\infty$ .

The homogeneity  $\|\alpha f\|_\infty = |\alpha| \|f\|_\infty$  is trivial if either  $\alpha = 0$  or  $\|f\|_\infty = 0$ . Assume then  $|\alpha|, \|f\|_\infty > 0$ . For any  $t \in (0, 1)$ ,  $t\|f\|_\infty < \|f\|_\infty$ , hence it is not an a.e. upper bound for  $|f|$ , so that  $\mu(|f| > t\|f\|_\infty) > 0$ , that is,  $\mu(|\alpha f| > t|\alpha| \|f\|_\infty) > 0$ . Therefore  $\|\alpha f\|_\infty \geq t|\alpha| \|f\|_\infty$  for all  $t \in (0, 1)$ , hence  $\|\alpha f\|_\infty \geq |\alpha| \|f\|_\infty$ . The reversed inequality follows trivially from (22), and the homogeneity of  $\|\cdot\|_\infty$  follows. We conclude that  $L^\infty$  is a normed space (over  $\mathbb{C}$ ) for the pointwise operations and the  $L^\infty$ -norm  $\|\cdot\|_\infty$ .

We verify its completeness as follows. Let  $\{f_n\}$  be a Cauchy sequence in  $L^\infty$ . In particular, it is a bounded set in  $L^\infty$ . Let then  $K = \sup_n \|f_n\|_\infty$ . By (22), the sets  $F_k := [|f_k| > K] \quad (k \in \mathbb{N})$  and

$$E_{n,m} := [|f_n - f_m| > \|f_n - f_m\|_\infty] \quad (n, m \in \mathbb{N})$$



are  $\mu$ -null, so their (countable) union  $E$  is null. For all  $x \in E^c$ ,

$$|f_n(x) - f_m(x)| \leq \|f_n - f_m\|_\infty \rightarrow 0$$

as  $n, m \rightarrow \infty$  and  $|f_n(x)| \leq K$ . By completeness of  $\mathbb{C}$ , the limit  $f(x) := \lim_n f_n(x)$  exists for all  $x \in E^c$  and  $|f(x)| \leq K$ . Defining  $f(x) = 0$  for all  $x \in E$ , we obtain a measurable function on  $X$  such that  $|f| \leq K$ , that is,  $f \in L^\infty$ . Given  $\epsilon > 0$ , let  $n_\epsilon \in \mathbb{N}$  be such that

$$\|f_n - f_m\|_\infty < \epsilon \quad (n, m > n_\epsilon).$$

Since  $|f_n(x) - f_m(x)| < \epsilon$  for all  $x \in E^c$  and  $n, m > n_\epsilon$ , letting  $m \rightarrow \infty$ , we obtain  $|f_n(x) - f(x)| \leq \epsilon$  for all  $x \in E^c$  and  $n > n_\epsilon$ , and since  $\mu(E) = 0$ ,

$$\|f_n - f\|_\infty \leq \epsilon \quad (n > n_\epsilon),$$

that is,  $f_n \rightarrow f$  in the  $L^\infty$ -metric. We proved

**Theorem 1.32.**  *$L^\infty$  is a Banach space.*

Defining the conjugate exponent of  $p = 1$  to be  $q = \infty$  (so that  $(1/p) + (1/q) = 1$  is formally valid in the usual sense), Holder's inequality remains true for this pair of conjugate exponents. Indeed, if  $f \in L^1$  and  $g \in L^\infty$ , then  $|fg| \leq \|g\|_\infty |f|$  a.e., and therefore  $fg \in L^1$  and

$$\|fg\|_1 \leq \|g\|_\infty \|f\|_1.$$

Formally

**Theorem 1.33.** *Holder's inequality (Theorem 1.26) is valid for conjugate exponents  $p, q \in [1, \infty]$ .*

## 1.6 Inner product

For the conjugate pair  $(p, q) = (2, 2)$ , Theorem 1.26 asserts that if  $f, g \in L^2$ , then the product  $f\bar{g}$  is integrable, so we may define

$$(f, g) := \int f\bar{g} d\mu. \tag{1}$$

( $\bar{g}$  denotes here the complex conjugate of  $g$ ). The function (or *form*)  $(\cdot, \cdot)$  has obviously the following properties on  $L^2 \times L^2$ :

- (i)  $(f, f) \geq 0$ , and  $(f, f) = 0$  if and only if  $f = 0$  (the zero element);
- (ii)  $(\cdot, g)$  is linear for each given  $g \in L^2$ ;
- (iii)  $(g, f) = \overline{(f, g)}$ .

Property (i) is called *positive definiteness* of the form  $(\cdot, \cdot)$ ; Properties (ii) and (iii) (together) are referred to as *sesquilinearity* or *hermitianity* of the form. We may also consider the weaker condition

$$(i') \quad (f, f) \geq 0 \text{ for all } f,$$

called (positive) *semi-definiteness* of the form.

**Definition 1.34.** Let  $X$  be a complex vector space (with elements  $x, y, \dots$ ). A *(semi)-inner product on  $X$*  is a (semi)-definite sesquilinear form  $(\cdot, \cdot)$  on  $X$ . The space  $X$  with a given (semi)-inner product is called a *(semi)-inner product space*.

If  $X$  is a semi-inner product space, the non-negative square root of  $(x, x)$  is denoted  $\|x\|$ .

Thus  $L^2$  is an inner product space for the inner product (1) and  $\|f\| := (f, f)^{1/2} = \|f\|_2$ . By Theorem 1.26 with  $p = q = 2$ ,

$$|(f, g)| \leq \|f\|_2 \|g\|_2 \quad (2)$$

for all  $f, g \in L^2$ . This special case of the Holder inequality is called the *Cauchy-Schwarz inequality*. We shall see below that it is valid in *any* semi-inner product space.

Observe that any sesquilinear form  $(\cdot, \cdot)$  is *conjugate linear* with respect to its second variable, that is, for each given  $x \in X$ ,

$$(x, \alpha u + \beta v) = \bar{\alpha}(x, u) + \bar{\beta}(x, v) \quad (3)$$

for all  $\alpha, \beta \in \mathbb{C}$  and  $u, v \in X$ .

In particular

$$(x, 0) = (0, y) = 0 \quad (4)$$

for all  $x, y \in X$ .

By (ii) and (3), for all  $\lambda \in \mathbb{C}$  and  $x, y \in X$ ,

$$(x + \lambda y, x + \lambda y) = (x, x) + \bar{\lambda}(x, y) + \lambda(y, x) + |\lambda|^2(y, y).$$

Since  $\lambda(y, x)$  is the conjugate of  $\bar{\lambda}(x, y)$  by (iii), we obtain the identity (for all  $\lambda \in \mathbb{C}$  and  $x, y \in X$ )

$$\|x + \lambda y\|^2 = \|x\|^2 + 2\Re[\bar{\lambda}(x, y)] + |\lambda|^2\|y\|^2. \quad (5)$$

In particular, for  $\lambda = 1$  and  $\lambda = -1$ , we have the identities

$$\|x + y\|^2 = \|x\|^2 + 2\Re(x, y) + \|y\|^2 \quad (6)$$

and

$$\|x - y\|^2 = \|x\|^2 - 2\Re(x, y) + \|y\|^2. \quad (7)$$

Adding, we obtain the so-called *parallelogram identity* for any s.i.p. (semi-inner product):

$$\|x + y\|^2 + \|x - y\|^2 = 2\|x\|^2 + 2\|y\|^2. \quad (8)$$

Subtracting (7) from (6), we obtain

$$4\Re(x, y) = \|x + y\|^2 - \|x - y\|^2. \quad (9)$$

If we replace  $y$  by  $iy$  in (9), we obtain

$$4\Im(x, y) = 4\Re[-i(x, y)] = 4\Re(x, iy) = \|x + iy\|^2 - \|x - iy\|^2. \quad (10)$$

By (9) and (10),

$$(x, y) = \frac{1}{4} \sum_{k=0}^3 i^k \|x + i^k y\|^2, \quad (11)$$

where  $i = \sqrt{-1}$ . This is the so-called *polarization identity* (which expresses the s.i.p. in terms of ‘induced norms’).

By (5),

$$0 \leq \|x\|^2 + 2\Re[\bar{\lambda}(x, y)] + |\lambda|^2 \|y\|^2 \quad (12)$$

for all  $\lambda \in \mathbb{C}$  and  $x, y \in X$ . If  $\|y\| > 0$ , take  $\lambda = -(x, y)/\|y\|^2$ ; then  $|(x, y)|^2/\|y\|^2 \leq \|x\|^2$ , and therefore

$$|(x, y)| \leq \|x\| \|y\|. \quad (13)$$

If  $\|y\| = 0$  but  $\|x\| > 0$ , interchange the roles of  $x$  and  $y$  and use (iii) to reach the same conclusion. If both  $\|x\|$  and  $\|y\|$  vanish, take  $\lambda = -(x, y)$  in (12): we get  $0 \leq -2|(x, y)|^2$ , hence  $|(x, y)| = 0 = \|x\| \|y\|$ , and we conclude that (13) is valid for *all*  $x, y \in X$ . This is the general Cauchy–Schwarz inequality for semi-inner products.

By (6) and (13),

$$\|x + y\|^2 \leq \|x\|^2 + 2|(x, y)| + \|y\|^2 \leq \|x\|^2 + 2\|x\| \|y\| + \|y\|^2 = (\|x\| + \|y\|)^2,$$

hence

$$\|x + y\| \leq \|x\| + \|y\|$$

for all  $x, y \in X$ . Taking  $x = 0$  in (5), we get  $\|\lambda y\| = |\lambda| \|y\|$  for all  $\lambda \in \mathbb{C}$  and  $y \in X$ . We conclude that  $\|\cdot\|$  is a semi-norm on  $X$ ; it is a norm if and only if the s.i.p. is an *inner product*, that is, if and only if it is *definite*. Thus, an inner product space  $X$  is a normed space for the norm  $\|x\| := (x, x)^{1/2}$  *induced* by its inner product (unless stated otherwise, this will be the standard norm for such spaces). In case  $X$  is *complete*, it is called a *Hilbert space*. Thus Hilbert spaces are special cases of Banach spaces.

The norm induced by the inner product (1) on  $L^2$  is the usual  $L^2$ -norm  $\|\cdot\|_2$ , so that, by Theorem 1.29,  $L^2$  is a Hilbert space.

## 1.7 Hilbert space: a first look

We consider some ‘geometric’ properties of Hilbert spaces.

**Theorem 1.35 (Distance theorem).** *Let  $X$  be a Hilbert space, and let  $K \subset X$  be non-empty, closed, and convex (i.e.  $(x+y)/2 \in K$  whenever  $x, y \in K$ ). Then for each  $x \in X$ , there exists a unique  $k \in K$  such that*

$$d(x, k) = d(x, K). \quad (1)$$

The notation  $d(x, y)$  is used for the metric induced by the norm,  $d(x, y) := \|x - y\|$ . As in any metric space,  $d(x, K)$  denotes the distance from  $x$  to  $K$ , that is,

$$d(x, K) := \inf_{y \in K} d(x, y). \quad (2)$$

**Proof.** Let  $d = d(x, K)$ . Since  $d^2 = \inf_{y \in K} \|x - y\|^2$ , there exist  $y_n \in K$  such that

$$(d^2 \leq) \|x - y_n\|^2 < d^2 + 1/n, \quad n = 1, 2, \dots \quad (3)$$

By the parallelogram identity,

$$\begin{aligned} \|y_n - y_m\|^2 &= \|(x - y_m) - (x - y_n)\|^2 \\ &= 2\|x - y_m\|^2 + 2\|x - y_n\|^2 - \|(x - y_m) + (x - y_n)\|^2. \end{aligned}$$

Rewrite the last term on the right-hand side in the form

$$4\|x - (y_m + y_n)/2\|^2 \geq 4d^2,$$

since  $(y_m + y_n)/2 \in K$ , by hypothesis. Hence by (3)

$$\|y_n - y_m\|^2 \leq 2/m + 2/n \rightarrow 0$$

as  $m, n \rightarrow \infty$ . Thus the sequence  $\{y_n\}$  is Cauchy. Since  $X$  is *complete*, the sequence converges in  $X$ , and its limit  $k$  is necessarily in  $K$  because  $y_n \in K$  for all  $n$  and  $K$  is closed. By continuity of the norm on  $X$ , letting  $n \rightarrow \infty$  in (3), we obtain  $\|x - k\| = d$ , as wanted.

To prove uniqueness, suppose  $k, k' \in K$  satisfy

$$\|x - k\| = \|x - k'\| = d.$$

Again by the parallelogram identity,

$$\begin{aligned} \|k - k'\|^2 &= \|(x - k') - (x - k)\|^2 \\ &= 2\|x - k'\|^2 + 2\|x - k\|^2 - \|(x - k') + (x - k)\|^2. \end{aligned}$$

As before, write the last term as  $4\|x - (k + k')/2\|^2 \geq 4d^2$  (since  $(k + k')/2 \in K$  by hypothesis). Hence

$$\|k - k'\|^2 \leq 2d^2 + 2d^2 - 4d^2 = 0,$$

and therefore  $k = k'$ .  $\square$

We say that the vector  $y \in X$  is *orthogonal* to the vector  $x$  if  $(x, y) = 0$ . In that case also  $(y, x) = \overline{(x, y)} = 0$ , so that the orthogonality relation is symmetric. For  $x$  given, let  $x^\perp$  denote the set of all vectors orthogonal to  $x$ . This is the kernel of the linear functional  $\phi = (\cdot, x)$ , that is, the set  $\phi^{-1}(\{0\})$ . As such a kernel, it is a subspace. Since  $|\phi(y) - \phi(z)| = |(y - z, x)| \leq \|y - z\| \|x\|$  by Schwarz' inequality,  $\phi$  is continuous, and therefore  $x^\perp = \phi^{-1}(\{0\})$  is closed. Thus  $x^\perp$  is a closed subspace. More generally, for any non-empty subset  $A$  of  $X$ , define

$$A^\perp := \bigcap_{x \in A} x^\perp = \{y \in Y; (y, x) = 0 \text{ for all } x \in A\}.$$

As the intersection of closed subspaces,  $A^\perp$  is a closed subspace of  $X$ .

**Theorem 1.36 (Orthogonal decomposition theorem).** *Let  $Y$  be a closed subspace of the Hilbert space  $X$ . Then  $X$  is the direct sum of  $Y$  and  $Y^\perp$ , that is, each  $x \in X$  has the unique orthogonal decomposition  $x = y + z$  with  $y \in Y$  and  $z \in Y^\perp$ .*

Note that the so-called *components*  $y$  and  $z$  of  $x$  (in  $Y$  and  $Y^\perp$ , respectively) are orthogonal.

**Proof.** As a closed subspace of  $X$ ,  $Y$  is a non-empty, closed, convex subset of  $X$ . By the distance theorem, there exists a unique  $y \in Y$  such that

$$\|x - y\| = d := d(x, Y).$$

Letting  $z := x - y$ , the existence part of the theorem will follow if we show that  $(z, u) = 0$  for all  $u \in Y$ . Since  $Y$  is a subspace, and  $Y \neq \{0\}$  without loss of generality, every  $u \in Y$  is a scalar multiple of a unit vector in  $Y$ , so it suffices to prove that  $(z, u) = 0$  for *unit vectors*  $u \in Y$ . For all  $\lambda \in \mathbb{C}$ , by the identity (5) (following Definition 1.34),

$$\|z - \lambda u\|^2 = \|z\|^2 - 2\Re[\bar{\lambda}(z, u)] + |\lambda|^2.$$

The left-hand side is

$$\|x - (y + \lambda u)\|^2 \geq d^2,$$

since  $y + \lambda u \in Y$ . Since  $\|z\| = d$ , we obtain

$$0 \leq -2\Re[\bar{\lambda}(z, u)] + |\lambda|^2.$$

Choose  $\lambda = (z, u)$ . Then  $0 \leq -|(z, u)|^2$ , so that  $(z, u) = 0$  as claimed.

If  $x = y + z = y' + z'$  are two decompositions with  $y, y' \in Y$  and  $z, z' \in Y^\perp$ , then  $y - y' = z' - z \in Y \cap Y^\perp$ , so that in particular  $y - y'$  is orthogonal to itself (i.e.  $(y - y', y - y') = 0$ ), which implies that  $y - y' = 0$ , whence  $y = y'$  and  $z = z'$ .  $\square$

We observed in passing that for each given  $y \in X$ , the function  $\phi := (\cdot, y)$  is a continuous linear functional on the inner product space  $X$ . For *Hilbert* spaces, this is the *general* form of continuous linear functionals:

**Theorem 1.37 ('Little' Riesz representation theorem).** *Let  $\phi : X \rightarrow \mathbb{C}$  be a continuous linear functional on the Hilbert space  $X$ . Then there exists a unique  $y \in X$  such that  $\phi = (\cdot, y)$ .*

**Proof.** If  $\phi = 0$  (the zero functional), take  $y = 0$ . Assume then that  $\phi \neq 0$ , so that its kernel  $Y$  is a closed subspace  $\neq X$ . Therefore  $Y^\perp \neq \{0\}$ , by Theorem 1.36. Let then  $z \in Y^\perp$  be a unit vector. Since  $Y \cap Y^\perp = \{0\}$ ,  $z \notin Y$ , so that  $\phi(z) \neq 0$ . For any given  $x \in X$ , we may then define

$$u := x - \frac{\phi(x)}{\phi(z)}z.$$

By linearity,

$$\phi(u) = \phi(x) - \frac{\phi(x)}{\phi(z)}\phi(z) = 0,$$

that is,  $u \in Y$ , and

$$x = u + \frac{\phi(x)}{\phi(z)}z \tag{4}$$

is the (unique) orthogonal decomposition of  $x$  (corresponding to the particular subspace  $Y$ , the kernel of  $\phi$ ). Define now  $y = \frac{\phi(x)}{\phi(z)}z (z \in Y^\perp)$ . By (4),

$$(x, y) = (u, y) + \frac{\phi(x)}{\phi(z)}\phi(z)(z, z) = \phi(x)$$

since  $(u, y) = 0$  and  $\|z\| = 1$ . This proves the existence part of the theorem. Suppose now that  $y, y' \in X$  are such that  $\phi(x) = (x, y) = (x, y')$  for all  $x \in X$ . Then  $(x, y - y') = 0$  for all  $x$ , hence in particular  $(y - y', y - y') = 0$ , which implies that  $y = y'$ .  $\square$

## 1.8 The Lebesgue–Radon–Nikodym theorem

We shall apply the Riesz representation theorem to prove the Lebesgue decomposition theorem and the Radon–Nikodym theorem for (positive) measures.

We start with a measure-theoretic lemma.

The positive measure space  $(X, \mathcal{A}, \mu)$  is  $\sigma$ -finite if there exists a *sequence* of mutually disjoint measurable sets  $X_j$  with union  $X$ , such that  $\mu(X_j) < \infty$  for all  $j$ .

**Lemma 1.38 (The averages lemma).** *Let  $(X, \mathcal{A}, \sigma)$  be a  $\sigma$ -finite positive measure space. Let  $g \in L^1(\sigma)$  be such that, for all  $E \in \mathcal{A}$  with  $0 < \sigma(E) < \infty$ , the ‘averages’*

$$A_E(g) := \frac{1}{\sigma(E)} \int_E g \, d\sigma$$

*are contained in some given closed set  $F \subset \mathbb{C}$ . Then  $g(x) \in F$   $\sigma$ -a.e.*

**Proof.** We need to prove that  $g^{-1}(F^c)$  is  $\sigma$ -null. Write the open set  $F^c$  as the countable union of the closed discs

$$\Delta_n := \{z \in \mathbb{C}; |z - a_n| \leq r_n\}, \quad n = 1, 2, \dots$$

Then

$$g^{-1}(F^c) = \bigcup_{n=1}^{\infty} g^{-1}(\Delta_n),$$

and it suffices to prove that  $E_\Delta := g^{-1}(\Delta)$  is  $\sigma$ -null whenever  $\Delta$  is a closed disc (with centre  $a$  and radius  $r$ ) contained in  $F^c$ .

Write  $X$  as the countable union of mutually disjoint measurable sets  $X_k$  with  $\sigma(X_k) < \infty$ . Set  $E_{\Delta,k} := E_\Delta \cap X_k$ , and suppose  $\sigma(E_{\Delta,k}) > 0$  for some  $\Delta$  as above and some  $k$ . Since  $|g(x) - a| \leq r$  on  $E := E_{\Delta,k}$ , and  $0 < \sigma(E) < \infty$ , we have

$$|A_E(g) - a| = |A_E(g - a)| \leq \frac{1}{\sigma(E)} \int_E |g - a| \, d\sigma \leq r,$$

so that  $A_E(g) \in \Delta \subset F^c$ , contradicting the hypothesis. Hence  $\sigma(E_{\Delta,k}) = 0$  for all  $k$  and therefore  $\sigma(E_\Delta) = 0$  for all  $\Delta$  as above.  $\square$

**Lemma 1.39.** *Let  $0 \leq \lambda \leq \sigma$  be finite measures on the measurable space  $(X, \mathcal{A})$ . Then there exists a measurable function  $g : X \rightarrow [0, 1]$  such that*

$$\int f \, d\lambda = \int f g \, d\sigma \tag{1}$$

*for all  $f \in L^2(\sigma)$ .*

**Proof.** By Definition 1.12, the relation  $\lambda \leq \sigma$  between positive measures implies that  $\int f \, d\lambda \leq \int f \, d\sigma$  for all non-negative measurable functions  $f$ . Hence  $L^2(\sigma) \subset L^2(\lambda) \subset L^1(\lambda)$ , by the second proposition following Theorem 1.26.)

For all  $f \in L^2(\sigma)$ , we have then by Schwarz’ inequality:

$$\left| \int f \, d\lambda \right| \leq \int |f| \, d\lambda \leq \int |f| \, d\sigma \leq \sigma(X)^{1/2} \|f\|_{L^2(\sigma)}.$$

Replacing  $f$  by  $f - h$  (with  $f, h \in L^2(\sigma)$ ), we get

$$\left| \int f \, d\lambda - \int h \, d\lambda \right| = \left| \int (f - h) \, d\lambda \right| \leq \sigma(X)^{1/2} \|f - h\|_{L^2(\sigma)},$$

so that the functional  $f \rightarrow \int f \, d\lambda$  is a continuous linear functional on  $L^2(\sigma)$ . By the Riesz representation theorem for the Hilbert space  $L^2(\sigma)$ , there exists an

element  $g_1 \in L^2(\sigma)$  such that this functional is  $(\cdot, g_1)$ . Letting  $g = \overline{g_1}$  ( $\in L^2(\sigma)$ ), we get the wanted relation (1).

Since  $I_E \in L^2(\sigma)$  (because  $\sigma$  is a finite measure), we have in particular

$$\lambda(E) = \int I_E d\lambda = \int_E g d\sigma$$

for all  $E \in \mathcal{A}$ . If  $\sigma(E) > 0$ ,

$$\frac{1}{\sigma(E)} \int_E g d\sigma = \frac{\lambda(E)}{\sigma(E)} \in [0, 1].$$

By the Averages Lemma 1.38,  $g(x) \in [0, 1]$   $\sigma$ -a.e., and we may then choose a representative of the equivalence class  $g$  with range in  $[0, 1]$ .  $\square$

**Terminology.** Let  $(X, \mathcal{A}, \lambda)$  be a positive measure space. We say that the set  $A \in \mathcal{A}$  *carries* the measure  $\lambda$  (or that  $\lambda$  is supported by  $E$ ) if  $\lambda(E) = \lambda(E \cap A)$  for all  $E \in \mathcal{A}$ .

This is of course equivalent to  $\lambda(E) = 0$  for all measurable subsets  $E$  of  $A^c$ .

Two (positive) measures  $\lambda_1, \lambda_2$  on  $(X, \mathcal{A})$  are *mutually singular* (notation  $\lambda_1 \perp \lambda_2$ ) if they are carried by *disjoint* measurable sets  $A_1, A_2$ . Equivalently, each measure is carried by a null set relative to the other measure.

On the other hand, if  $\lambda_2(E) = 0$  whenever  $\lambda_1(E) = 0$  (for  $E \in \mathcal{A}$ ), we say that  $\lambda_2$  is *absolutely continuous* with respect to  $\lambda_1$  (notation:  $\lambda_2 \ll \lambda_1$ ).

Equivalently,  $\lambda_2 \ll \lambda_1$  if and only if any (measurable) set that carries  $\lambda_1$  also carries  $\lambda_2$ .

**Theorem 1.40 (Lebesgue–Radon–Nikodym).** *Let  $(X, \mathcal{A}, \mu)$  be a  $\sigma$ -finite positive measure space, and let  $\lambda$  be a finite positive measure on  $(X, \mathcal{A})$ . Then*

(a)  $\lambda$  has the unique (so-called) Lebesgue decomposition

$$\lambda = \lambda_a + \lambda_s$$

with  $\lambda_a \ll \mu$  and  $\lambda_s \perp \mu$ ;

(b) there exists a unique  $h \in L^1(\mu)$  such that

$$\lambda_a(E) = \int_E h d\mu$$

for all  $E \in \mathcal{A}$ .

(Part (a) is the Lebesgue decomposition theorem; Part (b) is the Radon–Nikodym theorem.)

**Proof.** Case  $\mu(X) < \infty$ .

Let  $\sigma := \lambda + \mu$ . Then the finite positive measures  $\lambda, \sigma$  satisfy  $\lambda \leq \sigma$ , so that by Lemma 1.39, there exists a measurable function  $g : X \rightarrow [0, 1]$  such that (1) holds, that is, after rearrangement,

$$\int f(1 - g) d\lambda = \int fg d\mu \tag{2}$$



for all  $f \in L^2(\sigma)$ . Define

$$A := g^{-1}([0, 1)); \quad B := g^{-1}(\{1\}).$$

Then  $A, B$  are disjoint measurable sets with union  $X$ .

Taking  $f = I_B$  ( $\in L^2(\sigma)$ , since  $\sigma$  is a finite measure) in (2), we obtain  $\mu(B) = 0$  (since  $g = 1$  on  $B$ ). Therefore the measure  $\lambda_s$  defined on  $\mathcal{A}$  by

$$\lambda_s(E) := \lambda(E \cap B)$$

satisfies  $\lambda_s \perp \mu$ .

Define similarly  $\lambda_a(E) := \lambda(E \cap A)$ ; this is a positive measure on  $\mathcal{A}$ , mutually singular with  $\lambda_s$  (since it is carried by  $A = B^c$ ), and by additivity of measures,

$$\lambda(E) = \lambda(E \cap A) + \lambda(E \cap B) = \lambda_a(E) + \lambda_s(E),$$

so that the Lebesgue decomposition will follow if we show that  $\lambda_a \ll \mu$ . This follows trivially from the integral representation (b), which we proceed to prove.

For each  $n \in \mathbb{N}$  and  $E \in \mathcal{A}$ , take in (2)

$$f = f_n := (1 + g + \cdots + g^n)I_E.$$

(Since  $0 \leq g \leq 1$ ,  $f$  is a bounded measurable function, hence  $f \in L^2(\sigma)$ .) We obtain

$$\int_E (1 - g^{n+1}) d\lambda = \int_E (g + g^2 + \cdots + g^{n+1}) d\mu. \quad (3)$$

Since  $g = 1$  on  $B$ , the left-hand side equals  $\int_{E \cap A} (1 - g^{n+1}) d\lambda$ . However  $0 \leq g < 1$  on  $A$ , so that the integrands form a non-decreasing sequence of non-negative measurable functions converging pointwise to 1. By the monotone convergence theorem, the left-hand side of (3) converges therefore to  $\lambda(E \cap A) = \lambda_a(E)$ . The integrands on the right-hand side of (3) form a non-decreasing sequence of non-negative measurable functions converging pointwise to the (measurable) function

$$h := \sum_{n=1}^{\infty} g^n.$$

Again, by monotone convergence, the right-hand side of (3) converges to  $\int_E h d\mu$ , and the representation (b) follows. Taking in particular  $E = X$ , we get

$$\|h\|_{L^1(\mu)} = \int_X h d\mu = \lambda_a(X) = \lambda(A) < \infty,$$

so that  $h \in L^1(\mu)$ , and the existence part of the theorem is proved in case  $\mu(X) < \infty$ .

*General case.* Let  $X_j \in \mathcal{A}$  be mutually disjoint, with union  $X$ , such that  $0 < \mu(X_j) < \infty$ . Define

$$w = \sum_j \frac{1}{2^j \mu(X_j)} I_{X_j}.$$

This is a strictly positive  $\mu$ -integrable function, with  $\|w\|_1 = 1$ . Consider the positive measure

$$\nu(E) = \int_E w \, d\mu.$$

Then  $\nu(X) = \|w\|_1 = 1$ , and  $\nu \ll \mu$ . On the other hand, if  $\nu(E) = 0$ , then  $\sum_j (1/2^j \mu(X_j)) \mu(E \cap X_j) = 0$ , hence  $\mu(E \cap X_j) = 0$  for all  $j$ , and therefore  $\mu(E) = 0$ . This shows that  $\mu \ll \nu$  as well (one says that the measures  $\mu$  and  $\nu$  are *mutually absolutely continuous*, or *equivalent*).

Since  $\nu$  is a finite measure, the first part of the proof gives the decomposition  $\lambda = \lambda_a + \lambda_s$  with  $\lambda_a \ll \nu$  (hence  $\lambda_a \ll \mu$  by the trivial transitivity of the relation  $\ll$ ), and  $\lambda_s \perp \nu$  (hence  $\lambda_s \perp \mu$ , because  $\lambda_s$  is supported by a  $\nu$ -null set, which is also  $\mu$ -null, since  $\mu \ll \nu$ ). The first part of the proof gives also the representation (cf. Theorem 1.17)

$$\lambda_a(E) = \int_E h \, d\nu = \int_E h w \, d\mu = \int_E \tilde{h} \, d\mu,$$

where  $\tilde{h} := hw$  is non-negative, measurable, and

$$\|\tilde{h}\|_1 = \int_X \tilde{h} \, d\mu = \lambda_a(X) \leq \lambda(X) < \infty.$$

This completes the proof of the ‘existence part’ of the theorem in the general case.

To prove the *uniqueness* of the Lebesgue decomposition, suppose

$$\lambda = \lambda_a + \lambda_s = \lambda'_a + \lambda'_s,$$

with

$$\lambda_a, \lambda'_a \ll \mu \quad \text{and} \quad \lambda_s, \lambda'_s \perp \mu.$$

Let  $B$  be a  $\mu$ -null set that carries both  $\lambda_s$  and  $\lambda'_s$ . Then

$$\lambda_a(B) = \lambda'_a(B) = 0 \quad \text{and} \quad \lambda_s(B^c) = \lambda'_s(B^c) = 0,$$

so that for all  $E \in \mathcal{A}$ ,

$$\begin{aligned} \lambda_a(E) &= \lambda_a(E \cap B^c) = \lambda(E \cap B^c) \\ &= \lambda'_a(E \cap B^c) = \lambda'_a(E), \end{aligned}$$

hence also  $\lambda_s(E) = \lambda'_s(E)$ .

In order to prove the uniqueness of  $h$  in (b), suppose  $h, h' \in L^1(\mu)$  satisfy

$$\lambda_a(E) = \int_E h \, d\mu = \int_E h' \, d\mu.$$

Then  $h - h' \in L^1(\mu)$  satisfies  $\int_E (h - h') \, d\mu = 0$  for all  $E \in \mathcal{A}$ , and it follows from Proposition 1.22 that  $h - h' = 0$   $\mu$ -a.e., that is,  $h = h'$  as elements of  $L^1(\mu)$ .  $\square$

If the measure  $\lambda$  is absolutely continuous with respect to  $\mu$ , it has the trivial Lebesgue decomposition  $\lambda = \lambda + 0$ , with the zero measure as singular part. By

uniqueness, it follows that  $\lambda_a = \lambda$ , and therefore Part 2 of the theorem gives the representation  $\lambda(E) = \int_E h d\mu$  for all  $E \in \mathcal{A}$ . Conversely, such an integral representation of  $\lambda$  implies trivially that  $\lambda \ll \mu$  (if  $\mu(E) = 0$ , the function  $hI_E = 0$   $\mu$ -a.e., and therefore  $\lambda(E) = \int f I_E d\mu = 0$ ). Thus

**Theorem 1.41 (Radon–Nikodym).** *Let  $(X, \mathcal{A}, \mu)$  be a  $\sigma$ -finite positive measure space. A finite positive measure  $\lambda$  on  $\mathcal{A}$  is absolutely continuous with respect to  $\mu$  if and only if there exists  $h \in L^1(\mu)$  such that*

$$\lambda(E) = \int_E h d\mu \quad (E \in \mathcal{A}). \quad (*)$$

By Theorem 1.17, Relation  $(*)$  implies that

$$\int g d\lambda = \int gh d\mu \quad (**)$$

for all non-negative measurable functions  $g$  on  $X$ . Since we may take  $g = I_E$  in  $(**)$ , this last relation implies  $(*)$ . As mentioned after Theorem 1.17, these equivalent relations are *symbolically* written in the form  $d\lambda = h d\mu$ . It follows easily from Theorem 1.17 that in that case, if  $g \in L^1(\lambda)$ , then  $gh \in L^1(\mu)$  and  $(**)$  is valid for such (complex) functions  $g$ . The function  $h$  is called the *Radon–Nikodym derivative* of  $\lambda$  with respect to  $\mu$ , and is denoted  $d\lambda/d\mu$ .

## 1.9 Complex measures

**Definition 1.42.** Let  $(X, \mathcal{A})$  be an arbitrary measurable space. A *complex measure* on  $\mathcal{A}$  is a  $\sigma$ -additive function  $\mu : \mathcal{A} \rightarrow \mathbb{C}$ , that is,

$$\mu\left(\bigcup_n E_n\right) = \sum_n \mu(E_n) \quad (1)$$

for any sequence of mutually disjoint sets  $E_n \in \mathcal{A}$ .

Since the left-hand side of (1) is independent of the order of the sets  $E_n$  and is a complex number, the right-hand side converges in  $\mathbb{C}$  unconditionally, hence *absolutely*. Taking  $E_n = \emptyset$  for all  $n$ , the convergence of (1) shows that  $\mu(\emptyset) = 0$ . It follows that  $\mu$  is (finitely) additive, and since its values are complex numbers, it is ‘subtractive’ as well (i.e.  $\mu(E - F) = \mu(E) - \mu(F)$  whenever  $E, F \in \mathcal{A}$ ,  $F \subset E$ ).

A *partition* of  $E \in \mathcal{A}$  is a sequence of mutually disjoint sets  $A_k \in \mathcal{A}$  with union equal to  $E$ . We set

$$|\mu|(E) := \sup \sum_k |\mu(A_k)|, \quad (2)$$

where the supremum is taken over all partitions of  $E$ .

**Theorem 1.43.** *Let  $\mu$  be a complex measure on  $\mathcal{A}$ , and define  $|\mu|$  by (2). Then  $|\mu|$  is a finite positive measure on  $\mathcal{A}$  that dominates  $\mu$  (i.e.  $|\mu(E)| \leq |\mu|(E)$  for all  $E \in \mathcal{A}$ ).*

**Proof.** Let  $E = \bigcup E_n$  with  $E_n \in \mathcal{A}$  mutually disjoint ( $n \in \mathbb{N}$ ). For any partition  $\{A_k\}$  of  $E$ ,  $\{A_k \cap E_n\}_k$  is a partition of  $E_n$  ( $n = 1, 2, \dots$ ), so that

$$\sum_k |\mu(A_k \cap E_n)| \leq |\mu|(E_n), \quad n = 1, 2, \dots$$

We sum these inequalities over all  $n$ , interchange the order of summation in the double sum (of non-negative terms!), and use the triangle inequality to obtain:

$$\sum_n |\mu|(E_n) \geq \sum_k \left| \sum_n \mu(A_k \cap E_n) \right| = \sum_k |\mu(A_k)|,$$

since  $\{A_k \cap E_n\}_n$  is a partition of  $A_k$ , for each  $k \in \mathbb{N}$ . Taking now the supremum over all partitions  $\{A_k\}$  of  $E$ , it follows that

$$\sum_n |\mu|(E_n) \geq |\mu|(E). \quad (3)$$

On the other hand, given  $\epsilon > 0$ , there exists a partition  $\{A_{n,k}\}_k$  of  $E_n$  such that

$$\sum_k |\mu(A_{n,k})| > |\mu|(E_n) - \epsilon/2^n, \quad n = 1, 2, \dots$$

Since  $\{A_{n,k}\}_{n,k}$  is a partition of  $E$ , we obtain

$$|\mu|(E) \geq \sum_{n,k} |\mu(A_{n,k})| > \sum_n |\mu|(E_n) - \epsilon.$$

Letting  $\epsilon \rightarrow 0+$  and using (3), we conclude that  $|\mu|$  is  $\sigma$ -additive. Since  $|\mu|(\emptyset) = 0$  is trivial,  $|\mu|$  is indeed a positive measure on  $\mathcal{A}$ .  $\square$

In order to show that the measure  $|\mu|$  is finite, we need the following:

**Lemma.** *Let  $F \subset \mathbb{C}$  be a finite set. Then it contains a subset  $E$  such that*

$$\left| \sum_{z \in E} z \right| \geq \sum_{z \in F} |z|/4\sqrt{2}.$$

**Proof of lemma.** Let  $S$  be the sector

$$S = \{z = re^{i\theta}; r \geq 0, |\theta| \leq \pi/4\}.$$

For  $z \in S$ ,  $\Re z = |z| \cos \theta \geq |z| \cos \pi/4 = |z|/\sqrt{2}$ . Similarly, if  $z \in -S$ , then  $-z \in S$ , so that  $-\Re z = \Re(-z) \geq |-z|/\sqrt{2} = |z|/\sqrt{2}$ . If  $z \in iS$ , then  $-iz \in S$ ,

so that  $\Im z = \Re(-iz) \geq |-iz|/\sqrt{2} = |z|/\sqrt{2}$ . Similarly, if  $z \in -iS$ , one obtains as before  $-\Im z \geq |z|/\sqrt{2}$ . Denote

$$a := \sum_{z \in F} |z|; \quad a_k = \sum_{z \in F \cap (i^k S)} |z|, \quad k = 0, 1, 2, 3.$$

Since  $\mathbb{C} = \bigcup_{k=0}^3 i^k S$ , we have  $\sum_{k=0}^3 a_k \geq a$ , and therefore there exists  $k \in \{0, 1, 2, 3\}$  such that  $a_k \geq a/4$ . Fix such a  $k$  and define  $E = F \cap (i^k S)$ .

In case  $k = 0$ , that is, in case  $a_0 \geq a/4$ , we have

$$\begin{aligned} \left| \sum_{z \in E} z \right| &\geq \Re \sum_{z \in E} z = \sum_{z \in F \cap S} \Re z \\ &\geq \sum_{z \in F \cap S} |z|/\sqrt{2} = a_0/\sqrt{2} \geq a/4\sqrt{2}. \end{aligned}$$

Similarly, in case  $k = 2$ , replacing  $\Re$  by  $-\Re$  and  $S$  by  $-S$ , the same inequality is obtained. In cases  $k = 1 (k = 3)$ , we replace  $\Re$  and  $S$  by  $\Im (-\Im)$  and  $iS (-iS)$  respectively, in the above calculation. In all cases, we obtain  $|\sum_{z \in E} z| \geq a/4\sqrt{2}$ , as wanted.  $\square$

Returning to the proof of the finiteness of the measure  $|\mu|$ , suppose  $|\mu|(A) = \infty$  for some  $A \in \mathcal{A}$ . Then there exists a partition  $\{A_i\}$  of  $A$  such that  $\sum_i |\mu(A_i)| > 4\sqrt{2}(1 + |\mu(A)|)$ , and therefore there exists  $n$  such that

$$\sum_{i=1}^n |\mu(A_i)| > 4\sqrt{2}(1 + |\mu(A)|).$$

Take in the lemma

$$F = \{\mu(A_i); i = 1, \dots, n\},$$

let the corresponding subset  $E$  be associated with the set of indices  $J \subset \{1, \dots, n\}$ , and define

$$B := \bigcup_{i \in J} A_i \subset A.$$

Then

$$|\mu(B)| = \left| \sum_{i \in J} \mu(A_i) \right| \geq \sum_{i=1}^n |\mu(A_i)|/4\sqrt{2} > 1 + |\mu(A)|.$$

If  $C := A - B$ , then

$$|\mu(C)| = |\mu(A) - \mu(B)| \geq |\mu(B)| - |\mu(A)| > 1.$$

Also  $\infty = |\mu|(A) = |\mu|(B) + |\mu|(C)$  (since  $|\mu|$  is a measure), so one of the summands at least is infinite, and for *both* subsets, the  $\mu$ -measure has modulus  $> 1$ .

Thus, we proved that any  $A \in \mathcal{A}$  with  $|\mu|(A) = \infty$  is the disjoint union of subsets  $B_1, C_1 \in \mathcal{A}$ , with  $|\mu|(B_1) = \infty$  and  $|\mu|(C_1)| > 1$ .

Since  $|\mu|(B_1) = \infty$ ,  $B_1$  is the disjoint union of subsets  $B_2, C_2 \in \mathcal{A}$ , with  $|\mu|(B_2) = \infty$  and  $|\mu(C_2)| > 1$ . Continuing, we obtain two sequences  $\{B_n\}, \{C_n\} \subset \mathcal{A}$  with the following properties:

$$\begin{aligned} B_n &= B_{n+1} \cup C_{n+1} \quad (\text{disjoint union}); \\ |\mu|(B_n) &= \infty; \quad |\mu(C_n)| > 1; \quad n = 1, 2, \dots \end{aligned}$$

For  $i > j \geq 1$ , since  $B_{n+1} \subset B_n$  for all  $n$ , we have  $C_i \cap C_j \subset B_{i-1} \cap C_j \subset B_j \cap C_j = \emptyset$ , so  $C := \bigcup_n C_n$  is a disjoint union. Hence the series  $\sum_n \mu(C_n) = \mu(C) \in \mathbb{C}$  converges. In particular  $\mu(C_n) \rightarrow 0$ , contradicting the fact that  $|\mu(C_n)| > 1$  for all  $n = 1, 2, \dots$

Finally, since  $\{E, \emptyset, \emptyset, \dots\}$  is a partition of  $E \in \mathcal{A}$ , the inequality  $|\mu(E)| \leq |\mu|(E)$  follows from (2).  $\square$

**Definition 1.44.** The finite positive measure  $|\mu|$  is called the *total variation measure* of  $\mu$ , and  $\|\mu\| := |\mu|(X)$  is called the *total variation* of  $\mu$ .

Let  $M(X, \mathcal{A})$  denote the complex vector space of all complex measures on  $\mathcal{A}$  (with the ‘natural’ operations  $(\mu + \nu)(E) = \mu(E) + \nu(E)$  and  $(c\mu)(E) = c\mu(E)$ , for  $\mu, \nu \in M(X, \mathcal{A})$  and  $c \in \mathbb{C}$ ). With the total variation norm,  $M(X, \mathcal{A})$  is a normed space. We verify below its completeness:

**Proposition.**  $M(X, \mathcal{A})$  is a Banach space.

**Proof.** Let  $\{\mu_n\} \subset M := M(X, \mathcal{A})$  be Cauchy. For all  $E \in \mathcal{A}$ ,

$$|\mu_n(E) - \mu_m(E)| \leq \|\mu_n - \mu_m\| \rightarrow 0$$

as  $n, m \rightarrow \infty$ , so that

$$\mu(E) := \lim_n \mu_n(E)$$

exists. Clearly  $\mu$  is additive. Let  $E$  be the union of the mutually disjoint sets  $E_k \in \mathcal{A}$ ,  $k = 1, 2, \dots$ , let  $A_N = \bigcup_{k=1}^N E_k$ , and let  $\epsilon > 0$ . Let  $n_0 \in \mathbb{R}$  be such that

$$\|\mu_n - \mu_m\| < \epsilon \tag{*}$$

for all  $n, m > n_0$ . We have

$$\begin{aligned} |\mu(E) - \sum_{k=1}^N \mu(E_k)| &= |\mu(E) - \mu(A_N)| = |\mu(E - A_N)| \\ &\leq |(\mu - \mu_n)(E - A_N)| + |\mu_n(E - A_N)|. \end{aligned}$$

Since

$$|(\mu_n - \mu_m)(E - A_N)| \leq \|\mu_n - \mu_m\| < \epsilon$$

for all  $n, m > n_0$ , letting  $m \rightarrow \infty$ , we obtain  $|(\mu_n - \mu)(E - A_N)| \leq \epsilon$  for all  $n > n_0$  and all  $N \in \mathbb{N}$ . Therefore,

$$\left| \mu(E) - \sum_{k=1}^N \mu(E_k) \right| \leq \epsilon + \left| \mu_n(E) - \sum_{k=1}^N \mu_n(E_k) \right|$$

for all  $n > n_0$  and  $N \in \mathbb{N}$ . Fix  $n > n_0$  and let  $N \rightarrow \infty$ . Since  $\mu_n \in M$ , we obtain

$$\limsup_N \left| \mu(E) - \sum_{k=1}^N \mu(E_k) \right| \leq \epsilon,$$

so that  $\mu(E) = \sum_{k=1}^{\infty} \mu(E_k)$ . Thus  $\mu \in M$ .

Finally, we show that  $\|\mu - \mu_n\| \rightarrow 0$ . Let  $\{E_k\}$  be a partition of  $X$ . By (\*), for all  $N \in \mathbb{N}$  and  $n, m > n_0$ ,

$$\sum_{k=1}^N |(\mu_m - \mu_n)(E_k)| \leq \|\mu_m - \mu_n\| < \epsilon.$$

Letting  $m \rightarrow \infty$ , we get

$$\sum_{k=1}^N |(\mu - \mu_n)(E_k)| \leq \epsilon$$

for all  $N$  and  $n > n_0$ . Letting  $N \rightarrow \infty$ , we obtain  $\sum_{k=1}^{\infty} |(\mu - \mu_n)(E_k)| \leq \epsilon$  for all  $n > n_0$ , and since the partition was arbitrary, it follows that  $\|\mu - \mu_n\| \leq \epsilon$  for  $n > n_0$ .  $\square$

If  $\mu \in M(X, \mathcal{A})$  has *real* range, it is called a *real measure*. For example,  $\Re\mu$  (defined by  $(\Re\mu)(E) = \Re[\mu(E)]$ ) and  $\Im\mu$  are real measures for any complex measure  $\mu$ , and  $\mu = \Re\mu + i\Im\mu$ .

If  $\mu$  is a *real* measure, since  $|\mu(E)| \leq |\mu|(E)$  for all  $E \in \mathcal{A}$ , the measures

$$\mu^+ := (1/2)(|\mu| + \mu); \quad \mu^- := (1/2)(|\mu| - \mu)$$

are *finite positive measures*, called the *positive and negative variation measures*, respectively.

Clearly

$$\mu = \mu^+ - \mu^- \tag{4}$$

and

$$|\mu| = \mu^+ + \mu^-.$$

Representation (4) of a real measure as the difference of two finite positive measures is called the *Jordan decomposition* of the real measure  $\mu$ .

For a complex measure  $\lambda$ , write first  $\lambda = \nu + i\sigma$  with  $\nu := \Re\lambda$  and  $\sigma = \Im\lambda$ ; then write the Jordan decompositions of  $\nu$  and  $\sigma$ . It then follows that *any*

complex measure  $\lambda$  can be written as the linear combination

$$\lambda = \sum_{k=0}^3 i^k \lambda_k \quad (5)$$

of four finite positive measures  $\lambda_k$ .

If  $\mu$  is a positive measure and  $\lambda$  is a *complex* measure (both on the  $\sigma$ -algebra  $\mathcal{A}$ ), we say that  $\lambda$  is absolutely continuous with respect to  $\mu$  (notation:  $\lambda \ll \mu$ ) if  $\lambda(E) = 0$  whenever  $\mu(E) = 0, E \in \mathcal{A}$ ;  $\lambda$  is carried (or supported) by the set  $A \in \mathcal{A}$  if  $\lambda(E) = 0$  for all measurable subsets  $E$  of  $A^c$ ;  $\lambda$  is singular with respect to  $\mu$  if it is carried by a  $\mu$ -null set. Two complex measures  $\lambda_1, \lambda_2$  are *mutually singular* (notation  $\lambda_1 \perp \lambda_2$ ) if they are carried by disjoint measurable sets.

It follows immediately from (5) that Theorems 1.40 and 1.41 extend verbatim to the case of a *complex* measure  $\lambda$ . This is stated formally for future reference.

**Theorem 1.45.** *Let  $(X, \mathcal{A}, \mu)$  be a  $\sigma$ -finite positive measure space, and let  $\lambda$  be a complex measure on  $\mathcal{A}$ . Then*

(1)  *$\lambda$  has a unique Lebesgue decomposition*

$$\lambda = \lambda_a + \lambda_s$$

*with  $\lambda_a \ll \mu$  and  $\lambda_s \perp \mu$ ;*

(2) *there exists a unique  $h \in L^1(\mu)$  such that*

$$\lambda_a(E) = \int_E h d\mu \quad (E \in \mathcal{A});$$

(3)  *$\lambda \ll \mu$  if and only if there exists  $h \in L^1(\mu)$  such that  $\lambda(E) = \int_E h d\mu$  for all  $E \in \mathcal{A}$ .*

Another useful representation of a complex measure in terms of a finite positive measure is the following:

**Theorem 1.46.** *Let  $\mu$  be a complex measure on the measurable space  $(X, \mathcal{A})$ . Then there exists a measurable function  $h$  with  $|h| = 1$  on  $X$  such that  $d\mu = h d|\mu|$ .*

**Proof.** Since  $|\mu(E)| \leq |\mu|(E)$  for all  $E \in \mathcal{A}$ , it follows that  $\mu \ll |\mu|$ , and therefore, by Theorem 1.45 (since  $|\mu|$  is a *finite* positive measure), there exists  $h \in L^1(|\mu|)$  such that  $d\mu = h d|\mu|$ . For each  $E \in \mathcal{A}$  with  $|\mu|(E) > 0$ ,

$$\left| \frac{1}{|\mu|(E)} \int_E h d|\mu| \right| = \frac{|\mu(E)|}{|\mu|(E)} \leq 1.$$

By Lemma 1.38, it follows that  $|h| \leq 1$   $|\mu|$ -a.e.



We wish to show that  $|\mu|([|h| < 1]) = 0$ . Since  $[|h| < 1] = \bigcup_n [|h| < 1 - 1/n]$ , it suffices to show that  $|\mu|([|h| < r]) = 0$  for each  $r < 1$ . Denote  $A = [|h| < r]$  and let  $\{E_k\}$  be a partition of  $A$ . Then

$$\sum_k |\mu(E_k)| = \sum_k \left| \int_{E_k} h d|\mu| \right| \leq r \sum_k |\mu|(E_k) = r|\mu|(A).$$

Hence

$$|\mu|(A) \leq r|\mu|(A),$$

so that indeed  $|\mu|(A) = 0$ .

We conclude that  $|h| = 1$   $|\mu|$ -a.e., and since  $h$  is only determined a.e., we may replace it by a  $|\mu|$ -equivalent function which satisfies  $|h| = 1$  everywhere.  $\square$

If  $\mu$  is a complex measure on  $\mathcal{A}$  and  $f \in L^1(|\mu|)$ , there are two natural ways to define  $\int_X f d\mu$ . One way uses decomposition (5) of  $\mu$  as a linear combination of four finite positive measures  $\mu_k$ , which clearly satisfy  $\mu_k \leq |\mu|$ . Therefore  $f \in L^1(\mu_k)$  for all  $k = 0, \dots, 3$ , and we define  $\int_X f d\mu = \sum_{k=0}^3 i^k \int f d\mu_k$ . A second possible definition uses Theorem 1.46. Since  $|h| = 1$ ,  $fh \in L^1(|\mu|)$ , and we may define  $\int_X f d\mu = \int_X fh d|\mu|$ . One verifies easily that the two definitions above give the same value to the integral  $\int_X f d\mu$ . The integral thus defined is a linear functional on  $L^1(|\mu|)$ . As usual,  $\int_E f d\mu := \int_X f I_E d\mu$  for  $E \in \mathcal{A}$ .

By Theorem 1.46, every complex measure  $\lambda$  is of the form  $d\lambda = g d\mu$  for some (finite) positive measure  $\mu$  ( $= |\lambda|$ ) and a uniquely determined  $g \in L^1(\mu)$ .

Conversely, given a positive measure  $\mu$  and  $g \in L^1(\mu)$ , we may define  $d\lambda := g d\mu$  (as before, the meaning of this symbolic relation is that  $\lambda(E) = \int_E g d\mu$  for all  $E \in \mathcal{A}$ ). If  $\{E_k\}$  is a partition of  $E$ , and  $F_n = \bigcup_{k=1}^n E_k$ , then

$$\begin{aligned} \lambda(F_n) &= \int g I_{F_n} d\mu = \int \sum_{k=1}^n g I_{E_k} d\mu \\ &= \sum_{k=1}^n \lambda(E_k). \end{aligned}$$

Since  $g I_{F_n} \rightarrow g I_E$  pointwise as  $n \rightarrow \infty$  and  $|g I_{F_n}| \leq |g| \in L^1(\mu)$ , the dominated convergence theorem implies that the series  $\sum_{k=1}^\infty \lambda(E_k)$  converges to  $\int g I_E d\mu := \lambda(E)$ . Thus  $\lambda$  is a complex measure. The following theorem gives its total variation measure.

**Theorem 1.47.** *Let  $(X, \mathcal{A}, \mu)$  be a  $\sigma$ -finite positive measure space,  $g \in L^1(\mu)$ , and let  $\lambda$  be the complex measure  $d\lambda := g d\mu$ . Then  $d|\lambda| = |g| d\mu$ .*

**Proof.** By Theorem 1.46,  $d\lambda = h d|\lambda|$  with  $h$  measurable and  $|h| = 1$  on  $X$ . For all  $E \in \mathcal{A}$

$$\int_E \bar{h} g d\mu = \int_E \bar{h} d\lambda = \int_E \bar{h} h d|\lambda| = |\lambda|(E). \quad (*)$$

Therefore, whenever  $0 < \mu(E) < \infty$ ,

$$\frac{1}{\mu(E)} \int_E \bar{h}g \, d\mu = \frac{|\lambda|(E)}{\mu(E)} \geq 0.$$

By Lemma 1.38,  $\bar{h}g \geq 0$   $\mu$ -a.e. Hence  $\bar{h}g = |\bar{h}g| = |g|$   $\mu$ -a.e., and therefore, by (\*),

$$|\lambda|(E) = \int_E |g| \, d\mu \quad (E \in \mathcal{A}).$$

□

(The  $\sigma$ -finiteness hypothesis can be dropped: use Proposition 1.22 instead of Lemma 1.38.)

If  $\mu$  is a *real* measure and  $\mu = \mu^+ - \mu^-$  is its Jordan decomposition, the following theorem expresses the positive measures  $\mu^+$  and  $\mu^-$  in terms of a decomposition of the space  $X$  as the disjoint union of two measurable sets  $A$  and  $B$  that carry them (respectively). In particular,  $\mu^+ \perp \mu^-$ .

**Theorem 1.48 (Hahn decomposition).** *Let  $\mu$  be a real measure on the measurable space  $(X, \mathcal{A})$ . Then there exist disjoint sets  $A, B \in \mathcal{A}$  such that  $X = A \cup B$  and*

$$\mu^+(E) = \mu(E \cap A), \quad \mu^-(E) = -\mu(E \cap B)$$

for all  $E \in \mathcal{A}$ .

**Proof.** By Theorem 1.46,  $d\mu = h \, d|\mu|$  with  $h$  measurable and  $|h| = 1$  on  $X$ . For all  $E \in \mathcal{A}$  with  $|\mu|(E) > 0$ ,

$$\frac{1}{|\mu|(E)} \int_E h \, d|\mu| = \frac{\mu(E)}{|\mu|(E)} \in \mathbb{R}.$$

By the averages lemma, it follows that  $h$  is real  $|\mu|$ -a.e., and since it is only determined a.e., we may assume that  $h$  is real everywhere on  $X$ . However  $|h| = 1$ ; hence  $h(X) = \{-1, 1\}$ . Let  $A := [h = 1]$  and  $B = [h = -1]$ . Then  $X$  is the disjoint union of these measurable sets, and for all  $E \in \mathcal{A}$ ,

$$\begin{aligned} \mu^+(E) &:= (1/2)(|\mu|(E) + \mu(E)) \\ &= (1/2) \int_E (1 + h) \, d|\mu| \\ &= \int_{E \cap A} h \, d|\mu| = \mu(E \cap A). \end{aligned}$$

An analogous calculation shows that  $\mu^-(E) = -\mu(E \cap B)$ . □

## 1.10 Convergence

We consider in this section some modes of convergence and relations between them. In order to avoid repetitions,  $(X, \mathcal{A}, \mu)$  will denote throughout a positive measure space;  $f, f_n (n = 1, 2, \dots)$  are complex measurable functions on  $X$ .

**Definition 1.49.**

- (1)  $f_n$  converge to  $f$  *almost uniformly* if for any  $\epsilon > 0$ , there exists  $E \in \mathcal{A}$  with  $\mu(E) < \epsilon$ , such that  $f_n \rightarrow f$  uniformly on  $E^c$ .
- (2)  $\{f_n\}$  is *almost uniformly Cauchy* if for any  $\epsilon > 0$ , there exists  $E \in \mathcal{A}$  with  $\mu(E) < \epsilon$ , such that  $\{f_n\}$  is uniformly Cauchy on  $E^c$ .

**Remark 1.50.**

- (1) Taking  $\epsilon = 1/k$  with  $k = 1, 2, \dots$ , we see that if  $\{f_n\}$  is almost uniformly Cauchy, then there exist  $E_k \in \mathcal{A}$  such that  $\mu(E_k) < 1/k$  and  $\{f_n\}$  is uniformly Cauchy on  $E_k^c$ . Let  $E = \bigcap E_k$ ; then  $E \in \mathcal{A}$  and  $\mu(E) < 1/k$  for all  $k$ , so that  $\mu(E) = 0$ . If  $x \in E^c = \bigcup E_k^c$ , then  $x \in E_k^c$  for some  $k$ , so that  $\{f_n(x)\}$  is Cauchy, and consequently  $\exists \lim_n f_n(x) := f(x)$ . We may define  $f(x) = 0$  on  $E$ . The function  $f$  is measurable, and  $f_n \rightarrow f$  almost everywhere (since  $\mu(E) = 0$ ). For any  $\epsilon, \delta > 0$ , let  $F \in \mathcal{A}$ ,  $n_0 \in \mathbb{N}$  be such that  $\mu(F) < \epsilon$  and  $|f_n(x) - f_m(x)| < \delta$  for all  $x \in F$  and  $n, m > n_0$ . Setting  $G = F \cap E^c$ , we have  $\mu(G) < \epsilon$ , and letting  $m \rightarrow \infty$  in the last inequality, we get  $|f_n(x) - f(x)| \leq \delta$  for all  $x \in G$  and  $n > n_0$ . Thus  $f_n \rightarrow f$  uniformly on  $G$ , and consequently  $f_n \rightarrow f$  almost uniformly. This shows that *almost uniformly Cauchy sequences converge almost uniformly*; the converse follows trivially from the triangle inequality.
- (2) A trivial modification of the first argument above shows that if  $f_n \rightarrow f$  almost uniformly, then  $f_n \rightarrow f$  almost everywhere. In particular, the almost uniform limit  $f$  is uniquely determined up to equivalence.

**Definition 1.51.**

- (1) The sequence  $\{f_n\}$  converges to  $f$  *in measure* if for any  $\epsilon > 0$ ,

$$\lim_n \mu(|f_n - f| \geq \epsilon) = 0.$$

- (2) The sequence  $\{f_n\}$  is *Cauchy in measure* if for any  $\epsilon > 0$ ,

$$\lim_{n, m \rightarrow \infty} \mu(|f_n - f_m| \geq \epsilon) = 0.$$

**Remark 1.52.**

- (1) If  $f_n \rightarrow f$  and  $f_n \rightarrow f'$  in measure, then for any  $\epsilon > 0$ , the triangle inequality shows that

$$\mu(|f - f'| \geq \epsilon) \leq \mu(|f - f_n| \geq \epsilon/2) + \mu(|f_n - f'| \geq \epsilon/2) \rightarrow 0$$

as  $n \rightarrow \infty$ , that is,  $|f - f'| \geq \epsilon$  is  $\mu$ -null. Therefore  $[f \neq f'] = \bigcup_k [|f - f'| \geq 1/k]$  is  $\mu$ -null, and  $f = f'$  a.e. This shows that *limits in measure are uniquely determined (up to equivalence)*.

- (2) A similar argument based on the triangle inequality shows that if  $\{f_n\}$  converges in measure, then it is Cauchy in measure.

- (3) If  $f_n \rightarrow f$  almost uniformly, then  $f_n \rightarrow f$  in measure. Indeed, for any  $\epsilon, \delta > 0$ , there exists  $E \in \mathcal{A}$  and  $n_0 \in \mathbb{N}$ , such that  $\mu(E) < \delta$  and  $|f_n - f| < \epsilon$  on  $E^c$  for all  $n \geq n_0$ . Hence for all  $n \geq n_0$ ,  $[|f_n - f| \geq \epsilon] \subset E$ , and therefore  $\mu([|f_n - f| \geq \epsilon]) < \delta$ .

**Theorem 1.53.** *The sequence  $\{f_n\}$  converges in measure if and only if it is Cauchy in measure.*

**Proof.** By Remark 1.52.2, we need only to show the ‘if’ part of the theorem. Let  $\{f_n\}$  be Cauchy in measure. As in the proof of Lemma 1.30, we obtain integers  $1 \leq n_1 < n_2 < n_3, \dots$  such that  $\mu(E_k) < (1/2^k)$ , where

$$E_k := \left[ |f_{n_{k+1}} - f_{n_k}| \geq \frac{1}{2^k} \right].$$

The set  $F_m = \bigcup_{k \geq m} E_k$  has measure  $< \sum_{k \geq m} 2^{-k} = (1/2^{m-1})$ , and on  $F_m^c$  we have for  $j > i \geq m$

$$|f_{n_j} - f_{n_i}| \leq \sum_{k=i}^{j-1} |f_{n_{k+1}} - f_{n_k}| < \sum_{k=i}^{j-1} 2^{-k} < \frac{1}{2^{i-1}}.$$

This shows that  $\{f_{n_k}\}$  is almost uniformly Cauchy. By Remark 1.50,  $\{f_{n_k}\}$  converges almost uniformly to a measurable function  $f$ . Hence  $f_{n_k} \rightarrow f$  in measure, by Remark 1.52.3. For any  $\epsilon > 0$ , we have

$$[|f_n - f| \geq \epsilon] \subset [|f_n - f_{n_k}| \geq \epsilon/2] \cup [|f_{n_k} - f| \geq \epsilon/2]. \quad (1)$$

The measure of the first set on the right-hand side tends to zero when  $n, k \rightarrow \infty$ , since  $\{f_n\}$  is Cauchy in measure. The measure of the second set on the right-hand side of (1) tends to zero when  $k \rightarrow \infty$ , since  $f_{n_k} \rightarrow f$  in measure. Hence the measure of the set on the left-hand side of (1) tends to zero as  $n \rightarrow \infty$ .  $\square$

**Theorem 1.54.** *If  $f_n \rightarrow f$  in  $L^p$  for some  $p \in [1, \infty]$ , then  $f_n \rightarrow f$  in measure.*

**Proof.** For any  $\epsilon > 0$  and  $n \in \mathbb{N}$ , set  $E_n = [|f_n - f| \geq \epsilon]$ .

*Case  $p < \infty$ .* We have

$$\epsilon^p \mu(E_n) \leq \int_{E_n} |f_n - f|^p d\mu \leq \|f_n - f\|_p^p,$$

and consequently  $f_n \rightarrow f$  in  $L^p$  implies  $\mu(E_n) \rightarrow 0$ .

*Case  $p = \infty$ .* Let  $A = \bigcup_n A_n$ , where  $A_n = [|f_n - f| > \|f_n - f\|_\infty]$ . By definition of the  $L^\infty$ -norm, each  $A_n$  is null, and therefore  $A$  is null, and  $|f_n - f| \leq \|f_n - f\|_\infty$  on  $A^c$  for all  $n$  (hence  $f_n \rightarrow f$  uniformly on  $A^c$  and  $\mu(A) = 0$ ; in such a situation, one says that  $f_n \rightarrow f$  uniformly almost everywhere). If  $f_n \rightarrow f$  in  $L^\infty$ , there exists  $n_0$  such that  $\|f_n - f\|_\infty < \epsilon$  for all  $n \geq n_0$ . Thus  $|f_n - f| < \epsilon$  on  $A^c$  for all  $n \geq n_0$ , hence  $E_n \subset A$  for all  $n \geq n_0$ , and consequently  $E_n$  is null for all  $n \geq n_0$ .  $\square$

## 1.11 Convergence on finite measure space

On *finite* measure spaces, there exist some additional relations between the various types of convergence. A sample of such relations is discussed in the sequel.

**Theorem 1.55.** *Let  $(X, \mathcal{A}, \mu)$  be a finite measure space, and let  $f_n : X \rightarrow \mathbb{C}$  be measurable functions converging almost everywhere to the (measurable) function  $f$ . Then  $f_n \rightarrow f$  in measure.*

**Proof.** Translating the definition of *non-convergence* into set theoretic operations, we have (on any measure space!)

$$[f_n \text{ does not converge to } f] = \bigcup_{k \in \mathbb{N}} \limsup_n [|f_n - f| \geq 1/k].$$

This set is null (i.e.  $f_n \rightarrow f$  a.e.) if and only if  $\limsup_n [|f_n - f| \geq 1/k]$  is null for all  $k$ . Since  $\mu(X) < \infty$ , this is equivalent to

$$\lim_n \mu \left( \bigcup_{m \geq n} [|f_m - f| \geq 1/k] \right) = 0$$

for all  $k$ , which clearly implies that  $\lim_n \mu([|f_n - f| \geq 1/k]) = 0$  for all  $k$  (i.e.  $f_n \rightarrow f$  in measure).  $\square$

**Remark 1.56.**

- (1) Conversely, if  $f_n \rightarrow f$  in measure (in an arbitrary measure space), then there exists a subsequence  $f_{n_k}$  converging a.e. to  $f$  (cf. proof of Theorem 1.53).
- (2) If the *bounded* sequence  $\{f_n\}$  converges a.e. to  $f$ , then  $f_n \rightarrow f$  in  $L^p$  for any  $1 \leq p < \infty$  (by the proposition following Theorem 1.26; in an arbitrary measure space, the boundedness condition on the sequence must be replaced by its majoration by a fixed  $L^p$ -function, not necessarily constant).
- (3) If  $1 \leq r < p \leq \infty$ ,  $L^p$ -convergence implies  $L^r$ -convergence (by the second proposition following Theorem 1.26).

**Theorem 1.57 (Egoroff).** *Let  $(X, \mathcal{A}, \mu)$  be a finite positive measure space, and let  $\{f_n\}$  be measurable functions converging pointwise a.e. to the function  $f$ . Then  $f_n \rightarrow f$  almost uniformly.*

**Proof.** We first prove the following:

**Lemma** (Assumptions and notation as in theorem). *Given  $\epsilon, \delta > 0$ , there exist  $A \in \mathcal{A}$  with  $\mu(A) < \delta$  and  $N \in \mathbb{N}$  such that  $|f_n - f| < \epsilon$  on  $A^c$  for all  $n \geq N$ .*

**Proof of lemma.** Denote  $E_n := [|f_n - f| \geq \epsilon]$  and  $A_N := \bigcup_{n \geq N} E_n$ . Then  $\{A_N\}$  is a decreasing sequence of measurable sets, and since  $\mu$  is a finite measure,  $\mu(\bigcap_N A_N) = \lim_N \mu(A_N)$ . Clearly  $f_n(x)$  does not converge to  $f(x)$  when  $x \in \bigcap_N A_N$ , and since  $f_n \rightarrow f$  a.e., it follows that  $\mu(\bigcap_N A_N) = 0$ , that is,  $\lim_N \mu(A_N) = 0$ . Fix then  $N$  such that  $\mu(A_N) < \delta$  and choose  $A := A_N$ . Since  $A^c = \bigcap_{n \geq N} [|f_n - f| < \epsilon]$ , the set  $A$  satisfies the lemma's requirements.  $\square$

**Proof of theorem.** Given  $\epsilon, \delta > 0$ , apply the lemma with  $\epsilon_m = 1/m$  and  $\delta_m = (\delta/2^m)$ ,  $m = 1, 2, \dots$ . We get measurable sets  $A_m$  with  $\mu(A_m) < \delta_m$  and integers  $N_m$ , such that  $|f_n - f| < 1/m$  on  $A_m^c$  for all  $n \geq N_m$  ( $m = 1, 2, \dots$ ). Let  $A := \bigcup_m A_m$ ; then  $\mu(A) < \delta$ , and on  $A^c (= \bigcap_m A_m^c)$ , we have  $|f_n - f| < 1/m$  for all  $n \geq N_m$ ,  $m = 1, 2, \dots$ . Fix an integer  $m_0 > 1/\epsilon$ , and let  $N := N_{m_0}$ ; then  $|f_n - f| < \epsilon$  on  $A^c$  for all  $n \geq N$ .  $\square$

## 1.12 Distribution function

**Definition 1.58.** Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $f : X \rightarrow [0, \infty]$  be measurable. The *distribution function* of  $f$  is defined by

$$m(y) = \mu([f > y]) \quad (y > 0). \quad (1)$$

This is a non-negative non-increasing function on  $\mathbb{R}^+$ , so that  $m(\infty) := \lim_{y \rightarrow \infty} m(y)$  exists and is  $\geq 0$ . We shall *assume* in the sequel that  $m$  is *finite-valued* and  $m(\infty) = 0$ . The finiteness of  $m$  implies that

$$m(a) - m(b) = \mu([a < f \leq b]) \quad (0 < a < b < \infty).$$

Let  $\{y_n\}$  be any positive sequence increasing to  $\infty$ . If  $E_n := [f > y_n]$ , then  $E_{n+1} \subset E_n$  and  $\bigcap E_n = [f = \infty]$ . Since  $m$  is finite-valued, we have by Lemma 1.11

$$m(\infty) = \lim m(y_n) = \lim \mu(E_n) = \mu\left(\bigcap E_n\right) = \mu([f = \infty]).$$

Thus our second assumption above means that  $f$  is finite  $\mu$ -a.e.

Both assumptions above are satisfied in particular when  $\int_X f^p d\mu < \infty$  for some  $p \in [1, \infty)$ . This follows from the inequality

$$m(y) \leq \left( \frac{\|f\|_p}{y} \right)^p \quad (y > 0) \quad (2)$$

(cf. proof of Theorem 1.54).

**Theorem 1.59.** Suppose the distribution function  $m$  of the non-negative measurable function  $f$  is finite and vanishes at infinity. Then:

(1) For all  $p \in [1, \infty)$

$$\int_X f^p d\mu = - \int_0^\infty y^p dm(y), \quad (3)$$

where the integral on the right-hand side is the improper Riemann–Stieltjes integral

$$\lim_{a \rightarrow 0+; b \rightarrow \infty} \int_a^b y^p dm(y).$$

(2) If either one of the integrals  $-\int_0^\infty y^p dm(y)$  and  $p \int_0^\infty y^{p-1} m(y) dy$  is finite, then

$$\lim_{y \rightarrow 0} y^p m(y) = \lim_{y \rightarrow \infty} y^p m(y) = 0, \quad (4)$$

and the integrals coincide.

**Proof.** Let  $0 < a < b < \infty$  and  $n \in \mathbb{N}$ . Denote

$$\begin{aligned} y_j &= a + j \frac{b-a}{n2^n}, \quad j = 0, \dots, n2^n; \\ E_j &= [y_{j-1} < f \leq y_j]; \quad E_{a,b} = [a < f \leq b]; \\ s_n &= \sum_{j=1}^{n2^n} y_{j-1} I_{E_j}. \end{aligned}$$

The sequence  $\{s_n^p\}$  is a non-decreasing sequence of non-negative measurable functions with limit  $f^p$  (cf. proof of Theorem 1.8). By the Monotone Convergence theorem,

$$\begin{aligned} \int_{E_{a,b}} f^p d\mu &= \lim_n \int_{E_{a,b}} s_n^p d\mu = \lim_n \sum_{j=1}^{n2^n} y_{j-1}^p \mu(E_j) \\ &= - \lim_n \sum_{j=1}^{n2^n} y_{j-1}^p [m(y_j) - m(y_{j-1})] = - \int_a^b y^p dm(y). \end{aligned}$$

The first integral above converges to the (finite or infinite) limit  $\int_X f^p d\mu$  when  $a \rightarrow 0$  and  $b \rightarrow \infty$ . It follows that the last integral above converges to the same limit, that is, the improper Riemann–Stieltjes integral  $\int_0^\infty y^p dm(y)$  exists and (3) is valid.

Since  $-dm$  is a positive measure, we have

$$0 \leq a^p [m(a) - m(b)] = - \int_a^b a^p dm(y) \leq - \int_a^b y^p dm(y), \quad (5)$$

hence

$$0 \leq a^p m(a) \leq a^p m(b) - \int_a^b y^p dm(y). \quad (6)$$

Let  $b \rightarrow \infty$ . Since  $m(\infty) = 0$ ,

$$0 \leq a^p m(a) \leq - \int_a^\infty y^p dm(y). \quad (7)$$

In case  $\int_0^\infty y^p dm(y)$  is finite, letting  $a \rightarrow \infty$  in (7) shows that  $\lim_{a \rightarrow \infty} a^p m(a) = 0$ . Also letting  $a \rightarrow 0$  in (6) (with  $b$  fixed arbitrary) shows that

$$0 \leq \limsup_{a \rightarrow 0} a^p m(a) \leq - \int_0^b y^p dm(y).$$

Letting  $b \rightarrow 0$ , we conclude that  $\exists \lim_{a \rightarrow 0} a^p m(a) = 0$ , and (4) is verified.

An integration by parts gives

$$- \int_a^b y^p dm(y) = a^p m(a) - b^p m(b) + p \int_a^b y^{p-1} m(y) dy. \quad (8)$$

Letting  $a \rightarrow 0$  and  $b \rightarrow \infty$ , we obtain from (4) (in case  $\int_0^\infty y^p dm(y)$  is finite)

$$- \int_0^\infty y^p dm(y) = p \int_0^\infty y^{p-1} m(y) dy. \quad (9)$$

Consider finally the case when  $\int_0^\infty y^{p-1} m(y) dy < \infty$ . We have

$$\begin{aligned} (1 - 2^{-p})b^p m(b) &= [b^p - (b/2)^p]m(b) = m(b) \int_{b/2}^b py^{p-1} dy \\ &\leq p \int_{b/2}^b y^{p-1} m(y) dy \rightarrow 0 \end{aligned}$$

as  $b \rightarrow \infty$  or  $b \rightarrow 0$  (by Cauchy's criterion). Thus (4) is verified, and (9) was seen to follow from (4).  $\square$

**Corollary 1.60.** *Let  $f \in L^p(\mu)$  for some  $p \in [1, \infty)$ , and let  $m$  be the distribution function of  $|f|$ . Then*

$$\|f\|_p^p = - \int_0^\infty y^p dm(y) = p \int_0^\infty y^{p-1} m(y) dy.$$

## 1.13 Truncation

**Technique 1.61.** The technique of truncation of functions is useful in real methods of analysis. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $f : X \rightarrow \mathbb{C}$ . For each  $u > 0$ , we define the *truncation at  $u$  of  $f$*  by

$$f_u := fI_{[|f| \leq u]} + u(f/|f|)I_{[|f| > u]}. \quad (1)$$

Denote

$$f'_u := f - f_u = (f - u(f/|f|))I_{[|f| > u]} = (|f| - u)(f/|f|)I_{[|f| > u]}. \quad (2)$$

We have

$$|f_u| = |f|I_{[|f| \leq u]} + uI_{[|f| > u]} = \min(|f|, u), \quad (3)$$

$$|f'_u| = (|f| - u)I_{[|f| > u]}, \quad (4)$$

$$f = f_u + f'_u, \quad |f| = |f_u| + |f'_u|. \quad (5)$$



It follows in particular that  $f \in L^p(\mu)$  for some  $p \in [1, \infty]$  iff both  $f_u$  and  $f'_u$  are in  $L^p(\mu)$ . In this case  $f_u \in L^r(\mu)$  for any  $r \geq p$ , because  $|f_u/u| \leq 1$ , so that

$$u^{-r}|f_u|^r = |f_u/u|^r \leq |f_u/u|^p \leq u^{-p}|f|^p.$$

Similarly (still when  $f \in L^p$ ),  $f'_u \in L^r(\mu)$  for any  $r \leq p$ . Indeed, write

$$\int_X |f'_u|^r d\mu = \int_{[|f'_u| > 1]} + \int_{[0 < |f'_u| \leq 1]}.$$

For  $r \leq p$ , the first integral on the right-hand side is

$$\leq \int_{[|f'_u| > 1]} |f'_u|^p d\mu \leq \|f'_u\|_p^p;$$

the second integral on the right-hand side is

$$\begin{aligned} &\leq \mu([0 < |f'_u| \leq 1]) = \mu([0 < |f| - u \leq 1]) \\ &= \mu([u < |f| \leq u + 1]) = m(u) - m(u + 1). \end{aligned}$$

Thus for any  $r \leq p$ ,

$$\|f'_u\|_r^r \leq \|f'_u\|_p^p + m(u) - m(u + 1) < \infty,$$

as claimed.

Since  $|f_u| = \min(|f|, u)$ , we have  $[|f_u| > y] = [|f| > y]$  whenever  $0 < y < u$  and  $[|f_u| > y] = \emptyset$  whenever  $y \geq u$ . Therefore, if  $m_u$  and  $m$  are the distribution functions of  $|f_u|$  and  $|f|$ , respectively, we have

$$m_u(y) = m(y) \quad \text{for } 0 < y < u; \quad m_u(y) = 0 \quad \text{for } y \geq u. \quad (6)$$

For the distribution function  $m'_u$  of  $|f'_u|$ , we have the relation

$$m'_u(y) = m(y + u) \quad (y > 0), \quad (7)$$

since by (4)

$$m'_u(y) := \mu([|f'_u| > y]) = \mu([|f| - u > y]) = \mu([|f| > y + u]) = m(y + u).$$

By (6), (7), and Corollary 1.60, the following formulae are valid for any  $f \in L^p(\mu)$  ( $1 \leq p < \infty$ ) and  $u > 0$ :

$$\|f_u\|_r^r = r \int_0^u v^{r-1} m(v) dv \quad (r \geq p); \quad (8)$$

$$\|f'_u\|_r^r = r \int_u^\infty (v - u)^{r-1} m(v) dv \quad (r \leq p). \quad (9)$$

These formulae will be used in Section 5.40.

## Exercises

- Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $f$  be a non-negative measurable function on  $X$ . Let  $E := [f < 1]$ . Prove
  - $\mu(E) = \lim_n \int_E \exp(-f^n) d\mu$ .
  - $\sum_{n=1}^{\infty} \int_E f^n d\mu = \int_E (f/(1-f)) d\mu$ .
- Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $p, q$  be conjugate exponents. Prove that the map  $[f, g] \in L^p(\mu) \times L^q(\mu) \rightarrow fg \in L^1(\mu)$  is continuous.
- Let  $(X, \mathcal{A}, \mu)$  be a positive measure space,  $f_n : X \rightarrow \mathbb{C}$  measurable functions converging pointwise to  $f$ , and  $h : \mathbb{C} \rightarrow \mathbb{C}$  continuous and bounded. Prove that  $\lim_n \int_E h(f_n) d\mu = \int_E h(f) d\mu$  for each  $E \in \mathcal{A}$  with finite measure.
- Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $\mathcal{B} \subset \mathcal{A}$  be a  $\sigma$ -finite  $\sigma$ -algebra. If  $f \in L^1(\mathcal{A}) := L^1(X, \mathcal{A}, \mu)$ , consider the complex measure on  $\mathcal{B}$  defined by

$$\lambda_f(E) := \int_E f d\mu \quad (E \in \mathcal{B}).$$

Prove:

- There exists a unique element  $Pf \in L^1(\mathcal{B}) := L^1(X, \mathcal{B}, \mu)$  such that

$$\lambda_f(E) = \int_E (Pf) d\mu \quad (E \in \mathcal{B}).$$

- The map  $P : f \rightarrow Pf$  is a continuous linear map of  $L^1(\mathcal{A})$  onto the subspace  $L^1(\mathcal{B})$ , such that  $P^2 = P$  ( $P^2$  denotes the composition of  $P$  with itself). In particular,  $L^1(\mathcal{B})$  is a *closed* subspace of  $L^1(\mathcal{A})$ .
- Let  $(X, \mathcal{A}, \mu)$  be a finite positive measure space and  $f_n \in L^p(\mu)$  for all  $n \in \mathbb{N}$  (for some  $p \in [1, \infty)$ ). Suppose there exists a measurable function  $f : X \rightarrow \mathbb{C}$  such that  $\sup_n \sup_X |f_n - f| < \infty$  and  $f_n \rightarrow f$  in measure. Prove that  $f \in L^p(\mu)$  and  $f_n \rightarrow f$  in  $L^p$ -norm.
  - Let  $\lambda$  and  $\mu$  be positive  $\sigma$ -finite measures on the measurable space  $(X, \mathcal{A})$ . State and prove a version of the Lebesgue–Radon–Nikodym theorem for this situation.
  - Let  $\{\lambda_n\}$  be a sequence of complex measures on the measurable space  $(X, \mathcal{A})$  such that  $\sum_n \|\lambda_n\| < \infty$ . Prove
    - For each  $E \in \mathcal{A}$ , the series  $\sum_n \lambda_n(E)$  converges absolutely in  $\mathbb{C}$  and defines a complex measure  $\lambda$ ; the series  $\sum_n |\lambda_n|(E)$  converges in  $\mathbb{R}^+$ , and defines a finite positive measure  $\sigma$ , and  $\lambda \ll \sigma$ .

(b)

$$\frac{d\lambda}{d\sigma} = \sum_n \frac{d\lambda_n}{d\sigma}.$$

8. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $M := M(\mathcal{A})$  denote the vector space (over  $\mathbb{C}$ ) of all complex measures on  $\mathcal{A}$ . Set

$$M_a := \{\lambda \in M; \lambda \ll \mu\};$$

$$M_s := \{\lambda \in M; \lambda \perp \mu\}.$$

Prove:

- (a) If  $\lambda \in M$  is supported by  $E \in \mathcal{A}$ , then so is  $|\lambda|$ .
- (b)  $M_a$  and  $M_s$  are subspaces of  $M$  and  $M_a \perp M_s$  (in particular,  $M_a \cap M_s = \{0\}$ ).
- (c) If  $(X, \mathcal{A}, \mu)$  is  $\sigma$ -finite, then  $M = M_a \oplus M_s$ .
- (d)  $\lambda \in M_a$  iff  $|\lambda| \in M_a$  (and similarly for  $M_s$ ).
- (e) If  $\lambda_k \in M$  ( $k = 1, 2$ ), then  $\lambda_1 \perp \lambda_2$  iff  $|\lambda_1| \perp |\lambda_2|$ .
- (f)  $\lambda \ll \mu$  iff for each  $\epsilon > 0$ , there exists  $\delta > 0$  such that  $|\lambda(E)| < \epsilon$  for all  $E \in \mathcal{A}$  with  $\mu(E) < \delta$ .

(Hint: if the  $\epsilon, \delta$  condition fails, there exist  $E_n \in \mathcal{A}$  with  $\mu(E_n) < 1/2^n$  such that  $|\lambda(E_n)| \geq \epsilon$  (hence  $|\lambda|(E_n) \geq \epsilon$ ), for some  $\epsilon > 0$ ; consider the set  $E = \limsup E_n$ .)

9. Let  $(X, \mathcal{A}, \mu)$  be a *probability space* (i.e. a positive measure space such that  $\mu(X) = 1$ ). Let  $f, g$  be (complex) measurable functions. Prove that  $\|f\|_1 \|g\|_1 \geq \inf_X |fg|$ .
10. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space and  $f$  a complex measurable function on  $X$ .

- (a) If  $\mu(X) < \infty$ , prove that

$$\lim_{p \rightarrow \infty} \|f\|_p = \|f\|_\infty. \quad (*)$$

(The cases  $\|f\|_\infty = 0$  or  $\infty$  are trivial; we may then assume that  $\|f\|_\infty = 1$ ; given  $\epsilon$ , there exists  $E \in \mathcal{A}$  such that  $\mu(E) > 0$  and  $(1 - \epsilon)\mu(E)^{1/p} \leq \|f\|_p \leq \mu(X)^{1/p}$ .)

- (b) For an arbitrary positive measure space, if  $\|f\|_r < \infty$  for some  $r \in [1, \infty)$ , then  $(*)$  is valid.

(Consider the finite positive measure  $\nu(E) = \int_E |f|^r d\mu$ . We may assume as in Part (a) that  $\|f\|_\infty = 1$ . Verify that  $\|f\|_{L^\infty(\nu)} = 1$  and  $\|f\|_p = \|f\|_{L^{p-r}(\nu)}^{1-r/p}$  for all  $p \geq r + 1$ .)

11. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space,  $1 \leq p < \infty$ , and  $\epsilon > 0$ .
- (a) Suppose  $f_n, f$  are unit vectors in  $L^p(\mu)$  such that  $f_n \rightarrow f$  a.e. Consider the probability measure  $d\nu = |f|^p d\mu$ . Show that there exists  $E \in \mathcal{A}$  such that  $f_n/f \rightarrow 1$  uniformly on  $E$  and  $\nu(E^c) < \epsilon$ . (Hint: Egoroff's theorem.)
  - (b) For  $E$  as in Part (a), show that  $\limsup_n \int_{E^c} |f_n|^p d\mu < \epsilon$ .
  - (c) Deduce from Parts (a) and (b) that  $f_n \rightarrow f$  in  $L^p(\mu)$ -norm.
  - (d) If  $g_n, g \in L^p(\mu)$  are such that  $g_n \rightarrow g$  a.e. and  $\|g_n\|_p \rightarrow \|g\|_p$ , then  $g_n \rightarrow g$  in  $L^p(\mu)$ -norm. (Consider  $f_n = g_n/\|g_n\|_p$  and  $f = g/\|g\|_p$ .)
12. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space and  $f \in L^p(\mu)$  for some  $p \in [1, \infty)$ . Prove that the set  $[f \neq 0]$  has  $\sigma$ -finite measure.
13. Let  $(X, \mathcal{A})$  be a measurable space, and let  $f_n : X \rightarrow \mathbb{C}$ ,  $n \in \mathbb{N}$ , be measurable functions. Prove that the set of all points  $x \in X$  for which the complex sequence  $\{f_n(x)\}$  converges in  $\mathbb{C}$  is measurable.
14. Let  $(X, \mathcal{A})$  be a measurable space, and let  $E$  be a dense subset of  $\mathbb{R}$ . Suppose  $f : X \rightarrow \mathbb{R}$  is such that  $[f \geq c] \in \mathcal{A}$  for all  $c \in E$ . Prove that  $f$  is measurable.
15. Let  $(X, \mathcal{A})$  be a measurable space, and let  $f : X \rightarrow \mathbb{R}^+$  be measurable. Prove that there exist  $c_k > 0$  and  $E_k \in \mathcal{A}$  ( $k \in \mathbb{N}$ ) such that  $f = \sum_{k=1}^{\infty} c_k I_{E_k}$ . Conclude that for any positive measure  $\mu$  on  $\mathcal{A}$ ,  $\int f d\mu = \sum_{k=1}^{\infty} c_k \mu(E_k)$ ; in particular, if  $f \in L^1(\mu)$ , the series converges (in the strict sense) and  $\mu(E_k) < \infty$  for all  $k$ . (Hint: get  $s_n$  as in the approximation theorem, and observe that  $f = \sum_n (s_n - s_{n-1})$ .)
16. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space, and let  $\{E_k\} \subset \mathcal{A}$  be such that  $\sum_k \mu(E_k) < \infty$ . Prove that almost all  $x \in X$  lie in at most finitely many of the sets  $E_k$ . (Hint: The set of  $x$  that lie in infinitely many  $E_k$ s is  $\limsup E_k$ .)
17. Let  $X$  be a (complex) normed space. Define

$$f(x, y) = \frac{\|x + y\|^2 + \|x - y\|^2}{2\|x\|^2 + 2\|y\|^2} \quad (x, y \in X).$$

(We agree that the fraction is 1 when  $x = y = 0$ .) Prove:

- (a)  $1/2 \leq f \leq 2$ .
- (b)  $X$  is an inner product space iff  $f = 1$  (identically).

## 2

# Construction of measures

## 2.1 Semi-algebras

The purpose of this chapter is to construct measure spaces from more primitive objects. We start with a *semi-algebra*  $\mathcal{C}$  of subsets of a given set  $X$  and a *semi-measure*  $\mu$  defined on it.

**Definition 2.1.** Let  $X$  be a (non-empty) set. A *semi-algebra* of subsets of  $X$  (briefly, a semi-algebra *on*  $X$ ) is a subfamily  $\mathcal{C}$  of  $\mathbb{P}(X)$  with the following properties:

- (1) if  $A, B \in \mathcal{C}$ , then  $A \cap B \in \mathcal{C}$ ;  $\emptyset \in \mathcal{C}$ ;
- (2) if  $A \in \mathcal{C}$ , then  $A^c$  is the union of *finitely* many mutually disjoint sets in  $\mathcal{C}$ .

Any algebra is a semi-algebra, but not conversely. For example, the family

$$\mathcal{C} = \{(a, b]; a, b \in \mathbb{R}\} \cup \{(-\infty, b]; b \in \mathbb{R}\} \cup \{(a, \infty); a \in \mathbb{R}\} \cup \{\emptyset\}$$

is a semi-algebra on  $\mathbb{R}$ , but is not an algebra. Similar semi-algebras of *half-closed cells* arise naturally in the Euclidean space  $\mathbb{R}^k$ .

**Definition 2.2.** Let  $\mathcal{C}$  be a semi-algebra on  $X$ . A *semi-measure* on  $\mathcal{C}$  is a function

$$\mu : \mathcal{C} \rightarrow [0, \infty]$$

with the following properties:

- (1)  $\mu(\emptyset) = 0$ ;
- (2) if  $E_i \in \mathcal{C}$ ,  $i = 1, \dots, n$  are mutually disjoint with union  $E \in \mathcal{C}$ , then  $\mu(E) = \sum_i \mu(E_i)$ ;
- (3) if  $E_i \in \mathcal{C}$ ,  $i = 1, 2, \dots$  are mutually disjoint with union  $E \in \mathcal{C}$ , then  $\mu(E) \leq \sum_i \mu(E_i)$ .

If  $\mathcal{C}$  is a  $\sigma$ -algebra, any measure on  $\mathcal{C}$  is a semi-measure. A simple ‘natural’ example of a semi-measure on the semi-algebra of half-closed intervals on  $\mathbb{R}$  mentioned above is given by

$$\begin{aligned}\mu((a, b]) &= b - a, \quad a, b \in \mathbb{R}, \quad a < b; \\ \mu(\emptyset) &= 0; \quad \mu((-\infty, b]) = \mu((a, \infty)) = \infty.\end{aligned}$$

Let  $\mathcal{C}$  be a semi-algebra on  $X$ , and let  $\mathcal{A}$  be the family of all finite unions of mutually disjoint sets from  $\mathcal{C}$ . Then  $\emptyset \in \mathcal{A}$ ; if  $A = \bigcup E_i, B = \bigcup F_j \in \mathcal{A}$ , with  $E_i \in \mathcal{C}, i = 1, \dots, m$  disjoint and  $F_j \in \mathcal{C}, j = 1, \dots, n$  disjoint, then  $A \cap B = \bigcup_{i,j} E_i \cap F_j \in \mathcal{A}$  as a finite union of disjoint sets from  $\mathcal{C}$  by Condition (1) in Definition 2.1. Also  $A^c = \bigcap E_i^c \in \mathcal{A}$ , since  $E_i^c \in \mathcal{A}$  by Condition (2) in Definition 2.1, and we just saw that  $\mathcal{A}$  is closed under finite intersections. We conclude that  $\mathcal{A}$  is an *algebra* on  $X$  that includes  $\mathcal{C}$ , and it is obviously contained in any algebra on  $X$  that contains  $\mathcal{C}$ . Thus,  $\mathcal{A}$  is *the algebra generated by the semi-algebra  $\mathcal{C}$*  (i.e. the algebra, minimal under inclusion, that contains  $\mathcal{C}$ ).

**Definition 2.3.** Let  $\mathcal{A}$  be *any* algebra on the set  $X$ . A *measure on the algebra  $\mathcal{A}$*  is a function  $\mu : \mathcal{A} \rightarrow [0, \infty]$  such that  $\mu(\emptyset) = 0$ , and if  $E \in \mathcal{A}$  is the countable union of mutually disjoint sets  $E_i \in \mathcal{A}$ , then  $\mu(A) = \sum \mu(E_i)$  (i.e.  $\mu$  is countably additive *whenever this makes sense*).

**Theorem 2.4.** Let  $\mathcal{C}$  be a semi-algebra on the set  $X$ , let  $\mu$  be a semi-measure on  $\mathcal{C}$ , let  $\mathcal{A}$  be the algebra generated by  $\mathcal{C}$ , and extend  $\mu$  to  $\mathcal{A}$  by letting

$$\mu\left(\bigcup E_i\right) = \sum \mu(E_i) \quad (1)$$

for  $E_i \in \mathcal{C}, i = 1, \dots, n$ , mutually disjoint. Then  $\mu$  is a measure on the algebra  $\mathcal{A}$ .

**Proof.** First,  $\mu$  is well-defined by (1) on  $\mathcal{A}$ . Indeed, if  $E_i, F_j \in \mathcal{C}$  are such that  $A = \bigcup E_i = \bigcup F_j$  (finite *disjoint* unions), then each  $F_j \in \mathcal{C}$  is the finite disjoint union of the sets  $E_i \cap F_j \in \mathcal{C}$ ; by Condition (2) in Definition 2.2,

$$\mu(F_j) = \sum_i \mu(E_i \cap F_j),$$

and therefore

$$\sum \mu(F_j) = \sum_{i,j} \mu(E_i \cap F_j).$$

The symmetry of the right-hand side in  $E_i$  and  $F_j$  implies that it is also equal to  $\sum \mu(E_i)$ , and the definition (1) is indeed independent of the representation of  $A \in \mathcal{A}$  as a finite disjoint union of sets in  $\mathcal{C}$ .

It is now clear that  $\mu$  is *finitely* additive on  $\mathcal{A}$ , hence monotonic.

Let  $E \in \mathcal{A}$  be the disjoint union of  $E_i \in \mathcal{A}, i = 1, 2, \dots$ . For each  $n \in \mathbb{N}$ ,  $\bigcup_{i=1}^n E_i \subset E$ , hence

$$\sum_{i=1}^n \mu(E_i) = \mu\left(\bigcup_{i=1}^n E_i\right) \leq \mu(E),$$

and therefore

$$\sum_{i=1}^{\infty} \mu(E_i) \leq \mu(E). \quad (2)$$

Next, write  $E_i = \bigcup_j F_{ij}$ , a finite disjoint union of  $F_{ij} \in \mathcal{C}$ , for each fixed  $i$ , and similarly, since  $E \in \mathcal{A}$ ,  $E = \bigcup_k G_k$ , a finite disjoint union of sets  $G_k \in \mathcal{C}$ . Then  $G_k \in \mathcal{C}$  is the (countable) disjoint union of the sets  $F_{ij} \cap G_k \in \mathcal{C}$  (by Condition (1) in Definition 2.1), over all  $i, j$ . By Condition (3) in Definition 2.2., it follows that for all  $k$ ,

$$\mu(G_k) \leq \sum_{i,j} \mu(F_{ij} \cap G_k).$$

Hence

$$\begin{aligned} \mu(E) &:= \sum \mu(G_k) \leq \sum_{i,j,k} \mu(F_{ij} \cap G_k) \\ &= \sum_i \sum_{j,k} \mu(F_{ij} \cap G_k) = \sum_i \mu(E_i), \end{aligned}$$

by the definition (1) of  $\mu$  on  $\mathcal{A}$ , because  $E_i = \bigcup_{j,k} F_{ij} \cap G_k$ , a finite disjoint union of sets in  $\mathcal{C}$ . Together with (2), this proves that  $\mu$  is indeed a measure on the algebra  $\mathcal{A}$ .  $\square$

## 2.2 Outer measures

A measure on an algebra can be extended to an *outer measure* on  $\mathbb{P}(X)$ :

**Definition 2.5.** An outer measure on the set  $X$  (in fact, on  $\mathbb{P}(X)$ ) is a function

$$\mu^* : \mathbb{P}(X) \rightarrow [0, \infty]$$

with the following properties:

- (1)  $\mu^*(\emptyset) = 0$ ;
- (2)  $\mu^*$  is monotonic (i.e.  $\mu^*(E) \leq \mu^*(F)$  whenever  $E \subset F \subset X$ );
- (3)  $\mu^*$  is countably subadditive, that is

$$\mu^*\left(\bigcup E_i\right) \leq \sum \mu^*(E_i),$$

for any sequence  $\{E_i\} \subset \mathbb{P}(X)$ .

By (1) and (3), outer measures are *finitely* subadditive (i.e. (3) is valid for *finite* sequences  $\{E_i\}$  as well).

**Theorem 2.6.** Let  $\mu$  be a measure on the algebra  $\mathcal{A}$  of subsets of  $X$ . For any  $E \in \mathbb{P}(X)$ , let

$$\mu^*(E) := \inf \sum \mu(E_i),$$

where the infimum is taken over all sequences  $\{E_i\} \subset \mathcal{A}$  with  $E \subset \bigcup E_i$  (briefly, call such sequences ‘ $\mathcal{A}$ -covers of  $E$ ’). Then  $\mu^*$  is an outer measure on  $X$ , called the outer measure generated by  $\mu$ , and  $\mu^*|_{\mathcal{A}} = \mu$ .

**Proof.** We begin by showing that  $\mu^*|_{\mathcal{A}} = \mu$ . If  $E \in \mathcal{A}$ , then  $\{E, \emptyset, \emptyset, \dots\}$  is an  $\mathcal{A}$ -cover of  $E$ , hence  $\mu^*(E) \leq \mu(E)$ . Next, if  $\{E_i\}$  is any  $\mathcal{A}$ -cover of  $E$ , then for all  $n \in \mathbb{N}$ ,

$$F_n := E \cap E_n \cap E_{n-1}^c \cap \dots \cap E_1^c \in \mathcal{A}$$

(since  $\mathcal{A}$  is an algebra), and  $E \in \mathcal{A}$  is the disjoint union of the sets  $F_n \subset E_n$ . Therefore, since  $\mu$  is a measure on the algebra  $\mathcal{A}$ ,

$$\sum_n \mu(E_n) \geq \sum_n \mu(F_n) = \mu(E).$$

Taking the infimum over all  $\mathcal{A}$ -covers  $\{E_n\}$  of  $E$ , we obtain  $\mu^*(E) \geq \mu(E)$ , and the wanted equality follows.

In particular,  $\mu^*(\emptyset) = \mu(\emptyset) = 0$ .

If  $E \subset F \subset X$ , then every  $\mathcal{A}$ -cover of  $F$  is also an  $\mathcal{A}$ -cover of  $E$ ; this implies that  $\mu^*(E) \leq \mu^*(F)$ .

Let  $E_n \subset X, n \in \mathbb{N}$ , and  $E = \bigcup_n E_n$ . For  $\epsilon > 0$  given, and for each  $n \in \mathbb{N}$ , there exists an  $\mathcal{A}$ -cover  $\{E_{n,i}\}_i$  of  $E_n$  such that

$$\sum_i \mu(E_{n,i}) < \mu^*(E_n) + \epsilon/2^n.$$

Since  $\{E_{n,i}; n, i \in \mathbb{N}\}$  is an  $\mathcal{A}$ -cover of  $E$ , we have

$$\mu^*(E) \leq \sum_{n,i} \mu(E_{n,i}) \leq \sum_n \mu^*(E_n) + \epsilon,$$

and the arbitrariness of  $\epsilon$  implies that  $\mu^*$  is countably sub-additive.  $\square$

**Definition 2.7 (The Caratheodory measurability condition).** Let  $\mu^*$  be an outer measure on  $X$ . A set  $E \subset X$  is  $\mu^*$ -measurable if

$$\mu^*(A) = \mu^*(A \cap E) + \mu^*(A \cap E^c) \quad (1)$$

for every  $A \subset X$ .

We shall denote by  $\mathcal{M}$  the family of all  $\mu^*$ -measurable subsets of  $X$ .

By subadditivity of outer measures, (1) is equivalent to the inequality

$$\mu^*(A) \geq \mu^*(A \cap E) + \mu^*(A \cap E^c) \quad (2)$$

(for every  $A \subset X$ ). Since (2) is trivial when  $\mu^*(A) = \infty$ , we can use only subsets  $A$  of finite outer measure in the measurability test (2).

**Theorem 2.8.** Let  $\mu^*$  be an outer measure on  $X$ , let  $\mathcal{M}$  be the family of all  $\mu^*$ -measurable subsets of  $X$ , and let  $\bar{\mu} := \mu^*|_{\mathcal{M}}$ . Then  $(X, \mathcal{M}, \bar{\mu})$  is a complete positive measure space (called the measure space induced by the given outer measure).



**Proof.** If  $\mu^*(E) = 0$ , also  $\mu^*(A \cap E) = 0$  by monotonicity (for all  $A \subset X$ ), and (2) follows (again by monotonicity of  $\mu^*$ ). Hence  $E \in \mathcal{M}$  whenever  $\mu^*(E) = 0$ , and in particular  $\emptyset \in \mathcal{M}$ . By monotonicity, this implies also that the measure space of the theorem is automatically complete.

The symmetry of the Caratheodory condition in  $E$  and  $E^c$  implies that  $E^c \in \mathcal{M}$  whenever  $E \in \mathcal{M}$ .

Let  $E, F \in \mathcal{M}$ . Then for all  $A \subset X$ , it follows from (2) (first for  $F$  with the ‘test set’  $A$ , and then for  $E$  with the test set  $A \cap F^c$ ) and the finite subadditivity of  $\mu^*$ , that

$$\begin{aligned} \mu^*(A) &\geq \mu^*(A \cap F) + \mu^*(A \cap F^c) \\ &\geq \mu^*(A \cap F) + \mu^*(A \cap F^c \cap E) + \mu^*(A \cap F^c \cap E^c) \\ &\geq \mu^*([A \cap F] \cup [A \cap (E - F)]) + \mu^*(A \cap (E \cup F)^c) \\ &= \mu^*(A \cap (E \cup F)) + \mu^*(A \cap (E \cup F)^c), \end{aligned}$$

and we conclude that  $E \cup F \in \mathcal{M}$ , and so  $\mathcal{M}$  is an algebra on  $X$ . It follows in particular that any countable union  $E$  of sets from  $\mathcal{M}$  can be written as a *disjoint* countable union of sets  $E_i \in \mathcal{M}$ . Set  $F_n = \bigcup_{i=1}^n E_i (\subset E)$ ,  $n = 1, 2, \dots$ . Then  $F_n \in \mathcal{M}$ , and therefore, by (2) and monotonicity, we have for all  $A \subset X$

$$\begin{aligned} \mu^*(A) &\geq \mu^*(A \cap F_n) + \mu^*(A \cap F_n^c) \\ &\geq \mu^*(A \cap F_n) + \mu^*(A \cap E^c). \end{aligned} \tag{3}$$

By (1), since  $E_n \in \mathcal{M}$ , we have

$$\begin{aligned} \mu^*(A \cap F_n) &= \mu^*(A \cap F_n \cap E_n) + \mu^*(A \cap F_n \cap E_n^c) \\ &= \mu^*(A \cap E_n) + \mu^*(A \cap F_{n-1}). \end{aligned} \tag{4}$$

The recursion (4) implies that for all  $n \in \mathbb{N}$ ,

$$\mu^*(A \cap F_n) = \sum_{i=1}^n \mu^*(A \cap E_i). \tag{5}$$

Substitute (5) in (3), let  $n \rightarrow \infty$ , and use the  $\sigma$ -subadditivity of  $\mu^*$ ; whence

$$\begin{aligned} \mu^*(A) &\geq \sum_{i=1}^{\infty} \mu^*(A \cap E_i) + \mu^*(A \cap E^c) \\ &\geq \mu^*(A \cap E) + \mu^*(A \cap E^c). \end{aligned}$$

This shows that  $E \in \mathcal{M}$ , and we conclude that  $\mathcal{M}$  is a  $\sigma$ -algebra.

Choosing  $A = F_n$  in (5), we obtain (by monotonicity)

$$\mu^*(E) \geq \mu^*(F_n) = \sum_{i=1}^n \mu^*(E_i), \quad n = 1, 2, \dots$$

Letting  $n \rightarrow \infty$ , we see that

$$\mu^*(E) \geq \sum_{i=1}^{\infty} \mu^*(E_i).$$

Together with the  $\sigma$ -subadditivity of  $\mu^*$ , this proves the  $\sigma$ -additivity of  $\mu^*$  restricted to  $\mathcal{M}$ , as wanted.  $\square$

## 2.3 Extension of measures on algebras

Combining Theorems 2.6 and 2.8, we obtain the *Caratheodory extension theorem*.

**Theorem 2.9.** *Let  $\mu$  be a measure on the algebra  $\mathcal{A}$ , let  $\mu^*$  be the outer measure induced by  $\mu$ , and let  $(X, \mathcal{M}, \bar{\mu})$  be the (complete, positive) measure space induced by  $\mu^*$ . Then  $\mathcal{A} \subset \mathcal{M}$ , and  $\bar{\mu}$  extends  $\mu$  to a measure (on the  $\sigma$ -algebra  $\mathcal{M}$ ), which is finite ( $\sigma$ -finite) if  $\mu$  is finite ( $\sigma$ -finite, respectively).*

The measure  $\bar{\mu}$  is called the Caratheodory extension of  $\mu$ .

**Proof.** By Theorems 2.6 and 2.8, we need only to prove the inclusion  $\mathcal{A} \subset \mathcal{M}$ , for then

$$\bar{\mu}|_{\mathcal{A}} = (\mu^*|_{\mathcal{M}})|_{\mathcal{A}} = \mu^*|_{\mathcal{A}} = \mu.$$

Let then  $E \in \mathcal{A}$ , and let  $A \subset X$  be such that  $\mu^*(A) < \infty$ . For any given  $\epsilon > 0$ , there exists an  $\mathcal{A}$ -cover  $\{E_i\}$  of  $A$  such that

$$\mu^*(A) + \epsilon > \sum_i \mu(E_i). \quad (1)$$

Since  $E \in \mathcal{A}$ ,  $\{E_i \cap E\}$  and  $\{E_i \cap E^c\}$  are  $\mathcal{A}$ -covers of  $A \cap E$  and  $A \cap E^c$ , respectively, and therefore

$$\sum_i \mu(E_i \cap E) \geq \mu^*(A \cap E)$$

and

$$\sum_i \mu(E_i \cap E^c) \geq \mu^*(A \cap E^c).$$

Adding these relations and using the additivity of the measure  $\mu$  on the algebra  $\mathcal{A}$ , we obtain

$$\sum_i \mu(E_i) \geq \mu^*(A \cap E) + \mu^*(A \cap E^c). \quad (2)$$

By (1), (2), and the arbitrariness of  $\epsilon$ , we get

$$\mu^*(A) \geq \mu^*(A \cap E) + \mu^*(A \cap E^c)$$

for all  $A \subset X$ , so that  $E \in \mathcal{M}$ , and  $\mathcal{A} \subset \mathcal{M}$ .

If  $\mu$  is finite, then since  $X \in \mathcal{A}$  and  $\mu^*|_{\mathcal{A}} = \mu$ , we have  $\mu^*(X) = \mu(X) < \infty$ . The  $\sigma$ -finite case is analogous.  $\square$

If we start from a *semi-measure*  $\mu$  on a *semi-algebra*  $\mathcal{C}$ , we first extend it to a measure (same notation) on the *algebra*  $\mathcal{A}$  generated by  $\mathcal{C}$  (as in Theorem 2.4).

We then apply the Caratheodory extension theorem to obtain the complete positive measure space  $(X, \mathcal{M}, \bar{\mu})$  with  $\mathcal{A} \subset \mathcal{M}$  and  $\bar{\mu}$  extending  $\mu$ . Note that if  $\mu$  is finite ( $\sigma$ -finite) on  $\mathcal{C}$ , then its extension  $\bar{\mu}$  is finite ( $\sigma$ -finite, respectively).

## 2.4 Structure of measurable sets

Let  $\mu$  be a measure on the algebra  $\mathcal{A}$  on  $X$ . Denote by  $\mu^*$  the outer measure induced by  $\mu$ , and let  $\mathcal{M}$  be the  $\sigma$ -algebra of all  $\mu^*$ -measurable subsets of  $X$ . Consider the family  $\mathcal{A}_\sigma \subset \mathcal{M}$  of all countable unions of sets from  $\mathcal{A}$ . Note that if we start from a semi-algebra  $\mathcal{C}$  and  $\mathcal{A}$  is the algebra generated by it, then  $\mathcal{A}_\sigma = \mathcal{C}_\sigma$ .

**Lemma 2.10.** *For any  $E \subset X$  with  $\mu^*(E) < \infty$  and for any  $\epsilon > 0$ , there exists  $A \in \mathcal{A}_\sigma$  such that  $E \subset A$  and*

$$\mu^*(A) \leq \mu^*(E) + \epsilon.$$

**Proof.** By definition of  $\mu^*(E)$ , there exists an  $\mathcal{A}$ -cover  $\{E_i\}$  of  $E$  such that  $\sum \mu(E_i) \leq \mu^*(E) + \epsilon$ . Then  $A := \bigcup E_i \in \mathcal{A}_\sigma$ ,  $E \subset A$ , and

$$\mu^*(A) \leq \sum \mu^*(E_i) = \sum \mu(E_i) \leq \mu^*(E) + \epsilon,$$

as wanted. □

If  $\mathcal{B}$  is any family of subsets of  $X$ , denote by  $\mathcal{B}_\delta$  the family of all countable intersections of sets from  $\mathcal{B}$ . Let  $\mathcal{A}_{\sigma\delta} := (\mathcal{A}_\sigma)_\delta$ .

**Proposition 2.11.** *Let  $\mu, \mathcal{A}, \mu^*$  be as before. Then for each  $E \subset X$  with  $\mu^*(E) < \infty$ , there exists  $A \in \mathcal{A}_{\sigma\delta}$  such that  $E \subset A$  and  $\mu^*(E) = \mu^*(A) (= \bar{\mu}(A))$ .*

**Proof.** Let  $E$  be a subset of  $X$  with finite outer measure. For each  $n \in \mathbb{N}$ , there exists  $A_n \in \mathcal{A}_\sigma$  such that  $E \subset A_n$  and  $\mu^*(A_n) < \mu^*(E) + 1/n$  (by Lemma 2.10). Therefore  $A := \bigcap A_n \in \mathcal{A}_{\sigma\delta}$ ,  $E \subset A$ , and

$$\mu^*(E) \leq \mu^*(A) \leq \mu^*(A_n) \leq \mu^*(E) + 1/n$$

for all  $n$ , so that  $\mu^*(E) = \mu^*(A)$ . □

The structure of  $\mu^*$ -measurable sets is described in the next theorem.

**Theorem 2.12.** *Let  $\mu$  be a  $\sigma$ -finite measure on the algebra  $\mathcal{A}$  on  $X$ , and let  $\mu^*$  be the outer measure induced by it. Then  $E \subset X$  is  $\mu^*$ -measurable if and only if there exists  $A \in \mathcal{A}_{\sigma\delta}$  such that  $E \subset A$  and  $\mu^*(A - E) = 0$ .*

**Proof.** We observed in the proof of Theorem 2.8 that  $\mathcal{M}$  contains every set of  $\mu^*$ -measure zero. Thus, if  $E \subset A \in \mathcal{A}_{\sigma\delta}$  and  $\mu^*(A - E) = 0$ , then  $A - E \in \mathcal{M}$  and  $A \in \mathcal{M}$  (because  $\mathcal{A}_{\sigma\delta} \subset \mathcal{M}$ ), and therefore  $E = A - (A - E) \in \mathcal{M}$ .

Conversely, suppose  $E \in \mathcal{M}$ . By the  $\sigma$ -finiteness hypothesis, we may write  $X = \bigcup X_i$  with  $X_i \in \mathcal{A}$  mutually disjoint and  $\mu(X_i) < \infty$ . Let  $E_i := E \cap X_i$ .

By Lemma 2.10, there exist  $A_{ni} \in \mathcal{A}_\sigma$  such that  $E_i \subset A_{ni}$  and  $\mu^*(A_{ni}) \leq \mu^*(E_i) + (1/n2^i)$ , for all  $n, i \in \mathbb{N}$ . Set  $A_n := \bigcup_i A_{ni}$ . Then for all  $n$ ,  $A_n \in \mathcal{A}_\sigma$ ,  $E \subset A_n$ , and  $A_n - E \subset \bigcup_i (A_{ni} - E_i)$ , so that

$$\mu^*(A_n - E) \leq \sum_i \mu^*(A_{ni} - E_i) \leq \sum_i \frac{1}{n2^i} = 1/n.$$

Let  $A := \bigcap A_n$ . Then  $E \subset A$ ,  $A \in \mathcal{A}_{\sigma\delta}$ , and since  $A - E \subset A_n - E$  for all  $n$ ,  $\mu^*(A - E) = 0$ .  $\square$

We can use Lemma 2.10 to prove a *uniqueness* theorem for the extension of measures on algebras.

**Theorem 2.13 (Uniqueness of extension).** *Let  $\mu$  be a measure on the algebra  $\mathcal{A}$  on  $X$ , and let  $\bar{\mu}$  be the Caratheodory extension of  $\mu$  (as a measure on the  $\sigma$ -algebra  $\mathcal{M}$ , cf. Theorem 2.9). Consider the  $\sigma$ -algebra  $\mathcal{B}$  generated by  $\mathcal{A}$  (of course  $\mathcal{B} \subset \mathcal{M}$ ). If  $\mu_1$  is any measure that extends  $\mu$  to  $\mathcal{B}$ , then  $\mu_1(E) = \bar{\mu}(E)$  for any set  $E \in \mathcal{B}$  with  $\bar{\mu}(E) < \infty$ . If  $\mu$  is  $\sigma$ -finite, then  $\mu_1 = \bar{\mu}$  on  $\mathcal{B}$ .*

**Proof.** Since  $\mu_1 = \mu = \bar{\mu}$  on  $\mathcal{A}$ , and each set in  $\mathcal{A}_\sigma$  is a *disjoint* countable union of sets  $A_i \in \mathcal{A}$ , we have  $\mu_1 = \bar{\mu}$  on  $\mathcal{A}_\sigma$ .

Let  $E \in \mathcal{B}$  with  $\bar{\mu}(E) < \infty$ , and let  $\epsilon > 0$ . By Lemma 2.10, there exists  $A \in \mathcal{A}_\sigma$  such that  $E \subset A$  and  $\bar{\mu}(A) \leq \bar{\mu}(E) + \epsilon$ . Hence

$$\mu_1(E) \leq \mu_1(A) = \bar{\mu}(A) \leq \bar{\mu}(E) + \epsilon,$$

and therefore,  $\mu_1(E) \leq \bar{\mu}(E)$ , by the arbitrariness of  $\epsilon$ .

Note in passing that  $A - E \in \mathcal{B}$  with  $\bar{\mu}(A - E) \leq \epsilon < \infty$  (for any  $A$  as above). Therefore, we have in particular  $\mu_1(A - E) \leq \bar{\mu}(A - E) \leq \epsilon$ . Hence

$$\bar{\mu}(E) \leq \bar{\mu}(A) = \mu_1(A) = \mu_1(E) + \mu_1(A - E) \leq \mu_1(E) + \epsilon,$$

and the reverse inequality  $\bar{\mu}(E) \leq \mu_1(E)$  follows.

If  $\mu$  is  $\sigma$ -finite, write  $X$  as the disjoint union of  $X_i \in \mathcal{A}$  with  $\mu(X_i) < \infty$ ,  $i = 1, 2, \dots$ . Then, each  $E \in \mathcal{B}$  is the disjoint union of  $E_i := E \cap X_i$  with  $\mu(E_i) < \infty$ ; since  $\mu_1(E_i) = \bar{\mu}(E_i)$  for all  $i$ , also  $\mu_1(E) = \bar{\mu}(E)$ , by  $\sigma$ -additivity of both measures.  $\square$

## 2.5 Construction of Lebesgue–Stieltjes measures

We shall apply the general method of construction of measures described in the preceding sections to the special semi-algebra  $\mathcal{C}$  in the example following Definition 2.1, and to the semi-measure  $\mu$  induced by a given non-decreasing right-continuous function  $F : \mathbb{R} \rightarrow \mathbb{R}$ . Denote  $F(\infty) :=$

$\lim_{x \rightarrow \infty} F(x) (\in (-\infty, \infty])$ , and similarly  $F(-\infty) (\in [-\infty, \infty))$  (both limits exist, because  $F$  is non-decreasing). We define the semi-measure  $\mu$  (induced by  $F$ ) by

$$\begin{aligned}\mu(\emptyset) &= 0; \mu((a, b]) = F(b) - F(a) \quad (a, b \in \mathbb{R}, a < b); \\ \mu((-\infty, b]) &= F(b) - F(-\infty); \quad \mu((a, \infty)) = F(\infty) - F(a), a, b \in \mathbb{R}.\end{aligned}$$

The example following Definition 2.2 is the special case with  $F(x) = x, x \in \mathbb{R}$ .

We verify that the properties (2) and (3) of Definition 2.2 are satisfied.

Suppose  $(a, b]$  is the disjoint finite union of similar intervals. Then we may index the subintervals so that

$$a = a_1 < b_1 = a_2 < b_2 = a_3 < \cdots < b_n = b.$$

Therefore

$$\begin{aligned}\sum_{i=1}^n \mu((a_i, b_i]) &= \sum_{i=1}^n [F(b_i) - F(a_i)] = \sum_{i=1}^{n-1} [F(a_{i+1}) - F(a_i)] + F(b) - F(a_n) \\ &= F(b) - F(a) = \mu((a, b])\end{aligned}$$

for  $a, b \in \mathbb{R}, a < b$ . A similar argument for the cases  $(-\infty, b]$  and  $(a, \infty)$  completes the verification of Property (2). In order to verify Property (3), we show that whenever  $(a, b] \subset \bigcup_{i=1}^{\infty} (a_i, b_i]$ , then

$$F(b) - F(a) \leq \sum_i [F(b_i) - F(a_i)]. \quad (1)$$

This surely implies Property (3) for  $a, b$  finite. If  $(-\infty, b]$  is contained in such a union, then  $(-n, b]$  is contained in it as well, for all  $n \in \mathbb{N}$ , so that  $F(b) - F(-n)$  is majorized by the sum on the right-hand side of (1) for all  $n$ ; letting  $n \rightarrow \infty$ , we deduce that this sum majorizes  $\mu((-\infty, b])$ . A similar argument works for  $\mu((a, \infty))$ .

Let  $\epsilon > 0$ . By the right continuity of  $F$ , there exist  $c_i, i = 0, 1, 2, \dots$  such that

$$\begin{aligned}a &< c_0; \quad F(c_0) < F(a) + \epsilon; \\ b_i &< c_i; \quad F(c_i) < F(b_i) + \epsilon/2^i; \quad i = 1, 2, \dots\end{aligned} \quad (2)$$

We have  $[c_0, b] \subset \bigcup_{i=1}^{\infty} (a_i, c_i)$ , so that, by compactness, a *finite* number  $n$  of intervals  $(a_i, c_i)$  covers  $[c_0, b]$ . Thus  $c_0$  is in one of these intervals, say  $(a_1, c_1)$  (to simplify notation), that is,

$$a_1 < c_0 < c_1.$$

Assuming we got  $(a_i, c_i), 1 \leq i < k$  such that

$$a_i < c_{i-1} < c_i, \quad (3)$$

and  $c_{k-1} \leq b$  (that is  $c_{k-1} \in [c_0, b]$ ), there exists one of the  $n$  intervals above, say  $(a_k, c_k)$  to simplify notation, that contains  $c_{k-1}$ , so that (3) is valid for  $i = k$  as well. This (finite) inductive process will end after at most  $n$  steps (this will exhaust our finite cover), which means that for some  $k \leq n$ , we must get

$$b < c_k. \quad (4)$$

By (4), (3), and (2) (in this order),

$$\begin{aligned}
 \mu((a, b]) &:= F(b) - F(a) \leq F(c_k) + \sum_{i=2}^k [F(c_{i-1}) - F(a_i)] - F(c_0) + \epsilon \\
 &\leq \sum_{i=1}^k [F(c_i) - F(a_i)] + \epsilon \leq \sum_{i=1}^{\infty} [F(c_i) - F(a_i)] + \epsilon \\
 &\leq \sum_{i=1}^{\infty} [F(b_i) - F(a_i)] + 2\epsilon = \sum_i \mu((a_i, b_i]) + 2\epsilon,
 \end{aligned}$$

as wanted (by the arbitrariness of  $\epsilon$ ). By Theorems 2.4 and 2.9, the semi-measure  $\mu$  has an extension as a complete measure (also denoted  $\mu$ ) on a  $\sigma$ -algebra  $\mathcal{M}$  containing the Borel algebra  $\mathcal{B}$  (= the  $\sigma$ -algebra generated by  $\mathcal{C}$ , or by the algebra  $\mathcal{A}$ ). By Theorem 2.13, the extension is uniquely determined on  $\mathcal{B}$ . The (complete) measure space  $(\mathbb{R}, \mathcal{M}, \mu)$  is called the *Lebesgue–Stieltjes measure space induced by the given function  $F$*  (in the special case  $F(x) = x$ , this is the *Lebesgue measure space*). By Theorem 2.13, the ‘Lebesgue–Stieltjes measure  $\mu$  induced by  $F$ ’ is the unique measure on  $\mathcal{B}$  such that  $\mu((a, b]) = F(b) - F(a)$ . It is customary to write the integral  $\int f d\mu$  in the form  $\int f dF$ , and to call  $F$  the ‘distribution of  $\mu$ ’. Accordingly, in the special case of Lebesgue measure, the above integral is customarily written in the form  $\int f dx$ .

The Lebesgue measure  $\mu$  is the unique measure on  $\mathcal{B}$  such that  $\mu((a, b]) = b - a$  for all real  $a < b$ ; in particular, it is translation invariant on  $\mathcal{C}$ , hence on  $\mathcal{A}$  (if  $E \in \mathcal{A}$ , write  $E$  as a finite disjoint union of intervals  $(a_i, b_i]$ , then  $\mu(t + E) = \mu(\bigcup_i \{t + (a_i, b_i]\}) = \sum_i \mu((t + a_i, t + b_i]) = \sum \mu((a_i, b_i]) = \mu(E)$  for all real  $t$ ). Let  $\mu^*$  be the outer measure induced by  $\mu$ . Then for all  $E \subset \mathbb{R}$  and  $t \in \mathbb{R}$ ,  $\{E_i\}$  is an  $\mathcal{A}$ -cover of  $E$  if and only if  $\{t + E_i\}$  is an  $\mathcal{A}$ -cover of  $t + E$ , and therefore

$$\mu^*(t + E) := \inf \sum \mu(t + E_i) = \inf \sum \mu(E_i) = \mu^*(E).$$

In particular, if  $\mu^*(E) = 0$ , then  $\mu^*(t + E) = 0$  for all real  $t$ . By Theorem 2.12,  $E \in \mathcal{M}$  (that is, Lebesgue measurable on  $\mathbb{R}$ ) if and only if there exists  $A \in \mathcal{A}_{\sigma\delta}$  such that  $E \subset A$  and  $\mu^*(A - E) = 0$ . However, translations of unions and intersections of sets are unions and intersections of the translated sets (respectively). Thus, the existence of  $A \in \mathcal{A}_{\sigma\delta}$  as above implies that  $t + A \in \mathcal{A}_{\sigma\delta}$ ,  $t + E \subset t + A$ , and  $\mu^*((t + A) - (t + E)) = \mu^*(t + (A - E)) = 0$ , that is  $t + E \in \mathcal{M}$  (by Theorem 2.12), for all  $t \in \mathbb{R}$ . Since Lebesgue measure is the restriction of  $\mu^*$  to  $\mathcal{M}$ , we conclude from this discussion that the Lebesgue measure space  $(\mathbb{R}, \mathcal{M}, \mu)$  is *translation invariant*, which means that  $t + E \in \mathcal{M}$  and  $\mu(t + E) = \mu(E)$  for all  $t \in \mathbb{R}$  and  $E \in \mathcal{M}$ . In the terminology of Section 1.26, the map  $h(x) = x - t$  is a measurable map of  $(\mathbb{R}, \mathcal{M})$  onto itself (for each given  $t$ ), and the corresponding measure  $\nu(E) := \mu(h^{-1}(E)) = \mu(t + E) = \mu(E)$ . Therefore, by the Proposition there, for any non-negative measurable function and for any  $\mu$ -integrable

complex function  $f$  on  $\mathbb{R}$ ,

$$\int f d\mu = \int f_t d\mu,$$

where  $f_t(x) := f(x - t)$ . This is the translation invariance of the Lebesgue integral.

Consider the quotient group  $\mathbb{R}/\mathbb{Q}$  of the additive group  $\mathbb{R}$  by the subgroup  $\mathbb{Q}$  of rationals. Let  $A$  be an arbitrary bounded Lebesgue measurable subset of  $\mathbb{R}$  of positive measure. By the Axiom of Choice, there exists a set  $E \subset A$  that contains *precisely one point from each coset in  $\mathbb{R}/\mathbb{Q}$  that meets  $A$* . Since  $A$  is bounded,  $A \subset (-a, a)$  for some  $a \in (0, \infty)$ . We claim that  $A$  is contained in the disjoint union  $S := \bigcup_{r \in \mathbb{Q} \cap (-2a, 2a)} (r + E)$ . Indeed, if  $x \in A$ , there exists a unique  $y \in E$  such that  $x, y$  are in the same coset of  $\mathbb{Q}$ , that is,  $x - y = r \in \mathbb{Q}$ , hence  $x = r + y \in r + E$ ,  $|r| \leq |x| + |y| < 2a$ , so that indeed  $x \in S$ . If  $r, s \in \mathbb{Q} \cap (-2a, 2a)$  are distinct and  $x \in (r + E) \cap (s + E)$ , then there exist  $y, z \in E$  such that  $x = r + y = s + z$ . Hence,  $y - z = s - r \in \mathbb{Q} - \{0\}$ , which means that  $y, z$  are distinct points of  $E$  belonging to the same coset, contrary to the definition of  $E$ . Thus, the union  $S$  is indeed a *disjoint* union. Write  $\mathbb{Q} \cap (-2a, 2a) = \{r_k\}$ . Suppose  $E$  is (Lebesgue) measurable. Since  $r_k + E \subset r_k + A \subset (-3a, 3a)$  for all  $k$ , it follows that  $S$  is a measurable subset of  $(-3a, 3a)$ . Therefore, by  $\sigma$ -additivity and translation invariance of  $\mu$ ,

$$6a \geq \mu(S) = \sum_{k=1}^{\infty} \mu(r_k + E) \geq \sum_{k=1}^n \mu(E) = n\mu(E)$$

for all  $n \in \mathbb{N}$ , hence  $\mu(E) = 0$  and  $\mu(S) = 0$ . Since  $A \subset S$ , also  $\mu(A) = 0$ , contradicting our hypothesis. This shows that  $E$  is *not* Lebesgue measurable. Since any measurable set on  $\mathbb{R}$  of positive measure contains a *bounded* measurable subset of positive measure, we proved the following

**Proposition.** *Every (Lebesgue-) measurable subset of  $\mathbb{R}$  of positive measure contains a non-measurable subset.*

## 2.6 Riemann versus Lebesgue

Let  $-\infty < a < b < \infty$ , and let  $f : [a, b] \rightarrow \mathbb{R}$  be bounded. Denote

$$m = \inf_{[a, b]} f; \quad M = \sup_{[a, b]} f.$$

Given a ‘partition’  $P = \{x_k; k = 0, \dots, n\}$  of  $[a, b]$ , where  $a = x_0 < x_1 < \dots < x_n = b$ , we denote

$$m_k = \inf_{[x_{k-1}, x_k]} f; \quad M_k = \sup_{[x_{k-1}, x_k]} f;$$

$$L_P = \sum_{k=1}^n m_k(x_k - x_{k-1}); \quad U_P = \sum_{k=1}^n M_k(x_k - x_{k-1}).$$

Recall that the lower and upper Riemann integrals of  $f$  over  $[a, b]$  are defined as the supremum and infimum of  $L_P$  and  $U_P$  (respectively) over all ‘partitions’  $P$ , and  $f$  is Riemann integrable over  $[a, b]$  if these lower and upper integrals coincide (their common value is the Riemann integral, denoted  $\int_a^b f(x) dx$ ). For bounded complex functions  $f = u + iv$  with  $u, v$  real, one says that  $f$  is Riemann integrable iff both  $u$  and  $v$  are Riemann integrable, and  $\int_a^b f dx := \int_a^b u dx + i \int_a^b v dx$ .

**Proposition.** *If a bounded (complex) function on the real interval  $[a, b]$  is Riemann integrable, then it is Lebesgue integrable on  $[a, b]$ , and its Lebesgue integral  $\int_{[a,b]} f dx$  coincides with its Riemann integral  $\int_a^b f dx$ .*

**Proof.** It suffices to consider bounded *real* functions  $f$ . Given a partition  $P$ , consider the simple Borel functions

$$l_P = f(a)I_{\{a\}} + \sum_k m_k I_{(x_{k-1}, x_k]}; \quad u_P = f(a)I_{\{a\}} + \sum_k M_k I_{(x_{k-1}, x_k]}.$$

Then  $l_P \leq f \leq u_P$  on  $[a, b]$ , and

$$\int_{[a,b]} l_P dx = L_P; \quad \int_{[a,b]} u_P dx = U_P. \quad (1)$$

If  $f$  is Riemann integrable, there exists a sequence of partitions  $P_j$  of  $[a, b]$  such that  $P_{j+1}$  is a refinement of  $P_j$ ,  $\|P_j\| := \max_{x_k \in P_j} (x_k - x_{k-1}) \rightarrow 0$  as  $j \rightarrow \infty$ , and

$$\lim_j L_{P_j} = \lim_j U_{P_j} = \int_a^b f dx. \quad (2)$$

The sequences  $l_j := l_{P_j}$  and  $u_j := u_{P_j}$  are monotonic (non-decreasing and non-increasing, respectively) and bounded. Let then  $l := \lim_j l_j$  and  $u := \lim_j u_j$ . These are bounded Borel functions, and by (1), (2), and the Lebesgue dominated convergence theorem,

$$\int_{[a,b]} l dx = \lim_j \int_{[a,b]} l_j dx = \lim_j L_{P_j} = \int_a^b f dx, \quad (3)$$

and similarly  $\int_{[a,b]} u dx = \int_a^b f dx$ . In particular,  $\int_{[a,b]} (u - l) dx = 0$ , and since  $u - l \geq 0$ , it follows that  $u = l$  a.e.; however  $l \leq f \leq u$ , hence  $f = u = l$  a.e.; therefore  $f$  is Lebesgue measurable (hence Lebesgue integrable, since it is bounded) and  $\int_{[a,b]} f dx = \int_{[a,b]} l dx = \int_a^b f dx$  by (3).  $\square$

A similar proposition is valid for *absolutely convergent improper* Riemann integrals (on finite or infinite intervals). The easy proofs are omitted.

Let  $Q = \bigcup_j P_j$  (a countable set, hence a Lebesgue null set). If  $x \in [a, b]$  is *not* in  $Q$ ,  $f$  is continuous at  $x$  iff  $l(x) = u(x)$ . It follows from the preceding proof that if  $f$  is Riemann integrable, then it is *continuous* at almost all points not in  $Q$ , that is, almost everywhere in  $[a, b]$ . Conversely, if  $f$  is continuous a.e., then



$l = f = u$  a.e., hence  $\int_{[a,b]} l \, dx = \int_{[a,b]} u \, dx$ . Therefore, given  $\epsilon > 0$ , there exists  $j$  such that  $\int_{[a,b]} u_j \, dx - \int_{[a,b]} l_j \, dx < \epsilon$ , that is,  $U_{P_j} - L_{P_j} < \epsilon$ . This means that  $f$  is Riemann integrable on  $[a, b]$ . Formally:

**Proposition.** *Let  $f$  be a bounded complex function on  $[a, b]$ . Then  $f$  is Riemann integrable on  $[a, b]$  if and only if it is continuous almost everywhere in  $[a, b]$ .*

## 2.7 Product measure

Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be measure spaces. A *measurable rectangle* is a cartesian product  $A \times B$  with  $A \in \mathcal{A}$  and  $B \in \mathcal{B}$ . The set  $\mathcal{C}$  of all measurable rectangles is a semi-algebra, since

$$(A \times B) \cap (C \times D) = (A \cap C) \times (B \cap D)$$

and

$$(A \times B)^c = (A^c \times B) \cup (A \times B^c) \cup (A^c \times B^c),$$

where the union on the right is clearly disjoint. Define  $\lambda$  on  $\mathcal{C}$  by

$$\lambda(A \times B) = \mu(A)\nu(B).$$

We claim that  $\lambda$  is a semi-measure on  $\mathcal{C}$  (cf. Definition 2.2). Indeed, Property (1) is trivial, while Properties (2) and (3) follow from the stronger property:

If  $A_i \times B_i \in \mathcal{C}$  are mutually disjoint with union  $A \times B \in \mathcal{C}$ , then

$$\lambda(A \times B) = \sum_{i=1}^{\infty} \mu(A_i)\nu(B_i). \quad (1)$$

**Proof.** Let  $x \in A$ . For each  $y \in B$ , there exists a unique  $i$  such that the pair  $[x, y]$  belongs to  $A_i \times B_i$  (because the rectangles are mutually disjoint). Thus,  $B$  decomposes as the disjoint union

$$B = \bigcup_{\{i; x \in A_i\}} B_i.$$

Therefore

$$\nu(B) = \sum_{\{i; x \in A_i\}} \nu(B_i),$$

and so

$$\nu(B)I_A(x) = \sum_{i=1}^{\infty} \nu(B_i)I_{A_i}(x).$$

By Beppo Levi's theorem (1.16),

$$\begin{aligned} \lambda(A \times B) &:= \mu(A)\nu(B) = \int_X \nu(B)I_A(x) \, d\mu \\ &= \sum_i \int \nu(B_i)I_{A_i} \, d\mu = \sum_i \lambda(A_i \times B_i). \end{aligned}$$

□

By the Caratheodory extension theorem, there exists a complete measure space, which we shall denote

$$(X \times Y, \mathcal{A} \times \mathcal{B}, \mu \times \nu),$$

and call *the product of the given measure spaces*, such that  $\mathcal{C} \subset \mathcal{A} \times \mathcal{B}$  and

$$(\mu \times \nu)(A \times B) = \lambda(A \times B) := \mu(A)\nu(B)$$

for  $A \times B \in \mathcal{C}$ .

The central theorem of this section is the Fubini–Tonelli theorem, that relates the ‘double integral’ (relative to  $\mu \times \nu$ ) with the ‘iterated integrals’ (relative to  $\mu$  and  $\nu$  in either order). We need first some technical lemmas.

**Lemma 2.14.** *For each  $E \in \mathcal{C}_{\sigma\delta}$ , the sections  $E_x := \{y \in Y; [x, y] \in E\}$  ( $x \in X$ ) belong to  $\mathcal{B}$ .*

**Proof.** If  $E = A \times B \in \mathcal{C}$ , then  $E_x$  is either  $B$  (when  $x \in A$ ) or  $\emptyset$  (otherwise), so clearly belongs to  $\mathcal{B}$ . If  $E \in \mathcal{C}_\sigma$ , then  $E = \bigcup_i E_i$  with  $E_i \in \mathcal{C}$ ; hence

$$E_x = \bigcup_i (E_i)_x \in \mathcal{B}.$$

Similarly, if  $E \in \mathcal{C}_{\sigma\delta}$ , then  $E = \bigcap_i E_i$  with  $E_i \in \mathcal{C}_\sigma$ , and therefore

$$E_x = \bigcap_i (E_i)_x \in \mathcal{B}$$

for all  $x \in X$ . □

By the lemma, the function

$$g_E(x) := \nu(E_x) : X \rightarrow [0, \infty]$$

is well defined, for each  $E \in \mathcal{C}_{\sigma\delta}$ .

**Lemma 2.15.** *Suppose the measure space  $(X, \mathcal{A}, \mu)$  is complete. For each  $E \in \mathcal{C}_{\sigma\delta}$  with  $(\mu \times \nu)(E) < \infty$ , the function  $g_E(x) := \nu(E_x)$  is  $\mathcal{A}$ -measurable, and*

$$\int_X g_E d\mu = (\mu \times \nu)(E). \quad (2)$$

**Proof.** For an arbitrary  $E = A \times B \in \mathcal{C}$ ,

$$g_E = \nu(B)I_A$$

is clearly  $\mathcal{A}$ -measurable (since  $A \in \mathcal{A}$ ), and (2) is trivially true.

If  $E \in \mathcal{C}_\sigma$  (arbitrary), we may represent it as a *disjoint* union of  $E_i \in \mathcal{C}$  ( $i \in \mathbb{N}$ ), and therefore  $g_E = \sum_i g_{E_i}$  is  $\mathcal{A}$ -measurable, and by the Beppo Levi theorem and the  $\sigma$ -additivity of  $\mu \times \nu$ ,

$$\int_X g_E d\mu = \sum_i \int_X g_{E_i} d\mu = \sum_i (\mu \times \nu)(E_i) = (\mu \times \nu)(E).$$

Let now  $E \in \mathcal{C}_{\sigma\delta}$  with  $(\mu \times \nu)(E) < \infty$ . Thus  $E = \bigcap_i F_i$  with  $F_i \in \mathcal{C}_\sigma$ . By Lemma 2.10, there exists  $G \in \mathcal{C}_\sigma$  such that  $E \subset G$  and  $(\mu \times \nu)(G) < (\mu \times \nu)(E) + 1 < \infty$ . Then

$$E = E \cap G = \bigcap_i (F_i \cap G) = \bigcap_k E_k,$$

where (for  $k = 1, 2, \dots$ )

$$E_k := \bigcap_{i=1}^k (F_i \cap G).$$

Since  $\mathcal{C}_\sigma$  is an algebra,  $E_k \in \mathcal{C}_\sigma$ ,  $E_{k+1} \subset E_k$ , and  $E_1 \subset G$  has finite product measure. Therefore  $g_{E_k}$  is  $\mathcal{A}$ -measurable, and

$$\int_X g_{E_k} d\mu = (\mu \times \nu)(E_k) < \infty$$

for all  $k$ . In particular  $g_{E_1} < \infty$   $\mu$ -a.e.

For  $x$  such that  $g_{E_1}(x) = \nu((E_1)_x) < \infty$ , we have by Lemma 1.11:

$$g_E(x) := \nu(E_x) = \nu\left(\bigcap_k (E_k)_x\right) = \lim_k \nu((E_k)_x) = \lim_k g_{E_k}(x).$$

Hence  $g_{E_k} \rightarrow g_E$   $\mu$ -a.e. Since the measure space  $(X, \mathcal{A}, \mu)$  is complete by hypothesis, it follows that  $g_E$  is  $\mathcal{A}$ -measurable. Also  $0 \leq g_{E_k} \leq g_{E_1}$  for all  $k$ , and  $\int_X g_{E_1} d\mu < \infty$ . Therefore, by Lebesgue's dominated convergence theorem and Lemma 1.11,

$$\int_X g_E d\mu = \lim_k \int_X g_{E_k} d\mu = \lim_k (\mu \times \nu)(E_k) = (\mu \times \nu)(E).$$

□

We now extend the above lemma to *all*  $E \in \mathcal{A} \times \mathcal{B}$  with *finite* product measure.

**Lemma 2.16.** *Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be complete measure spaces. Let  $E \in \mathcal{A} \times \mathcal{B}$  have finite product measure. Then the sections  $E_x$  are  $\mathcal{B}$ -measurable for  $\mu$ -almost all  $x$ ; the ( $\mu$ -a.e. defined and finite) function  $g_E(x) := \nu(E_x)$  is  $\mathcal{A}$ -measurable, and*

$$\int_X g_E d\mu = (\mu \times \nu)(E).$$

**Proof.** By Proposition 2.11, since  $E$  has finite product measure, there exists  $F \in \mathcal{C}_{\sigma\delta}$  such that  $E \subset F$  and  $(\mu \times \nu)(F) = (\mu \times \nu)(E) < \infty$ . Let  $G := F - E$ . Then  $G \in \mathcal{A} \times \mathcal{B}$  has zero product measure (since  $E$  and  $F$  have equal *finite* product measure). Again by Proposition 2.11, there exists  $H \in \mathcal{C}_{\sigma\delta}$  such that  $G \subset H$  and  $(\mu \times \nu)(H) = 0$ . By Lemma 2.15,  $g_H$  is  $\mathcal{A}$ -measurable and  $\int_X g_H d\mu = (\mu \times \nu)(H) = 0$ . Therefore,  $\nu(H_x) := g_H(x) = 0$   $\mu$ -a.e. Since  $G_x \subset H_x$ , it follows from the completeness of the measure space  $(Y, \mathcal{B}, \nu)$  that, for  $\mu$ -almost all  $x$ ,  $G_x$

is  $\mathcal{B}$ -measurable and  $\nu(G_x) = 0$ . Since  $E = F - G$ , it follows that for  $\mu$ -almost all  $x$ ,  $E_x$  is  $\mathcal{B}$ -measurable and  $\nu(E_x) = \nu(F_x)$ , that is,  $g_E = g_F$  ( $\mu$ -a.e.) is  $\mathcal{A}$ -measurable (by Lemma 2.15), and

$$\int_X g_E d\mu = \int_X g_F d\mu = (\mu \times \nu)(F) = (\mu \times \nu)(E).$$

□

Note that for any  $E \subset X \times Y$  and  $x \in X$ ,

$$I_{E_x} = I_E(x, \cdot).$$

Therefore, if  $E \in \mathcal{A} \times \mathcal{B}$  has finite product measure, then for  $\mu$ -almost all  $x$ , the function  $I_E(x, \cdot)$  is  $\mathcal{B}$ -measurable, with integral (over  $Y$ ) equal to  $\nu(E_x) := g_E(x) < \infty$ , that is, for  $\mu$ -almost all  $x$ ,

$$I_E(x, \cdot) \in L^1(\nu), \quad (i)$$

and its integral ( $= g_E$ ) is  $\mathcal{A}$ -measurable, with integral (over  $X$ ) equal to  $(\mu \times \nu)(E) < \infty$ , that is,

$$\int_Y I_E(x, \cdot) d\nu \in L^1(\mu), \quad (ii)$$

and

$$\begin{aligned} \int_X \left[ \int_Y I_E(x, \cdot) d\nu \right] d\mu &= \int_X g_E d\mu = (\mu \times \nu)(E) \\ &= \int_{X \times Y} I_E d(\mu \times \nu). \end{aligned} \quad (iii)$$

If  $f$  is a simple non-negative function in  $L^1(\mu \times \nu)$ , we may write  $f = \sum c_k I_{E_k}$  (finite sum), with  $c_k > 0$  and  $(\mu \times \nu)(E_k) < \infty$ . Then for  $\mu$ -almost all  $x$ ,  $f(x, \cdot)$  is a linear combination of  $L^1(\nu)$ -functions (by (i)), hence belongs to  $L^1(\nu)$ ; its integral (over  $Y$ ) is a linear combination of the  $g_{E_k} \in L^1(\mu)$ , hence belongs to  $L^1(\mu)$ , and by (iii),

$$\int_X \left[ \int_Y f(x, \cdot) d\nu \right] d\mu = \sum_k c_k \int_{X \times Y} I_{E_k} d(\mu \times \nu) = \int_{X \times Y} f d(\mu \times \nu).$$

If  $f \in L^1(\mu \times \nu)$  is non-negative, by Theorem 1.8, we get simple measurable functions

$$0 \leq s_1 \leq s_2 \leq \cdots \leq f$$

such that  $\lim s_n = f$ . Necessarily,  $s_n \in L^1(\mu \times \nu)$ , so by the preceding conclusions, for  $\mu$ -almost all  $x$ ,  $s_n(x, \cdot)$  are  $\mathcal{B}$ -measurable, and their integrals (over  $Y$ ) are  $\mathcal{A}$ -measurable; therefore, for  $\mu$ -almost all  $x$ ,  $f(x, \cdot)$  is  $\mathcal{B}$ -measurable, and by the monotone convergence theorem,

$$\int_Y f(x, \cdot) d\nu = \lim \int_Y s_n(x, \cdot) d\nu,$$

so that the integrals on the left are  $\mathcal{A}$ -measurable. Applying the monotone convergence theorem to the sequence on the right, we have by (iii) for  $s_n$ ,

$$\begin{aligned} \int_X \left[ \int_Y f(x, \cdot) d\nu \right] d\mu &= \lim_n \int_X \left[ \int_Y s_n(x, \cdot) d\nu \right] d\mu = \lim_n \int_{X \times Y} s_n d(\mu \times \nu) \\ &= \int_{X \times Y} f d(\mu \times \nu) < \infty \end{aligned}$$

(by another application of the monotone convergence theorem). In particular,  $\int_Y f(x, \cdot) d\nu \in L^1(\mu)$  and therefore,  $f(x, \cdot) \in L^1(\nu)$  for  $\mu$ -almost all  $x$ .

For  $f \in L^1(\mu \times \nu)$  complex, decompose  $f = u^+ - u^- + iv^+ - iv^-$  to obtain the conclusions (i)–(iii) for  $f$  instead of  $I_E$ . Finally, we may interchange the roles of  $x$  and  $y$ . Collecting, we proved the following.

**Theorem 2.17 (Fubini's theorem).** *Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be complete (positive) measure spaces, and let  $f \in L^1(\mu \times \nu)$ . Then*

- (i) *for  $\mu$ -almost all  $x$ ,  $f(x, \cdot) \in L^1(\nu)$  and for  $\nu$ -almost all  $y$ ,  $f(\cdot, y) \in L^1(\mu)$ ;*
- (ii)  *$\int_Y f(x, \cdot) d\nu \in L^1(\mu)$  and  $\int_X f(\cdot, y) d\mu \in L^1(\nu)$ ;*
- (iii)  *$\int_X [\int_Y f(x, \cdot) d\nu] d\mu = \int_{X \times Y} f d(\mu \times \nu) = \int_Y [\int_X f(\cdot, y) d\mu] d\nu$ .*

When we need to verify the hypothesis  $f \in L^1(\mu \times \nu)$  (i.e. the *finiteness* of the integral  $\int_{X \times Y} |f| d(\mu \times \nu)$ ), the following theorem on *non-negative* functions is useful.

**Theorem 2.18 (Tonelli's theorem).** *Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be complete  $\sigma$ -finite measure spaces, and let  $f \geq 0$  be  $\mathcal{A} \times \mathcal{B}$ -measurable. Then (i) and (ii) in Fubini's theorem (2.17) are valid with the relation ' $\in L^1(\cdots)$ ' replaced by the expression '*is measurable*', and (iii) is valid.*

**Proof.** The integrability of  $f \geq 0$  was used in the preceding proof to guarantee that the measurable simple functions  $s_n$  be in  $L^1(\mu \times \nu)$ , that is, that they vanish outside a measurable set of *finite* (product) measure, so that the preceding step, based on Lemma 2.16, could be applied. In our case, the product measure space  $Z = X \times Y$  is  $\sigma$ -finite. Write  $Z = \bigcup_n Z_n$  with  $Z_n \in \mathcal{A} \times \mathcal{B}$  of finite product measure and  $Z_n \subset Z_{n+1}$ . With  $s_n$  as before, the 'corrected' simple functions  $s'_n := s_n I_{Z_n}$  meet the said requirements.  $\square$

## Exercises

1. Calculate (with appropriate justification):

- (a)  $\lim_{n \rightarrow \infty} \int_{\mathbb{R}} (e^{-x^2/n}) / (1 + x^2) dx$ .
- (b)  $\lim_{t \rightarrow 0+} \int_0^{\pi/2} \sin[(\pi/2)e^{-tx^2}] \cos x dx$ .
- (c)  $\int_0^1 \int_0^\infty [y \arctan(xy)] / [(1 + x^2 y^2)(1 + y^2)] dy dx$ .

2. Let  $L^1(\mathbb{R})$  be the Lebesgue space with respect to the Lebesgue measure on  $\mathbb{R}$ . If  $f \in L^1(\mathbb{R})$ , define

$$F_u(t) = \int_{\mathbb{R}} \frac{\sin(t-s)u}{(t-s)u} f(s) ds \quad (u > 0, t \in \mathbb{R}).$$

Prove:

- (a) For each  $u > 0$ , the function  $F_u : \mathbb{R} \rightarrow \mathbb{C}$  is well defined, continuous, and bounded by  $\|f\|_1$ .
- (b)  $\lim_{u \rightarrow \infty} F_u = 0$  and  $\lim_{u \rightarrow 0+} F_u = \int_{\mathbb{R}} f(s) ds$  pointwise.
3. Let  $h : [0, \infty) \rightarrow [0, \infty)$  have a non-negative continuous derivative,  $h(0) = 0$ , and  $h(\infty) = \infty$ . Prove that

$$\int_0^\infty \int_{[h' \geq s]} \exp(-h(t)^2) dt ds = \sqrt{\pi}/2.$$

4. Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be complete  $\sigma$ -finite positive measure spaces, and  $p \in [1, \infty)$ . Consider the map

$$[f, g] \in L^p(\mu) \times L^p(\nu) \rightarrow F(x, y) := f(x)g(y).$$

Prove:

- (a)  $F \in L^p(\mu \times \nu)$  and
- $$\|F\|_{L^p(\mu \times \nu)} = \|f\|_{L^p(\mu)} \|g\|_{L^p(\nu)}.$$
- (b) The map  $[f, g] \rightarrow F$  is continuous from  $L^p(\mu) \times L^p(\nu)$  to  $L^p(\mu \times \nu)$ .
5. Let  $f : \mathbb{R}^2 \rightarrow \mathbb{C}$  be Lebesgue measurable, such that  $|f(x, y)| \leq M e^{-x^2} I_{[-|x|, |x|]}(y)$  on  $\mathbb{R}^2$ , for some constant  $M > 0$ . Prove:
- (a)  $f \in L^p(\mathbb{R}^2)$  for all  $p \in [1, \infty)$ , and  $\|f\|_{L^p(\mathbb{R}^2)} \leq M(2/p)^{1/p}$ .
- (b) Suppose  $h : \mathbb{R} \rightarrow \mathbb{C}$  is continuous and vanishes outside the interval  $[-1, 1]$ . Define  $f : \mathbb{R}^2 \rightarrow \mathbb{C}$  by  $f(x, y) = e^{-x^2} h(y/x)$  for  $x \neq 0$  and  $f(0, y) = 0$ . Then  $\int_{\mathbb{R}^2} f dx dy = \int_{-1}^1 h(t) dt$ .
6. Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ . Prove:
- (a) If  $f(x, \cdot)$  is Borel for all real  $x$  and  $f(\cdot, y)$  is continuous for all real  $y$ , then  $f$  is Borel on  $\mathbb{R}^2$ .
- (b) If  $f(x, \cdot)$  is Lebesgue measurable for all  $x$  in some dense set  $E \subset \mathbb{R}$  and  $f(\cdot, y)$  is continuous for almost all  $y \in \mathbb{R}$ , then  $f$  is Lebesgue measurable on  $\mathbb{R}^2$ .

## Convolution and Fourier transform

7. If  $E \subset \mathbb{R}$ , denote

$$\tilde{E} := \{(x, y) \in \mathbb{R}^2; x - y \in E\}$$

and

$$\mathcal{S} := \{E \subset \mathbb{R}; \tilde{E} \in \mathcal{B}(\mathbb{R}^2)\},$$

where  $\mathcal{B}(\mathbb{R}^2)$  is the Borel  $\sigma$ -algebra on  $\mathbb{R}^2$ . Prove:

- (a)  $\mathcal{S}$  is a  $\sigma$ -algebra on  $\mathbb{R}$  which contains the open sets (hence  $\mathcal{B}(\mathbb{R}) \subset \mathcal{S}$ ).
- (b) If  $f$  is a Borel function on  $\mathbb{R}$ , then  $f(x - y)$  is a Borel function on  $\mathbb{R}^2$ .
- (c) If  $f, g$  are integrable Borel functions on  $\mathbb{R}$ , then  $f(x - y)g(y)$  is an integrable Borel function on  $\mathbb{R}^2$  and its  $L^1(\mathbb{R}^2)$ -norm is equal to the product of the  $L^1(\mathbb{R})$  norms of  $f$  and  $g$ .
- (d) Let  $L^1(\mathbb{R})$  and  $L^1(\mathbb{R}^2)$  be the Lebesgue spaces for the Lebesgue measure spaces on  $\mathbb{R}$  and  $\mathbb{R}^2$  respectively. If  $f, g \in L^1(\mathbb{R})$ , then  $f(x - y)g(y) \in L^1(\mathbb{R}^2)$ ,

$$\|f(x - y)g(y)\|_{L^1(\mathbb{R}^2)} = \|f\|_1 \|g\|_1,$$

and

$$\int_{\mathbb{R}} |f(x - y)g(y)| dy < \infty \quad (1)$$

for almost all  $x$ .

- (e) For  $x$  such that (1) holds, define

$$(f * g)(x) = \int_{\mathbb{R}} f(x - y)g(y) dy. \quad (2)$$

Show that the (almost everywhere defined and finite-valued) function  $f * g$  (called the *convolution* of  $f$  and  $g$ ) is in  $L^1(\mathbb{R})$ , and

$$\|f * g\|_1 \leq \|f\|_1 \|g\|_1. \quad (3)$$

- (f) For  $f \in L^1(\mathbb{R})$ , define its *Fourier transform*  $Ff$  by

$$(Ff)(t) = \int_{\mathbb{R}} f(x)e^{-ixt} dx \quad (t \in \mathbb{R}). \quad (4)$$

Show that  $Ff : \mathbb{R} \rightarrow \mathbb{C}$  is continuous, bounded by  $\|f\|_1$ , and  $F(f * g) = (Ff)(Fg)$  for all  $f, g \in L^1(\mathbb{R})$ .

- (g) If  $f = I_{(a,b]}$  for  $-\infty < a < b < \infty$ , then

$$\lim_{|t| \rightarrow \infty} (Ff)(t) = 0. \quad (5)$$

- (h) Show that the *step functions* (i.e. finite linear combinations of indicators of disjoint intervals  $(a_k, b_k]$ ) are dense in  $C_c(\mathbb{R})$  (the normed space of

continuous complex functions on  $\mathbb{R}$  with compact support, with pointwise operations and supremum norm), and hence also in  $L^p(\mathbb{R})$  for any  $1 \leq p < \infty$ .

(k) Prove (5) for any  $f \in L^1(\mathbb{R})$ . (This is the *Riemann–Lebesgue lemma*.)

(l) Generalize the above statements to functions on  $\mathbb{R}^k$ .

8. Let  $p \in [1, \infty)$  and let  $q$  be its conjugate exponent. Let  $K : \mathbb{R}^2 \rightarrow \mathbb{C}$  be Lebesgue measurable such that

$$\tilde{K}(y) := \int_{\mathbb{R}} |K(x, y)| dx \in L^q(\mathbb{R}).$$

Denote

$$(Tf)(x) = \int_{\mathbb{R}} K(x, y)f(y) dy$$

Prove:

(a)  $\int_{\mathbb{R}} \int_{\mathbb{R}} |K(x, y)f(y)| dy dx \leq \|\tilde{K}\|_q \|f\|_p$  for all  $f \in L^p(\mathbb{R})$ . Conclude that  $\tilde{K}(x, \cdot)f \in L^1(\mathbb{R})$  for almost all  $x$ , and therefore  $Tf$  is well defined a.e. (when  $f \in L^p$ ).

(b)  $T$  is a continuous (linear) map of  $L^p(\mathbb{R})$  into  $L^1(\mathbb{R})$ , and  $\|Tf\|_1 \leq \|\tilde{K}\|_q \|f\|_p$ .

9. Apply Fubini's theorem to the function  $e^{-xy} \sin x$  in order to prove the (Dirichlet) formula

$$\int_0^\infty \frac{\sin x}{x} dx = \pi/2.$$



# 3

## Measure and topology

In this chapter, the space  $X$  will be a topological space, and we shall be interested in constructing a measure space  $(X, \mathcal{M}, \mu)$  with a ‘natural’ affinity to the given topology.

We recall first some basic topological concepts.

### 3.1 Partition of unity

A *Hausdorff space* (or  $T_2$ -space) is a topological space  $(X, \tau)$  in which distinct points have disjoint open neighbourhoods. A Hausdorff space  $X$  is *locally compact* if each point in  $X$  has a *compact neighbourhood*. A Hausdorff space  $X$  can be imbedded (homeomorphically) as a dense subspace of a compact space  $Y$  (the *Alexandroff one-point compactification of  $X$* ), and  $Y$  is Hausdorff if and only if  $X$  is locally compact. In that case,  $Y$  is *normal* (as a compact Hausdorff space), and Urysohn’s lemma is valid in  $Y$ , that is, given disjoint closed sets  $A, B$  in  $Y$ , there exists a continuous function  $h : Y \rightarrow [0, 1]$  such that  $h(A) = \{0\}$  and  $h(B) = \{1\}$ . Theorem 3.1 below ‘translates’ this result to  $X$ . We need the following important concept: for any complex continuous function  $f$  on  $X$ , the *support* of  $f$  (denoted  $\text{supp } f$ ) is defined as the *closure* of  $[f^{-1}(\{0\})]^c$ .

**Theorem 3.1 (Urysohn’s lemma for locally compact Hausdorff space).**  
*Let  $X$  be a locally compact Hausdorff space, let  $U \subset X$  be open, and let  $K \subset U$  be compact.*

*Then there exists a continuous function  $f : X \rightarrow [0, 1]$ , with compact support such that  $\text{supp } f \subset U$  and  $f(K) = \{1\}$ .*

**Proof.** Let  $Y$  be the Alexandroff one-point compactification of  $X$ . The set  $U$  is open in  $X$ , hence in  $Y$ . The set  $K$  is compact in  $X$ , hence in  $Y$ , and is therefore closed in  $Y$  (since  $Y$  is Hausdorff). Since  $Y$  is normal, and the closed set  $K$  is contained in the open set  $U$ , there exists an open set  $V$  in  $Y$  such that  $K \subset V$  and  $\text{cl}_Y(V) \subset U$  (where  $\text{cl}_Y$  denotes the closure operator in  $Y$ ). Therefore,  $K \subset V \cap X := W$ ,  $W$  is open in  $X$ , and  $\text{cl}_X(W) = \text{cl}_Y(V) \cap X \subset U$ .

In the sequel, all closures are *closures in  $X$* .

Since  $X$  is locally compact, each  $x \in K$  has an open neighbourhood  $N_x$  with compact closure. Then  $N_x \cap W$  is an open neighbourhood of  $x$ , with closure contained in  $\text{cl}(N_x) \cap \text{cl}(W)$ , which is compact (since  $\text{cl}(N_x)$  is compact), and is contained in  $U$ . By compactness of  $K$ , we obtain finitely many points  $x_i \in K$  such that

$$K \subset \bigcup_{i=1}^p (N_{x_i} \cap W) := N.$$

The open set  $N$  has closure equal to the union of the compact sets  $\text{cl}(N_{x_i} \cap W)$ , which is compact and contained in  $U$ . We proved that whenever  $K$  is a compact subset of the open set  $U$  in  $X$ , there exists an open set  $W$  with compact closure such that

$$K \subset W \subset \text{cl}(W) \subset U. \quad (1)$$

(We wrote  $W$  instead of  $N$ .)

The sets  $K$  and  $Y - W$  are disjoint closed sets in  $Y$ . By Urysohn's lemma for normal spaces, there exists a continuous function  $h : Y \rightarrow [0, 1]$  such that  $h = 0$  on  $Y - W$  and  $h(K) = \{1\}$ . Let  $f := h|_X$  (the restriction of  $h$  to  $X$ ). Then  $f : X \rightarrow [0, 1]$  is continuous,  $f(K) = \{1\}$ , and since  $[f \neq 0] \subset W$ , we have by (1)

$$\text{supp } f \subset \text{cl}(W) \subset U.$$

In particular,  $f$  has compact support (since  $\text{supp } f$  is a closed subset of the compact set  $\text{cl}(W)$ ).  $\square$

**Notation 3.2.** We shall denote the space of all complex (real) continuous functions with compact support on the locally compact Hausdorff space  $X$  by  $C_c(X)$  ( $C_c^{\mathbb{R}}(X)$ , respectively). By Theorem 3.1, this is a *non-trivial* normed vector space over  $\mathbb{C}$  (over  $\mathbb{R}$ , respectively), with the *uniform norm*

$$\|f\|_u := \sup_X |f|.$$

The *positive cone*

$$C_c^+(X) := \{f \in C_c^{\mathbb{R}}(X); f \geq 0\}$$

will play a central role in this section.

Actually, Theorem 3.1 asserts that for any *open* set  $U \neq \emptyset$ , the set

$$\Omega(U) := \{f \in C_c^+(X); f \leq 1, \text{supp } f \subset U\} \quad (2)$$

is  $\neq \{0\}$ .

The following theorem generalizes Theorem 3.1 to the case of a finite open cover of the compact set  $K$ . Any set of functions  $\{h_1, \dots, h_n\}$  with the properties described in Theorem 3.3 is called a *partition of unity in  $C_c(X)$  subordinate to the open cover  $\{V_1, \dots, V_n\}$  of the compact set  $K$* .

**Theorem 3.3.** *Let  $X$  be a locally compact Hausdorff space, let  $K \subset X$  be compact, and let  $V_1, \dots, V_n$  be open subsets of  $X$  such that*

$$K \subset V_1 \cup \dots \cup V_n.$$

*Then there exist  $h_i \in \Omega(V_i)$ ,  $i = 1, \dots, n$ , such that*

$$1 = h_1 + \dots + h_n \quad \text{on } K.$$

**Proof.** For each  $x \in K$ , there exists an index  $i(x)$  (between 1 and  $n$ ) such that  $x \in V_{i(x)}$ . By (1) (applied to the compact set  $\{x\}$  contained in the open set  $V_{i(x)}$ ), there exists an open set  $W_x$  with compact closure such that

$$x \in W_x \subset \text{cl}(W_x) \subset V_{i(x)}. \quad (3)$$

By the compactness of  $K$ , there exist  $x_1, \dots, x_m \in K$  such that

$$K \subset W_{x_1} \cup \dots \cup W_{x_m}.$$

Define for each  $i = 1, \dots, n$

$$H_i := \bigcup \{ \text{cl}(W_{x_j}); \text{cl}(W_{x_j}) \subset V_i \}.$$

As a finite union of compact sets,  $H_i$  is compact, and contained in  $V_i$ . By Theorem 3.1, there exist  $f_i \in \Omega(V_i)$ , such that  $f_i = 1$  on  $H_i$ . Take  $h_1 = f_1$  and for  $k = 2, \dots, n$ , consider the continuous functions

$$h_k = f_k \prod_{i=1}^{k-1} (1 - f_i).$$

An immediate induction on  $k$  (up to  $n$ ) shows that

$$h_1 + \dots + h_k = 1 - \prod_{i=1}^k (1 - f_i).$$

Since  $f_i = 1$  on  $H_i$ , the product  $\prod_{i=1}^n (1 - f_i)$  vanishes on  $\bigcup_{i=1}^n H_i$ , and this union contains the union of the  $W_{x_j}$ ,  $j = 1, \dots, m$ , hence contains  $K$ . Therefore,  $h_1 + \dots + h_n = 1$  on  $K$ . The support of  $h_k$  is contained in the support of  $f_k$ , hence in  $V_k$ , and  $0 \leq h_k \leq 1$  trivially.  $\square$

## 3.2 Positive linear functionals

**Definition 3.4.** A linear functional  $\phi : C_c(X) \rightarrow \mathbb{C}$  is said to be *positive* if  $\phi(f) \geq 0$  for all  $f \in C_c^+(X)$ .

This is clearly equivalent to the *monotonicity* condition:  $\phi(f) \geq \phi(g)$  whenever  $f \geq g$  ( $f, g \in C_c^{\mathbb{R}}(X)$ ).

Let  $V \in \tau$ . The indicator  $I_V$  is continuous if and only if  $V$  is also closed. In that case,  $I_V \in \Omega(V)$ , and  $f \leq I_V$  for all  $f \in \Omega(V)$ . By monotonicity of  $\phi$ ,  $0 \leq \phi(f) \leq \phi(I_V)$  for all  $f \in \Omega(V)$ , and therefore  $0 \leq \sup_{f \in \Omega(V)} \phi(f) \leq \phi(I_V)$ . Since  $I_V \in \Omega(V)$ , we actually have an identity (in our special case). The set function  $\phi(I_V)$  is a ‘natural’ candidate for a measure of  $V$  (associated to the given functional). Since the supremum expression makes sense for arbitrary open sets, we take it as the definition of our ‘measure’ (first defined on  $\tau$ ).

**Definition 3.5.** Let  $(X, \tau)$  be a locally compact Hausdorff space, and let  $\phi$  be a positive linear functional on  $C_c(X)$ . Set

$$\mu(V) := \sup_{f \in \Omega(V)} \phi(f) \quad (V \in \tau).$$

Note that by definition

$$\phi(f) \leq \mu(V)$$

whenever  $f \in \Omega(V)$ .

**Lemma 3.6.**  $\mu$  is non-negative, monotonic, and subadditive (on  $\tau$ ), and  $\mu(\emptyset) = 0$ .

**Proof.** For each  $f \in \Omega(V) \subset C_c^+(X)$ ,  $\phi(f) \geq 0$ , and therefore  $\mu(V) \geq 0$  (for all  $V \in \tau$ ). Since  $\Omega(\emptyset) = \{0\}$  and  $\phi(0) = 0$ , we trivially have  $\mu(\emptyset) = 0$ . If  $V \subset W$  (with  $V, W \in \tau$ ), then  $\Omega(V) \subset \Omega(W)$ , so that  $\mu(V) \leq \mu(W)$ .

Next, let  $V_i \in \tau, i = 1, \dots, n$ , with union  $V$ . Fix  $f \in \Omega(V)$ . Let then  $h_1, \dots, h_n$  be a partition of unity in  $C_c(X)$  subordinate to the open covering  $V_1, \dots, V_n$  of the compact set  $\text{supp } f$ . Then

$$f = \sum_{i=1}^n h_i f \quad h_i f \in \Omega(V_i), \quad i = 1, \dots, n.$$

Therefore

$$\phi(f) = \sum_{i=1}^n \phi(h_i f) \leq \sum_{i=1}^n \mu(V_i).$$

Taking the supremum over all  $f \in \Omega(V)$ , we obtain

$$\mu(V) \leq \sum_{i=1}^n \mu(V_i).$$

□

We now extend the definition of  $\mu$  to  $\mathbb{P}(X)$ .

**Definition 3.7.** For any  $E \in \mathbb{P}(X)$ , we set

$$\mu^*(E) = \inf_{E \subset V \in \tau} \mu(V).$$

If  $E \in \tau$ , then  $E \subset E \in \tau$ , so that  $\mu^*(E) \leq \mu(E)$ . On the other hand, whenever  $E \subset V \in \tau$ ,  $\mu(E) \leq \mu(V)$  (by Lemma 3.6), so that  $\mu(E) \leq \mu^*(E)$ . Thus  $\mu^* = \mu$  on  $\tau$ , and  $\mu^*$  is indeed an extension of  $\mu$ .

**Lemma 3.8.**  $\mu^*$  is an outer measure.

**Proof.** First, since  $\emptyset \in \tau$ ,  $\mu^*(\emptyset) = \mu(\emptyset) = 0$ .

If  $E \subset F \subset X$ , then  $E \subset V \in \tau$  whenever  $F \subset V \in \tau$ , and therefore

$$\mu^*(E) := \inf_{E \subset V \in \tau} \mu(V) \leq \inf_{F \subset V \in \tau} \mu(V) := \mu^*(F),$$

proving the monotonicity of  $\mu^*$ .

Let  $\{E_i\}$  be any sequence of subsets of  $X$ . If  $\mu^*(E_i) = \infty$  for *some*  $i$ , then  $\sum_i \mu^*(E_i) = \infty \geq \mu^*(\bigcup_i E_i)$  trivially. Assume therefore that  $\mu^*(E_i) < \infty$  for *all*  $i$ . Let  $\epsilon > 0$ . By Definition 3.7, there exist open sets  $V_i$  such that

$$E_i \subset V_i; \quad \mu(V_i) < \mu^*(E_i) + \epsilon/2^i, \quad i = 1, 2, \dots$$

Let  $E$  and  $V$  be the unions of the sets  $E_i$  and  $V_i$ , respectively. If  $f \in \Omega(V)$ , then  $\{V_i\}$  is an open cover of the compact set  $\text{supp } f$ , and there exists therefore  $n \in \mathbb{N}$  such that  $\text{supp } f \subset V_1 \cup \dots \cup V_n$ , that is,  $f \in \Omega(V_1 \cup \dots \cup V_n)$ . Hence, by Lemma 3.6,

$$\phi(f) \leq \mu(V_1 \cup \dots \cup V_n) \leq \mu(V_1) + \dots + \mu(V_n) \leq \sum_{i=1}^{\infty} \mu^*(E_i) + \epsilon.$$

Taking the supremum over all  $f \in \Omega(V)$ , we get

$$\mu(V) \leq \sum_i \mu^*(E_i) + \epsilon.$$

Since  $E \subset V \in \tau$ , we get by definition

$$\mu^*(E) \leq \mu(V) \leq \sum_i \mu^*(E_i) + \epsilon,$$

and the  $\sigma$ -subadditivity of  $\mu^*$  follows from the arbitrariness of  $\epsilon$ .  $\square$

At this stage, we could appeal to Caratheodory's theory (cf. Chapter 2) to obtain the wanted measure space. We prefer, however, to give a construction independent of Chapter 2, and strongly linked to the topology of the space.

Denote by  $\mathcal{K}$  the family of all *compact* subsets of  $X$ .

**Lemma 3.9.**  $\mu^*$  is finite and additive on  $\mathcal{K}$ .

**Proof.** Let  $K \in \mathcal{K}$ . By Theorem 3.1, there exists  $f \in C_c(X)$  such that  $0 \leq f \leq 1$  and  $f = 1$  on  $K$ . Let  $V := [f > 1/2]$ . Then  $K \subset V \in \tau$ , so that  $\mu^*(K) \leq \mu(V)$ . On the other hand, for all  $h \in \Omega(V)$ ,  $h \leq 1 < 2f$  on  $V$ , so that  $\mu(V) := \sup_{h \in \Omega(V)} \phi(h) \leq \phi(2f)$  (by monotonicity of  $\phi$ ), and therefore  $\mu^*(K) \leq \phi(2f) < \infty$ .

Next, let  $K_i \in \mathcal{K}$  ( $i = 1, 2$ ) be disjoint, and let  $\epsilon > 0$ . Then  $K_1 \subset K_2^c \in \tau$ . By (1) in the proof of Theorem 3.1, there exists  $V_1$  open with compact closure such

that  $K_1 \subset V_1$  and  $\text{cl}(V_1) \subset K_2^c$ . Hence  $K_2 \subset [\text{cl}(V_1)]^c := V_2 \in \tau$ , and  $V_2 \subset V_1^c$ . Thus,  $V_i$  are disjoint open sets containing  $K_i$  (respectively).

Since  $K_1 \cup K_2$  is compact,  $\mu^*(K_1 \cup K_2) < \infty$ , and therefore, by definition of  $\mu^*$ , there exists an open set  $W$  such that  $K_1 \cup K_2 \subset W$  and

$$\mu(W) < \mu^*(K_1 \cup K_2) + \epsilon. \quad (1)$$

By definition of  $\mu$  on open sets, there exist  $f_i \in \Omega(W \cap V_i)$  such that

$$\mu(W \cap V_i) < \phi(f_i) + \epsilon, \quad i = 1, 2. \quad (2)$$

Since  $K_i \subset W \cap V_i, i = 1, 2$ , it follows from (2) that

$$\mu^*(K_1) + \mu^*(K_2) \leq \mu(W \cap V_1) + \mu(W \cap V_2) < \phi(f_1) + \phi(f_2) + 2\epsilon = \phi(f_1 + f_2) + 2\epsilon.$$

However,  $f_1 + f_2 \in \Omega(W)$ . Hence, by (1),

$$\mu^*(K_1) + \mu^*(K_2) < \mu(W) + 2\epsilon < \mu^*(K_1 \cup K_2) + 3\epsilon.$$

The arbitrariness of  $\epsilon$  and the subadditivity of  $\mu^*$  give  $\mu^*(K_1 \cup K_2) = \mu^*(K_1) + \mu^*(K_2)$ .  $\square$

**Definition 3.10.** The *inner measure* of  $E \in \mathbb{P}(X)$  is defined by

$$\mu_*(E) := \sup_{\{K \in \mathcal{K}; K \subset E\}} \mu^*(K).$$

By monotonicity of  $\mu^*$ , we have  $\mu_* \leq \mu^*$ , and equality (and finiteness) is valid on  $\mathcal{K}$  (cf. Lemma 3.9). We consider then the family

$$\mathcal{M}_0 := \{E \in \mathbb{P}(X); \mu_*(E) = \mu^*(E) < \infty\}. \quad (3)$$

We just observed that

$$\mathcal{K} \subset \mathcal{M}_0. \quad (4)$$

Another important subfamily of  $\mathcal{M}_0$  consists of the *open sets of finite measure*  $\mu$ :

**Lemma 3.11.**  $\tau_0 := \{V \in \tau; \mu(V) < \infty\} \subset \mathcal{M}_0$ .

**Proof.** Let  $V \in \tau_0$  and  $\epsilon > 0$ . Since  $\mu(V) < \infty$ , it follows from the definition of  $\mu$  that there exists  $f \in \Omega(V)$  such that  $\mu(V) - \epsilon < \phi(f)$ . Let  $K := \text{supp } f$ . Whenever  $K \subset W \in \tau$ , we have necessarily  $f \in \Omega(W)$ , and therefore  $\phi(f) \leq \mu(W)$ . Hence

$$\begin{aligned} \mu(V) - \epsilon < \phi(f) &\leq \inf_{\{W \in \tau; K \subset W\}} \mu(W) \\ &:= \mu^*(K) \leq \mu_*(V) \leq \mu^*(V) = \mu(V), \end{aligned}$$

and so  $\mu_*(V) = \mu(V) (= \mu^*(V))$  by the arbitrariness of  $\epsilon$ .  $\square$

**Lemma 3.12.**  $\mu^*$  is  $\sigma$ -additive on  $\mathcal{M}_0$ , that is, for any sequence of mutually disjoint sets  $E_i \in \mathcal{M}_0$  with union  $E$ ,

$$\mu^*(E) = \sum_i \mu^*(E_i). \quad (5)$$

Furthermore,  $E \in \mathcal{M}_0$  if  $\mu^*(E) < \infty$ . In particular,  $\mathcal{M}_0$  is closed under finite disjoint unions.

**Proof.** By Lemma 3.8, it suffices to prove the inequality  $\geq$  in (5). Since this inequality is trivial when  $\mu^*(E) = \infty$ , we may assume that  $\mu^*(E) < \infty$ .

Let  $\epsilon > 0$  be given. For all  $i = 0, 1, \dots$ , since  $E_i \in \mathcal{M}_0$ , there exist  $H_i \in \mathcal{K}$  such that  $H_i \subset E_i$  and

$$\mu^*(E_i) < \mu^*(H_i) + \epsilon/2^i. \quad (6)$$

The sets  $H_i$  are necessarily mutually disjoint. Define

$$K_n = \bigcup_{i=1}^n H_i \quad n = 1, 2, \dots$$

Then  $K_n \subset E$ , and by Lemma 3.9 and (6),

$$\sum_{i=1}^n \mu^*(E_i) < \sum_{i=1}^n \mu^*(H_i) + \epsilon = \mu^*(K_n) + \epsilon \leq \mu^*(E) + \epsilon.$$

The arbitrariness of  $\epsilon$  proves the wanted inequality  $\geq$  in (5). Now  $\mu^*(E)$  is the finite sum of the series  $\sum \mu^*(E_i)$ ; hence, given  $\epsilon > 0$ , we may choose  $n \in \mathbb{N}$  such that  $\mu^*(E) < \sum_{i=1}^n \mu^*(E_i) + \epsilon$ . For that  $n$ , if the compact set  $K_n$  is defined as before, we get  $\mu^*(E) < \mu^*(K_n) + 2\epsilon$ , and therefore  $\mu^*(E) = \mu_*(E)$ , that is,  $E \in \mathcal{M}_0$ .  $\square$

**Lemma 3.13.**  $\mathcal{M}_0$  is a ring of subsets of  $X$ , that is, it is closed under the operations  $\cup, \cap, -$  between sets. Furthermore, if  $E \in \mathcal{M}_0$ , then for each  $\epsilon > 0$ , there exist  $K \in \mathcal{K}$  and  $V \in \tau_0$  such that

$$K \subset E \subset V; \quad \mu(V - K) < \epsilon. \quad (7)$$

**Proof.** Note that  $V - K$  is open, so that (7) makes sense. We prove it first. By definition of  $\mu^*$  and  $\mathcal{M}_0$ , there exist  $V \in \tau$  and  $K \in \mathcal{K}$  such that  $K \subset E \subset V$  and

$$\mu(V) - \epsilon/2 < \mu^*(E) < \mu^*(K) + \epsilon/2.$$

In particular  $\mu(V) < \infty$  and  $\mu(V - K) \leq \mu(V) < \infty$ , so that  $V, V - K \in \tau_0$ . By Lemma 3.11,  $V - K \in \mathcal{M}_0$ . Since also  $K \in \mathcal{K} \subset \mathcal{M}_0$ , it follows from Lemma 3.12 that

$$\mu^*(K) + \mu^*(V - K) = \mu(V) < \mu^*(K) + \epsilon.$$

Since  $\mu^*(K) < \infty$ , we obtain  $\mu^*(V - K) < \epsilon$ .

Now, let  $E_i \in \mathcal{M}_0, i = 1, 2$ . Given  $\epsilon > 0$ , pick  $K_i, V_i$  as in (7). Since

$$E_1 - E_2 \subset V_1 - K_2 \subset (V_1 - K_1) \cup (K_1 - V_2) \cup (V_2 - K_2),$$

and the sets on the right are disjoint sets in  $\mathcal{M}_0$ , we have by Lemma 3.12,

$$\mu^*(E_1 - E_2) < \mu^*(K_1 - V_2) + 2\epsilon.$$

Since  $K_1 - V_2 (= K_1 \cap V_2^c)$  is a *compact* subset of  $E_1 - E_2$ , it follows that  $\mu^*(E_1 - E_2) < \mu_*(E_1 - E_2) + 2\epsilon$  (and of course  $\mu^*(E_1 - E_2) \leq \mu^*(E_1) < \infty$ ), so that  $E_1 - E_2 \in \mathcal{M}_0$ .

Now  $E_1 \cup E_2 = (E_1 - E_2) \cup E_2 \in \mathcal{M}_0$  as the disjoint union of sets in  $\mathcal{M}_0$  (cf. Lemma 3.12), and  $E_1 \cap E_2 = E_1 - (E_1 - E_2) \in \mathcal{M}_0$  since  $\mathcal{M}_0$  is closed under difference.  $\square$

**Definition 3.14.**

$$\mathcal{M} = \{E \in \mathbb{P}(X); E \cap K \in \mathcal{M}_0 \text{ for all } K \in \mathcal{K}\}.$$

If  $E$  is a *closed* set, then  $E \cap K \in \mathcal{K} \subset \mathcal{M}_0$  for all  $K \in \mathcal{K}$ , so that  $\mathcal{M}$  contains all closed sets.

**Lemma 3.15.**  $\mathcal{M}$  is a  $\sigma$ -algebra containing the Borel algebra  $\mathcal{B}$  (of  $X$ ), and

$$\mathcal{M}_0 = \{E \in \mathcal{M}; \mu^*(E) < \infty\}. \quad (8)$$

Furthermore, the restriction  $\mu := \mu^*|_{\mathcal{M}}$  is a measure.

**Proof.** We first prove (8). If  $E \in \mathcal{M}_0$ , then since  $\mathcal{K} \subset \mathcal{M}_0$ , we surely have  $E \cap K \in \mathcal{M}_0$  for all  $K \in \mathcal{K}$  (by Lemma 3.13), so that  $E \in \mathcal{M}$  (and of course  $\mu^*(E) < \infty$  by definition).

On the other hand, suppose  $E \in \mathcal{M}$  and  $\mu^*(E) < \infty$ . Let  $\epsilon > 0$ . By definition, there exists  $V \in \tau$  such that  $E \subset V$  and  $\mu(V) < \mu^*(E) + 1 < \infty$ . By Lemma 3.11,  $V \in \mathcal{M}_0$ . Applying Lemma 3.13 (7) to  $V$ , we obtain a set  $K \in \mathcal{K}$  such that  $K \subset V$  and  $\mu^*(V - K) < \epsilon$ . Since  $E \cap K \in \mathcal{M}_0$  (by definition of  $\mathcal{M}$ ), there exists  $H \in \mathcal{K}$  such that  $H \subset E \cap K$  and

$$\mu^*(E \cap K) < \mu^*(H) + \epsilon.$$

Now  $E \subset (E \cap K) \cup (V - K)$ , so that by Lemma 3.8,

$$\mu^*(E) \leq \mu^*(E \cap K) + \mu^*(V - K) < \mu^*(H) + 2\epsilon \leq \mu_*(E) + 2\epsilon.$$

The arbitrariness of  $\epsilon$  implies that  $\mu^*(E) \leq \mu_*(E)$ , so that  $E \in \mathcal{M}_0$ , and (8) is proved.

Since  $\mathcal{M}$  contains all closed sets (see observation following Definition 3.14), we may conclude that  $\mathcal{B} \subset \mathcal{M}$  once we know that  $\mathcal{M}$  is a  $\sigma$ -algebra.

If  $E \in \mathcal{M}$ , then for all  $K \in \mathcal{K}$ ,

$$E^c \cap K = K - (E \cap K) \in \mathcal{M}_0$$

by definition and Lemma 3.13. Hence  $E^c \in \mathcal{M}$ .



Let  $E_i \in \mathcal{M}$ ,  $i = 1, 2, \dots$ , with union  $E$ . Then for each  $K \in \mathcal{K}$ ,

$$E \cap K = \bigcup_i E_i \cap K = \bigcup_i F_i,$$

where

$$F_i := (E_i \cap K) - \bigcup_{j < i} (E_j \cap K)$$

are mutually disjoint sets in  $\mathcal{M}_0$  (by definition of  $\mathcal{M}$  and Lemma 3.13). Since  $\mu^*(E \cap K) \leq \mu^*(K) < \infty$  (by Lemma 3.9), it follows from Lemma 3.12 that  $E \cap K \in \mathcal{M}_0$ , and we conclude that  $E \in \mathcal{M}$ .

Finally, let  $E_i \in \mathcal{M}$  be mutually disjoint with union  $E$ . If  $\mu^*(E_i) = \infty$  for some  $i$ , then also  $\mu^*(E) = \infty$  by monotonicity, and  $\mu^*(E) = \sum_i \mu^*(E_i)$  trivially. Suppose then that  $\mu^*(E_i) < \infty$  for all  $i$ . By (8), it follows that  $E_i \in \mathcal{M}_0$  for all  $i$ , and the wanted  $\sigma$ -additivity of  $\mu = \mu^*|_{\mathcal{M}}$  follows from Lemma 3.12.  $\square$

We call  $(X, \mathcal{M}, \mu)$  the *measure space associated with the positive linear functional  $\phi$* . Integration in the following discussion is performed over this measure space.

**Lemma 3.16.** *For all  $f \in C_c^+(X)$ ,*

$$\phi(f) \leq \int_X f d\mu.$$

**Proof.** Fix  $f \in C_c^+(X)$ , let  $K$  be its (compact) support, and let  $0 \leq a < b$  be such that  $[a, b]$  contains the (compact) range of  $f$ . Given  $\epsilon > 0$ , choose points

$$0 \leq y_0 \leq a < y_1 < \dots < y_n = b$$

such that  $y_k - y_{k-1} < \epsilon$ , and set

$$E_k = [y_{k-1} < f \leq y_k] \quad k = 1, \dots, n.$$

Since  $f$  is continuous with support  $K$ , the sets  $E_k$  are disjoint Borel sets with union  $K$ . By definition of our measure space, there exist open sets  $V_k$  such that

$$E_k \subset V_k; \quad \mu(V_k) < \mu(E_k) + \epsilon/n$$

for  $k = 1, \dots, n$ . Since  $f \leq y_k$  on  $E_k$ , it follows from the continuity of  $f$  that there exist open sets  $U_k$  such that

$$E_k \subset U_k; \quad f < y_k + \epsilon \text{ on } U_k.$$

Taking  $W_k := V_k \cap U_k$ , we have for all  $k = 1, \dots, n$

$$E_k \subset W_k; \quad \mu(W_k) < \mu(E_k) + \epsilon/n; \quad f < y_k + \epsilon \text{ on } W_k.$$

Let  $\{h_k; k = 1, \dots, n\}$  be a partition of unity in  $C_c(X)$  subordinate to the open covering  $\{W_k; k = 1, \dots, n\}$  of  $K$ . Then

$$f = \sum_{k=1}^n h_k f,$$

and for all  $k = 1, \dots, n$ ,

$$h_k f \leq h_k (y_k + \epsilon)$$

(since  $h_k \in \Omega(W_k)$  and  $f < y_k + \epsilon$  on  $W_k$ ),

$$\phi(h_k) \leq \mu(W_k)$$

(since  $h_k \in \Omega(W_k)$ ), and

$$y_k = y_{k-1} + (y_k - y_{k-1}) < f + \epsilon \text{ on } E_k.$$

Therefore

$$\begin{aligned} \phi(f) &= \sum_{k=1}^n \phi(h_k f) \leq \sum_k (y_k + \epsilon) \phi(h_k) \leq \sum_k (y_k + \epsilon) \mu(W_k) \\ &\leq \sum_k (y_k + \epsilon) [\mu(E_k) + \epsilon/n] \leq \sum_k y_k \mu(E_k) + \epsilon \mu(K) + \sum_k (y_k + \epsilon) \epsilon/n \\ &\leq \sum_k \int_{E_k} (f + \epsilon) d\mu + \epsilon \mu(K) + (b + \epsilon) \epsilon = \int_X f d\mu + \epsilon [2\mu(K) + b + \epsilon]. \end{aligned}$$

Since  $\mu(K) < \infty$ , the lemma follows from the arbitrariness of  $\epsilon$ .  $\square$

**Lemma 3.17.** *For all  $f \in C_c(X)$ ,*

$$\phi(f) = \int_X f d\mu.$$

**Proof.** By linearity, it suffices to prove the lemma for *real*  $f \in C_c(X)$ . Given such  $f$ , let  $K$  be its (compact) support, and let  $M = \sup |f|$ . For any  $\epsilon > 0$ , choose  $V$  open such that  $K \subset V$  and  $\mu(V) < \mu(K) + \epsilon$ ; then choose  $h \in \Omega(V)$  such that  $\mu(V) < \phi(h) + \epsilon$ . By Urysohn's lemma, there is a function  $k \in \Omega(V)$  such that  $k = 1$  on  $K$ . Let  $g = \max\{h, k\}$  ( $= (1/2)(h + k + |h - k|) \in C_c^+(X)$ ). Then  $g \in \Omega(V)$ ,  $g = 1$  on  $K$ , and  $\mu(V) < \phi(g) + \epsilon$ . Define  $F = f + Mg$ . Then  $F \in C_c^+(X)$  and  $F = f + M$  on  $K$ . By Lemma 3.16,

$$\phi(F) \leq \int_X F d\mu,$$

that is, since  $g \in \Omega(V)$ ,

$$\begin{aligned} \phi(f) + M\phi(g) &\leq \int_X f d\mu + M \int_X g d\mu \leq \int_X f d\mu + M\mu(V) \\ &\leq \int_X f d\mu + M[\phi(g) + \epsilon]. \end{aligned}$$

Hence, by the arbitrariness of  $\epsilon$ ,

$$\phi(f) \leq \int_X f \, d\mu$$

for all real  $f \in C_c(X)$ . Replacing  $f$  by  $-f$ , we also have

$$-\phi(f) = \phi(-f) \leq \int_X (-f) \, d\mu = - \int_X f \, d\mu,$$

so that  $\phi(f) = \int_X f \, d\mu$ . □

### 3.3 The Riesz–Markov representation theorem

**Theorem 3.18 (Riesz–Markov).** *Let  $(X, \tau)$  be a locally compact Hausdorff space, and let  $\phi$  be a positive linear functional on  $C_c(X)$ . Let  $(X, \mathcal{M}, \mu)$  be the measure space associated with  $\phi$ . Then*

$$\phi(f) = \int_X f \, d\mu \quad f \in C_c(X). \quad (*)$$

In addition, the following properties are valid:

- (1)  $\mathcal{B}(X) \subset \mathcal{M}$ .
- (2)  $\mu$  is finite on  $\mathcal{K}$  (the compact subsets of  $X$ ).
- (3)  $\mu(E) = \inf_{E \subset V \in \tau} \mu(V)$  for all  $E \in \mathcal{M}$ .
- (4)  $\mu(E) = \sup_{\{K \in \mathcal{K}; K \subset E\}} \mu(K)$  (i) for all  $E \in \tau$ , and (ii) for all  $E \in \mathcal{M}$  with finite measure.
- (5) the measure space  $(X, \mathcal{M}, \mu)$  is complete.

Furthermore, the measure  $\mu$  is uniquely determined on  $\mathcal{M}$  by (\*), (2), (3), and (4)-(i).

**Proof.** Properties (\*), (1), (2), and (4)-(ii) are valid by Lemma 3.17, 3.15, 3.9, and 3.15 (together with Definition 3.10 and the following notation (3)), respectively. Property (3) follows from Definition 3.7, since  $\mu := \mu^*|_{\mathcal{M}}$ .

If  $E \in \mathcal{M}$  has measure zero, and  $F \subset E$ , then  $\mu^*(F) = 0$  and  $\mu^*(K) = 0$  for all  $K \in \mathcal{K}, K \subset F$  (by monotonicity), so that  $\mu_*(F) = 0 = \mu^*(F) < \infty$ , that is,  $F \in \mathcal{M}_0 \subset \mathcal{M}$ , and (5) is proved.

We prove (4)-(i). Let  $V \in \tau$ . If  $\mu(V) < \infty$ , then  $V \in \mathcal{M}_0$  by Lemma 3.11, and (4)-(i) follows from the definition of  $\mathcal{M}_0$ . Assume then that  $\mu(V) = \infty$ . By Definition 3.5, for each  $n \in \mathbb{N}$ , there exists  $f_n \in \Omega(V)$  such that  $\phi(f_n) > n$ . Let  $K_n := \text{supp}(f_n)$ . Then for all  $n$ ,

$$\mu_*(V) \geq \mu(K_n) \geq \int_{K_n} f_n \, d\mu = \phi(f_n) > n,$$

so that  $\mu_*(V) = \infty = \mu(V)$ , and (4)-(i) is valid for  $V$ .

Suppose  $\nu$  is any positive measure on  $\mathcal{M}$  satisfying Properties (\*), (2), (3), and (4)-(i). Let  $\epsilon > 0$  and  $K \in \mathcal{K}$ . By (2) and (3), there exists  $V \in \tau$  such that  $K \subset V$  and  $\nu(V) < \nu(K) + \epsilon$ . By Urysohn's lemma (3.1), there exists  $f \in \Omega(V)$  such that  $f = 1$  on  $K$ . Hence  $I_K \leq f \leq I_V$ , and therefore, by (\*) for both  $\mu$  and  $\nu$ ,

$$\mu(K) = \int_X I_K d\mu \leq \int_X f d\mu = \phi(f) = \int_X f d\nu \leq \int_X I_V d\nu = \nu(V) < \nu(K) + \epsilon.$$

Hence  $\mu(K) \leq \nu(K)$ , and so  $\mu(K) = \nu(K)$  by symmetry. By (4)-(i), it follows that  $\mu = \nu$  on  $\tau$ , hence on  $\mathcal{M}$ , by (3).  $\square$

In case  $X$  is  $\sigma$ -compact, the following additional structural properties are valid for the measure space associated with  $\phi$ .

**Theorem 3.19.** *Let  $X$  be a Hausdorff, locally compact,  $\sigma$ -compact space, and let  $(X, \mathcal{M}, \mu)$  be the measure space associated with the positive linear functional  $\phi$  on  $C_c(X)$ . Then:*

- (1) *For all  $E \in \mathcal{M}$  and  $\epsilon > 0$ , there exist  $F$  closed and  $V$  open such that*

$$F \subset E \subset V; \quad \mu(V - F) < \epsilon.$$

- (2) *Properties (3) and (4) in Theorem 3.18 are valid for all  $E \in \mathcal{M}$  (this fact is formulated by the expression:  $\mu$  is regular. One says also that  $\mu|_{\mathcal{B}(X)}$  is a regular Borel measure).*

- (3) *For all  $E \in \mathcal{M}$ , there exist an  $\mathcal{F}_\sigma$  set  $A$  and a  $\mathcal{G}_\delta$  set  $B$  such that*

$$A \subset E \subset B; \quad \mu(B - A) = 0$$

*(i.e., every set in  $\mathcal{M}$  is the union of an  $\mathcal{F}_\sigma$  set and a null set).*

**Proof.** The  $\sigma$ -compactness hypothesis means that  $X = \bigcup_i K_i$  with  $K_i$  compact. Let  $\epsilon > 0$  and  $E \in \mathcal{M}$ . By 3.18(2),  $\mu(K_i \cap E) \leq \mu(K_i) < \infty$ , and therefore, by 3.18(3), there exist open sets  $V_i$  such that

$$K_i \cap E \subset V_i; \quad \mu(V_i - (K_i \cap E)) < \epsilon/2^{i+1}, \quad i = 1, 2, \dots$$

Set  $V = \bigcup_i V_i$ . Then  $V$  is open, contains  $E$ , and

$$\mu(V - E) \leq \mu\left(\bigcup_i (V_i - (K_i \cap E))\right) < \epsilon/2.$$

Replacing  $E$  by  $E^c$ , we obtain in the same fashion an open set  $W$  containing  $E^c$  such that  $\mu(W - E^c) < \epsilon/2$ . Setting  $F := W^c$ , we obtain a closed set contained in  $E$  such that  $\mu(E - F) < \epsilon/2$ , and (1) follows.

Next, for an arbitrary closed set  $F$ , we have  $F = \bigcup_i (K_i \cap F)$ . Let  $H_n = \bigcup_{i=1}^n K_i \cap F$ . Then  $H_n$  is compact for each  $n$ ,  $H_n \subset F$ , and  $\mu(H_n) \rightarrow \mu(F)$ . Therefore, Property (4) in Theorem 3.18 is valid for *closed* sets. If  $E \in \mathcal{M}$ , the

first part of the proof gives us a closed subset  $F$  of  $E$  such that  $\mu(E - F) < 1$ . If  $\mu(E) = \infty$ , also  $\mu(F) = \infty$ , and therefore

$$\sup_{\{K \in \mathcal{K}; K \subset E\}} \mu(K) \geq \sup_{\{K \in \mathcal{K}; K \subset F\}} \mu(K) = \mu(F) = \infty = \mu(E).$$

Together with (3) and (4)-(ii) in Theorem 3.18, this means that Properties (3) and (4) in 3.18 are valid for all  $E \in \mathcal{M}$ .

Finally, for any  $E \in \mathcal{M}$ , take  $\epsilon = 1/n$  ( $n = 1, 2, \dots$ ) in (1); this gives us closed sets  $F_n$  and open sets  $V_n$  such that

$$F_n \subset E \subset V_n; \quad \mu(V_n - F_n) < 1/n, \quad n = 1, 2, \dots$$

Set  $A = \bigcup F_n$  and  $B = \bigcap V_n$ . Then  $A \in \mathcal{F}_\sigma$ ,  $B \in \mathcal{G}_\delta$ ,  $A \subset E \subset B$ , and since  $B - A \subset V_n - F_n$ , we have  $\mu(B - A) < 1/n$  for all  $n$ , so that  $\mu(B - A) = 0$ .  $\square$

### 3.4 Lusin's theorem

For the measure space of Theorem 3.18, the relation between  $\mathcal{M}$ -measurable functions and continuous functions is described in the following

**Theorem 3.20 (Lusin).** *Let  $X$  be a locally compact Hausdorff space, and let  $(X, \mathcal{M}, \mu)$  be a measure space such that  $\mathcal{B}(X) \subset \mathcal{M}$  and Properties (2), (3), and (4)-(ii) of Theorem 3.18 are satisfied. Let  $A \in \mathcal{M}$ ,  $\mu(A) < \infty$ , and let  $f : X \rightarrow \mathbb{C}$  be measurable and vanish on  $A^c$ . Then, for any  $\epsilon > 0$ , there exists  $g \in C_c(X)$  such that  $\mu([f \neq g]) < \epsilon$ . In case  $f$  is bounded, one may choose  $g$  such that  $\|g\|_u \leq \|f\|_u$ .*

**Proof.** Suppose the theorem proved for *bounded* functions  $f$  (satisfying the hypothesis of the theorem). For an arbitrary  $f$  (as in the theorem), the sets  $E_n := [f] \geq n$ ,  $n = 1, 2, \dots$  form a decreasing sequence of measurable subsets of  $A$ . Since  $\mu(A) < \infty$ , it follows from Lemma 1.11 that  $\lim \mu(E_n) = \mu(\bigcap E_n) = \mu(\emptyset) = 0$ . Therefore, we may choose  $n$  such that  $\mu(E_n) < \epsilon/2$ . The function  $f_n := f I_{E_n^c}$  satisfies the hypothesis of the theorem and is also bounded (by  $n$ ). By our assumption, there exists  $g \in C_c(X)$  such that  $\mu([g \neq f_n]) < \epsilon/2$ . Therefore

$$\begin{aligned} \mu([g \neq f]) &= \mu([g \neq f] \cap E_n) + \mu([g \neq f_n] \cap E_n^c) \\ &\leq \mu(E_n) + \mu([g \neq f_n]) < \epsilon. \end{aligned}$$

Next, we may restrict our attention to *non-negative* functions  $f$  as above. Indeed, in the general case, we may write  $f = \sum_{k=0}^3 i^k u_k$  with  $u_k$  non-negative, measurable, bounded, and vanishing on  $A^c$ . By the special case we assumed, there exist  $g_k \in C_c(X)$  such that  $\mu([g_k \neq u_k]) < \epsilon/4$ . Let  $E = \bigcup_{k=0}^3 [g_k \neq u_k]$  and  $g := \sum_{k=0}^3 i^k g_k$ . Then  $g \in C_c(X)$ , and since  $[g \neq f] \subset E$ , we have indeed  $\mu([g \neq f]) < \epsilon$ .

Let then  $0 \leq f < M$  satisfy the hypothesis of the theorem. Replacing  $f$  by  $f/M$ , we may assume that  $0 \leq f < 1$ .

Since  $\mu(A) < \infty$ , Property (4)-(ii) gives us a *compact* set  $K \subset A$  such that  $\mu(A - K) < \epsilon/2$ . Suppose the theorem is true for  $A$  compact. The function  $f_K := fI_K$  is measurable with range in  $[0, 1)$  and vanishes outside  $K$ . By the theorem for compact  $A$ , there exists  $g \in C_c(X)$  such that  $\mu([g \neq f_K]) < \epsilon/2$ . Then

$$\begin{aligned} \mu([g \neq f]) &= \mu([g \neq f_K] \cap (K \cup A^c)) + \mu([g \neq f] \cap K^c \cap A) \\ &\leq \mu([g \neq f_K]) + \mu(A - K) < \epsilon. \end{aligned}$$

It remains to prove the theorem for  $f$  measurable with range in  $[0, 1)$ , that vanishes on the complement of a *compact* set  $A$ .

By Theorem 1.8, there exist measurable simple functions

$$0 \leq \phi_1 \leq \phi_2 \leq \dots \leq f$$

such that  $f = \lim \phi_n$ . Therefore,  $f = \sum_n \psi_n$ , where  $\psi_1 = \phi_1$ ,  $\psi_n := \phi_n - \phi_{n-1} = 2^{-n}I_{E_n}$  (for  $n > 1$ ), and  $E_n$  are measurable subsets of  $A$  (so that  $\mu(E_n) < \infty$ ). Since  $A$  is a compact subset of the locally compact Hausdorff space  $X$ , there exists an open set  $V$  with compact closure such that  $A \subset V$  (cf. (1) in the proof of Theorem 3.1). By Properties (3) and (4)-(ii) of the measure space (since  $\mu(E_n) < \infty$ ), there exist  $K_n$  compact and  $V_n$  open such that

$$K_n \subset E_n \subset V_n \subset V,$$

and

$$\mu(V_n - K_n) < \epsilon/2^n, \quad n = 1, 2, \dots$$

By Urysohn's lemma (3.1), there exist  $h_n \in C_c(X)$  such that  $0 \leq h_n \leq 1$ ,  $h_n = 1$  on  $K_n$ , and  $h_n = 0$  on  $V_n^c$ . Set

$$g = \sum_n 2^{-n}h_n.$$

The series is majorized by the convergent series of constants  $\sum 2^{-n}$ , hence converges uniformly on  $X$ ; therefore  $g$  is continuous. For all  $n$ ,  $V_n \subset \text{cl}(V)$  and  $g$  vanishes on the set  $\bigcap V_n^c = (\bigcup V_n)^c$ , which contains  $(\text{cl}(V))^c$ ; thus the support of  $g$  is contained in the compact set  $\text{cl}(V)$ , and so  $g \in C_c(X)$ . Since  $2^{-n}h_n = \psi_n$  on  $K_n \cup V_n^c$ , we have

$$[g \neq f] \subset \bigcup_n [2^{-n}h_n \neq \psi_n] \subset \bigcup_n (V_n - K_n),$$

and therefore

$$\mu([g \neq f]) < \sum_n \epsilon/2^n = \epsilon.$$

We show finally how to 'correct'  $g$  so that  $\|g\|_u \leq \|f\|_u$  when  $f$  is a *bounded* function satisfying the hypothesis of the theorem. Suppose  $g \in C_c(X)$  is such that  $\mu([g \neq f]) < \epsilon$ . Let  $E = [|g| \leq \|f\|_u]$ . Define

$$g_1 = gI_E + (g/|g|)\|f\|_u I_{E^c}.$$

Then  $g_1$  is continuous (!),  $\|g_1\|_u \leq \|f\|_u$ , and since  $g_1(x) = 0$  iff  $g(x) = 0$ ,  $g_1$  has compact support. Since  $[g = f] \subset E$ , we have  $[g = f] \subset [g_1 = f]$ , hence  $\mu([g_1 \neq f]) \leq \mu([g \neq f]) < \epsilon$ .  $\square$

**Corollary 3.21.** *Let  $(X, \mathcal{M}, \mu)$  be a measure space as in Theorem 3.20. Then for each  $p \in [1, \infty)$ ,  $C_c(X)$  is dense in  $L^p(\mu)$ .*

In the terminology of Definition 1.28, Corollary 3.21 establishes that  $L^p(\mu)$  is the *completion* of  $C_c(X)$  in the  $\|\cdot\|_p$ -metric.

**Proof.** Since  $\mathcal{B}(X) \subset \mathcal{M}$ , Borel functions are  $\mathcal{M}$ -measurable; in particular, continuous functions are  $\mathcal{M}$ -measurable. If  $f \in C_c(X)$  and  $K := \text{supp } f$ , then  $\int_X |f|^p d\mu \leq \|f\|_u^p \mu(K) < \infty$  by Property (2). Thus  $C_c(X) \subset L^p(\mu)$  for all  $p \in [1, \infty)$ . By Theorem 1.27, it suffices to prove that for each *simple* measurable function  $\phi$  vanishing outside a measurable set  $A$  of finite measure and for each  $\epsilon > 0$ , there exists  $g \in C_c(X)$  such that  $\|\phi - g\|_p < \epsilon$ . By Theorem 3.20 applied to  $\phi$ , there exists  $g \in C_c(X)$  such that  $\mu([\phi \neq g]) < (\epsilon/(2\|\phi\|_u))^p$  and  $\|g\|_u \leq \|\phi\|_u$ . Then

$$\|\phi - g\|_p^p = \int_{[\phi \neq g]} |\phi - g|^p d\mu \leq (2\|\phi\|_u)^p \mu([\phi \neq g]) < \epsilon^p,$$

as wanted.  $\square$

By Lemma 1.30, we obtain

**Corollary 3.22.** *Let  $(X, \mathcal{M}, \mu)$  be a measure space as in Theorem 3.20. Let  $f \in L^p(\mu)$  for some  $p \in [1, \infty)$ . Then there exists a sequence  $\{g_n\} \subset C_c(X)$  that converges to  $f$  almost everywhere.*

In view of the observation following the statement of Corollary 3.21, it is interesting to find the completion of  $C_c(X)$  with respect to the  $\|\cdot\|_u$ -metric. We start with a definition.

**Definition 3.23.** Let  $X$  be a locally compact Hausdorff space. Then  $C_0(X)$  will denote the space of all complex continuous functions  $f$  on  $X$  with the following property:

(\*) for each  $\epsilon > 0$ , there exists a compact subset  $K \subset X$  such that  $|f| < \epsilon$  on  $K^c$ .

A function with Property (\*) is said to *vanish at infinity*.

Under pointwise operations,  $C_0(X)$  is a complex vector space, that contains  $C_c(X)$ . If  $f \in C_0(X)$  and  $K$  is as in (\*) with  $\epsilon = 1$ , then  $\|f\|_u \leq \sup_K |f| + 1 < \infty$ , and it follows that  $C_0(X)$  is a normed space for the uniform norm.

**Theorem 3.24.**  *$C_0(X)$  is the completion of  $C_c(X)$ .*

**Proof.** Let  $\{f_n\} \subset C_0(X)$  be Cauchy. Then  $f := \lim f_n$  exists pointwise uniformly on  $X$ , so that  $f$  is continuous on  $X$  and  $\|f_n - f\|_u \rightarrow 0$ . Given  $\epsilon > 0$ , let  $n_0 \in \mathbb{N}$  be such that  $\|f_n - f\|_u < \epsilon/2$  for all  $n > n_0$ . Fix  $n > n_0$  and a compact

set  $K$  such that  $|f_n| < \epsilon/2$  on  $K^c$  (cf. (\*)). Then  $|f| \leq |f - f_n| + |f_n| < \epsilon$  on  $K^c$ , so that  $f \in C_0(X)$ , and we conclude that  $C_0(X)$  is complete.

Given  $f \in C_0(X)$  and  $\epsilon > 0$ , let  $K$  be as in (\*). By Urysohn's Lemma (3.1), there exists  $h \in C_c(X)$  such that  $0 \leq h \leq 1$  on  $X$  and  $h = 1$  on  $K$ . Then  $hf \in C_c(X)$ ,  $|f - hf| = (1 - h)|f| = 0$  on  $K$ , and  $|f - hf| < |f| < \epsilon$  on  $K^c$ , so that  $\|f - hf\|_u < \epsilon$ . This shows that  $C_c(X)$  is dense in  $C_0(X)$ .  $\square$

**Example 3.25.** Consider the special case  $X = \mathbb{R}^k$ , the  $k$ -dimensional Euclidean space. If  $f \in C_c(\mathbb{R}^k)$  and  $T$  is any closed cell containing  $\text{supp } f$ , let  $\phi(f)$  be the Riemann integral of  $f$  on  $T$ . Then  $\phi$  is a well-defined positive linear functional on  $C_c(\mathbb{R}^k)$ . Let  $(\mathbb{R}^k, \mathcal{M}, m)$  be the associated measure space as in Theorem 3.18. Then, by Theorem 3.18, the integral  $\int_{\mathbb{R}^k} f \, dm$  coincides with the Riemann integral of  $f$  for all  $f \in C_c(\mathbb{R}^k)$ .

For  $n \in \mathbb{N}$  large enough and  $a < b$  real, let  $f_{n,a,b} : \mathbb{R} \rightarrow [0, 1]$  denote the function equal to zero outside  $[a + 1/n, b - 1/n]$ , to 1 in  $[a + 2/n, b - 2/n]$ , and linear elsewhere. Then

$$\int_a^b f_{n,a,b} \, dx = b - a - 3/n.$$

If  $T = \{x \in \mathbb{R}^k; a_i \leq x_i \leq b_i; i = 1, \dots, k\}$ , consider the function  $F_{n,T} = \prod_{i=1}^k f_{n,a_i,b_i} \in C_c(\mathbb{R}^k)$ . Then

$$F_{n,T} \leq I_T \leq F_{n,T_n},$$

where  $T_n = \{x \in \mathbb{R}^k; a_i - 2/n \leq x_i \leq b_i + 2/n\}$ . Therefore, by Fubini's theorem for the Riemann integral on cells,

$$\prod_{i=1}^k (b_i - a_i - 3/n) = \int F_{n,T} \, dx_1 \dots dx_k \leq m(T) \leq \int F_{n,T_n} = \prod_{i=1}^k (b_i - a_i + 1/n).$$

Letting  $n \rightarrow \infty$ , we conclude that

$$m(T) = \prod_{i=1}^k (b_i - a_i) := \text{vol}(T).$$

By Theorem 2.13 and the subsequent constructions of Lebesgue's measure on  $\mathbb{R}$  and of the product measure, the measure  $m$  coincides with Lebesgue's measure on the Borel subsets of  $\mathbb{R}^k$ .

### 3.5 The support of a measure

**Definition 3.26.** Let  $(X, \mathcal{M}, \mu)$  be as in Theorem 3.18. Let  $V$  be the union of all the open  $\mu$ -null sets in  $X$ . The *support* of  $\mu$  is the complement  $V^c$  of  $V$ , and is denoted by  $\text{supp } \mu$ .



Since  $V$  is open,  $\text{supp } \mu$  is a closed subset of  $X$ . Also, by Property (4)-(i) of  $\mu$  (cf. Theorem 3.18),

$$\mu(V) = \sup_{K \in \mathcal{K}; K \subset V} \mu(K). \quad (1)$$

If  $K$  is a compact subset of  $V$ , the open  $\mu$ -null sets are an open cover of  $K$ , and there exist therefore finitely many  $\mu$ -null sets that cover  $K$ ; hence  $\mu(K) = 0$ , and it follows from (1) that  $\mu(V) = 0$ . Thus  $S = \text{supp } \mu$  is *the smallest closed set with a  $\mu$ -null complement*.

For any  $f \in L^1(\mu)$ , we have

$$\int_X f \, d\mu = \int_S f \, d\mu. \quad (2)$$

If  $f \in C_c(X)$  is non-negative and  $\int_X f \, d\mu = 0$ , then  $f = 0$  identically on the support  $S$  of  $\mu$ . Indeed, suppose there exists  $x_0 \in S$  such that  $f(x_0) \neq 0$ . Then there exists an open neighbourhood  $U$  of  $x_0$  such that  $f \neq 0$  on  $U$ . Let  $K$  be any compact subset of  $U$ . Then  $c := \min_K f > 0$ , and

$$0 = \int_X f \, d\mu \geq \int_K f \, d\mu \geq c\mu(K).$$

Hence  $\mu(K) = 0$ , and therefore  $\mu(U) = 0$  by Property (4)-(i) of  $\mu$  (cf. Theorem 3.18). Thus  $U \subset S^c$ , which implies the contradiction  $x_0 \in S^c$ .

Together with (2), this shows that  $\int_X f \, d\mu = 0$  for a non-negative function  $f \in C_c(X)$  if and only if  $f$  vanishes identically on  $\text{supp } \mu$ .

### 3.6 Measures on $\mathbb{R}^k$ ; differentiability

**Notation 3.27.** If  $E \subset \mathbb{R}^k$ , we denote the diameter of  $E$  (i.e.  $\sup_{x,y \in E} d(x,y)$ ) by  $\delta(E)$ . Let  $\mu$  be a real or a positive Borel measure on  $\mathbb{R}^k$ , and let  $m$  denote the Lebesgue measure on  $\mathbb{R}^k$ . Fix  $x \in \mathbb{R}^k$ , and consider the quotients  $\mu(E)/m(E)$  for all open cubes  $E$  containing  $x$ . The *upper derivative* of  $\mu$  at  $x$  is defined by

$$(\bar{D}\mu)(x) = \limsup_{\delta(E) \rightarrow 0} \frac{\mu(E)}{m(E)} := \lim_{r \rightarrow 0} \sup_{\delta(E) < r} \frac{\mu(E)}{m(E)}.$$

The *lower derivative* of  $\mu$  at  $x$ , denoted  $(\underline{D}\mu)(x)$ , is defined similarly by replacing  $\limsup$  and  $\sup$  by  $\liminf$  and  $\inf$ , respectively.

Since  $\sup_{\delta(E) < r} \mu(E)/m(E)$  is an increasing function of  $r$ ,  $(\bar{D}\mu)(x)$  is well defined. The same is true of  $(\underline{D}\mu)(x)$ , and we have trivially  $(\underline{D}\mu)(x) \leq (\bar{D}\mu)(x)$ . In case these quantities are *equal and finite*, one says that  $\mu$  is *differentiable at  $x$* ; the common value is denoted  $(D\mu)(x)$ , and is called *the derivative of  $\mu$  at  $x$* .

If  $f(x) := \sup_{x \in E; \delta(E) < r} \mu(E)/m(E) > c$  for some real  $c$  and some  $r > 0$ , there exists an *open* cube  $E_0$  containing  $x$  with  $\delta(E_0) < r$  such that

$\mu(E_0)/m(E_0) > c$ ; this inequality is true for all  $y \in E_0$ , and therefore, for each  $y \in E_0$ , the above supremum *over all open cubes  $E$  containing  $y$  with  $\delta(E) < r$*  is  $> c$ . This shows that  $[f > c]$  is open, and therefore  $f$  is a Borel function of  $x$ . Consequently  $\bar{D}\mu$  is a Borel function.

If  $\mu_k$  ( $k = 1, 2$ ) are real Borel measures with *finite* upper derivatives at  $x$ , then

$$\bar{D}(\mu_1 + \mu_2) \leq \bar{D}\mu_1 + \bar{D}\mu_2$$

at every point  $x$ ; for  $\underline{D}$ , the inequality is reversed. It follows in particular that if both  $\mu_k$  are differentiable at  $x$ , the same is true of  $\mu := \mu_1 + \mu_2$ , and  $(D\mu)(x) = (D\mu_1)(x) + (D\mu_2)(x)$ .

The concepts of differentiability and derivative are extended to complex measures in the usual way.

The next theorem relates  $D\mu$  to the Radon–Nikodym derivative  $d\mu_a/dm$  of the absolutely continuous part  $\mu_a$  of  $\mu$  in its Lebesgue decomposition with respect to  $m$  (cf. Theorem 1.45).

**Theorem 3.28.** *Let  $\mu$  be a complex Borel measure on  $\mathbb{R}^k$ . Then  $\mu$  is differentiable  $m$ -a.e., and  $D\mu = d\mu_a/dm$  (as elements of  $L^1(\mathbb{R}^k)$ ).*

It follows in particular that  $\mu \perp m$  iff  $D\mu = 0$   $m$ -a.e., and  $\mu \ll m$  iff  $\mu(E) = \int_E (D\mu)dm$  for all  $E \in \mathcal{B} := \mathcal{B}(\mathbb{R}^k)$ .

**Proof.** 1. Consider first a *positive* Borel measure  $\mu$  which is *finite on compact sets*.

Fix  $A \in \mathcal{B}$  and  $c > 0$ , and assume that the Borel set

$$A_c := A \cap [\bar{D}\mu > c] \tag{1}$$

(cf. Section 3.27) has *positive* Lebesgue measure.

Since  $m$  is regular, there exists a compact set  $K \subset A_c$  such that  $m(K) > 0$ . Fix  $r > 0$ . For each  $x \in K$ , there exists an open cube  $E$  with  $\delta(E) < r$  such that  $x \in E$  and  $\mu(E)/m(E) > c$ . By compactness of  $K$ , we may choose finitely many of these cubes, say  $E_1, \dots, E_n$ , such that  $K \subset \bigcup_i E_i$  and  $\delta(E_i) \geq \delta(E_{i+1})$ . We pick a disjoint subfamily of  $E_i$  as follows:  $i_1 = 1$ ;  $i_2$  is the first index  $> i_1$  such that  $E_{i_2}$  does not meet  $E_{i_1}$ ;  $i_3$  is the first index  $> i_2$  such that  $E_{i_3}$  does not meet  $E_{i_1}$  and  $E_{i_2}$ ; etc. . . Let  $V_j$  be the closed ball centred at the centre  $p_j$  of  $E_{i_j}$  with diameter  $3\delta(E_{i_j})$ . If  $\gamma_k$  denotes the ratio of the volumes of a ball and a cube in  $\mathbb{R}^k$  with the same diameter, then  $m(V_j) = \gamma_k 3^k m(E_{i_j})$ .

For each  $i = 1, \dots, n$ , there exists  $i_j \leq i$  such that  $E_i$  meets  $E_{i_j}$ , say at some point  $q$ . Then for all  $y \in E_i$ ,

$$d(y, p_j) \leq d(y, q) + d(q, p_j) \leq \delta(E_i) + \delta(E_{i_j})/2 \leq 3\delta(E_{i_j})/2,$$

since  $i_j \leq i$  implies that  $\delta(E_i) \leq \delta(E_{i_j})$ . Hence  $E_i \subset V_j$ , and

$$K \subset \bigcup_i E_i \subset \bigcup_j V_j.$$

Therefore

$$\begin{aligned} m(K) &\leq \sum_j m(V_j) = \gamma_k 3^k \sum_j m(E_{i_j}) < \gamma_k 3^k c^{-1} \sum_j \mu(E_{i_j}) \\ &= \gamma_k 3^k c^{-1} \mu\left(\bigcup_j E_{i_j}\right). \end{aligned}$$

Each  $E_{i_j}$  is an open cube of diameter  $< r$  containing some point of  $K$ ; therefore

$$\bigcup_j E_{i_j} \subset \{y; d(y, K) < r\} := K_r.$$

The (open) set  $K_r$  has compact closure, and therefore  $\mu(K_r) < \infty$  by hypothesis, and by the preceding calculation

$$m(K) \leq \gamma_k 3^k c^{-1} \mu(K_r). \quad (2)$$

Take  $r = 1/N$  ( $N \in \mathbb{N}$ );  $\{K_{1/N}\}_{N \in \mathbb{N}}$  is a decreasing sequence of open sets of finite  $\mu$ -measure with intersection  $K$ ; therefore  $\mu(K) = \lim_N \mu(K_{1/N})$ , and it follows from (2) that  $m(K) \leq \gamma_k 3^k c^{-1} \mu(K)$ . Hence

$$\mu(A_c) \geq \mu(K) \geq \gamma_k^{-1} 3^{-k} c m(K) > 0.$$

We proved therefore that  $m(A_c) > 0$  implies  $\mu(A_c) > 0$ . Consequently, if  $\mu(A) = 0$  (so that  $\mu(A_c) = 0$  for all  $c > 0$ ), then  $m(A_c) = 0$  for all  $c > 0$ . Since  $A \cap [\bar{D}\mu > 0] = \bigcup_{p=1}^{\infty} A_{1/p}$ , it then follows that  $m(A \cap [\bar{D}\mu > 0]) = 0$ . But  $\bar{D}\mu \geq 0$  since  $\mu$  is a positive measure. Therefore  $\bar{D}\mu = 0$   $m$ -a.e. on  $A$  (for each  $A \in \mathcal{B}$  with  $\mu(A) = 0$ ). Hence  $0 \leq \underline{D}\mu \leq \bar{D}\mu = 0$   $m$ -a.e. on  $A$ , and we conclude that  $D\mu$  exists and equals zero  $m$ -a.e. on  $A$  (if  $\mu(A) = 0$ ).

If  $\mu \perp m$ , there exists  $A \in \mathcal{B}$  such that  $\mu(A) = 0$  and  $m(A^c) = 0$ .

Then  $m([\bar{D}\mu > 0] \cap A) = 0$  and trivially  $m([\bar{D}\mu > 0] \cap A^c) = 0$ . Hence  $m([\bar{D}\mu > 0]) = 0$ , and therefore  $D\mu = 0$   $m$ -a.e.

If  $\mu$  is a complex Borel measure, we use its canonical (Jordan) decomposition  $\mu = \sum_{k=0}^3 i^k \mu_k$ , where  $\mu_k$  are finite positive Borel measures. If  $\mu \perp m$ , also  $\mu_k \perp m$  for all  $k$ , hence  $D\mu_k = 0$   $m$ -a.e. for  $k = 0, \dots, 3$ , and consequently  $D\mu = \sum_{k=0}^3 i^k D\mu_k = 0$   $m$ -a.e.

2. Let  $\mu$  be a *real* Borel measure *absolutely continuous* with respect to  $m$  (restricted to  $\mathcal{B}$ ), and let  $h = d\mu/dm$  be the Radon–Nikodym derivative ( $h$  is real  $m$ -a.e., and since it is only determined  $m$ -a.e., we may assume that  $h$  is a *real* (Borel) function (in  $L^1(\mathbb{R}^k)$ ). We claim that

$$m([h < \bar{D}\mu]) = 0. \quad (3)$$

Assuming the claim and replacing  $\mu$  by  $-\mu$  (so that  $h$  is replaced by  $-h$ ), since  $\bar{D}(-\mu) = -\underline{D}\mu$ , we obtain  $m([h > \underline{D}\mu]) = 0$ . Consequently

$$h \leq \underline{D}\mu \leq \bar{D}\mu \leq h \quad m\text{-a.e.},$$

that is,  $\mu$  is differentiable and  $D\mu = h$   $m$ -a.e. The case of a complex Borel measure  $\mu \ll m$  follows trivially from the real case. Finally, if  $\mu$  is an arbitrary complex Borel measure, we use the Lebesgue decomposition  $\mu = \mu_a + \mu_s$  as in Theorem 1.45. It follows that  $\mu$  is differentiable and  $D\mu = D\mu_a + D\mu_s = d\mu_a/dm$   $m$ -a.e. (cf. Part 1 of the proof), as wanted.

To prove (3) it suffices to show that  $E_r := [h < r < \bar{D}\mu] (= [h < r] \cap [\bar{D}\mu > r])$  is  $m$ -null for any rational number  $r$ , because  $[h < \bar{D}\mu] = \bigcup_{r \in \mathbb{Q}} E_r$ . Fix  $r \in \mathbb{Q}$ , and consider the *positive* Borel measure

$$\lambda(E) := \int_{E \cap [h \geq r]} (h - r) dm \quad (E \in \mathcal{B}). \quad (4)$$

Since  $h \in L^1(m)$ ,  $\lambda$  is finite on compact sets, and  $\lambda([h < r]) = 0$ . By Part 1 of the proof, it follows that  $D\lambda = 0$   $m$ -a.e. on  $[h < r]$ . For any  $E \in \mathcal{B}$ ,

$$\begin{aligned} \mu(E) &= \int_E h dm = \int_E [(h - r) + r] dm = \int_E (h - r) dm + rm(E) \\ &= \int_{E \cap [h > r]} (h - r) dm + \int_{E \cap [h \leq r]} (h - r) dm + rm(E) \leq \lambda(E) + rm(E). \end{aligned}$$

Given  $x \in \mathbb{R}^k$ , we have then for any open cube  $E$  containing  $x$

$$\frac{\mu(E)}{m(E)} \leq \frac{\lambda(E)}{m(E)} + r.$$

Taking the supremum over all such  $E$  with  $\delta(E) < s$  and letting then  $s \rightarrow 0$ , we obtain

$$(\bar{D}\mu)(x) \leq (\bar{D}\lambda)(x) + r = r$$

$m$ -a.e. on  $[h < r]$ . Equivalently,  $m([h < r] \cap [\bar{D}\mu > r]) = 0$ . □

**Corollary 3.29.** *If  $f \in L^1(\mathbb{R}^k)$ , then*

$$\lim_{\delta(E) \rightarrow 0} m(E)^{-1} \int_E |f(y) - f(x)| dy = 0 \quad (5)$$

for almost all  $x \in \mathbb{R}^k$ . (The limit is over open cubes containing  $x$ .)

In particular, the averages of  $f$  over open cubes  $E$  containing  $x$  converge almost everywhere to  $f(x)$  as  $\delta(E) \rightarrow 0$ .

**Proof.** For each  $c \in \mathbb{Q} + i\mathbb{Q}$  and  $N \in \mathbb{N}$ , consider the finite positive Borel measure

$$\mu_N(E) := \int_{E \cap B(0, N)} |f - c| dy \quad (E \in \mathcal{B}),$$

where  $B(0, N) = \{y \in \mathbb{R}^k; |y| < N\}$ . By Theorem 3.28,  $\mu_N(E)/m(E) \rightarrow |f(x) - c| I_{B(0, N)}(x)$   $m$ -a.e. when the open cubes  $E$  containing  $x$  satisfy  $\delta(E) \rightarrow 0$ . Denote the ‘exceptional  $m$ -null set’ by  $G_{c, N}$ , and let

$$G := \bigcup \{G_{c, N}; c \in \mathbb{Q} + i\mathbb{Q}, N \in \mathbb{N}\}.$$

We have  $m(G) = 0$ , and the proof will be completed by showing that (5) is valid for each  $x \notin G$ .

Let  $x \notin G$  and  $\epsilon > 0$ . By the density of  $\mathbb{Q} + i\mathbb{Q}$  in  $\mathbb{C}$ , there exists  $c \in \mathbb{Q} + i\mathbb{Q}$  such that  $|f(x) - c| < \epsilon$ . Choose  $N > |x| + 1$ . All open cubes containing  $x$  with diameter  $< 1$  are contained in  $B(0, N)$ , and therefore  $\mu_N(E)/m(E) \rightarrow |f(x) - c|$  when  $\delta(E) \rightarrow 0$ . Since

$$\begin{aligned} m(E)^{-1} \int_E |f(y) - f(x)| dy &\leq m(E)^{-1} \int_{E(=E \cap B(0, N))} |f(y) - c| dy \\ &\quad + m(E)^{-1} \int_E |f(x) - c| dy \leq \frac{\mu_N(E)}{m(E)} + \epsilon, \end{aligned}$$

it follows that

$$\limsup_{\delta(E) \rightarrow 0} m(E)^{-1} \int_E |f(y) - f(x)| dy \leq |f(x) - c| + \epsilon < 2\epsilon.$$

The arbitrariness of  $\epsilon$  shows that the above limsup is 0 for all  $x \notin G$ , and therefore the limit of the averages exists and equals zero for all  $x \notin G$ .  $\square$

## Exercises

### Translations in $L^p$

1. Let  $L^p$  be the Lebesgue space on  $\mathbb{R}^k$  with respect to Lebesgue measure. For each  $t \in \mathbb{R}^k$ , let

$$[T(t)f](x) = f(x + t) \quad (f \in L^p; x \in \mathbb{R}^k).$$

This so-called ‘translation operator’ is a linear isometry of  $L^p$  onto itself. Prove that  $T(t)f \rightarrow f$  in  $L^p$ -norm as  $t \rightarrow 0$ , for each  $f \in L^p$  ( $1 \leq p < \infty$ ). (Hint: use Corollary 3.21 and an ‘ $\epsilon/3$  argument’.)

### Automatic regularity

2. Let  $X$  be a locally compact Hausdorff space in which every open set is  $\sigma$ -compact (e.g. an Euclidean space). Then every positive Borel measure  $\lambda$  which is finite on compact sets is regular. (Hint: consider the positive linear functional  $\phi(f) := \int_X f d\lambda$ . If  $(X, \mathcal{M}, \mu)$  is the associated measure space as in Theorem 3.18, show that  $\lambda = \mu$  on open sets and use Theorem 3.19.)

### Hardy inequality

3. Let  $1 < p < \infty$ , and let  $L^p(\mathbb{R}^+)$  denote the Lebesgue space for  $\mathbb{R}^+ := (0, \infty)$  with respect to the Lebesgue measure. For  $f \in L^p(\mathbb{R}^+)$ , define

$$(Tf)(x) = (1/x) \int_0^x f(t) dt \quad (x \in \mathbb{R}^+).$$

Prove:

- (a)  $Tf$  is well defined, and  $|(Tf)(x)| \leq x^{-1/p} \|f\|_p$ .  
 (b) Denote by  $D$ ,  $M$ , and  $I$  the differentiation, multiplication by  $x$ , and identity operators, respectively (on appropriate domains). Verify the identities

$$MDT = I - T \quad \text{on } C_c^+(\mathbb{R}^+), \quad (1)$$

where multiplication of operators is their composition.

$$\|Tf\|_p^p = q \int_0^\infty f(Tf)^{p-1} dx \quad (2)$$

for all  $f \in C_c^+(\mathbb{R}^+)$ , where  $q$  is the conjugate exponent of  $p$ . (Hint: integrate by parts.)

- (c)  $\|Tf\|_p \leq q \|f\|_p \quad f \in C_c^+(\mathbb{R}^+)$ .  
 (d) Extend the (Hardy) inequality (c) to all  $f \in L^p(\mathbb{R}^+)$ . (Hint: use Corollary 3.21.)  
 (e) Show that  $\sup_{0 \neq f \in L^p} \|Tf\|_p / \|f\|_p = q$ . (Hint: consider the functions  $f_n(x) = x^{-1/p} I_{[1, n]}(\cdot)$ )

## Absolutely continuous and singular functions

4. Recall that a function  $f : \mathbb{R} \rightarrow \mathbb{C}$  has *bounded variation* if its *total variation function*  $v_f$  is *bounded*, where

$$v_f(x) := \sup_P \sum_k |f(x_k) - f(x_{k-1})| < \infty,$$

and  $P = \{x_k; k = 0, \dots, n\}$ ,  $x_{k-1} < x_k$ ,  $x_n = x$  (the supremum is taken over all such ‘partitions’  $P$  of  $(-\infty, x]$ ).

The *total variation* of  $f$  is  $V(f) := \sup_{\mathbb{R}} v_f$ .

It follows from a theorem of Jordan that such a function has a ‘canonical’ (Jordan) decomposition  $f = \sum_{k=0}^3 i^k f_k$  where  $f_k$  are non-decreasing real function. Therefore  $f$  has one-sided limits at every point. We say that  $f$  is *normalized* if it is left-continuous and  $f(-\infty) = 0$ .

- (a) Let  $\mu$  be a complex Borel measure on  $\mathbb{R}$ . Show that  $f(x) := \mu((-\infty, x))$  is a normalized function of bounded variation (briefly,  $f$  is NBV).  
 (b) Conversely, if  $f$  is NBV and  $\mu$  is the corresponding Lebesgue–Stieltjes measure (constructed through the Jordan decomposition of  $f$  as in Chapter 2, with left continuity replacing right continuity), then  $\mu$  (restricted to  $\mathcal{B} := \mathcal{B}(\mathbb{R})$ ) is a complex Borel measure such that  $f(x) = \mu((-\infty, x))$  for all  $x \in \mathbb{R}$ . (Also  $v_f(x) = |\mu|((-\infty, x))$  and  $V(f) = \|\mu\|$ .)

- (c)  $f : \mathbb{R} \rightarrow \mathbb{C}$  is *absolutely continuous* if for each  $\epsilon > 0$  there exists  $\delta > 0$  such that whenever  $\{(a_k, b_k); k = 1, \dots, n\}$  is a finite family of disjoint intervals of total length  $< \delta$ , we have  $\sum_k |f(b_k) - f(a_k)| < \epsilon$ . If  $f$  is NBV and  $\mu$  is the Borel measure associated to  $f$  as in Part b., then  $\mu \ll m$  iff  $f$  is absolutely continuous (cf. Theorem 3.28 and Exercise 8f in Chapter 1).
- (d) Let  $h \in L^1 := L^1(\mathbb{R})$ ,  $f(x) = \int_{-\infty}^x h(t) dt$ , and  $\mu(E) = \int_E h(t) dt$  ( $E \in \mathcal{B}$ ). Conclude from Parts (a) and (c) that  $f$  is absolutely continuous and  $D\mu = h$   $m$ -a.e. (Cf. Theorem 3.28.)
- (e) Let  $\mu$  and  $f$  be as in Part (a), and let  $x \in \mathbb{R}$  be fixed. Show that  $(D\mu)(x)$  exists iff  $f'(x)$  exists and  $f'(x) = (D\mu)(x)$ . In particular, if  $\mu \perp m$ , then  $f' = 0$   $m$ -a.e. (such a function is called a *singular function*). (Cf. Theorem 3.28.)
- (f) With  $h$  and  $f$  as in Part (d), conclude from Parts (d) and (e) (and Theorem 3.28) that  $f' = h$   $m$ -a.e.
- (g) If  $f$  is NBV, show that  $f'$  exists  $m$ -a.e. and is in  $L^1$ , and  $f(x) = f_s(x) + \int_{-\infty}^x f'(t) dt$  where  $f_s$  is a singular NBV function. (Apply Parts (b), (e), and (f), and the Lebesgue decomposition.)

## Cantor functions

5. Let  $\{r_n\}_{n=0}^\infty$  be a positive decreasing sequence with  $r_0 = 1$ . Denote  $r = \lim_n r_n$ . Let  $C_0 = [0, 1]$ , and for  $n \in \mathbb{N}$ , let  $C_n$  be the union of the  $2^n$  disjoint closed intervals of length  $r_n/2^n$  obtained by removing open intervals at the center of the  $2^{n-1}$  intervals comprising  $C_{n-1}$  (note that the removed intervals have length  $(r_{n-1} - r_n)/2^{n-1} > 0$  and  $m(C_n) = r_n$ ). Let  $C = \bigcap_n C_n$ .
- (a)  $C$  is a compact set of Lebesgue measure  $r$ .
- (b) Let  $g_n = r_n^{-1} I_{C_n}$  and  $f_n(x) = \int_0^x g_n(t) dt$ . Then  $f_n$  is continuous, non-decreasing, constant on each open interval comprising  $C_n^c$ ,  $f_n(0) = 0$ ,  $f_n(1) = 1$ , and  $f_n$  converge *uniformly* in  $[0, 1]$  to some function  $f$ . The function  $f$  is continuous, non-decreasing, has range equal to  $[0, 1]$ , and  $f' = 0$  on  $C^c$ . (In particular, if  $r = 0$ ,  $f' = 0$   $m$ -a.e., but  $f$  is *not* constant. Such so-called *Cantor functions* are examples of continuous non-decreasing non-constant singular functions.)

## Semi-continuity

6. Let  $X$  be a locally compact Hausdorff space. A function  $f : X \rightarrow \overline{\mathbb{R}}$  is *lower semi-continuous* (l.s.c.) if  $[f > c]$  is open for all real  $c$ ;  $f$  is *upper semi-continuous* (u.s.c.) if  $[f < c]$  is open for all real  $c$ . Prove:
- (a)  $f$  is continuous iff it is both l.s.c. and u.s.c.

- (b) If  $f$  is l.s.c. (u.s.c.) and  $\alpha$  is a positive constant, then  $\alpha f$  is l.s.c. (u.s.c., respectively). Also  $-f$  is u.s.c. (l.s.c., respectively).
  - (c) If  $f, g$  are l.s.c. (u.s.c.), then  $f + g$  is l.s.c. (u.s.c., respectively).
  - (d) The supremum (infimum) of any family of l.s.c. (u.s.c.) functions is l.s.c. (u.s.c., respectively).
  - (e) If  $\{f_n\}$  is a sequence of non-negative l.s.c. functions, then  $f := \sum_n f_n$  is l.s.c.
  - (f) The indicator  $I_A$  is l.s.c. (u.s.c.) if  $A \subset X$  is open (closed, respectively).
7. Let  $(X, \mathcal{M}, \mu)$  be a positive measure space as in the Riesz–Markov theorem.
- (a) Let  $0 \leq f \in L^1(\mu)$  and  $\epsilon > 0$ . Represent  $f = \sum_{j=1}^{\infty} c_j I_{E_j}$  as in Exercise 15, Chapter 1, and choose  $K_j$  compact and  $V_j$  open such that  $K_j \subset E_j \subset V_j$  and  $\mu(V_j - K_j) < \epsilon/c_j 2^{j+1}$ . Fix  $n$  such that  $\sum_{j>n} c_j \mu(E_j) < \epsilon/2$  and define  $u = \sum_{j=1}^n c_j I_{K_j}$  and  $v = \sum_{j=1}^{\infty} c_j I_{V_j}$ . Prove that  $u$  is u.s.c.,  $v$  is l.s.c.,  $u \leq f \leq v$ , and  $\int_X (v - u) d\mu < \epsilon$ .
  - (b) Generalize the above conclusion to any *real* function  $f \in L^1(\mu)$ . (This is the *Vitali–Caratheodory theorem*.) (Hint: Exercise 6)

## Fundamental theorem of calculus

8. Let  $f : [a, b] \rightarrow \mathbb{R}$  be differentiable at every point of  $[a, b]$ , and suppose  $f' \in L^1 := L^1([a, b])$  (with respect to Lebesgue measure  $dt$ ). Denote  $\int_a^b f'(t) dt = c$  and fix  $\epsilon > 0$ . By Exercise 7 above, there exists  $v$  l.s.c. such that  $f' \leq v$  and  $\int_a^b v dt < c + \epsilon$ . Fix a constant  $r > 0$  such that  $r(b - a) < c + \epsilon - \int_a^b v dt$ , and let  $g = v + r$ . Observe that  $g$  is l.s.c.,  $g > f'$ , and  $\int_a^b g dt < c + \epsilon$ . By the l.s.c. property of  $g$  and the differentiability of  $f$ , we may associate to each  $x \in [a, b]$  a number  $\delta(x)$  such that  $g(t) > f'(x)$  and  $f(t) - f(x) < (t - x)[f'(x) + \epsilon]$  for all  $t \in (x, x + \delta(x))$ .

Define

$$F(x) = \int_a^x g(t) dt - f(x) + f(a) + \epsilon(x - a).$$

( $F$  is clearly continuous and  $F(a) = 0$ .)

- (a) Show that  $F(t) > F(x)$  for all  $t \in (x, x + \delta(x))$ .
- (b) Conclude that  $F(b) \geq 0$ , and consequently  $f(b) - f(a) < c + \epsilon(1 + b - a)$ . Hence  $f(b) - f(a) \leq c$ .
- (c) Conclude that  $\int_a^b f'(t) dt = f(b) - f(a)$ . (Hint: replace  $f$  by  $-f$  in the conclusion of Part b)



### Approximation almost everywhere by continuous functions

9. Let  $(X, \mathcal{M}, \mu)$  be a positive measure space as in the Riesz–Markov theorem. Let  $f : X \rightarrow \mathbb{C}$  be a bounded measurable function vanishing outside some measurable set of finite measure. Prove that there exists a sequence  $\{g_n\} \subset C_c(X)$  such that  $\|g_n\|_u \leq \|f\|_u$  and  $g_n \rightarrow f$  almost everywhere. (Hint: Lusin and Exercise 16 of Chapter 1.)

# 4

## Continuous linear functionals

The general form of continuous linear functionals on Hilbert space was described in Theorem 1.37. In the present chapter, we shall obtain the general form of continuous linear functionals on some of the normed spaces we have encountered.

### 4.1 Linear maps

We consider first some basic facts about arbitrary linear maps between normed spaces.

**Definition 4.1.** Let  $X, Y$  be normed spaces (over  $\mathbb{C}$ , to fix the ideas), and let  $T : X \rightarrow Y$  be a linear map (it is customary to write  $Tx$  instead of  $T(x)$ , and norms are denoted by  $\|\cdot\|$  in any normed space, unless some distinction is absolutely necessary). One says that  $T$  is *bounded* if

$$\|T\| := \sup_{x \neq 0} \frac{\|Tx\|}{\|x\|} < \infty.$$

Equivalently,  $T$  is bounded iff there exists  $M \geq 0$  such that

$$\|Tx\| \leq M \|x\| \quad (x \in X), \tag{1}$$

and  $\|T\|$  is the smallest constant  $M$  for which (1) is valid. In particular

$$\|Tx\| \leq \|T\| \|x\| \quad (x \in X). \tag{2}$$

The homogeneity of  $T$  shows that the following conditions are equivalent:

- (a)  $T$  is ‘bounded’;
- (b) the map  $T$  is bounded (in the usual sense) on the ‘unit ball’  $B_X := \{x \in X; \|x\| < 1\}$ ;

- (c) the map  $T$  is bounded on the ‘closed unit ball’  $\bar{B}_X := \{x \in X; \|x\| \leq 1\}$ ;
- (d) the map  $T$  is bounded on the ‘unit sphere’  $S_X := \{x \in X; \|x\| = 1\}$ .

In addition, one has (for  $T$  bounded):

$$\|T\| = \sup_{x \in \bar{B}_X} \|Tx\| = \sup_{x \in \bar{B}_X} \|Tx\| = \sup_{x \in S_X} \|Tx\|. \quad (3)$$

By (3), the set  $B(X, Y)$  of all bounded linear maps from  $X$  to  $Y$  is a complex vector space for the pointwise operations, and  $\|\cdot\|$  is a norm on  $B(X, Y)$ , called the *operator norm* or the *uniform norm*.

**Theorem 4.2.** *Let  $X, Y$  be normed spaces, and  $T : X \rightarrow Y$  be linear. Then the following properties are equivalent:*

- (i)  $T \in B(X, Y)$ ;
- (ii)  $T$  is uniformly continuous on  $X$ ;
- (iii)  $T$  is continuous at some point  $x_0 \in X$ .

**Proof.** Assume (i). Then for all  $x, y \in X$ ,

$$\|Tx - Ty\| = \|T(x - y)\| \leq \|T\| \|x - y\|,$$

which clearly implies (ii) (actually, this is the stronger property:  $T$  is *Lipschitz* with *Lipschitz constant*  $\|T\|$ ).

Trivially, (ii) implies (iii). Finally, if (iii) holds, there exists  $\delta > 0$  such that

$$\|Tx - Tx_0\| < 1$$

whenever  $\|x - x_0\| < \delta$ .

By linearity of  $T$ , this is equivalent to:  $\|Tz\| < 1$  whenever  $z \in X$  and  $\|z\| < \delta$ . Since  $\|\delta x\| < \delta$  for all  $x \in B_X$ , it follows that  $\delta\|Tx\| = \|T(\delta x)\| < 1$ , that is,  $\|Tx\| < 1/\delta$  on  $B_X$ , hence  $\|T\| \leq 1/\delta$ .  $\square$

**Notation 4.3.** Let  $X$  be a (complex) normed space. Then

$$B(X) := B(X, X);$$

$$X^* := B(X, \mathbb{C}).$$

Elements of  $B(X)$  will be called bounded *operators* on  $X$ ; elements of  $X^*$  will be called bounded *linear functionals* on  $X$ , and will be denoted usually by  $x^*, y^*, \dots$

Since the norm on  $\mathbb{C}$  is the absolute value, the norm of  $x^* \in X^*$  as defined in Definition 4.1 takes the form

$$\|x^*\| = \sup_{x \neq 0} |x^*x|/\|x\| = \sup_{x \in B_X} |x^*x|;$$

also, (2) takes the form

$$|x^*x| \leq \|x^*\| \|x\| \quad (x \in X).$$

The normed space  $X^*$  is called the (normed) dual or the *conjugate space* of  $X$ .

**Theorem 4.4.** *Let  $X, Y$  be normed spaces. If  $Y$  is complete, then  $B(X, Y)$  is complete.*

**Proof.** Suppose  $Y$  is complete, and let  $\{T_n\} \subset B(X, Y)$  be a Cauchy sequence. For each  $x \in X$ ,

$$\|T_n x - T_m x\| = \|(T_n - T_m)x\| \leq \|T_n - T_m\| \|x\| \rightarrow 0$$

when  $n, m \rightarrow \infty$ , that is,  $\{T_n x\}$  is Cauchy in  $Y$ . Since  $Y$  is complete, the limit  $\lim_n T_n x$  exists in  $Y$ . We denote it by  $Tx$ . By the basic properties of limits, the map  $T : X \rightarrow Y$  is linear. Given  $\epsilon > 0$ , there exists  $n_0 \in \mathbb{N}$  such that

$$\|T_n - T_m\| < \epsilon, \quad n, m > n_0.$$

Therefore

$$\|T_n x - T_m x\| < \epsilon \|x\|, \quad n, m > n_0, \quad x \in X.$$

Letting  $m \rightarrow \infty$ , we get by continuity of the norm

$$\|T_n x - Tx\| \leq \epsilon \|x\|, \quad n > n_0, \quad x \in X.$$

In particular  $T_n - T \in B(X, Y)$ , and thus  $T = T_n - (T_n - T) \in B(X, Y)$ , and  $\|T_n - T\| \leq \epsilon$  for all  $n > n_0$ . This shows that  $B(X, Y)$  is complete.  $\square$

Since  $\mathbb{C}$  is complete, we have

**Corollary 4.5.** *The conjugate space of any normed space is complete.*

## 4.2 The conjugates of Lebesgue spaces

**Theorem 4.6.**

- (i) *Let  $(X, \mathcal{A}, \mu)$  be a positive measure space. Let  $1 < p < \infty$ , let  $q$  be the conjugate exponent, and let  $\phi \in L^p(\mu)^*$ . Then there exists a unique element  $g \in L^q(\mu)$  such that*

$$\phi(f) = \int_X fg \, d\mu \quad (f \in L^p(\mu)). \quad (1)$$

*Moreover, the map  $\phi \rightarrow g$  is an isometric isomorphism of  $L^p(\mu)^*$  and  $L^q(\mu)$ .*

- (ii) *In case  $p = 1$ , the result is valid if the measure space is  $\sigma$ -finite.*

**Proof.** *Uniqueness.* If  $g, g'$  are as in the theorem, and  $h := g - g'$ , then  $h \in L^q(\mu)$  and

$$\int_X fh \, d\mu = 0 \quad (f \in L^p(\mu)). \quad (2)$$

The function  $\theta : \mathbb{C} \rightarrow \mathbb{C}$  defined by

$$\theta(z) = |z|/z \quad \text{for } z \neq 0; \quad \theta(0) = 0$$

is Borel, so that, in case  $1 < p < \infty$ , the function  $f := |h|^{q-1}\theta(h)$  is measurable, and

$$\int_X |f|^p d\mu = \int_X |h|^{(q-1)p} d\mu = \int_X |h|^q d\mu < \infty.$$

Hence by (2) for this function  $f$ ,

$$0 = \int_X |h|^{q-1}\theta(h)h d\mu = \int_X |h|^q d\mu,$$

and consequently  $h$  is the zero element of  $L^q(\mu)$ .

In case  $p = 1$ , take in (2)  $f = I_E$ , where  $E \in \mathcal{A}$  and  $0 < \mu(E) < \infty$  (so that  $f \in L^1(\mu)$ ). Then  $0 = (1/\mu(E)) \int_E h d\mu$  for all such  $E$ , and therefore  $h = 0$  a.e. by the Averages lemma (Lemma 1.38) (since we assume that  $X$  is  $\sigma$ -finite in case  $p = 1$ ).

*Existence.* Let  $(X, \mathcal{A}, \mu)$  be an arbitrary positive measure space, and  $1 \leq p < \infty$ . If  $g \in L^q(\mu)$  and we define  $\psi(f)$  by the right-hand side of (1), then Holder's inequality (Theorem 1.26) implies that  $\psi$  is a well-defined linear functional on  $L^p(\mu)$ , and

$$|\psi(f)| \leq \|g\|_q \|f\|_p \quad (f \in L^p(\mu)),$$

so that  $\psi \in L^p(\mu)^*$  and

$$\|\psi\| \leq \|g\|_q. \quad (3)$$

In order to prove the *existence* of  $g$  as in the theorem, it suffices to prove the following:

**Claim.** *There exists a complex measurable function  $g$  such that*

$$\|g\|_q \leq \|\phi\| \quad (4)$$

and

$$\phi(I_E) = \int_E g d\mu \quad (E \in \mathcal{A}_0), \quad (5)$$

where  $\mathcal{A}_0 := \{E \in \mathcal{A}; \mu(E) < \infty\}$ .

Indeed, Relation (5) means that (1) is valid for  $f = I_E$ , for all  $E \in \mathcal{A}_0$ ; by linearity of  $\phi$  and  $\psi$ , (1) is then valid for all simple functions in  $L^p(\mu)$ . Since these functions are dense in  $L^p(\mu)$  (Theorem 1.27), the conclusion  $\phi = \psi$  follows from the continuity of both functionals on  $L^p(\mu)$ , and the relation  $\|g\|_q = \|\phi\|$  follows then from (3) and (4).

**Proof of the claim.** *Case of a finite measure space  $(X, \mathcal{A}, \mu)$ .* In that case,  $I_E \in L^p(\mu)$  for any  $E \in \mathcal{A}$ , and  $\|I_E\|_p = \mu(E)^{1/p}$ . Consider the trivially additive set function

$$\lambda(E) := \phi(I_E) \quad (E \in \mathcal{A}).$$

If  $\{E_k\} \subset \mathcal{A}$  is a sequence of mutually disjoint sets with union  $E$ , set  $A_n = \bigcup_{k=1}^n E_k$ . Then

$$\|I_{A_n} - I_E\|_p = \|I_{E-A_n}\|_p = \mu(E - A_n)^{1/p} \rightarrow 0$$

as  $n \rightarrow \infty$ , since  $\{E - A_n\}$  is a decreasing sequence of measurable sets with empty intersection (cf. Lemma 1.11). Since  $\phi$  is continuous on  $L^p(\mu)$ , it follows that

$$\begin{aligned}\lambda(E) &:= \phi(I_E) = \lim_n \phi(I_{A_n}) = \lim_n \lambda(A_n) \\ &= \lim_n \sum_{k=1}^n \lambda(E_k) = \sum_{k=1}^{\infty} \lambda(E_k),\end{aligned}$$

so that  $\lambda$  is a complex measure.

If  $\mu(E) = 0$  for some  $E \in \mathcal{A}$ , then  $\|I_E\|_p = 0$ , and therefore  $\lambda(E) := \phi(I_E) = 0$  by linearity of  $\phi$ . This means that  $\lambda \ll \mu$ , and therefore, by the Radon–Nikodym theorem, there exists  $g \in L^1(\mu)$  such that

$$\phi(I_E) = \int_E g \, d\mu = \int_X I_E g \, d\mu \quad (E \in \mathcal{A}).$$

Thus (5) is valid, with  $g$  integrable. We show that this *modified* version of (5) implies (4) (hence the claim).

By linearity of  $\phi$  and the integral, it follows from (5) (modified version) that (1) is valid for all simple measurable functions  $f$ . If  $f$  is a *bounded measurable function*, there exists a sequence of simple measurable functions  $s_n$  such that  $\|s_n - f\|_u \rightarrow 0$  (cf. Theorem 1.8). Then

$$\|s_n - f\|_p \leq \|s_n - f\|_u \mu(X)^{1/p} \rightarrow 0,$$

and therefore, by continuity of  $\phi$ ,

$$\phi(f) = \lim_n \phi(s_n) = \lim_n \int_X s_n g \, d\mu.$$

Also

$$\left| \int_X s_n g \, d\mu - \int_X f g \, d\mu \right| \leq \|s_n - f\|_u \|g\|_1 \rightarrow 0,$$

and we conclude that (1) is valid for all *bounded* measurable functions  $f$ .

*Case  $p = 1$ .* For any  $E \in \mathcal{A}$  with  $\mu(E) > 0$ ,

$$\left| \frac{1}{\mu(E)} \int_E g \, d\mu \right| = \frac{|\phi(I_E)|}{\mu(E)} \leq \|\phi\| \frac{\|I_E\|_1}{\mu(E)} = \|\phi\|.$$

Therefore  $|g| \leq \|\phi\|$  a.e. (by the Averages Lemma), that is,

$$\|g\|_{\infty} \leq \|\phi\|,$$

as desired.

*Case  $1 < p < \infty$ .* Let  $E_n := [|g| \leq n]$  ( $n = 1, 2, \dots$ ). Define  $f_n := I_{E_n} |g|^{q-1} \theta(g)$ , with  $\theta$  as in the beginning of the proof. Then  $f_n$  are *bounded* measurable functions, so that by (1) for such functions,

$$\int_{E_n} |g|^q \, d\mu = \int_X f_n g \, d\mu = \phi(f_n) = |\phi(f_n)| \leq \|\phi\| \|f_n\|_p.$$

However, since  $|f_n|^p = I_{E_n}|g|^{(q-1)p} = I_{E_n}|g|^q$ , it follows that

$$\|f_n\|_p^p = \int_{E_n} |g|^q d\mu.$$

Therefore

$$\left( \int_{E_n} |g|^q d\mu \right)^{1-1/p} \leq \|\phi\|,$$

that is,

$$\left( \int_X I_{E_n} |g|^q d\mu \right)^{1/q} \leq \|\phi\|.$$

Since  $0 \leq I_{E_1}|g|^q \leq I_{E_2}|g|^q \leq \dots$  and  $\lim_n I_{E_n}|g|^q = |g|^q$ , the Monotone Convergence Theorem implies that  $\|g\|_q \leq \|\phi\|$ , as wanted.

*Case of a  $\sigma$ -finite measure space;  $1 \leq p < \infty$ .* We use the function  $w$  and the equivalent finite measure  $d\nu = w d\mu$  (satisfying  $\nu(X) = 1$ ), as defined in the proof of Theorem 1.40. Define

$$V_p : L^p(\nu) \rightarrow L^p(\mu)$$

by

$$V_p f = w^{1/p} f.$$

Then

$$\|V_p f\|_{L^p(\mu)}^p = \int_X |f|^p w d\mu = \int_X |f|^p d\nu = \|f\|_{L^p(\nu)}^p,$$

so that  $V_p$  is a linear isometry of  $L^p(\nu)$  onto  $L^p(\mu)$ . Consequently,  $\phi \circ V_p \in L^p(\nu)^*$ , and  $\|\phi \circ V_p\| = \|\phi\|$  (where the norms are those of the respective dual spaces). Since  $\nu$  is a finite measure, there exists (by the preceding case) a measurable function  $g_1$  such that

$$\|g_1\|_{L^q(\nu)} \leq \|\phi \circ V_p\| = \|\phi\|, \quad (6)$$

and

$$(\phi \circ V_p)(f) = \int_X f g_1 d\nu \quad (f \in L^p(\nu)). \quad (7)$$

Thus, for all  $E \in \mathcal{A}_0$ ,

$$\phi(I_E) = (\phi \circ V_p)(w^{-1/p} I_E) = \int_X w^{-1/p} I_E g_1 d\nu = \int_E w^{1/q} g_1 d\mu. \quad (8)$$

In case  $p > 1$  (so that  $1 < q < \infty$ ), set  $g = w^{1/q} g_1 (= V_q g_1)$ . Then (5) is valid, and by (6),

$$\|g\|_{L^q(\mu)} = \|g_1\|_{L^q(\nu)} \leq \|\phi\|,$$

as desired.

In case  $p = 1$  (so that  $q = \infty$ ), we have by (8)  $\phi(I_E) = \int_E g_1 d\mu$ . Thus (5) is valid with  $g = g_1$ , and since the measures  $\mu$  and  $\nu$  are equivalent, we have by (6)

$$\|g\|_{L^\infty(\mu)} = \|g_1\|_{L^\infty(\nu)} \leq \|\phi\|,$$

as wanted.

*Case of an arbitrary measure space;  $1 < p < \infty$ .* For each  $E \in \mathcal{A}_0$ , consider the finite measure space  $(E, \mathcal{A} \cap E, \mu)$ , and let  $L^p(E)$  be the corresponding  $L^p$ -space. We can identify  $L^p(E)$  (isomorphically and isometrically) with the subspace of  $L^p(\mu)$  of all elements vanishing on  $E^c$ , and therefore the restriction  $\phi_E := \phi|_{L^p(E)}$  belongs to  $L^p(E)^*$  and  $\|\phi_E\| \leq \|\phi\|$ . By the finite measure case, there exists  $g_E \in L^q(E)$  such that

$$\|g_E\|_{L^q(E)} = \|\phi_E\| (\leq \|\phi\|)$$

and

$$\phi_E(f) = \int_E f g_E d\mu \quad \text{for all } f \in L^p(E).$$

If  $E, F \in \mathcal{A}_0$ , then for all measurable subsets  $G$  of  $E \cap F$ ,  $I_G \in L^p(E \cap F) \subset L^p(E)$ , so that  $\phi_E(I_G) = \phi_{E \cap F}(I_G)$ , and therefore

$$\begin{aligned} \int_G (g_E - g_{E \cap F}) d\mu &= \int_E I_G g_E d\mu - \int_{E \cap F} I_G g_{E \cap F} d\mu \\ &= \phi_E(I_G) - \phi_{E \cap F}(I_G) = 0. \end{aligned}$$

By Proposition 1.22 (applied to the *finite* measure space  $(E \cap F, \mathcal{A} \cap (E \cap F), \mu)$ ) and by symmetry,  $g_E = g_{E \cap F} = g_F$  a.e. on  $E \cap F$ . It follows that for any mutually disjoint sets  $E, F \in \mathcal{A}_0$ ,  $g_{E \cup F}$  coincides a.e. with  $g_E$  on  $E$  and with  $g_F$  on  $F$ , and therefore

$$\begin{aligned} \|g_{E \cup F}\|_q^q &= \int_{E \cup F} |g_{E \cup F}|^q d\mu \\ &= \int_E |g_E|^q d\mu + \int_F |g_F|^q d\mu = \|g_E\|_q^q + \|g_F\|_q^q, \end{aligned}$$

that is,  $\|g_E\|_q^q$  is an additive function of  $E$  on  $\mathcal{A}_0$ . Let

$$K := \sup_{E \in \mathcal{A}_0} \|g_E\|_q (\leq \|\phi\|),$$

and let then  $\{E_n\}$  be a non-decreasing sequence in  $\mathcal{A}_0$  such that  $\|\phi_{E_n}\| \rightarrow K$ . Set  $F := \bigcup_n E_n$ .

If  $E \in \mathcal{A}_0$  and  $E \cap F = \emptyset$ , then since  $E$  and  $E_n$  are disjoint for all  $n$ , it follows from the additivity of the set function  $\|g_E\|_q^q$  that

$$\|g_E\|_q^q = \|g_{E \cup E_n}\|_q^q - \|g_{E_n}\|_q^q \leq K^q - \|\phi_{E_n}\|_q^q \rightarrow 0.$$



Hence  $\|g_E\|_q = 0$  for all  $E \in \mathcal{A}_0$  disjoint from  $F$ , that is,  $g_E = 0$  a.e. for such  $E$ . Consequently, for  $E \in \mathcal{A}_0$  arbitrary, we have a.e. on  $E - F$   $g_E = g_{E-F} = 0$ , and therefore

$$\phi(I_E) = \phi_E(I_E) = \int_E g_E d\mu = \int_{E \cap F} g_E d\mu = \int_{E \cap F} g_{E \cap F} d\mu. \quad (9)$$

Since  $g_{E_n} = g_{E_{n+1}}$  a.e. on  $E_n$ , the limit  $g := \lim_n g_{E_n}$  exists a.e. and vanishes on  $F^c$ ; it is measurable, and by the Monotone Convergence Theorem,

$$\|g\|_{L^q(\mu)} = \lim_n \|g_{E_n}\|_{L^q(\mu)} = \lim_n \|\phi_{E_n}\| = K \leq \|\phi\|,$$

and (4) is verified. Fix  $n$ . For all  $k \geq n$ ,  $g_{E \cap F} = g_{E_k}$  a.e. on  $(E \cap F) \cap E_k = E \cap E_k$ , hence (a.e.) on  $E \cap E_n$ . Therefore  $g_{E \cap F} = g$  a.e. on  $E \cap E_n$  for all  $n$ , hence (a.e.) on  $E \cap F$ , and consequently (5) follows from (9). This completes the proof of the claim.  $\square$

### 4.3 The conjugate of $C_c(X)$

Let  $X$  be a locally compact Hausdorff space, and consider the normed space  $C_c(X)$  with the uniform norm

$$\|f\| = \|f\|_u := \sup_X |f| \quad (f \in C_c(X)).$$

If  $\mu$  is a complex Borel measure on  $X$ , write  $d\mu = h d|\mu|$ , where  $|\mu|$  is the total variation measure corresponding to  $\mu$  and  $h$  is a uniquely determined Borel function with  $|h| = 1$  (cf. Theorem 1.46). Set

$$\psi(f) := \int_X f d\mu := \int_X f h d|\mu| \quad (f \in C_c(X)). \quad (1)$$

Then

$$|\psi(f)| \leq \int_X |f| d|\mu| \leq |\mu|(X) \|f\|, \quad (2)$$

so that  $\psi$  is a well-defined, clearly linear, continuous functional on  $C_c(X)$ , with norm

$$\|\psi\| \leq \|\mu\| := |\mu|(X). \quad (3)$$

We shall prove that *every* continuous linear functional  $\phi$  on  $C_c(X)$  is of this form for a uniquely determined *regular* complex Borel measure  $\mu$ , and  $\|\phi\| = \|\mu\|$ . This will be done by using Riesz–Markov Representation Theorem 3.18 for *positive* linear functionals on  $C_c(X)$ . Our first step will be to associate a positive linear functional  $|\phi|$  to each given  $\phi \in C_c(X)^*$ .

**Definition 4.7.** Let  $\phi \in C_c(X)^*$ . The *total variation functional*  $|\phi|$  is defined by

$$|\phi|(f) := \sup\{|\phi(h)|; h \in C_c(X), |h| \leq f\} \quad (0 \leq f \in C_c(X));$$

$$|\phi|(u + iv) = |\phi|(u^+) - |\phi|(u^-) + i|\phi|(v^+) - i|\phi|(v^-) \quad (u, v \in C_c^R(X)).$$

**Theorem 4.8.** *The total variation functional  $|\phi|$  of  $\phi \in C_c(X)^*$  is a positive linear functional on  $C_c(X)$ , and it satisfies the inequality*

$$|\phi(f)| \leq |\phi|(|f|) \leq \|\phi\| \|f\| \quad (f \in C_c(X)).$$

**Proof.** Let  $C_c^+(X) := \{f \in C_c(X); f \geq 0\}$ . It is clear from Definition 4.7 that

$$0 \leq |\phi|(f) \leq \|\phi\| \|f\| < \infty, \quad (4)$$

$|\phi|$  is monotonic on  $C_c^+(X)$  and  $|\phi|(cf) = c|\phi|(f)$  (and in particular  $|\phi|(0) = 0$ ) for all  $c \geq 0$  and  $f \in C_c^+(X)$ . We show that  $|\phi|$  is additive on  $C_c^+(X)$ .

Let  $\epsilon > 0$  and  $f_k \in C_c^+(X)$  be given ( $k = 1, 2$ ). By definition, there exist  $h_k \in C_c(X)$  such that  $|h_k| \leq f_k$  and  $|\phi|(f_k) \leq |\phi|(h_k)| + \epsilon/2$ ,  $k = 1, 2$ . Therefore, writing the complex numbers  $\phi(h_k)$  in polar form, we obtain

$$\begin{aligned} 0 &\leq |\phi|(f_1) + |\phi|(f_2) \leq |\phi|(h_1)| + |\phi|(h_2)| + \epsilon \\ &= e^{-i\theta_1} \phi(h_1) + e^{-i\theta_2} \phi(h_2) + \epsilon = \phi(e^{-i\theta_1} h_1 + e^{-i\theta_2} h_2) + \epsilon \\ &\leq |\phi|(f_1 + f_2) + \epsilon, \end{aligned}$$

because

$$|e^{-i\theta_1} h_1 + e^{-i\theta_2} h_2| \leq |h_1| + |h_2| \leq f_1 + f_2.$$

Hence  $|\phi|$  is ‘super-additive’ on  $C_c^+(X)$ .

Next, let  $h \in C_c(X)$  satisfy  $|h| \leq f_1 + f_2 := f$ . Let  $V = [f > 0]$ . Define for  $k = 1, 2$

$$h_k = (f_k/f)h \quad \text{on } V; \quad h_k = 0 \quad \text{on } V^c.$$

The functions  $h_k$  are continuous on  $V$  and  $V^c$ . If  $x$  is a boundary point of  $V$ , then  $x \notin V$  (since  $V$  is open), so that  $f(x) = 0$  and  $h_k(x) = 0$ . Let  $\{x_\alpha\} \subset V$  be a net converging to  $x$ . Then by continuity of  $h$ , we have for  $k = 1, 2$ :

$$|h_k(x_\alpha)| \leq |h(x_\alpha)| \rightarrow |h(x)| \leq |f(x)| = 0,$$

so that  $\lim_\alpha h_k(x_\alpha) = 0 = h_k(x)$ . This shows that  $h_k$  are continuous on  $X$ . Trivially,  $\text{supp } h_k \subset \text{supp } f_k$ , so that  $h_k \in C_c(X)$ , and by definition,  $|h_k| \leq f_k$  and  $h = h_1 + h_2$ . Therefore

$$|\phi(h)| = |\phi(h_1) + \phi(h_2)| \leq |\phi|(f_1) + |\phi|(f_2).$$

Taking the supremum over all  $h \in C_c(X)$  such that  $|h| \leq f$ , we obtain that  $|\phi|$  is subadditive. Together with the super-additivity obtained before, this proves that  $|\phi|$  is additive.

Next, consider  $|\phi|$  over  $C_c^R(X)$ . The homogeneity over  $\mathbb{R}$  is easily verified. Additivity is proved as in Theorem 1.19. Let  $f = f^+ - f^-$  and  $g = g^+ - g^-$  be functions in  $C_c^R(X)$ , and let  $h = h^+ - h^- := f + g = f^+ - f^- + g^+ - g^-$ . Then  $h^+ + f^- + g^- = f^+ + g^+ + h^-$ , so that by the additivity of  $|\phi|$  on  $C_c^+(X)$ , we obtain

$$|\phi|(h^+) + |\phi|(f^-) + |\phi|(g^-) = |\phi|(f^+) + |\phi|(g^+) + |\phi|(h^-),$$

and since all summands are finite, it follows that

$$|\phi|(h) := |\phi|(h^+) - |\phi|(h^-) = |\phi|(f^+) - |\phi|(f^-) + |\phi|(g^+) - |\phi|(g^-) := |\phi|(f) + |\phi|(g).$$

The linearity of  $|\phi|$  over  $C_c(X)$  now follows easily from the definition. Thus  $|\phi|$  is a positive linear functional on  $C_c(X)$  (cf. (4)). By (4) for the function  $|f|$ ,  $|\phi|(|f|) \leq \|\phi\| \|f\|$ . Also, since  $h = f$  belongs to the set of functions used in the definition of  $|\phi|(|f|)$ , we have  $|\phi(f)| \leq |\phi|(|f|)$ .  $\square$

## 4.4 The Riesz representation theorem

**Theorem 4.9.** *Let  $X$  be a locally compact Hausdorff space, and let  $\phi \in C_c(X)^*$ . Then there exists a unique regular complex Borel measure  $\mu$  on  $X$  such that*

$$\phi(f) = \int_X f \, d\mu \quad (f \in C_c(X)). \quad (1)$$

Furthermore,

$$\|\phi\| = \|\mu\|. \quad (2)$$

‘Regularity’ of the complex measure  $\mu$  means *by definition* that its total variation measure  $|\mu|$  is regular.

**Proof.** We apply Theorem 3.18 to the positive linear functional  $|\phi|$ . Denote by  $\lambda$  the positive Borel measure obtained by restricting the measure associated with  $|\phi|$  (by Theorem 3.18) to the Borel algebra  $\mathcal{B}(X) \subset \mathcal{M}$ . Then

$$|\phi|(f) = \int_X f \, d\lambda \quad (f \in C_c(X)). \quad (3)$$

By Definition 3.5 and Theorem 4.8,

$$\lambda(X) = \sup\{|\phi|(f); 0 \leq f \leq 1\} \leq \|\phi\| \quad (4)$$

In particular, every Borel set in  $X$  has finite  $\lambda$ -measure, and therefore, by Theorem 3.18 (cf. (3) and (4)(ii)),  $\lambda$  is *regular*.

By Theorem 4.8 and (3), for all  $f \in C_c(X)$ ,

$$|\phi(f)| \leq |\phi|(|f|) = \int_X |f| \, d\lambda = \|f\|_{L^1(\lambda)}.$$

This shows that  $\phi$  is a continuous linear functional on the subspace  $C_c(X)$  of  $L^1(\lambda)$ , with norm  $\leq 1$ . By Theorem 3.21,  $C_c(X)$  is dense in  $L^1(\lambda)$ , and it follows that  $\phi$  has a unique extension as an element of  $L^1(\lambda)^*$  with norm  $\leq 1$ . By Theorem 4.6, there exists a unique element  $g \in L^\infty(\lambda)$  such that

$$\phi(f) = \int_X fg \, d\lambda \quad (f \in C_c(X)) \quad (5)$$

and  $\|g\|_\infty \leq 1$ .

Define  $d\mu = g d\lambda$ . Then  $\mu$  is a complex Borel measure satisfying (1). By Theorem 1.47 and (4),

$$|\mu|(X) = \int_X |g| d\lambda \leq \lambda(X) \leq \|\phi\|.$$

By (3) of Section 4.3, the reversed inequality is a consequence of (1), so that (2) follows.

Gathering some of the above inequalities, we have

$$\|\phi\| = |\mu|(X) = \int_X |g| d\lambda \leq \lambda(X) \leq \|\phi\|.$$

Thus  $\lambda(X) = \int_X |g| d\lambda$ , that is,  $\int_X (1 - |g|) d\lambda = 0$ . Since  $1 - |g| \geq 0$   $\lambda$ -a.e., it follows that  $|g| = 1$  a.e., and since  $g$  is only a.e.-determined, we may choose  $g$  such that  $|g| = 1$  identically on  $X$ .

For all Borel sets  $E$ ,  $|\mu|(E) = \int_E |g| d\lambda = \lambda(E)$ , which proves that  $|\mu| = \lambda$ . In particular,  $\mu$  is regular.

In order to prove uniqueness, we observe that the sum  $\nu$  of two finite positive regular Borel measures  $\nu_k$  is regular. Indeed, given  $\epsilon > 0$  and  $E \in \mathcal{B}(X)$ , there exist  $K_k$  compact and  $V_k$  open such that  $K_k \subset E \subset V_k$  and

$$\nu_k(V_k) - \epsilon/2 \leq \nu_k(E) \leq \nu_k(K_k) + \epsilon/2.$$

Then  $K := K_1 \cup K_2 \subset E \subset V := V_1 \cap V_2$ ,  $K$  is compact,  $V$  is open, and by monotonicity of positive measures,

$$\nu(V) - \epsilon \leq \nu_1(V_1) + \nu_2(V_2) - \epsilon \leq \nu(E) \leq \nu_1(K_1) + \nu_2(K_2) + \epsilon \leq \nu(K) + \epsilon.$$

Suppose now that the representation (1) is valid for the regular complex measures  $\mu_1$  and  $\mu_2$ . Then  $\int_X f d\mu = 0$  for all  $f \in C_c(X)$ , for  $\mu = \mu_1 - \mu_2$ . We must show that  $\|\mu\| = 0$  (i.e.  $\mu = 0$ ). Since  $|\mu_k|$  are finite positive regular Borel measures, the positive Borel measure  $\nu := |\mu_1| + |\mu_2|$  is regular. Write  $d\mu = h d|\mu|$ , where  $h$  is a Borel function with  $|h| = 1$  (cf. Theorem 1.46). Since  $\nu$  is regular, it follows from Theorem 3.21 that there exists a sequence  $\{f_n\} \subset C_c(X)$  that converges to  $\bar{h}$  in the  $L^1(\nu)$ -metric. Since  $\bar{h}h = 1$ ,  $|\mu| = |\mu_1 - \mu_2| \leq |\mu_1| + |\mu_2| := \nu$ , and  $\int_X f_n h d|\mu| = \int_X f_n d\mu = 0$ , we obtain

$$\begin{aligned} \|\mu\| &:= |\mu|(X) = \left| \int_X f_n h d|\mu| - \int_X \bar{h} h d|\mu| \right| = \left| \int_X (f_n - \bar{h}) h d|\mu| \right| \\ &\leq \int_X |f_n - \bar{h}| d|\mu| \leq \int_X |f_n - \bar{h}| d\nu = \|f_n - \bar{h}\|_{L^1(\nu)} \rightarrow 0 \end{aligned}$$

as  $n \rightarrow \infty$ . Hence  $\|\mu\| = 0$ . □

**Remark 4.10.** If  $S = \text{supp}|\mu|$  (cf. Definition 3.26), we have

$$\|\phi\| = \|\mu\| := |\mu|(X) = |\mu|(S)$$

and

$$\int_X f d\mu = \int_S f d\mu \quad (f \in L^1(|\mu|)).$$

The second formula follows from Theorem 1.46 and Definition 3.26(2). Indeed, write  $d\mu = h d|\mu|$  where  $h$  is a Borel measurable function with  $|h| = 1$  on  $X$  (cf. Theorem 1.46). Then for all  $f \in L^1(|\mu|)$ , we have (cf. Definition 3.26(2))

$$\int_X f d\mu := \int_X fh d|\mu| = \int_S fh d|\mu| := \int_S f d\mu.$$

## 4.5 Haar measure

As an application of the Riesz–Markov Representation Theorem for positive linear functionals (Theorem 3.18), we shall construct a (left) translation-invariant positive measure on any locally compact topological group.

A *topological group* is a group  $G$  with a Hausdorff topology for which the group operations (multiplication and inverse) are continuous. It follows that for each fixed  $a \in G$ , the *left (right) translation*  $x \rightarrow ax$  ( $x \rightarrow xa$ ) is a homeomorphism of  $G$  onto itself. For any open neighbourhood  $V$  of the identity  $e$ , the set  $aV$  ( $Va$ ) is an open neighbourhood of  $a$ .

Suppose  $G$  is locally compact, and  $f, g \in C_c^+ := C_c^+(G) := \{f \in C_c(G); f \geq 0, f \text{ not identically zero}\}$ . Fix  $0 < \alpha < \|g\| := \|g\|_u$ . There exists  $a \in G$  such that  $g(a) > \alpha$ , and therefore there exists an open neighbourhood of  $e$ ,  $V$ , such that  $g(x) \geq \alpha$  for all  $x \in aV$ . By compactness of  $\text{supp } f$ , there exist  $x_1, \dots, x_n \in G$  such that  $\text{supp } f \subset \bigcup_{k=1}^n x_k V$ . Set  $s_k := ax_k^{-1}$ . Then for  $x \in x_k V$ ,  $s_k x \in aV$ , and therefore  $g(s_k x) \geq \alpha$ . If  $x \in \text{supp } f$ , there exists  $k \in \{1, \dots, n\}$  such that  $x \in x_k V$ , so that (for this  $k$ )

$$f(x) \leq \|f\| \leq \frac{\|f\|}{\alpha} g(s_k x) \leq \sum_{i=1}^n c_i g(s_i x), \quad (1)$$

where  $c_i = \|f\|/\alpha$  for all  $i = 1, \dots, n$ . Since (1) is trivial on  $(\text{supp } f)^c$ , we see that there exist  $n \in \mathbb{N}$  and  $(c_1, \dots, c_n, s_1, \dots, s_n) \in (\mathbb{R}^+)^n \times G^n$  such that

$$f(x) \leq \sum_{i=1}^n c_i g(s_i x) \quad (x \in G). \quad (2)$$

Denote by  $\Omega(f : g)$  the non-empty set of such rows (with  $n$  varying) and let

$$(f : g) = \inf \sum_{i=1}^n c_i, \quad (*)$$

where the infimum is taken over all  $(c_1, \dots, c_n)$  such that  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f : g)$  for some  $n$  and  $s_i$ .

We verify some elementary properties of the functional  $(f : g)$  for  $g$  fixed as above.

Let  $f_s(x) := f(sx)$  for  $s \in G$  fixed ( $f_s$  is the so-called *left  $s$ -translate of  $f$* ). If  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f : g)$ , then  $f_s(x) = f(sx) \leq \sum_i c_i g(s_i sx)$  for all  $x \in G$ , hence  $(c_1, \dots, c_n, s_1 s, \dots, s_n s) \in \Omega(f_s : g)$ , and consequently  $(f_s : g) \leq \sum_{i=1}^n c_i$ . Taking the infimum over all such rows, we get  $(f_s : g) \leq (f : g)$ . But then  $(f : g) = ((f_s)_{s^{-1}} : g) \leq (f_s : g)$ , and we conclude that

$$(f_s : g) = (f : g) \quad (3)$$

for all  $s \in G$  (i.e. the functional  $(\cdot : g)$  is *left translation invariant*).

In the following arguments,  $\epsilon$  denotes an arbitrary positive number.

If  $c > 0$  and  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f : g)$  is such that  $\sum c_i < (f : g) + \epsilon$ , then  $(cf)(x) \leq \sum_i c c_i g(s_i x)$  for all  $x \in G$ , and therefore  $(cf : g) \leq \sum_i c c_i < c(f : g) + c\epsilon$ . The arbitrariness of  $\epsilon$  implies that  $(cf : g) \leq c(f : g)$ . Applying this inequality to the function  $cf$  and the constant  $1/c$  (instead of  $f$  and  $c$ , respectively), we obtain the reversed inequality. Hence

$$(cf : g) = c(f : g) \quad (c > 0). \quad (4)$$

Let  $x_0 \in G$  be such that  $f(x_0) = \max_G f$  (since  $f$  is continuous with compact support, such a point  $x_0$  exists). Then for any  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f : g)$ ,

$$\|f\| = f(x_0) \leq \sum_i c_i g(s_i x_0) \leq \|g\| \sum_i c_i.$$

Hence

$$\frac{\|f\|}{\|g\|} \leq (f : g). \quad (5)$$

Next, consider three functions  $f_1, f_2, g \in C_c^+$ . If  $f_1 \leq f_2$ , one has trivially  $\Omega(f_2 : g) \subset \Omega(f_1 : g)$ , and therefore

$$f_1 \leq f_2 \quad \text{implies} \quad (f_1 : g) \leq (f_2 : g). \quad (6)$$

There exist  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f_1 : g)$  and  $(d_1, \dots, d_m, t_1, \dots, t_m) \in \Omega(f_2 : g)$  such that  $\sum c_i < (f_1 : g) + \epsilon/2$  and  $\sum d_j < (f_2 : g) + \epsilon/2$ . Then for all  $x \in G$ ,

$$f_1(x) + f_2(x) \leq \sum_i c_i g(s_i x) + \sum_j d_j g(t_j x) = \sum_{k=1}^{n+m} c'_k g(s'_k x),$$

where  $c'_k = c_k, s'_k = s_k$  for  $k = 1, \dots, n$ , and  $c'_k = d_{k-n}, s'_k = t_{k-n}$  for  $k = n+1, \dots, n+m$ . Thus  $(c'_1, \dots, c'_{n+m}, s'_1, \dots, s'_{n+m}) \in \Omega(f_1 + f_2 : g)$ , and therefore

$$(f_1 + f_2 : g) \leq \sum_{k=1}^{n+m} c'_k = \sum_{k=1}^n c_k + \sum_{j=1}^m d_j < (f_1 : g) + (f_2 : g) + \epsilon.$$

This proves that

$$(f_1 + f_2 : g) \leq (f_1 : g) + (f_2 : g). \quad (7)$$

Let  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(f : g)$  and  $(d_1, \dots, d_m, t_1, \dots, t_m) \in \Omega(g : h)$ , where  $f, g, h \in C_c^+$ . Then for all  $x \in G$ ,

$$f(x) \leq \sum_i c_i g(s_i x) \leq \sum_i c_i \sum_j d_j h(t_j s_i x) = \sum_{i,j} c_i d_j h(t_j s_i x),$$

that is,  $(c_i d_j, t_j s_i)_{i=1, \dots, n; j=1, \dots, m} \in \Omega(f : h)$ , and consequently

$$(f : h) \leq \sum_{i,j} c_i d_j = \left( \sum_i c_i \right) \left( \sum_j d_j \right).$$

Taking the infimum of the right-hand side over all the rows involved, we conclude that

$$(f : h) \leq (f : g)(g : h). \quad (8)$$

With  $g$  fixed, denote

$$\Lambda_h f := \frac{(f : h)}{(g : h)}. \quad (9)$$

Since  $\Lambda_h$  is a constant multiple of  $(f : h)$ , the function  $f \rightarrow \Lambda_h f$  satisfies (3), (4), (6), and (7).

By (8),  $\Lambda_h f \leq (f : g)$ . Also

$$(g : f) \Lambda_h f = (g : f) \frac{(f : h)}{(g : h)} \geq \frac{(g : h)}{(g : h)} = 1.$$

Hence

$$\frac{1}{(g : f)} \leq \Lambda_h f \leq (f : g). \quad (10)$$

By (10),  $\Lambda_h$  is a point in the compact Hausdorff space

$$\Delta := \prod_{f \in C_c^+} \left[ \frac{1}{(g : f)}, (f : g) \right]$$

(cf. Tychonoff's theorem). Consider the system  $\mathcal{V}$  of all (open) neighbourhoods of the identity. For each  $V \in \mathcal{V}$ , let  $\Sigma_V$  be the closure in  $\Delta$  of the set  $\{\Lambda_h; h \in C_V^+\}$ , where  $C_V^+ := C_V^+(G)$  consists of all  $h \in C_c^+$  with support in  $V$ . Then  $\Sigma_V$  is a non-empty compact subset of  $\Delta$ . If  $V_1, \dots, V_n \in \mathcal{V}$  and  $V := \bigcap_{i=1}^n V_i$ , then  $C_V^+ \subset \bigcap_i C_{V_i}^+$ , and therefore  $\Sigma_V \subset \bigcap_i \Sigma_{V_i}$ . In particular, the family of compact sets  $\{\Sigma_V; V \in \mathcal{V}\}$  has the finite intersection property, and consequently

$$\bigcap_{V \in \mathcal{V}} \Sigma_V \neq \emptyset.$$

Let  $\Lambda$  be any point in this intersection, and extend the functional  $\Lambda$  to  $C_c := C_c(G)$  in the obvious way ( $\Lambda 0 = 0$ ;  $\Lambda f = \Lambda f^+ - \Lambda f^-$  for real  $f \in C_c$ , and  $\Lambda(u + iv) = \Lambda u + i\Lambda v$  for real  $u, v \in C_c$ ).

**Theorem 4.11.**  $\Lambda$  is a non-zero left translation invariant positive linear functional on  $C_c$ .

**Proof.** Since  $\Lambda \in \Delta$ , we have

$$\Lambda f \in \left[ \frac{1}{(g : f)}, (f : g) \right]$$

for all  $f \in C_c^+$ , so that in particular  $\Lambda f > 0$  for such  $f$ , and  $\Lambda$  is not identically zero.

For any  $V \in \mathcal{V}$ , we have  $\Lambda \in \Sigma_V$ ; hence every basic neighbourhood  $N$  of  $\Lambda$  in  $\Delta$  meets the set  $\{\Lambda_h; h \in C_V^+\}$ . Recall that

$$N = N(\Lambda; f_1, \dots, f_n; \epsilon) := \{\Phi \in \Delta; |\Phi f_i - \Lambda f_i| < \epsilon; i = 1, \dots, n\},$$

where  $f_i \in C_c^+$ . Thus, for any  $V \in \mathcal{V}$  and  $f_1, \dots, f_n \in C_c^+$ , there exists  $h \in C_V^+$  such that

$$|\Lambda_h f_i - \Lambda f_i| < \epsilon \quad (i = 1, \dots, n). \quad (11)$$

Given  $f \in C_c^+$  and  $c > 0$ , apply (11) with  $f_1 = f$  and  $f_2 = cf$ . By Property (4) for  $\Lambda_h$ , we have

$$|\Lambda(cf) - c\Lambda f| \leq |\Lambda(cf) - \Lambda_h(cf)| + c|\Lambda_h f - \Lambda f| < (1 + c)\epsilon,$$

so that  $\Lambda(cf) = c\Lambda f$  by the arbitrariness of  $\epsilon$ .

A similar argument (using Relation (3) for  $\Lambda_h$ ) shows that  $\Lambda f_s = \Lambda f$  for all  $f \in C_c^+$  and  $s \in G$ .

In order to prove the additivity of  $\Lambda$  on  $C_c^+$ , we use the following:

**Lemma.** Let  $f_1, f_2 \in C_c^+(G)$  and  $\epsilon > 0$ . Then there exists  $V \in \mathcal{V}$  such that

$$\Lambda_h f_1 + \Lambda_h f_2 \leq \Lambda_h(f_1 + f_2) + \epsilon$$

for all  $h \in C_V^+(G)$ .

**Proof of lemma.** Let  $f = f_1 + f_2$ , and fix  $k \in C_c^+(G)$  such that  $k = 1$  on  $\{x \in G; f(x) > 0\}$ . For  $g$  fixed as above, let

$$\delta := \frac{\epsilon}{4(k : g)}; \quad \eta := \min \left\{ \frac{\epsilon}{4(f : g)}, 1/2 \right\}.$$

Thus

$$2\eta(f : g) \leq \epsilon/2; \quad 2\eta \leq 1; \quad 2\delta(k : g) \leq \epsilon/2. \quad (12)$$

For  $i = 1, 2$ , let  $h_i := f_i/F$ , where  $F := f + \delta k$  ( $h_i = 0$  at points where  $F = 0$ ). The functions  $h_i$  are well defined, and continuous with compact support; it follows that there exists  $V \in \mathcal{V}$  such that

$$|h_i(x) - h_i(y)| < \eta \quad (i = 1, 2)$$

for all  $x, y \in G$  such that  $y^{-1}x \in V$  (uniform continuity of  $h_i$ !).



Let  $h \in C_V^+(G)$ . Let  $(c_1, \dots, c_n, s_1, \dots, s_n) \in \Omega(F : h)$  and  $x \in G$ . If  $j \in \{1, \dots, n\}$  is such that  $h(s_j x) \neq 0$ , then  $s_j x \in V$ , and therefore  $|h_i(x) - h_i(s_j^{-1})| < \eta$  for  $i = 1, 2$ . Hence

$$h_i(x) \leq |h_i(x) - h_i(s_j^{-1})| + h_i(s_j^{-1}) < h_i(s_j^{-1}) + \eta.$$

Therefore, for  $i = 1, 2$ ,

$$\begin{aligned} f_i(x) = F(x)h_i(x) &\leq \sum_{\{j; h(s_j x) \neq 0\}} c_j h(s_j x) h_i(x) \\ &\leq \sum_{\{j; h(s_j x) \neq 0\}} c_j [h_i(s_j^{-1}) + \eta] h(s_j x) \leq \sum_{j=1}^n c_j^i h(s_j x), \end{aligned}$$

where  $c_j^i := c_j [h_i(s_j^{-1}) + \eta]$ . Hence  $(f_i : h) \leq \sum_j c_j^i$ , and since  $h_1 + h_2 = f/F \leq 1$ , we obtain

$$(f_1 : h) + (f_2 : h) \leq \sum_j c_j (1 + 2\eta).$$

Taking the infimum of the right-hand side over all rows in  $\Omega(F : h)$ , we conclude that

$$\begin{aligned} (f_1 : h) + (f_2 : h) &\leq (F : h)(1 + 2\eta) \leq [(f : h) + \delta(k : h)](1 + 2\eta) \quad \text{by (7) and (4)} \\ &= (f : h) + 2\eta(f : h) + \delta(1 + 2\eta)(k : h). \end{aligned}$$

Dividing by  $(g : h)$ , we obtain

$$\Lambda_h f_1 + \Lambda_h f_2 \leq \Lambda_h f + 2\eta \Lambda_h f + \delta(1 + 2\eta) \Lambda_h k.$$

By (10) and (12), the second term on the right-hand side is  $\leq 2\eta(f : g) \leq \epsilon/2$ , and the third term is  $\leq 2\delta(k : g) \leq \epsilon/2$ , as desired.

We return to the proof of the theorem.

Given  $\epsilon > 0$  and  $f_1, f_2 \in C_c^+$ , if  $V \in \mathcal{V}$  is chosen as in the lemma, then for any  $h \in C_V^+$ , we have (by (7) for  $\Lambda_h$ )

$$|\Lambda_h(f_1 + f_2) - (\Lambda_h f_1 + \Lambda_h f_2)| \leq \epsilon. \quad (13)$$

Apply (11) to the functions  $f_1, f_2$ , and  $f_3 = f := f_1 + f_2$ , with  $V$  as in the lemma. Then for  $h$  as in (11), it follows from (13) that

$$\begin{aligned} |\Lambda f - (\Lambda f_1 + \Lambda f_2)| &\leq |\Lambda f - \Lambda_h f| + |\Lambda_h f - (\Lambda_h f_1 + \Lambda_h f_2)| \\ &\quad + |\Lambda_h f_1 - \Lambda f_1| + |\Lambda_h f_2 - \Lambda f_2| < 4\epsilon, \end{aligned}$$

and the additivity of  $\Lambda$  on  $C_c^+$  follows from the arbitrariness of  $\epsilon$ .

The desired properties of  $\Lambda$  on  $C_c$  follow as in the proof of Theorem 4.8.  $\square$

**Theorem 4.12.** *If  $\Lambda'$  is any left translation invariant positive linear functional on  $C_c(G)$ , then  $\Lambda' = c\Lambda$  for some constant  $c \geq 0$ .*

**Proof.** If  $\Lambda' = 0$ , take  $c = 0$ . So we may assume  $\Lambda' \neq 0$ . Since both  $\Lambda$  and  $\Lambda'$  are uniquely determined by their values on  $C_c^+$  (by linearity), it suffices to show that  $\Lambda'/\Lambda$  is constant on  $C_c^+$ . Thus, given  $f, g \in C_c^+$ , we must show that  $\Lambda'f/\Lambda f = \Lambda'g/\Lambda g$ .

Let  $K$  be the (compact) support of  $f$ ; since  $G$  is locally compact, there exists an open set  $W$  with compact closure such that  $K \subset W$ . For each  $x \in K$ , there exists  $W_x \in \mathcal{V}$  such that the  $x$ -neighbourhood  $xW_x$  is contained in  $W$ . By continuity of the group operation, there exists  $V_x \in \mathcal{V}$  such that  $V_x V_x \subset W_x$ . By compactness of  $K$ , there exist  $x_1, \dots, x_n \in K$  such that  $K \subset \bigcup_{i=1}^n x_i V_{x_i}$ . Let  $V_1 = \bigcap_{i=1}^n V_{x_i}$ . Then  $V_1 \in \mathcal{V}$  and

$$KV_1 \subset \bigcup_{i=1}^n x_i V_{x_i} V_1 \subset \bigcup_{i=1}^n x_i V_{x_i} V_{x_i} \subset \bigcup_{i=1}^n x_i W_{x_i} \subset W.$$

Similarly, there exists  $V_2 \in \mathcal{V}$  such that  $V_2 K \subset W$ .

Let  $\epsilon > 0$ . By uniform continuity of  $f$ , there exist  $V_3, V_4 \in \mathcal{V}$  such that, for all  $x \in G$ ,

$$|f(x) - f(sx)| < \epsilon/2 \quad \text{for all } s \in V_3$$

and

$$|f(x) - f(xt)| < \epsilon/2 \quad \text{for all } t \in V_4.$$

Let  $U := \bigcap_{j=1}^4 V_j$  and  $V := U \cap U^{-1}$  (where  $U^{-1} := \{x^{-1}; x \in U\}$ ). Then  $V \in \mathcal{V}$  has the following properties:

$$KV \subset W; \quad VK \subset W; \quad V^{-1} = V; \quad (14)$$

$$|f(sx) - f(xt)| < \epsilon \quad \text{for all } x \in G, \quad s, t \in V. \quad (15)$$

We shall need to integrate (15) with respect to  $x$  over  $G$ ; since the constant  $\epsilon$  is not integrable (unless  $G$  is compact), we fix a function  $k \in C_c^+$  such that  $k = 1$  on  $W$ ; necessarily

$$f(sx) = f(sx)k(x) \quad \text{and} \quad f(xs) = f(xs)k(x) \quad (16)$$

for all  $x \in G$  and  $s \in V$ . (This is trivial for  $x \in W$  since  $k = 1$  on  $W$ . If  $x \notin W$ , then  $x \notin KV$  and  $x \notin VK$  by (14); if  $sx \in K$  for some  $s \in V$ , then  $s^{-1} \in V$ , and consequently  $x = s^{-1}(sx) \in VK$ , a contradiction. Hence  $sx \notin K$ , and similarly  $xs \notin K$ , for all  $s \in V$ . Therefore both relations in (16) reduce to  $0 = 0$  when  $x \notin W$  and  $s \in V$ .)

By (15) and (16)

$$|f(xs) - f(sx)| \leq \epsilon k(x) \quad (17)$$

for all  $x \in G$  and  $s \in V$ .

Fix  $h' \in C_V^+$ , and let  $h(x) := h'(x) + h'(x^{-1})$ . Let  $\mu, \mu'$  be the unique positive measures associated with  $\Lambda$  and  $\Lambda'$ , respectively (cf. Theorem 3.18). Since  $h(x^{-1}y)f(y) \in C_c(G \times G) \subset L^1(\mu \times \mu')$ , we have by Fubini's theorem and the relation  $h(x^{-1}y) = h(y^{-1}x)$ :

$$\iint h(y^{-1}x)f(y) d\mu'(x) d\mu(y) = \iint h(x^{-1}y)f(y) d\mu(y) d\mu'(x). \quad (18)$$

By left translation invariance of  $\Lambda'$ , the left-hand side of (18) is equal to

$$\int \left( \int h(y^{-1}x) d\mu'(x) \right) f(y) d\mu(y) = \iint h(x) d\mu'(x) f(y) d\mu(y) = \Lambda' h \Lambda f. \quad (19)$$

By left translation invariance of  $\Lambda$ , the right-hand side of (18) is equal to

$$\int \left( \int h(y) f(xy) d\mu(y) \right) d\mu'(x),$$

and therefore  $\Lambda' h \Lambda f$  equals this last integral. On the other hand, by left translation invariance of  $\Lambda'$ ,

$$\begin{aligned} \int \left( \int h(y) f(yx) d\mu'(x) \right) d\mu(y) &= \int h(y) \left( \int f(yx) d\mu'(x) \right) d\mu(y) \\ &= \int h(y) \int f(x) d\mu'(x) d\mu(y) = \Lambda h \Lambda' f. \end{aligned}$$

Since  $h$  has support in  $V$ , we conclude from these calculations and from (17) that

$$\begin{aligned} |\Lambda' h \Lambda f - \Lambda h \Lambda' f| &= \left| \int_{x \in G} \int_{y \in V} h(y) [f(xy) - f(yx)] d\mu(y) d\mu'(x) \right| \\ &\leq \epsilon \int_{x \in G} \int_{y \in V} h(y) k(x) d\mu(y) d\mu'(x) = \epsilon \Lambda h \Lambda' k. \end{aligned} \quad (20)$$

Similarly, for  $g$  instead of  $f$ , and  $k'$  associated to  $g$  as  $k$  was to  $f$ , we obtain

$$|\Lambda' h \Lambda g - \Lambda h \Lambda' g| \leq \epsilon \Lambda h \Lambda' k'. \quad (21)$$

By (20) and (21) divided, respectively, by  $\Lambda h \Lambda f$  and  $\Lambda h \Lambda g$ , we have

$$\left| \frac{\Lambda' h}{\Lambda h} - \frac{\Lambda' f}{\Lambda f} \right| \leq \epsilon \frac{\Lambda' k}{\Lambda f}$$

and

$$\left| \frac{\Lambda' h}{\Lambda h} - \frac{\Lambda' g}{\Lambda g} \right| \leq \epsilon \frac{\Lambda' k'}{\Lambda g}.$$

Consequently

$$\left| \frac{\Lambda' f}{\Lambda f} - \frac{\Lambda' g}{\Lambda g} \right| \leq \epsilon \left( \frac{\Lambda' k}{\Lambda f} + \frac{\Lambda' k'}{\Lambda g} \right),$$

and the desired conclusion  $\Lambda' f / \Lambda f = \Lambda' g / \Lambda g$  follows from the arbitrariness of  $\epsilon$ .  $\square$

**Definition 4.13.** The unique (up to a constant factor) left translation invariant positive linear functional  $\Lambda$  on  $C_c(G)$  is called the (left) Haar functional for  $G$ . The measure  $\mu$  corresponding to  $\Lambda$  through Theorem 3.18 is the (left) Haar measure for  $G$ .

If  $G$  is compact, its (unique up to a constant factor) left Haar measure (which is finite by Theorem 3.18(2)) is *normalized* so that  $G$  has measure 1.

In an analogous way, there exists a unique (up to a constant factor) right translation invariant positive measure (as in Theorem 3.18)  $\lambda$  on  $G$ :

$$\int_G f^t d\lambda = \int_G f d\lambda \quad (f \in C_c(G); t \in G), \quad (22)$$

where  $f^t(x) := f(xt)$ .

Given the left Haar functional  $\Lambda$  on  $G$  and  $t \in G$ , define the functional  $\Lambda^t$  on  $C_c$  by  $\Lambda^t f := \Lambda f^t$ . Then  $\Lambda^t$  is a left translation invariant positive linear functional (because  $\Lambda^t(f_s) = \Lambda(f_s)^t = \Lambda(f^t)_s = \Lambda(f^t) := \Lambda^t f$ ), and therefore, by Theorem 4.12,

$$\Lambda^t = c(t)\Lambda \quad (23)$$

for some positive number  $c(t)$ . The function  $c(\cdot)$  is called the *modular function* of  $G$ . Since  $(\Lambda^t)^s = (\Lambda)^{st}$ , we have

$$c(st)\Lambda = (\Lambda)^{st} = c(s)\Lambda^t = c(s)c(t)\Lambda,$$

that is,  $c(\cdot)$  is a homomorphism of  $G$  into the multiplicative group of positive reals. We say that  $G$  is *unimodular* if  $c(\cdot) = 1$ . If  $G$  is compact, applying (23) to the function  $1 \in C_c(G)$ , we get  $c(\cdot) = 1$ . If  $G$  is abelian, we have  $f^t = f_t$ , hence  $\Lambda^t f = \Lambda(f_t) = \Lambda f$  for all  $f \in C_c$ , and therefore  $c(\cdot) = 1$ . Thus compact groups and (locally compact) abelian groups are unimodular.

If  $G$  is unimodular, the left Haar functional  $\Lambda$  is also *inverse invariant*, that is, letting  $\tilde{f}(x) := f(x^{-1})$ , one has  $\Lambda \tilde{f} = \Lambda f$  for all  $f \in C_c$ . Indeed, define  $\tilde{\Lambda}$  by  $\tilde{\Lambda} f = \Lambda \tilde{f}$  ( $f \in C_c$ ). Then  $\tilde{\Lambda}$  is a non-zero positive linear functional on  $C_c$ ; it is left translation invariant because  $(\tilde{f}_s)(x) = f_s(x^{-1}) = f(sx^{-1}) = f((xs^{-1})^{-1}) = \tilde{f}(xs^{-1}) = (\tilde{f})^{s^{-1}}(x)$ , and therefore by (23),

$$\tilde{\Lambda} f_s = \Lambda(\tilde{f}_s) = \Lambda^{s^{-1}} \tilde{f} = c(s^{-1})\Lambda \tilde{f} = \tilde{\Lambda} f.$$

By Theorem 4.12, there exists a positive constant  $\alpha$  such that  $\tilde{\Lambda} = \alpha\Lambda$ . Since  $f = \tilde{\tilde{f}}$  for all  $f$ , we have  $\tilde{\tilde{\Lambda}} = \Lambda$ , hence  $\alpha^2 = 1$ , and therefore  $\tilde{\Lambda} = \Lambda$ .

In terms of the Haar measure  $\mu$ , the inverse invariance of  $\Lambda$  takes the form

$$\int_G f(x^{-1}) d\mu(x) = \int_G f(x) d\mu(x) \quad (f \in C_c(G)) \quad (24)$$

for any *unimodular* (locally compact) group  $G$ .

## Exercises

1. Let  $X, Y$  be Banach spaces,  $Z$  a dense subspace of  $X$ , and  $T \in B(Z, Y)$ . Then there exists a unique  $\tilde{T} \in B(X, Y)$  such that  $\tilde{T}|_Z = T$ . Moreover, the map  $T \rightarrow \tilde{T}$  is an isometric isomorphism of  $B(Z, Y)$  onto  $B(X, Y)$ .
2. Let  $X$  be a locally compact Hausdorff space. Prove that  $C_0(X)^*$  is isometrically isomorphic to  $M_r(X)$ , the space of all *regular* complex Borel measures on  $X$ . (Hint: Theorems 3.24 and 4.9, and Exercise 1.)
3. Let  $X_k$   $k = 1, \dots, n$  be normed spaces, and consider  $\prod_k X_k$  as a normed space with the norm  $\|[x_1, \dots, x_n]\| = \sum_k \|x_k\|$ . Prove that there exists an isometric isomorphism of  $(\prod_k X_k)^*$  and  $\prod_k X_k^*$  with the norm  $\|[x_1^*, \dots, x_n^*]\| = \max_k \|x_k^*\|$ . (Hint: given  $\phi \in (\prod_k X_k)^*$ , define  $x_k^* x_k = \phi([0, \dots, x_k, 0, \dots, 0])$  for  $x_k \in X_k$ . Note that  $\phi([x_1, \dots, x_n]) = \sum_k x_k^* x_k$ .)
4. Let  $X$  be a locally compact Hausdorff space. Let  $Y$  be a normed space, and  $T \in B(C_c(X), Y)$ . Prove that there exists a unique  $P : \mathcal{B}(X) \rightarrow Y^{**} := (Y^*)^*$  such that  $P(\cdot)y^* \in M_r(X)$  for each  $y^* \in Y^*$  and

$$y^* T f = \int_X f d(P(\cdot)y^*)$$

for all  $f \in C_c(X)$  and  $y^* \in Y^*$ . Moreover  $\|P(\cdot)y^*\| = \|y^* \circ T\|$  for the appropriate norms (for all  $y^* \in Y^*$ ) and  $\|P(\delta)\| \leq \|T\|$  for all  $\delta \in \mathcal{B}(X)$ .

## Convolution on $L^p$

5. Let  $L^p$  denote the Lebesgue spaces on  $\mathbb{R}^k$  with respect to Lebesgue measure. Prove that if  $f \in L^1$  and  $g \in L^p$ , then  $f * g \in L^p$  and  $\|f * g\|_p \leq \|f\|_1 \|g\|_p$ . (Hint: use Theorems 4.6, 2.18, 1.33, and the translation invariance of Lebesgue measure; cf. Exercise 7, Chapter 2, in its  $\mathbb{R}^k$  version.)

## Approximate identities

6. Let  $m$  denote the normalized Lebesgue measure on  $[-\pi, \pi]$ . Let  $K_n : [-\pi, \pi] \rightarrow [0, \infty)$  be Lebesgue measurable functions such that  $\int_{-\pi}^{\pi} K_n dm = 1$  and

$$\sup_{\delta \leq |x| \leq \pi} K_n(x) \rightarrow 0 \quad (*)$$

as  $n \rightarrow \infty$ , for all  $\delta > 0$ . (Any sequence  $\{K_n\}$  with these properties is called an *approximate identity*.) Extend  $K_n$  to  $\mathbb{R}$  as  $2\pi$ -periodic functions.

Consider the convolutions

$$(K_n * f)(x) := \int_{-\pi}^{\pi} K_n(x-t)f(t) dm(t) = \int_{-\pi}^{\pi} f(x-t)K_n(t) dm(t)$$

with  $2\pi$ -periodic functions  $f$  on  $\mathbb{R}$ . Prove:

- (a) If  $f$  is continuous,  $K_n * f \rightarrow f$  uniformly on  $[-\pi, \pi]$ . (Hint:  $\int_{-\pi}^{\pi} = \int_{|t| < \delta} + \int_{\delta \leq |t| \leq \pi}$ .)
  - (b) If  $f \in L^p := L^p(-\pi, \pi)$  for some  $p \in [1, \infty)$ , then  $K_n * f \rightarrow f$  in  $L^p$ . (Hint: use the density of  $C([-\pi, \pi])$  in  $L^p$ , cf. Corollary 3.21, Part (a), and Exercise 5.)
  - (c) If  $f \in L^\infty$ , then  $K_n * f \rightarrow f$  in the *weak\**-topology on  $L^\infty$  (cf. Theorem 4.6); this means that  $\int (K_n * f)g \, dm \rightarrow \int fg \, dm$  for all  $g \in L^1$ .
7. Consider the measure space  $(\mathbb{N}, \mathbb{P}(\mathbb{N}), \mu)$ , where  $\mu$  is the *counting measure* ( $\mu(E)$  is the number of points in  $E$  if  $E$  is a finite subset of  $\mathbb{N}$  and  $= \infty$  otherwise). The space  $l^p := L^p(\mathbb{N}, \mathbb{P}(\mathbb{N}), \mu)$  is the space of all complex sequences  $x := \{x(n)\}$  such that  $\|x\|_p := (\sum |x(n)|^p)^{1/p} < \infty$  (in case  $p < \infty$ ) or  $\|x\|_\infty := \sup |x(n)| < \infty$  (in case  $p = \infty$ ). As a special case of Theorem 4.6, if  $p \in [1, \infty)$  and  $q$  is its conjugate exponent, then  $(l^p)^*$  is isometrically isomorphic to  $l^q$  through the map  $x^* \in (l^p)^* \rightarrow y \in l^q$ , where  $y := \{y(n)\}$  is the unique element of  $l^q$  such that  $x^*x = \sum x(n)y(n)$  for all  $x \in l^p$ . *Prove this directly!* (Hint: consider the unit vectors  $e_m \in l^p$  with  $e_m(n) = \delta_{n,m}$ , the Kronecker delta.)
  8. Consider  $\mathbb{N}$  with the discrete topology, and let  $c_0 := C_0(\mathbb{N})$  (this is the space of all complex sequences  $x := \{x_n\} = \{x(n)\}$  with  $\lim x_n = 0$ ). As a special case of Exercise 2, if  $x^* \in c_0^*$ , there exists a unique complex Borel measure  $\mu$  on  $\mathbb{N}$  such that  $x^*x = \sum_n x(n)\mu(\{n\})$ . Denote  $y(n) = \mu(\{n\})$ . Then  $\|y\|_1 = \sum |\mu(\{n\})| \leq |\mu|(\mathbb{N}) = \|\mu\| = \|x^*\|$ , that is,  $y \in l^1$  and  $\|y\|_1 \leq \|x^*\|$ . The reversed inequality is trivial. This shows that  $c_0^*$  is isometrically isometric to  $l^1$  through the map  $x^* \rightarrow y$ , where  $x^*x = \sum_n x(n)y(n)$ . *Prove this directly!*
  9. Let  $c$  denote the space of all *convergent* complex sequences  $x = \{x(n)\}$  with pointwise operations and the supremum norm. Show that  $c$  is a Banach space and  $c^*$  is isometrically isomorphic to  $l^1$ . (Hint: given  $x^* \in c^*$ ,  $x^*|_{c_0} \in c_0^*$ ; apply Exercise 8, and note that for each  $x \in c$ ,  $x - (\lim x)e \in c_0$ , where  $e(\cdot) = 1$ .)
  10. Let  $(X, \mathcal{A}, \mu)$  be a positive measure space,  $q \in (1, \infty]$ , and  $p = q/(q-1)$ . Prove that for all  $h \in L^q(\mu)$

$$\|h\|_q = \sup \left| \sum_k \alpha_k \int_{E_k} h \, d\mu \right|,$$

where the supremum is taken over all finite sums with  $\alpha_k \in \mathbb{C}$  and  $E_k \in \mathcal{A}$  with  $0 < \mu(E_k) < \infty$ , such that  $\sum |\alpha_k|^p \mu(E_k) \leq 1$ . (In case  $q = \infty$ , assume that the measure space is  $\sigma$ -finite.)

# 5

## Duality

We studied in preceding chapters the conjugate space  $X^*$  for various special normed spaces. Our purpose in the present chapter is to examine  $X^*$  and its relationship to  $X$  for a *general* normed space  $X$ .

### 5.1 The Hahn–Banach theorem

Let  $X$  be a vector space over  $\mathbb{R}$ . Suppose  $p : X \rightarrow \mathbb{R}$  is subadditive and homogeneous for non-negative scalars. A linear functional  $f$  on a subspace  $Y$  of  $X$  is  $p$ -dominated if  $f(y) \leq p(y)$  for all  $y \in Y$ . The starting point of this section is the following:

**Lemma 5.1 (The Hahn–Banach lemma).** *Let  $f$  be a  $p$ -dominated linear functional on the subspace  $Y$  of  $X$ . Then there exists a  $p$ -dominated linear functional  $F$  on  $X$  such that  $F|_Y = f$ .*

**Proof.** A  $p$ -dominated extension of  $f$  is a  $p$ -dominated linear functional  $g$  on a subspace  $D(g)$  of  $X$  containing  $Y$ , such that  $g|_Y = f$ . The family  $\mathcal{F}$  of all  $p$ -dominated extensions of  $f$  is partially ordered by setting  $g \leq h$  (for  $g, h \in \mathcal{F}$ ) if  $h$  is an extension of  $g$ . Each totally ordered subfamily  $\mathcal{F}_0$  of  $\mathcal{F}$  has an upper bound in  $\mathcal{F}$ , namely, the functional  $w$  whose domain is the subspace  $D(w) := \bigcup_{g \in \mathcal{F}_0} D(g)$ , and for  $x \in D(w)$  (so that  $x \in D(g)$  for some  $g \in \mathcal{F}_0$ ),  $w(x) = g(x)$ . Note that  $w$  is well defined, that is, its domain is indeed a subspace of  $X$  and the value  $w(x)$  is independent of the particular  $g$  such that  $x \in D(g)$ , thanks to the *total* ordering of  $\mathcal{F}_0$ . By Zorn's lemma,  $\mathcal{F}$  has a *maximal element*  $F$ . To complete the proof, we wish to show that  $D(F) = X$ . Suppose that  $D(F)$  is a proper subspace of  $X$ , and let then  $z_0 \in X - D(F)$ . Let  $Z$  be the subspace spanned by  $D(F)$  and  $z_0$ . The general element of  $Z$  has the form  $z = u + \alpha z_0$  with  $u \in D(F)$  and  $\alpha \in \mathbb{R}$  *uniquely* determined (indeed, if  $z = u' + \alpha' z_0$  is another such representation with  $\alpha' \neq \alpha$ , then  $z_0 = (\alpha' - \alpha)^{-1}(u - u') \in D(F)$ , a contradiction; thus  $\alpha' = \alpha$ , and therefore  $u' = u$ ). For any choice of  $\lambda \in \mathbb{R}$ , the functional  $h$  with domain  $Z$ , defined by  $h(z) = F(u) + \alpha\lambda$  is a well-defined linear

functional such that  $h|_{D(F)} = F$ . If we show that  $\lambda$  *can be chosen* such that the corresponding  $h$  is  $p$ -dominated, then  $h \in \mathcal{F}$  with domain  $Z$  *properly* containing  $D(F)$ , a contradiction to the maximality of the element  $F$  of  $\mathcal{F}$ .

Since  $F$  is  $p$ -dominated, we have for all  $u', u'' \in D(F)$

$$\begin{aligned} F(u') + F(u'') &= F(u' + u'') \leq p(u' + u'') \\ &= p([u' + z_0] + [u'' - z_0]) \leq p(u' + z_0) + p(u'' - z_0), \end{aligned}$$

that is,

$$F(u'') - p(u'' - z_0) \leq p(u' + z_0) - F(u') \quad (u', u'' \in D(F)).$$

Pick *any*  $\lambda$  *between* the supremum of the numbers on the left-hand side and the infimum of the numbers on the right-hand side. Then for all  $u', u'' \in D(F)$ ,

$$F(u') + \lambda \leq p(u' + z_0) \quad \text{and} \quad F(u'') - \lambda \leq p(u'' - z_0).$$

Taking  $u' = u/\alpha$  if  $\alpha > 0$  and  $u'' = u/(-\alpha)$  if  $\alpha < 0$  and multiplying the inequalities by  $\alpha$  and  $-\alpha$ , respectively, it follows from the homogeneity of  $p$  for non-negative scalars that

$$F(u) + \alpha\lambda \leq p(u + \alpha z_0) \quad (u \in D(F))$$

for *all* real  $\alpha$ , that is,  $h(z) \leq p(z)$  for all  $z \in Z$ . □

**Theorem 5.2 (The Hahn–Banach theorem).** *Let  $Y$  be a subspace of the normed space  $X$ , and let  $y^* \in Y^*$ . Then there exists  $x^* \in X^*$  such that  $x^*|_Y = y^*$  and  $\|x^*\| = \|y^*\|$ .*

**Proof.** *Case of real scalar field:* Take

$$p(x) := \|y^*\| \|x\| \quad (x \in X).$$

This function is subadditive and homogeneous, and

$$y^*y \leq |y^*y| \leq \|y^*\| \|y\| := p(y) \quad (y \in Y).$$

By Lemma 5.1, there exists a  $p$ -dominated linear functional  $F$  on  $X$  such that  $F|_Y = y^*$ . Thus, for all  $x \in X$ ,

$$F(x) \leq \|y^*\| \|x\|$$

and

$$-F(x) = F(-x) \leq p(-x) = \|y^*\| \|x\|,$$

that is,

$$|F(x)| \leq \|y^*\| \|x\|.$$

This shows that  $F := x^* \in X^*$  and  $\|x^*\| \leq \|y^*\|$ . Since the reversed inequality is trivial for *any* linear extension of  $y^*$ , the theorem is proved in the case of real scalars.



*Case of complex scalar field:* Take  $f := \Re y^*$  in Lemma 5.1. Then  $f(iy) = \Re[y^*(iy)] = \Re[iy^*y] = -\Im(y^*y)$ , and therefore

$$y^*y = f(y) - i f(iy) \quad (y \in Y). \quad (1)$$

For  $p$  as before, the functional  $f$  is  $p$ -dominated and *linear* on the vector space  $Y$  over the field  $\mathbb{R}$  (indeed,  $f(y) \leq |y^*y| \leq p(y)$  for all  $y \in Y$ ). By Lemma 5.1, there exists a  $p$ -dominated linear functional  $F : X \rightarrow \mathbb{R}$  (over *real* scalars!) such that  $F|_Y = f$ . Define

$$x^*x := F(x) - iF(ix) \quad (x \in X).$$

By (1),  $x^*|_Y = y^*$ . Clearly,  $x^*$  is additive and homogeneous for *real* scalars. Also, for all  $x \in X$ ,

$$x^*(ix) = F(ix) - iF(-x) = i[F(x) - iF(ix)] = ix^*x,$$

and it follows that  $x^*$  is homogeneous over  $\mathbb{C}$ .

Given  $x \in X$ , write  $x^*x = \rho\bar{\omega}$  with  $\rho \geq 0$  and  $\omega \in \mathbb{C}$  with modulus one. Then

$$\begin{aligned} |x^*x| &= \omega x^*x = x^*(\omega x) = \Re[x^*(\omega x)] \\ &= F(\omega x) \leq \|y^*\| \|\omega x\| = \|y^*\| \|x\|. \end{aligned}$$

Thus  $x^* \in X^*$  with norm  $\leq \|y^*\|$  (hence  $= \|y^*\|$ , since  $x^*$  is an extension of  $y^*$ ).  $\square$

**Corollary 5.3.** *Let  $Y$  be a subspace of the normed space  $X$ , and let  $x \in X$  be such that*

$$d := d(x, Y) := \inf_{y \in Y} \|x - y\| > 0.$$

*Then there exists  $x^* \in X^*$  with  $\|x^*\| = 1/d$ , such that*

$$x^*|_Y = 0, \quad x^*x = 1.$$

**Proof.** Let  $Z$  be the linear span of  $Y$  and  $x$ . Since  $d(x, Y) > 0$ ,  $x \notin Y$ , so that the general element  $z \in Z$  has the *unique* representation  $z = y + \alpha x$  with  $y \in Y$  and  $\alpha \in \mathbb{C}$ . Define then  $z^*z = \alpha$ . This is a well-defined linear functional on  $Z$ ,  $z^*|_Y = 0$ , and  $z^*x = 1$ . Also  $z^*$  is bounded, since

$$\begin{aligned} \|z^*\| &:= \sup_{0 \neq z \in Z} \frac{|\alpha|}{\|z\|} \\ &= \sup_{\alpha \neq 0; y \in Y} \frac{1}{\|(y + \alpha x)/\alpha\|} = \frac{1}{\inf_{y \in Y} \|x - y\|} = 1/d. \end{aligned}$$

By the Hahn–Banach theorem, there exists  $x^* \in X^*$  with norm  $= \|z^*\| = 1/d$  that extends  $z^*$ , whence  $x^*|_Y = 0$  and  $x^*x = 1$ .  $\square$

Note that if  $Y$  is a *closed* subspace of  $X$ , the condition  $d(x, Y) > 0$  is equivalent to  $x \notin Y$ . If  $Y \neq X$ , such an  $x$  exists, and therefore, by Corollary 5.3, there exists a *non-zero*  $x^* \in X^*$  such that  $x^*|_Y = 0$ . Formally

**Corollary 5.4.** *Let  $Y \neq X$  be a closed subspace of the normed space  $X$ . Then there exists a non-zero  $x^* \in X^*$  that vanishes on  $Y$ .*

For a not necessarily closed subspace  $Y$ , we apply the last corollary to its closure  $\bar{Y}$  (which is a closed subspace). By continuity, vanishing of  $x^*$  on  $Y$  is equivalent to its vanishing on  $\bar{Y}$ , and we obtain, therefore, the following useful criterion for *non-density*:

**Corollary 5.5.** *Let  $Y$  be a subspace of the normed space  $X$ . Then  $Y$  is not dense in  $X$  if and only if there exists a non-zero  $x^* \in X^*$  that vanishes on  $Y$ .*

For reference, we also state this criterion as a *density criterion*:

**Corollary 5.6.** *Let  $Y$  be a subspace of the normed space  $X$ . Then  $Y$  is dense in  $X$  if and only if the vanishing of an  $x^* \in X^*$  on  $Y$  implies  $x^* = 0$ .*

**Corollary 5.7.** *Let  $X$  be a normed space, and let  $0 \neq x \in X$ . Then there exists  $x^* \in X^*$  such that  $\|x^*\| = 1$  and  $x^*x = \|x\|$ . In particular,  $X^*$  separates points, that is, if  $x, y$  are distinct vectors in  $X$ , then there exists a functional  $x^* \in X^*$  such that  $x^*x \neq x^*y$ .*

**Proof.** We take  $Y = \{0\}$  in Corollary 5.3. Then  $d(x, Y) = \|x\| > 0$ , so that there exists  $z^* \in X^*$  such that  $\|z^*\| = 1/\|x\|$  and  $z^*x = 1$ . Let  $x^* := \|x\|z^*$ . Then  $x^*x = \|x\|$  and  $\|x^*\| = 1$  as wanted.

If  $x, y$  are distinct vectors, we apply the preceding result to the non-zero vector  $x - y$ ; we then obtain  $x^* \in X^*$  such that  $\|x^*\| = 1$  and  $x^*x - x^*y = x^*(x - y) = \|x - y\| \neq 0$ .  $\square$

**Corollary 5.8.** *Let  $X$  be a normed space. Then for each  $x \in X$ ,*

$$\|x\| = \sup_{x^* \in X^*; \|x^*\|=1} |x^*x|.$$

**Proof.** The relation being trivial for  $x = 0$ , we assume  $x \neq 0$ , and apply Corollary 5.7 to obtain an  $x^*$  with unit norm such that  $x^*x = \|x\|$ . Therefore, the supremum above is  $\geq \|x\|$ . Since the reverse inequality is a consequence of the definition of the norm of  $x^*$ , the result follows.  $\square$

Given  $x \in X$ , the functional

$$\kappa x : x^* \rightarrow x^*x$$

on  $X^*$  is linear, and Corollary 5.8 establishes that  $\|\kappa x\| = \|x\|$ . This means that  $\hat{x} := \kappa x$  is a continuous linear functional on  $X^*$ , that is, an element of  $(X^*)^* := X^{**}$ .

The map  $\kappa : X \rightarrow X^{**}$  is linear (since for all  $x^* \in X^*$ ,  $[\kappa(x + \alpha x')](x^*) = x^*(x + \alpha x') = x^*x + \alpha x^*x' = [\kappa x + \alpha \kappa x'](x^*)$ ) and *isometric* (since  $\|\kappa x - \kappa x'\| = \|\kappa(x - x')\| = \|x - x'\|$ ). The *isometric isomorphism*  $\kappa$  is called the *canonical* (or *natural*) *embedding* of  $X$  in the *second dual*  $X^{**}$ .

Note that  $X$  is complete iff its isometric image  $\hat{X} := \kappa X$  is complete, and since conjugate spaces are always complete,  $\kappa X$  is complete iff it is a *closed* subspace of  $X^{**}$ . Thus, a normed space is complete iff its canonical embedding  $\kappa X$  is a *closed* subspace of  $X^{**}$ . In case  $\kappa X = X^{**}$ , we say that  $X$  is *reflexive*. Our observations show in particular that a reflexive space is *necessarily* complete.

## 5.2 Reflexivity

**Theorem 5.9.** *A closed subspace of a reflexive Banach space is reflexive.*

**Proof.** Let  $X$  be a reflexive Banach space and let  $Y$  be a closed subspace of  $X$ . The restriction map

$$\psi : x^* \rightarrow x^*|_Y \quad (x^* \in X^*)$$

is a norm-decreasing linear map of  $X^*$  into  $Y^*$ . For each  $y^{**} \in Y^{**}$ , the function  $y^{**} \circ \psi$  belongs to  $X^{**}$ ; we thus have the (continuous linear) map

$$\chi : Y^{**} \rightarrow X^{**} \quad \chi(y^{**}) = y^{**} \circ \psi.$$

Let  $\kappa$  denote the canonical imbedding of  $X$  onto  $X^{**}$  (recall that  $X$  is reflexive!), and consider the (continuous linear) map

$$\kappa^{-1} \circ \chi : Y^{**} \rightarrow X.$$

We claim that its range  $Z$  is *contained in*  $Y$ . Indeed, suppose  $z \in Z$  but  $z \notin Y$ . Since  $Y$  is a closed subspace of  $X$ , there exists  $x^* \in X^*$  such that  $x^*Y = \{0\}$  and  $x^*z = 1$ . Then  $\psi(x^*) = 0$ , and since  $z = (\kappa^{-1} \circ \chi)(y^{**})$  for some  $y^{**} \in Y^{**}$ , we have

$$1 = x^*z = (\kappa z)(x^*) = [\chi(y^{**})](x^*) = [y^{**} \circ \psi](x^*) = y^{**}(0) = 0,$$

a contradiction. Thus

$$\kappa^{-1} \circ \chi : Y^{**} \rightarrow Y.$$

Given  $y^{**} \in Y^{**}$ , consider then the element

$$y := [\kappa^{-1} \circ \chi](y^{**}) \in Y.$$

For any  $y^* \in Y^*$ , let  $x^* \in X^*$  be an extension of  $y^*$  (cf. Hahn–Banach theorem). Then

$$y^{**}(y^*) = y^{**}(\psi(x^*)) = [\chi(y^{**})](x^*) = [\kappa(y)](x^*) = x^*(y) = y^*(y) = (\kappa_Y y)(y^*),$$

where  $\kappa_Y$  denotes the canonical imbedding of  $Y$  into  $Y^{**}$ . This shows that  $y^{**} = \kappa_Y y$ , so that  $\kappa_Y$  is *onto*, as wanted.  $\square$

**Theorem 5.10.** *If  $X$  and  $Y$  are isomorphic Banach spaces, then  $X$  is reflexive if and only if  $Y$  is reflexive.*

**Proof.** Let  $T : X \rightarrow Y$  be an isomorphism (i.e. a linear homeomorphism). Assume  $Y$  reflexive; all we need to show is that  $X$  is reflexive.

Given  $y^* \in Y^*$  and *any*  $T \in B(X, Y)$ , the composition  $y^* \circ T$  is a continuous linear functional on  $X$ , which we denote  $T^*y^*$ . This defines a map  $T^* \in B(Y^*, X^*)$ , called the (Banach) *adjoint* of  $T$ . One verifies easily that if  $T^{-1} \in B(Y, X)$ , then  $(T^*)^{-1}$  exists and equals  $(T^{-1})^*$ .

For simplicity of notation, we shall use the ‘hat notation’ ( $\hat{x}$  and  $\hat{y}$ ) for elements of  $X$  and  $Y$ , without specifying the space in the hat symbol.

Let  $x^{**} \in X^{**}$  be given. Then  $x^{**} \circ T^* \in Y^{**}$ , and since  $Y$  is reflexive, there exists a unique  $y \in Y$  such that

$$x^{**} \circ T^* = \hat{y}.$$

Let  $x = T^{-1}y$ . Then for all  $x^* \in X^*$ ,

$$\begin{aligned} \hat{x}x^* &= x^*x = x^*T^{-1}y = ((T^*)^{-1}x^*)y \\ &= \hat{y}((T^*)^{-1}x^*) = x^{**}[T^*((T^*)^{-1}x^*)] = x^{**}x^*, \end{aligned}$$

that is,  $x^{**} = \hat{x}$ . □

**Theorem 5.11.** *A Banach space is reflexive if and only if its conjugate is reflexive.*

**Proof.** Let  $X$  be a reflexive Banach space, and let  $\kappa$  be its canonical embedding onto  $X^{**}$ .

For any  $\phi \in (X^*)^{**} = (X^{**})^*$  the map  $\phi \circ \kappa$  is a continuous linear functional  $x^* \in X^*$ , and for any  $x^{**} \in X^{**}$ , letting  $x := \kappa^{-1}x^{**}$ , we have

$$\phi(x^{**}) = \phi(\kappa(x)) = x^*x = (\kappa x)(x^*) = x^{**}x^* = (\kappa_{X^*}x^*)(x^{**}).$$

This shows that  $\kappa_{X^*}$  is onto, that is  $X^*$  is reflexive.

Conversely, if  $X^*$  is reflexive, then  $X^{**}$  is reflexive by the first part of the proof. Since  $\kappa X$  is a *closed subspace* of  $X^{**}$ , it is reflexive by Theorem 5.9. Therefore,  $X$  is reflexive since it is isomorphic to  $\kappa X$ , by Theorem 5.10. □

We show below that Hilbert space and  $L^p$ -spaces (for  $1 < p < \infty$ ) are reflexive.

A map  $T : X \rightarrow Y$  between complex vector spaces is said to be *conjugate-homogeneous* if

$$T(\lambda x) = \bar{\lambda}Tx \quad (x \in X; \lambda \in \mathbb{C}).$$

An additive conjugate-homogeneous map is called a *conjugate-linear* map. In particular, we may talk of *conjugate-isomorphisms*.

**Lemma 5.12.** *If  $X$  is a Hilbert space (over  $\mathbb{C}$ ), then there exists an isometric conjugate-isomorphism  $V : X^* \rightarrow X$ , such that*

$$x^*x = (x, Vx^*) \quad (x \in X) \quad (1)$$

for all  $x^* \in X^*$ .

**Proof.** If  $x^* \in X^*$ , the ‘Little’ Riesz Representation theorem (Theorem 1.37) asserts that there exists a unique element  $y \in X$  such that  $x^*x = (x, y)$  for all  $x \in X$ . Denote  $y = Vx^*$ , so that  $V : X^* \rightarrow X$  is uniquely determined by the identity (1).

It follows from (1) that  $V$  is conjugate-linear. If  $Vx^* \neq 0$ , we have by (1) and Schwarz’ inequality

$$\begin{aligned} \|Vx^*\| &= \frac{(Vx^*, Vx^*)}{\|Vx^*\|} = \frac{|x^*(Vx^*)|}{\|Vx^*\|} \\ &\leq \|x^*\| = \sup_{x \neq 0} \frac{|x^*x|}{\|x\|} = \sup_{x \neq 0} \frac{|(x, Vx^*)|}{\|x\|} \leq \|Vx^*\|, \end{aligned}$$

so that  $\|Vx^*\| = \|x^*\|$  (this is trivially true also in case  $Vx^* = 0$ , since then  $x^* = 0$  by (1)).

Being *conjugate*-linear and norm-preserving,  $V$  is isometric, hence continuous and injective. It is also onto, because any  $y \in X$  induces the functional  $x^*$  defined by  $x^*x = (x, y)$  for all  $x \in X$ , and clearly  $Vx^* = y$  by the uniqueness of the Riesz representation.  $\square$

**Theorem 5.13.** *Hilbert space is reflexive.*

**Proof.** Denote by  $J$  the conjugation operator in  $\mathbb{C}$ .

Given  $x^{**} \in X^{**}$  (for  $X$  a complex Hilbert space), the map

$$J \circ x^{**} \circ V^{-1} : X \rightarrow \mathbb{C}$$

is continuous and linear. Denote it by  $x_0^*$ . Let  $x := Vx_0^*$ . Then for all  $x^* \in X^*$ ,

$$\begin{aligned} \hat{x}x^* &= x^*x = (x, Vx^*) = (Vx_0^*, Vx^*) = \overline{(Vx^*, Vx_0^*)} \\ &= (J \circ x_0^* \circ V)x^* = x^{**}x^*, \end{aligned}$$

that is,  $x^{**} = \hat{x}$ .  $\square$

In particular, finite dimensional spaces  $\mathbb{C}^n$  are reflexive. Also  $L^2(\mu)$  (for any positive measure space  $(X, \mathcal{A}, \mu)$ ) is reflexive. Theorem 5.14 establishes the reflexivity of all  $L^p$ -spaces for  $1 < p < \infty$ .

**Theorem 5.14.** *Let  $(X, \mathcal{A}, \mu)$  be a positive measure space. Then the space  $L^p(\mu)$  is reflexive for  $1 < p < \infty$ .*

**Proof.** Let  $q = p/(p-1)$ . Write  $L^p := L^p(\mu)$  and

$$\langle f, g \rangle := \int_X fg \, d\mu \quad (f \in L^p, g \in L^q).$$

By Theorem 4.6, there exists an isometric isomorphism

$$V_p : (L^p)^* \rightarrow L^q$$

such that

$$x^* f = \langle f, V_p x^* \rangle \quad (f \in L^p)$$

for all  $x^* \in (L^p)^*$ .

Given  $x^{**} \in (L^p)^{**}$ , the map  $x^{**} \circ (V_p)^{-1}$  is a continuous linear functional on  $L^q$ ; therefore

$$f := V_q \circ x^{**} \circ (V_p)^{-1} \in L^p.$$

Let  $x^* \in (L^p)^*$ , and write  $g := V_p x^* \in L^q$ ; we have

$$\begin{aligned} \hat{f}(x^*) &= x^* f = \langle f, g \rangle = [(V_q)^{-1} f](g) \\ &= x^{**}(V_p^{-1} g) = x^{**} x^*. \end{aligned}$$

This shows that  $x^{**} = \hat{f}$ . □

The theorem is false in general for  $p = 1$  and  $p = \infty$ .

Also the space  $C_0(X)$  (for a locally compact Hausdorff space  $X$ ) is not reflexive in general. We shall not prove these facts here.

### 5.3 Separation

We now consider applications of the Hahn–Banach lemma to *separation* of convex sets in vector spaces.

Let  $X$  be a vector space over  $\mathbb{C}$  or  $\mathbb{R}$ . A *convex combination* of vectors  $x_k \in X$  ( $k = 1, \dots, n$ ) is any vector of the form

$$\sum_{k=1}^n \alpha_k x_k \quad \left( \alpha_k \geq 0; \sum_k \alpha_k = 1 \right).$$

A subset  $K \subset X$  is *convex* if it contains the convex combinations of any two vectors in it.

Equivalently, a set  $K \subset X$  is convex if it is invariant under the operation of taking convex combinations of its elements. Indeed, invariance under convex combinations of pairs of elements is precisely the definition of convexity. On the other hand, if  $K$  is convex, one can prove the said invariance by induction on the number  $n$  of vectors. Assuming invariance for  $n \geq 2$  vectors, consider any convex combination  $z = \sum_{k=1}^{n+1} \alpha_k x_k$  of vectors  $x_k \in K$ . If  $\alpha := \sum_{k=1}^n \alpha_k = 0$ , then  $z = x_{n+1} \in K$  trivially. So assume  $\alpha > 0$ ; since  $\alpha_{n+1} = 1 - \alpha$ , we have

$$z = \alpha \sum_{k=1}^n \frac{\alpha_k}{\alpha} x_k + (1 - \alpha) x_{n+1} \in K,$$

by the induction hypothesis and the convexity of  $K$ .

The intersection of a family of convex sets is clearly convex. The *convex hull* of a set  $M$  (denoted  $\text{co}(M)$ ) is the intersection of all convex sets containing  $M$ . It is the smallest convex set containing  $M$  ('smallest' with respect to set inclusion), and consists of all the convex combinations of vectors in  $M$ .

If  $M, N$  are convex subsets of  $X$  and  $\alpha, \beta$  are scalars, then  $\alpha M + \beta N$  is convex. Also  $TM$  is convex for any linear map  $T : X \rightarrow Y$  between vector spaces.

Let  $M \subset X$ ; the point  $x \in M$  is an *internal point* of  $M$  if for each  $y \in X$ , there exists  $\epsilon = \epsilon(y)$  such that  $x + \alpha y \in M$  for all  $\alpha \in \mathbb{C}$  with  $|\alpha| \leq \epsilon$ . Clearly, an internal point of  $M$  is also an internal point of any  $N$  such that  $M \subset N \subset X$ .

Suppose  $0$  is an internal point of the *convex* set  $M$ . Then for each  $y \in X$ , there exists  $\epsilon := \epsilon(y)$  such that  $y/\rho \in M$  for all  $\rho \geq 1/\epsilon$ . If  $y/\rho \in M$ , then by convexity  $y/(\rho/\alpha) = (1 - \alpha)0 + \alpha(y/\rho) \in M$  for all  $0 < \alpha \leq 1$ . Since  $\rho \leq \rho/\alpha < \infty$  for such  $\alpha$ , this means that  $y/\rho' \in M$  for all  $\rho' \geq \rho$ , that is, the set  $\{\rho > 0; y/\rho \in M\}$  is a subray of  $\mathbb{R}^+$  that contains  $1/\epsilon$ . Let  $\kappa(y)$  be the left endpoint of that ray, that is,

$$\kappa(y) := \inf\{\rho > 0; y/\rho \in M\}.$$

Then  $\kappa(y) \leq 1/\epsilon$ , so that  $0 \leq \kappa(y) < \infty$ , and  $\kappa(y) \leq 1$  for  $y \in M$  (equivalently, if  $\kappa(y) > 1$ , then  $y \in M^c$ ). If  $\alpha > 0$ ,

$$\alpha\{\rho > 0; y/\rho \in M\} = \{\alpha\rho; (\alpha y)/(\alpha\rho) \in M\},$$

and it follows that  $\kappa(\alpha y) = \alpha\kappa(y)$  (this is also trivially true for  $\alpha = 0$ , since  $0 \in M$ ).

If  $x, y \in X$  and  $\rho > \kappa(x), \sigma > \kappa(y)$ , then  $x/\rho, y/\sigma \in M$ , and since  $M$  is convex,

$$\frac{x+y}{\rho+\sigma} = \frac{\rho}{\rho+\sigma}x/\rho + \frac{\sigma}{\rho+\sigma}y/\sigma \in M.$$

Hence  $\kappa(x+y) \leq \rho + \sigma$ , and therefore  $\kappa(x+y) \leq \kappa(x) + \kappa(y)$ .

We conclude that  $\kappa : X \rightarrow [0, \infty)$  is a subadditive positive-homogeneous functional, referred to as the *Minkowski functional* of the convex set  $M$ .

**Lemma 5.15.** *Let  $M \subset X$  be convex with  $0$  internal, and let  $\kappa$  be its Minkowski functional. Then  $\kappa(x) < 1$  if and only if  $x$  is an internal point of  $M$ , and  $\kappa(x) > 1$  if and only if  $x$  is an internal point of  $M^c$ .*

**Proof.** Let  $x$  be an internal point of  $M$ , and let  $\epsilon(\cdot)$  be as in the definition. Then  $x + \epsilon(x)x \in M$ , and therefore

$$\kappa(x) \leq \frac{1}{1 + \epsilon(x)} < 1.$$

Conversely, suppose  $\kappa(x) < 1$ . Then  $x = x/1 \in M$ . Let  $y \in X$ . Since  $0$  is internal for  $M$ , there exists  $\epsilon_0 > 0$  (depending on  $y$ ) such that  $\beta y \in M$  for  $|\beta| \leq \epsilon_0$ . In particular  $\kappa(\epsilon_0 \omega y) \leq 1$  for all  $\omega \in \mathbb{C}$  with  $|\omega| = 1$ . Now (with any  $\alpha = |\alpha|\omega$ ),

$$\kappa(x + \alpha y) \leq \kappa(x) + \kappa(\alpha y) = \kappa(x) + \frac{|\alpha|}{\epsilon_0} \kappa(\epsilon_0 \omega y) \leq \kappa(x) + |\alpha|/\epsilon_0 < 1$$

for  $|\alpha| \leq \epsilon$ , with  $\epsilon < [1 - \kappa(x)]\epsilon_0$ . Hence  $x + \alpha y \in M$  for  $|\alpha| \leq \epsilon$ , so that  $x$  is an internal point of  $M$ .

If  $x$  is an internal point of  $M^c$ , there exists  $\epsilon > 0$  such that  $x - \epsilon x \in M^c$ . If however  $\kappa(x) \leq 1$ , then  $1/(1 - \epsilon) > 1 \geq \kappa(x)$ , and therefore  $x - \epsilon x \in M$ , contradiction. Thus  $\kappa(x) > 1$ .

Conversely, suppose  $\kappa(x) > 1$ . Let  $y \in X$ , and choose  $\epsilon_0$  as above. Then

$$\kappa(x + \alpha y) \geq \kappa(x) - \frac{|\alpha|}{\epsilon_0} \kappa(\epsilon_0 \omega y) \geq \kappa(x) - |\alpha|/\epsilon_0 > 1$$

if  $|\alpha| \leq \epsilon$  with  $\epsilon < [\kappa(x) - 1]\epsilon_0$ . This shows that  $x + \alpha y \in M^c$  for  $|\alpha| \leq \epsilon$ , and so  $x$  is an internal point of  $M^c$ .  $\square$

By the lemma,  $\kappa(x) = 1$  if and only if  $x$  is not internal for  $M$  and for  $M^c$ ; such a point is called a *bounding point* for  $M$  (or for  $M^c$ ).

We shall apply the Hahn–Banach lemma (Lemma 5.1) with  $p = \kappa$  to obtain the following:

**Theorem 5.16 (Separation theorem).** *Let  $M, N$  be disjoint non-empty convex sets in the vector space  $X$  (over  $\mathbb{C}$  or  $\mathbb{R}$ ), and suppose  $M$  has an internal point. Then there exists a non-zero linear functional  $f$  on  $X$  such that*

$$\sup \Re f(M) \leq \inf \Re f(N).$$

(one says that  $f$  separates  $M$  and  $N$ ).

**Proof.** Suppose that the theorem is valid for vector spaces over  $\mathbb{R}$ . If  $X$  is a vector space over  $\mathbb{C}$ , we may consider it as a vector space over  $\mathbb{R}$ , and get an  $\mathbb{R}$ -linear non-zero functional  $\phi : X \rightarrow \mathbb{R}$  such that  $\sup \phi(M) \leq \inf \phi(N)$ . Setting  $f(x) := \phi(x) - i\phi(ix)$  as in the proof of Theorem 5.2, we obtain a non-zero  $\mathbb{C}$ -linear functional on  $X$  such that  $\sup \Re f(M) = \sup \phi(M) \leq \inf \phi(N) = \inf \Re f(N)$ , as wanted.

This shows that we need to prove the theorem for vector spaces over  $\mathbb{R}$  only. We may also assume that  $0$  is an internal point of  $M$ . Indeed, suppose the theorem is valid in that case. By assumption,  $M$  has an internal point  $x$ . Thus for each  $y \in X$ , there exists  $\epsilon > 0$  such that  $x + \alpha y \in M$  for all  $|\alpha| \leq \epsilon$ . Equivalently,  $0 + \alpha y \in M - x$  for all such  $\alpha$ , that is,  $0$  is internal for  $M - x$ . The sets  $N - x$  and  $M - x$  are disjoint convex sets, and the theorem (for the special case  $0$  internal to  $M - x$ ) implies the existence of a non-zero linear functional  $f$  such that  $\sup f(M - x) \leq \inf f(N - x)$ . Therefore  $\sup f(M) \leq \inf f(N)$  as desired.

Fix  $z \in N$  and let  $K := M - N + z$ . Then  $K$  is convex,  $M \subset K$ , and therefore  $0$  is an internal point of  $K$ . Let  $\kappa$  be the Minkowski functional of  $K$ . Since  $M$  and  $N$  are disjoint,  $z \notin K$ , and therefore  $\kappa(z) \geq 1$ .

Define  $f_0 : \mathbb{R}z \rightarrow \mathbb{R}$  by  $f_0(\lambda z) = \lambda \kappa(z)$ . Then  $f_0$  is linear and  $\kappa$ -dominated (since for  $\lambda \geq 0$ ,  $f_0(\lambda z) := \lambda \kappa(z) = \kappa(\lambda z)$ , and for  $\lambda < 0$ ,  $f_0(\lambda z) < 0 \leq \kappa(\lambda z)$ ). By the Hahn–Banach lemma (Lemma 5.1), there exists a  $\kappa$ -dominated linear



extension  $f : X \rightarrow \mathbb{R}$  of  $f_0$ . Then  $f(z) = f_0(z) = \kappa(z) \geq 1$ , and  $f(x) \leq \kappa(x) \leq 1$  for all  $x \in K$ . This means that  $f$  is a non-zero linear functional on  $X$  such that

$$f(M) - f(N) + f(z) = f(M - N + z) = f(K) \leq 1 \leq f(z),$$

that is,  $f(M) \leq f(N)$ . □

## 5.4 Topological vector spaces

We consider next a vector space  $X$  with a Hausdorff topology such that the vector space operations are continuous (such a space is called a *topological vector space*). The function

$$f : (x, y, \alpha) \in X \times X \times [0, 1] \rightarrow \alpha x + (1 - \alpha)y \in X$$

is continuous. The set  $M \subset X$  is convex if and only if  $f(M \times M \times [0, 1]) \subset M$ . Therefore, by continuity of  $f$ , if  $M$  is convex, we have

$$f(\bar{M} \times \bar{M} \times [0, 1]) = f(\overline{M \times M \times [0, 1]}) \subset \overline{f(M \times M \times [0, 1])} \subset \bar{M},$$

which proves that the closure of a convex set is convex. A trivial modification of the proof shows that the closure of a subspace is a subspace.

Let  $M^\circ$  denote the interior of  $M$ . We show that for  $0 < \alpha < 1$  and  $M \subset X$  convex,

$$\alpha M^\circ + (1 - \alpha)\bar{M} \subset M^\circ. \quad (1)$$

In particular, it follows from (1) that  $M^\circ$  is convex.

Let  $x \in M^\circ$ , and let  $V$  be a neighbourhood of  $x$  contained in  $M$ . Since addition and multiplication by a non-zero scalar are homeomorphisms of  $X$  onto itself,  $U := V - x$  is a neighbourhood of 0 and  $y + \beta U$  is a neighbourhood of  $y$  for any  $0 \neq \beta \in \mathbb{R}$ . Therefore, if  $y \in \bar{M}$ , there exists  $y_\beta \in M \cap (y + \beta U)$ . Thus there exists  $u \in U$  such that  $y = y_\beta - \beta u$ . Then, given  $\alpha \in (0, 1)$  and choosing  $\beta = \alpha/(\alpha - 1)$ , we have by convexity of  $M$

$$\begin{aligned} \alpha x + (1 - \alpha)y &= \alpha x + (1 - \alpha)y_\beta + (\alpha - 1)\beta u \\ &= \alpha(x + u) + (1 - \alpha)y_\beta \in \alpha V + (1 - \alpha)y_\beta \\ &:= V_\beta \subset \alpha M + (1 - \alpha)M \subset M, \end{aligned}$$

where  $V_\beta$  is clearly open. This proves that  $\alpha x + (1 - \alpha)y \in M^\circ$ , as wanted.

If  $M^\circ \neq \emptyset$  and we fix  $x \in M^\circ$ , the continuity of the vector space operations imply that

$$y = \lim_{\alpha \rightarrow 0+} \alpha x + (1 - \alpha)y \quad (y \in \bar{M}),$$

and it follows from (1) that  $M^\circ$  is dense in  $M$ .

With notation as before, it follows from the continuity of multiplication by scalars that for any  $y \in X$ , there exists  $\epsilon = \epsilon(y)$  such that  $\alpha y \in U$  for all  $\alpha \in \mathbb{C}$

with  $|\alpha| \leq \epsilon$ ; thus, for these  $\alpha, x + \alpha y \in x + U = V \subset M$ . This shows that interior points of  $M$  are internal for  $M$ .

It follows that bounding points for  $M$  are boundary points of  $M$ .

Conversely, if  $x \in M$  is internal and  $M^\circ \neq \emptyset$ , pick  $m \in M^\circ$ . Then there exists  $\epsilon > 0$  such that

$$(1 + \epsilon)x - \epsilon m = x + \epsilon(x - m) := m' \in M.$$

Therefore, by (1),

$$x = \frac{\epsilon}{1 + \epsilon}m + \frac{1}{1 + \epsilon}m' \in M^\circ.$$

Thus internal points for  $M$  are interior points of  $M$  (when the interior is not empty).

Still for  $M$  convex with non-empty interior, suppose  $y$  is a boundary point of  $M$ . Pick  $x \in M^\circ$ . For  $0 < \alpha < 1, y + \alpha(x - y) = \alpha x + (1 - \alpha)y \in M^\circ$  by (1), and therefore  $y$  is not internal for  $M^c$ . It is not internal for  $M$  as well (since internal points are interior!). Thus  $y$  is a bounding point for  $M$ . Collecting, we have

**Lemma 5.17.** *Let  $M$  be a convex set with non-empty interior in a topological vector space. A point is internal (bounding) of  $M$  if and only if it is an interior (boundary) point of  $M$ .*

**Lemma 5.18.** *Let  $X$  be a topological vector space, let  $M, N$  be non-empty subsets of  $X$  with  $M^\circ \neq \emptyset$ . If  $f$  is a linear functional on  $X$  that separates  $M$  and  $N$ , then  $f$  is continuous.*

**Proof.** Since  $(\Im f)(x) = -(\Re f)(ix)$  in the case of complex vector spaces, it suffices to show that  $\Re f$  is continuous, and this reduces the complex case to the real case. Let then  $f : X \rightarrow \mathbb{R}$  be linear such that  $\sup f(M) := \delta \leq \inf f(N)$ . Let  $m \in M^\circ$  and  $n \in N$ . Let then  $U$  be a symmetric neighbourhood of 0 (i.e.  $-U = U$ ) such that  $m + U \subset M$  (if  $V$  is any 0-neighbourhood such that  $m + V \subset M$ , we may take  $U = V \cap (-V)$ ). Then  $0 \in -U = U \subset M - m$ , and therefore, for any  $u \in U$ ,

$$f(u) \leq \sup f(M) - f(m) = \delta - f(m) \leq f(n) - f(m),$$

and the same inequality holds for  $-u$ . In particular (taking  $u = 0$ ),  $f(n) - f(m) \geq 0$ . Pick any  $\rho > f(n) - f(m)$ . Then  $f(u) < \rho$  and also  $-f(u) = f(-u) < \rho$ , that is,  $|f(u)| < \rho$  for all  $u \in U$ . Hence, given  $\epsilon > 0$ , the 0-neighbourhood  $(\epsilon/\rho)U$  is mapped by  $f$  into  $(-\epsilon, \epsilon)$ , which proves that  $f$  is continuous at 0. However continuity at 0 is equivalent to continuity for linear maps between topological vector spaces, as is readily seen by translation.  $\square$

Combining Lemmas 5.17 and 5.18 with Theorem 5.16, we obtain the following *separation theorem* for topological vector spaces:

**Theorem 5.19 (Separation theorem).** *In a topological vector space, any two disjoint non-empty convex sets, one of which has non-empty interior, can be separated by a non-zero continuous linear functional.*

If we have *strict* inequality in Theorem 5.16, the functional  $f$  *strictly separates* the sets  $M$  and  $N$  (it is necessarily non-zero). A strict separation theorem is stated below for a *locally convex* topological vector space (t.v.s.), that is, a t.v.s. whose topology has a *base consisting of convex sets*.

**Theorem 5.20 (Strict separation theorem).** *Let  $X$  be a locally convex t.v.s. Let  $M, N$  be non-empty disjoint convex sets in  $X$ . Suppose  $M$  is compact and  $N$  is closed. Then there exists a continuous linear functional on  $X$  which strictly separates  $M$  and  $N$ .*

**Proof.** Observe first that  $M - N$  is *closed*. Indeed, if a net  $\{m_i - n_i\}$  ( $m_i \in M; n_i \in N; i \in I$ ) converges to  $x \in X$ , then since  $M$  is compact, a subnet  $\{m_{i'}\}$  converges to some  $m \in M$ . By continuity of vector space operations, the net  $\{n_{i'}\} = \{m_{i'} - (m_{i'} - n_{i'})\}$  converges to  $m - x$ , and since  $N$  is closed,  $m - x := n \in N$ . Therefore  $x = m - n \in M - N$  and  $M - N$  is closed. It is also convex.

Since  $M, N$  are disjoint, the point 0 is in the open set  $(M - N)^c$ , and since  $X$  is locally convex, there exists a convex neighbourhood of 0,  $U$ , disjoint from  $M - N$ . By Theorem 5.19 (applied to the sets  $M - N$  and  $U$ ), there exists a non-zero continuous linear functional  $f$  separating  $M - N$  and  $U$ :

$$\sup \Re f(U) \leq \inf \Re f(M - N).$$

Since  $f \neq 0$ , there exists  $y \in X$  such that  $f(y) = 1$ . By continuity of multiplication by scalars, there exists  $\epsilon > 0$  such that  $\epsilon y \in U$ . Then

$$\epsilon = \Re f(\epsilon y) \leq \sup \Re f(U) \leq \inf \Re f(M - N),$$

that is,  $\Re f(n) + \epsilon \leq \Re f(m)$  for all  $m \in M$  and  $n \in N$ . Thus

$$\sup \Re f(N) < \sup \Re f(N) + \epsilon \leq \inf \Re f(M).$$

□

Taking  $M = \{p\}$ , we get the following

**Corollary 5.21.** *Let  $X$  be a locally convex t.v.s., let  $N$  be a (non-empty) closed convex set in  $X$ , and  $p \notin N$ . Then there exists a continuous linear functional  $f$  strictly separating  $p$  and  $N$ . In particular (with  $N = \{q\}, q \neq p$ ), the continuous linear functionals on  $X$  separate the points of  $X$  (i.e. if  $p, q$  are any distinct points of  $X$ , then there exists a continuous linear functional  $f$  on  $X$  such that  $f(p) \neq f(q)$ ).*

## 5.5 Weak topologies

We shall consider topologies induced on a given vector space  $X$  by families of linear functionals on it. Let  $\Gamma$  be a *separating* vector space of linear functionals on  $X$ . Equivalently, if  $\Gamma x = \{0\}$ , then  $x = 0$ . The  $\Gamma$ -topology of  $X$  is *the weakest*

topology on  $X$  for which all  $f \in \Gamma$  are continuous. A base for this topology consists of all sets of the form

$$N(x; \Delta, \epsilon) = \{y \in X; |f(y) - f(x)| < \epsilon \text{ for all } f \in \Delta\},$$

where  $x \in X$ ,  $\Delta \subset \Gamma$  is finite, and  $\epsilon > 0$ . The net  $\{x_i; i \in I\}$  converges to  $x$  in the  $\Gamma$ -topology if and only if  $f(x_i) \rightarrow f(x)$  for all  $f \in \Gamma$ . The vector space operations are  $\Gamma$ -continuous, and the sets in the basis are clearly convex, so that  $X$  with the  $\Gamma$ -topology (sometimes denoted  $X_\Gamma$ ) is a locally convex t.v.s. Let  $X_\Gamma^*$  denote the space of all continuous linear functionals on  $X_\Gamma$ . By definition of the  $\Gamma$ -topology,  $\Gamma \subset X_\Gamma^*$ . We show below that we actually have equality between these sets.

**Lemma 5.22.** *Let  $f_1, \dots, f_n, g$  be linear functionals on the vector space  $X$  such that*

$$\bigcap_{i=1}^n \ker f_i \subset \ker g.$$

*Then  $g \in \text{span} \{f_1, \dots, f_n\}$  ( $:=$  the linear span of  $f_1, \dots, f_n$ ).*

**Proof.** Consider the linear map

$$T : X \rightarrow \mathbb{C}^n, \quad Tx = (f_1(x), \dots, f_n(x)) \in \mathbb{C}^n.$$

Define  $\phi : TX \rightarrow \mathbb{C}$  by  $\phi(Tx) = g(x)$  ( $x \in X$ ). If  $Tx = Ty$ , then  $x - y \in \bigcap_i \ker f_i \subset \ker g$ , hence  $g(x) = g(y)$ , which shows that  $\phi$  is well defined. It is clearly linear, and has therefore an extension as a linear functional  $\tilde{\phi}$  on  $\mathbb{C}^n$ . The form of  $\tilde{\phi}$  is  $\tilde{\phi}(\lambda_1, \dots, \lambda_n) = \sum_i \alpha_i \lambda_i$  with  $\alpha_i \in \mathbb{C}$ . In particular, for all  $x \in X$ ,

$$g(x) = \phi(Tx) = \tilde{\phi}(Tx) = \sum_i \alpha_i f_i(x).$$

□

**Theorem 5.23.**  $X_\Gamma^* = \Gamma$ .

**Proof.** It suffices to prove that if  $0 \neq g$  is a  $\Gamma$ -continuous linear functional on  $X$ , then  $g \in \Gamma$ . Let  $U$  be the unit disc in  $\mathbb{C}$ . If  $g$  is  $\Gamma$ -continuous, there exists a basic neighbourhood  $N = N(0; f_1, \dots, f_n; \epsilon)$  of zero (in the  $\Gamma$ -topology) such that  $g(N) \subset U$ . If  $x \in \bigcap_i \ker f_i := Z$ , then  $x \in N$ , and therefore  $|g(x)| < 1$ . But  $Z$  is a subspace of  $X$ , hence  $kx \in Z$  for all  $k \in \mathbb{N}$ , and so  $k|g(x)| < 1$  for all  $k$ . This shows that  $Z \subset \ker g$ , and therefore, by Lemma 5.22,  $g \in \text{span} \{f_1, \dots, f_n\} \subset \Gamma$ . □

The following special  $\Gamma$ -topologies are especially important:

- (1) If  $X$  is a Banach space and  $X^*$  is its conjugate space, the  $X^*$ -topology for  $X$  is called *the weak topology for  $X$*  (the usual norm topology is also called *the strong topology*, and is clearly stronger than the weak topology).
- (2) If  $X^*$  is the conjugate of the Banach space  $X$ , the  $\hat{X}$ -topology for  $X^*$  is called *the weak\*-topology for  $X^*$* . It is in general weaker than the weak

topology (i.e. the  $X^{**}$ -topology) on  $X^*$ . The basis given above (in case of the  $weak^*$ -topology) consists of the sets

$$N(x^*; x_1, \dots, x_n; \epsilon) = \{y^* \in X^*; |y^* x_k - x^* x_k| < \epsilon\}$$

with  $x^* \in X^*, x_k \in X, \epsilon > 0, n \in \mathbb{N}$ .

A net  $\{x_i^*\}$  converges  $weak^*$  to  $x^*$  if and only if  $x_i^* x \rightarrow x^* x$  for all  $x \in X$  (this is pointwise convergence of the functions  $x_i^*$  to  $x^*$  on  $X$ !).

**Theorem 5.24 (Alaoglu's theorem).** *Let  $X$  be a Banach space. Then the (strongly) closed unit ball of  $X^*$*

$$S^* := \{x^* \in X^*; \|x^*\| \leq 1\}$$

*is compact in the  $weak^*$  topology.*

**Proof.** Let

$$\Delta(x) = \{\lambda \in \mathbb{C}; |\lambda| \leq \|x\|\} \quad (x \in X),$$

and

$$\Delta = \prod_{x \in X} \Delta(x)$$

with the Cartesian product topology. By Tychonoff's theorem,  $\Delta$  is compact.

If  $f \in S^*, f(x) \in \Delta(x)$  for each  $x \in X$ , so that  $f \in \Delta$ , that is,  $S^* \subset \Delta$ . Convergence in the relative  $\Delta$ -topology on  $S^*$  is pointwise convergence at all points  $x \in X$ , and this is precisely  $weak^*$ -convergence in  $S^*$ . The theorem will then follow if we show that  $S^*$  is closed in  $\Delta$ . Suppose  $\{f_i; i \in I\}$  is a net in  $X^*$  converging in  $\Delta$  to some  $f$ . This means that  $f_i(x) \rightarrow f(x)$  for all  $x \in X$ . Therefore, for each  $x, y \in X$  and  $\lambda \in \mathbb{C}$ ,

$$\begin{aligned} f(x + \lambda y) &= \lim_i f_i(x + \lambda y) = \lim_i [f_i(x) + \lambda f_i(y)] \\ &= \lim_i f_i(x) + \lambda \lim_i f_i(y) = f(x) + \lambda f(y), \end{aligned}$$

and since  $|f(x)| \leq \|x\|$ , we conclude that  $f \in S^*$ . □

**Theorem 5.25 (Goldstine's theorem).** *Let  $S$  and  $S^{**}$  be the strongly closed unit balls in  $X$  and  $X^{**}$ , respectively, and let  $\kappa : S \rightarrow S^{**}$  be the canonical embedding (cf. comments following Corollary 5.8). Then  $\kappa S$  is  $weak^*$ -dense in  $S^{**}$ .*

**Proof.** Let  $\overline{\kappa S}$  denote the  $weak^*$ -closure of  $\kappa S$ . Proceeding by contradiction, suppose  $x^{**} \in S^{**}$  is not in  $\overline{\kappa S}$ . We apply Corollary 5.21 in the locally convex t.v.s.  $X^{**}$  with the  $weak^*$ -topology. There exists then a ( $weak^*$ -)continuous linear functional  $F$  on  $X^{**}$  and a real number  $\lambda$  such that

$$\Re F(x^{**}) > \lambda > \sup_{x \in S} \Re F(\hat{x}), \quad (1)$$

where  $\hat{x} := \kappa x$ .

The *weak\**-topology on  $X^{**}$  is the  $\Gamma$ -topology on  $X^{**}$  where  $\Gamma$  consists of all the linear functionals on  $X^{**}$  of the form  $x^{**} \rightarrow x^{**}x^*$ , with  $x^* \in X^*$ . By Theorem 5.23, the functional  $F$  is of this form, for a suitable  $x^*$ . We can then rewrite (1) as follows:

$$\Re x^{**}x^* > \lambda > \sup_{x \in S} \Re x^*x. \quad (2)$$

For any  $x \in S$ , write  $|x^*x| = \omega x^*x$  with  $|\omega| = 1$ . Then by (2), since  $\omega x \in S$  whenever  $x \in S$  and  $\|x^{**}\| \leq 1$ ,

$$|x^*x| = x^*(\omega x) = \Re x^*(\omega x) < \lambda < \Re x^{**}x^* \leq |x^{**}x^*| \leq \|x^*\|.$$

Hence

$$\|x^*\| = \sup_{x \in S} |x^*x| \leq \lambda < \|x^*\|,$$

contradiction. □

In contrast to Theorem 5.24, we have

**Theorem 5.26.** *The closed unit ball  $S$  of a Banach space  $X$  is weakly compact if and only if  $X$  is reflexive.*

**Proof.** Observe that  $\kappa : S \rightarrow S^{**}$  is continuous when  $S$  and  $S^{**}$  are endowed with the (relative) *weak* and *weak\** topologies, respectively. (The net  $\{x_i\}_{i \in I}$  converges weakly to  $x$  in  $S$  if and only if  $x^*x_i \rightarrow x^*x$  for all  $x^* \in X^*$ , i.e.  $\hat{x}_i(x^*) \rightarrow \hat{x}(x^*)$  for all  $x^*$ , which is equivalent to  $\kappa(x_i) \rightarrow \kappa(x)$  *weak\**. This shows actually that  $\kappa$  is a homeomorphism of  $S$  with the relative *weak* topology and  $\kappa S$  with the relative *weak\** topology.) Therefore, if we assume that  $S$  is weakly compact, it follows that  $\kappa S$  is *weak\**-compact. Thus  $\kappa S$  is *weak\**-closed, and since it is *weak\**-dense in  $S^{**}$  (by Theorem 5.25), it follows that  $\kappa S = S^{**}$ . By linearity of  $\kappa$ , we then conclude that  $\kappa X = X^{**}$ , so that  $X$  is reflexive.

Conversely, if  $X$  is reflexive,  $\kappa S = S^{**}$  (since  $\kappa$  is norm-preserving). But  $S^{**}$  is *weak\**-compact by Theorem 5.24 (applied to the conjugate space  $X^{**}$ ). Since  $\kappa$  is a homeomorphism of  $S$  (with the weak topology) and  $\kappa S$  (with the *weak\** topology), as we observed above, it follows that  $S$  is weakly compact. □

It is natural to ask about compactness of  $S$  in the strong topology (i.e. the norm-topology). We have

**Theorem 5.27.** *The strongly closed unit ball  $S$  of a Banach space  $X$  is strongly compact if and only if  $X$  is finite dimensional.*

**Proof.** If  $X$  is an  $n$ -dimensional Banach space, there exists a (linear) homeomorphism  $\tau : X \rightarrow \mathbb{C}^n$ . Then  $\tau S$  is closed and bounded in  $\mathbb{C}^n$ , hence compact, by the Heine–Borel theorem. Therefore  $S$  is compact in  $X$ .

Conversely, if  $S$  is (strongly) compact, its open covering  $\{B(x, 1/2); x \in S\}$  by balls has a finite subcovering  $\{B(x_k, 1/2); k = 1, \dots, n\}$  ( $x_k \in S$ ). Let  $Y$  be the linear span of the vectors  $x_k$ . Then  $Y$  is a closed subspace of  $X$  of dimension

$\leq n$ . Suppose  $x^* \in X^*$  vanishes on  $Y$ . Given  $x \in S$ , there exists  $k, 1 \leq k \leq n$ , such that  $x \in B(x_k, 1/2)$ . Then

$$|x^*x| = |x^*(x - x_k)| \leq \|x^*\| \|x - x_k\| \leq \|x^*\|/2,$$

and therefore

$$\|x^*\| = \sup_{x \in S} |x^*x| \leq \|x^*\|/2,$$

this shows that  $\|x^*\| = 0$ , and so  $X = Y$  by Corollary 5.4.  $\square$

It follows from Theorems 5.24 and 5.27 that the weak\*-topology is strictly weaker than the strong topology on  $X^*$  when  $X$  (hence  $X^*$ ) is infinite dimensional. Similarly, by Theorems 5.26 and 5.27, the weak topology on an infinite-dimensional reflexive Banach space is strictly weaker than the strong topology.

## 5.6 Extremal points

As an application of the strict separation theorem (cf. Corollary 5.21), we shall prove the Krein–Milman theorem on extremal points.

Let  $X$  be a vector space (over  $\mathbb{C}$  or  $\mathbb{R}$ ). If  $x, y \in X$ , denote

$$\overline{xy} := \{\alpha x + (1 - \alpha)y; 0 < \alpha < 1\}.$$

Let  $K \subset X$ . A non-empty subset  $A \subset K$  is *extremal* in  $K$  if

$$[x, y \in K; \overline{xy} \cap A \neq \emptyset] \quad \text{implies} \quad [x, y \in A].$$

If  $A = \{a\}$  (a singleton) is extremal in  $K$ , we say that  $a$  is an *extremal point* of  $K$ : the criterion for this is

$$[x, y \in K; a \in \overline{xy}] \quad \text{implies} \quad [x = y = a].$$

Trivially, any non-empty  $K$  is extremal in itself. If  $B$  is extremal in  $A$  and  $A$  is extremal in  $K$ , then  $B$  is extremal in  $K$ . The non-empty intersection of a family of extremal sets in  $K$  is an extremal set in  $K$ .

From now on, let  $X$  be a *locally convex t.v.s.* and  $K \subset X$ . If  $A$  is a compact extremal set in  $K$  which contains no proper compact extremal subset, we call it a *minimal compact extremal set* in  $K$ .

**Lemma 5.28.** *A minimal compact extremal set  $A$  in  $K$  is a singleton.*

**Proof.** Suppose  $A$  contains two distinct points  $a, b$ . There exists  $x^* \in X^*$  such that  $f := \Re x^*$  assumes distinct values at these points (cf. Corollary 5.21). Let  $\rho = \min_A f$ ; since  $A$  is compact, the minimum  $\rho$  is attained on a non-empty subset  $B \subset A$ , and  $B \neq A$  since  $f$  is not constant on  $A$  ( $f(a) \neq f(b)$ !). The set  $B$  is a *closed* subset of the compact set  $A$ , and is therefore compact. We complete the proof by showing that  $B$  is extremal in  $A$  (hence in  $K$ , contradicting the

minimality assumption on  $A$ ). Let  $x, y \in A$  be such that  $\overline{xy} \cap B \neq \emptyset$ . Then there exists  $\alpha \in (0, 1)$  such that

$$\rho = f(\alpha x + (1 - \alpha)y) = \alpha f(x) + (1 - \alpha)f(y). \quad (1)$$

We have  $f(x) \geq \rho$  and  $f(y) \geq \rho$ ; if either inequality is strict, we get the contradiction  $\rho > \rho$  in (1). Hence  $f(x) = f(y) = \rho$ , that is,  $x, y \in B$ .  $\square$

**Lemma 5.29.** *If  $K \neq \emptyset$  is compact, then it has extremal points.*

**Proof.** Let  $\mathcal{A}$  be the family of all compact extremal subsets of  $K$ . It is non-empty, since  $K \in \mathcal{A}$ , and partially ordered by set inclusion. If  $\mathcal{B} \subset \mathcal{A}$  is totally ordered, then  $\bigcap \mathcal{B}$  is a non-empty compact extremal set in  $K$ , that is, belongs to  $\mathcal{A}$ , and is a lower bound for  $\mathcal{B}$ . By Zorn's lemma,  $\mathcal{A}$  has a minimal element, which is a singleton  $\{a\}$  by Lemma 5.28. Thus  $K$  has the extremal point  $a$ .  $\square$

If  $E \subset X$ , its *closed convex hull*  $\overline{\text{co}}(E)$  is defined as the closure of its convex hull  $\text{co}(E)$ .

**Theorem 5.30 (Krein–Milman's theorem).** *Let  $X$  be a locally convex t.v.s., and let  $K \subset X$  be compact. Let  $E$  be the set of extremal points of  $K$ . Then  $K \subset \overline{\text{co}}(E)$ .*

**Proof.** We may assume  $K \neq \emptyset$ , and therefore  $E \neq \emptyset$  by Lemma 5.29. Hence  $N := \overline{\text{co}}(E)$  is a non-empty closed convex set. Suppose there exists  $x \in K$  such that  $x \notin N$ . By Corollary 5.21, there exists  $x^* \in X^*$  such that  $f(x) < \inf_N f$  (where  $f = \Re x^*$ ). Let

$$B = \{k \in K; f(k) = \rho := \min_K f\}.$$

Then  $B$  is extremal in  $K$  (cf. proof of Lemma 5.28). Also  $B$  is a non-empty closed subset of the compact set  $K$ , hence is a non-empty compact set, and has therefore an extremal point  $b$ , by Lemma 5.29. Therefore,  $b$  is an extremal point of  $K$ , that is,  $b \in E \subset N$ . Hence

$$\rho = f(b) \geq \inf_N f > f(x) \geq \min_K f = \rho,$$

contradiction.  $\square$

**Corollary 5.31.** *(With assumptions and notation as in Theorem 5.30.)*

$$\overline{\text{co}}(K) = \overline{\text{co}}(E).$$

*In particular, a compact convex set in a locally convex t.v.s. is the closed convex hull of its extremal points.*

Consider for example the strongly closed unit ball  $S^*$  of the conjugate  $X^*$  of a normed space  $X$ . By Theorem 5.24, it is *weak*<sup>\*</sup>-compact and trivially convex. It is therefore the *weak*<sup>\*</sup>-closed convex hull of its extremal points.



The above remark will be applied as follows. Let  $X$  be a compact Hausdorff space, let  $C(X)$  be the space of all complex continuous functions on  $X$ , and let  $M(X)$  be the space of all regular complex Borel measures on  $X$  (cf. Theorem 4.9). By Theorem 4.9 (for  $X$  compact!), the space  $M(X)$  is isometrically isomorphic to the dual space  $C(X)^*$ . Its strongly closed unit ball  $M(X)_1$  is then *weak\**-compact (by Theorem 5.24).

If  $\mathcal{A}$  is any subset of  $C(X)$ , let

$$Y := \left\{ \mu \in M(X); \int_X f d\mu = 0 \quad (f \in \mathcal{A}) \right\}. \quad (2)$$

Clearly  $Y$  is *weak\**-closed, and therefore  $K := Y \cap M(X)_1$  is *weak\**-compact and trivially convex. It follows from Corollary 5.31 that  $K$  is the *weak\**-closed convex hull of its extremal points.

If  $\mathcal{A}$  is a closed subspace of  $C(X)$  ( $\mathcal{A} \neq C(X)$ ), it follows from Corollary 5.4 and Theorem 4.9 that  $Y \neq \{0\}$  (and  $K$  is the strongly closed unit ball of the Banach space  $Y$ ).

**Lemma 5.32.** *Let  $S$  be the strongly closed unit ball of a normed space  $Y \neq \{0\}$ , and let  $a$  be an extremal point of  $S$ . Then  $\|a\| = 1$ .*

**Proof.** Since  $Y \neq \{0\}$ , there exists  $0 \neq y \in S$ . Then  $0 \neq -y \in S$  and  $0 = (1/2)y + (1/2)(-y)$ , so that  $0$  is not extremal for  $S$ . Therefore  $a \neq 0$ . Define then  $b = a/\|a\| (\in S)$ . If  $\|a\| < 1$ , write  $a = \|a\|b + (1 - \|a\|)0$ , which is a proper convex combination of two elements of  $S$  distinct from  $a$ , contradicting the hypothesis that  $a$  is extremal. Hence  $\|a\| = 1$ .  $\square$

With notations and hypothesis as in the paragraph preceding the lemma, let  $\mu \in K$  be an extremal point of  $K$  (so that  $\|\mu\| = 1$ ), and let  $E = \text{supp}|\mu|$  (cf. Definition 3.26). Then  $E \neq \emptyset$  (since  $\|\mu\| = 1$ !), and by Remark 4.10,

$$\int_X f d\mu = \int_E f d\mu \quad (f \in C(X)). \quad (3)$$

**Lemma 5.33.** *Let  $\mathcal{A} \neq C(X)$  be a closed subalgebra of  $C(X)$  containing the identity 1. For  $K$  as above, let  $\mu$  be an extremal point of  $K$ , and let  $E = \text{supp}|\mu|$ . If  $f \in \mathcal{A}$  is real on  $E$ , then  $f$  is constant on  $E$ .*

**Proof.** Assume first that  $f \in \mathcal{A}$  has range in  $(0, 1)$  over  $E$ , and consider the measures  $d\sigma = f d\mu$  and  $d\tau = (1 - f) d\mu$ . Write  $d\mu = h d|\mu|$  with  $h$  measurable and  $|h| = 1$  (cf. Theorem 1.46). By Theorem 1.47,

$$d|\sigma| = |fh| d|\mu| = f d|\mu|; \quad d|\tau| = |(1 - f)h| d|\mu| = (1 - f) d|\mu|, \quad (4)$$

hence

$$\|\sigma\| = \int_X f d|\mu| = \int_E f d|\mu|, \quad (5)$$

and similarly

$$\|\tau\| = \int_E (1 - f) d|\mu|. \quad (6)$$

Therefore

$$\|\sigma\| + \|\tau\| = \int_E d|\mu| = |\mu|(E) = \|\mu\| = 1. \quad (7)$$

Since  $f$  and  $1 - f$  do not vanish identically on the support  $E$  of  $|\mu|$ , it follows from (5) and (6) and the discussion in Section 3.26 that  $\|\sigma\| > 0$  and  $\|\tau\| > 0$ . The measures  $\sigma' = \sigma/\|\sigma\|$  and  $\tau' = \tau/\|\tau\|$  are in  $M(X)_1$ , and for all  $g \in \mathcal{A}$ ,

$$\int_X g d\sigma' = \frac{1}{\|\sigma\|} \int_X g f d\mu = 0,$$

and similarly

$$\int_X g d\tau' = \frac{1}{\|\tau\|} \int_X g(1 - f) d\mu = 0,$$

since  $\mathcal{A}$  is an algebra. This means that  $\sigma'$  and  $\tau'$  are in  $K$ , and clearly

$$\mu = \|\sigma\|\sigma' + \|\tau\|\tau',$$

which is (by (7)) a proper convex combination. Since  $\mu$  is an extremal point of  $K$ , it follows that  $\mu = \sigma'$ . Therefore

$$\int_X g(f - \|\sigma\|) d\mu = 0$$

for all bounded Borel functions  $g$  on  $X$ . Choose in particular  $g = (f - \|\sigma\|)\bar{h}$ . Then

$$\int_E (f - \|\sigma\|)^2 d|\mu| = 0,$$

and consequently (cf. discussion following Definition 3.26)  $f = \|\sigma\|$  identically on  $E$ .

If  $f \in \mathcal{A}$  is real on  $E$ , there exist  $\alpha \in \mathbb{R}$  and  $\beta > 0$  such that  $0 < \beta(f - \alpha) < 1$  on  $E$ . Since  $1 \in \mathcal{A}$ , the function  $f_0 := \beta(f - \alpha)$  belongs to  $\mathcal{A}$  and has range in  $(0, 1)$ . By the first part of the proof,  $f_0$  is constant on  $E$ , and therefore  $f$  is constant on  $E$ .  $\square$

A non-empty subset  $E \subset X$  with the property of the conclusion of Lemma 5.33 (i.e. any function of  $\mathcal{A}$  that is real on  $E$  is necessarily constant on  $E$ ) is called an *antisymmetric set* (for  $\mathcal{A}$ ). If  $x, y \in X$  are contained in some antisymmetric set, they are said to be equivalent (with respect to  $\mathcal{A}$ ). This is an equivalence relation. Let  $\mathcal{E}$  be the family of all equivalence classes. If  $E$  is antisymmetric, and  $\tilde{E}$  is the equivalence class of any  $p \in E$ , then  $E \subset \tilde{E} \in \mathcal{E}$ .

**Theorem 5.34 (Bishop's antisymmetry theorem).** *Let  $X$  be a compact Hausdorff space. Let  $\mathcal{A}$  be a closed subalgebra of  $C(X)$  containing the constant function 1. Define  $\mathcal{E}$  as above. Suppose  $g \in C(X)$  has the property:*

(B) *For each  $E \in \mathcal{E}$ , there exists  $f \in \mathcal{A}$  such that  $g = f$  on  $E$ .*

*Then  $g \in \mathcal{A}$ .*

**Proof.** We may assume that  $\mathcal{A} \neq C(X)$ . Suppose that  $g \in C(X)$  satisfies Condition (B), but  $g \notin \mathcal{A}$ . By the Hahn–Banach theorem and the Riesz Representation theorem, there exists  $\nu \in M(X)_1$  such that  $\int_X h d\nu = 0$  for all  $h \in \mathcal{A}$  and  $\int_X g d\nu \neq 0$ . In particular  $K \neq \emptyset$  (since  $\nu \in K$ ), and is the *weak\**-closed convex hull of its extremal points. Let  $\mu$  be an extremal point of  $K$ . By Lemma 5.33, the set  $E := \text{supp}|\mu|$  is antisymmetric for  $\mathcal{A}$ . Let  $\tilde{E} \in \mathcal{E}$  be as defined above. By Condition (B), there exists  $f \in \mathcal{A}$  such that  $g = f$  on  $\tilde{E}$ , hence on  $\text{supp}|\mu| := E \subset \tilde{E}$ . Therefore

$$\int_X g d\mu = \int_X f d\mu = 0$$

since  $\mu \in K \subset Y$  and  $f \in \mathcal{A}$ . Since  $K$  is the *weak\**-closed convex hull of its extremal points and  $\nu \in K$ , it follows that  $\int_X g d\nu = 0$ , contradiction!  $\square$

We say that  $\mathcal{A}$  is *selfadjoint* if  $\bar{f} \in \mathcal{A}$  whenever  $f \in \mathcal{A}$ . This implies of course that  $\Re f$  and  $\Im f$  are in  $\mathcal{A}$  whenever  $f \in \mathcal{A}$  (and conversely);  $\mathcal{A}$  *separates points* (of  $X$ ) if whenever  $x, y \in X$  are distinct points, there exists  $f \in \mathcal{A}$  such that  $f(x) \neq f(y)$ .

**Corollary 5.35 (Stone–Weierstrass theorem).** *Let  $X$  be a compact Hausdorff space, and let  $\mathcal{A}$  be a closed selfadjoint subalgebra of  $C(X)$  containing 1 and separating points of  $X$ . Then  $\mathcal{A} = C(X)$ .*

**Proof.** If  $E$  is antisymmetric for  $\mathcal{A}$  and  $f \in \mathcal{A}$ , the real functions  $\Re f, \Im f \in \mathcal{A}$  are necessarily constant on  $E$ , and therefore  $f$  is constant on  $E$ . Since  $\mathcal{A}$  separates points, it follows that  $E$  is a singleton. Hence equivalent points must coincide, and so  $\mathcal{E}$  consists of all singletons. But then Condition (B) is trivially satisfied by *any*  $g \in C(X)$ : given  $\{p\} \in \mathcal{E}$ , choose  $f = g(p)1 \in \mathcal{A}$ , then surely  $g = f$  on  $\{p\}$ .  $\square$

## 5.7 The Stone–Weierstrass theorem

The Stone–Weierstrass theorem is one of the fundamental theorems of Functional Analysis, and it is worth giving it also an elementary proof, independent of the machinery developed above.

Let  $C_R(X)$  denote the algebra (over  $\mathbb{R}$ ) of all *real* continuous functions on  $X$ , and let  $\mathcal{A}$  be a subalgebra (over  $\mathbb{R}$ ). Since  $h(u) := u^{1/2} \in C_R([0, 1])$  and  $h(0) = 0$ , the classical Weierstrass Approximation theorem establishes the existence of polynomials  $p_n$  without free coefficient, converging to  $h$  uniformly on  $[0, 1]$ . Given  $f \in \mathcal{A}$ , the function  $u(x) := (f(x)^2/\|f\|^2) : X \rightarrow [0, 1]$  belongs to  $\mathcal{A}$ , and therefore  $p_n \circ u \in \mathcal{A}$  converge uniformly on  $X$  to  $h(u(x)) = |f(x)|/\|f\|$ . Hence  $|f| = \|f\| \cdot (|f|/\|f\|) \in \bar{\mathcal{A}}$ , where  $\bar{\mathcal{A}}$  denotes the closure of  $\mathcal{A}$  in  $C_R(X)$  with respect to the uniform norm.

If  $f, g \in \mathcal{A}$ , since

$$\max(f, g) = \frac{1}{2}[f + g + |f - g|]; \quad \min(f, g) = \frac{1}{2}[f + g - |f - g|],$$

it follows from the preceding conclusion that  $\max(f, g)$  and  $\min(f, g)$  belong to  $\bar{\mathcal{A}}$  as well.

Formally:

**Lemma 5.36.** *If  $\mathcal{A}$  is a subalgebra of  $C_R(X)$ , then  $|f|$ ,  $\max(f, g)$  and  $\min(f, g)$  belong to the uniform closure  $\bar{\mathcal{A}}$ , for any  $f, g \in \mathcal{A}$ .*

**Lemma 5.37.** *Let  $\mathcal{A}$  be a separating subspace of  $C_R(X)$  containing 1, then for any distinct points  $x_1, x_2 \in X$  and any  $\alpha_1, \alpha_2 \in \mathbb{R}$ , there exists  $h \in \mathcal{A}$  such that  $h(x_k) = \alpha_k, k = 1, 2$ .*

**Proof.** By hypothesis, there exists  $g \in \mathcal{A}$  such that  $g(x_1) \neq g(x_2)$ . Take

$$h(x) := \alpha_1 + \frac{\alpha_2 - \alpha_1}{g(x_2) - g(x_1)}[g(x) - g(x_1)].$$

□

We state now the Stone–Weierstrass theorem as an approximation theorem (real case first).

**Theorem 5.38.** *Let  $X$  be a compact Hausdorff space. Let  $\mathcal{A}$  be a separating subalgebra of  $C_R(X)$  containing 1. Then  $\mathcal{A}$  is dense in  $C_R(X)$ .*

**Proof.** Let  $f \in C_R(X)$  and  $\epsilon > 0$  be given. Fix  $x_0 \in X$ . For any  $x' \in X$ , there exists  $f' \in \mathcal{A}$  such that

$$f'(x_0) = f(x_0); \quad f'(x') \leq f(x') + \epsilon/2$$

(cf. Lemma 5.37).

By continuity of  $f$  and  $f'$ , there exists an open neighbourhood  $V(x')$  of  $x'$  such that  $f' \leq f + \epsilon$  on  $V(x')$ . By compactness of  $X$ , there exist  $x_k \in X, k = 1, \dots, n$ , such that

$$X = \bigcup_{k=1}^n V(x_k).$$

Let  $f_k \in \mathcal{A}$  be the function  $f'$  corresponding to the point  $x' = x_k$  as above, and let

$$g := \min(f_1, \dots, f_n).$$

Then  $g \in \bar{\mathcal{A}}$  by Lemma 5.36, and

$$g(x_0) = f(x_0); \quad g \leq f + \epsilon \quad \text{on } X.$$

By continuity of  $f$  and  $g$ , there exists an open neighbourhood  $W(x_0)$  of  $x_0$  such that

$$g \geq f - \epsilon \quad \text{on } W(x_0).$$

We now vary  $x_0$  (over  $X$ ). The open cover  $\{W(x_0); x_0 \in X\}$  of  $X$  has a finite subcover  $\{W_1, \dots, W_m\}$ , corresponding to functions  $g_1, \dots, g_m \in \bar{\mathcal{A}}$  as above. Thus

$$g_i \leq f + \epsilon \quad \text{on } X$$

and

$$g_i \geq f - \epsilon \quad \text{on } W_i.$$

Define

$$h = \max(g_1, \dots, g_m).$$

Then  $h \in \bar{\mathcal{A}}$  by Lemma 5.36, and

$$f - \epsilon \leq h \leq f + \epsilon \quad \text{on } X.$$

Therefore  $f \in \bar{\mathcal{A}}$ . □

**Theorem 5.39 (The Stone–Weierstrass Theorem).** *Let  $X$  be a compact Hausdorff space, and let  $\mathcal{A}$  be a separating selfadjoint subalgebra of  $C(X)$  containing 1. Then  $\mathcal{A}$  is dense in  $C(X)$ .*

**Proof.** Let  $\mathcal{A}_R$  be the algebra (over  $\mathbb{R}$ ) of all *real* functions in  $\mathcal{A}$ . It contains 1. Let  $x_1, x_2$  be distinct points of  $X$ , and let then  $h \in \mathcal{A}$  be such that  $h(x_1) \neq h(x_2)$ . Then either  $\Re h(x_1) \neq \Re h(x_2)$  or  $\Im h(x_1) \neq \Im h(x_2)$  (or both). Since  $\Re h = (h + \bar{h})/2 \in \mathcal{A}_R$  (since  $\mathcal{A}$  is selfadjoint), and similarly  $\Im h \in \mathcal{A}_R$ , it follows that  $\mathcal{A}_R$  is separating.

Let  $f \in C(X)$ . Then  $\Re f \in C_R(X)$ , and therefore, by Theorem 5.38, there exists a sequence  $g_n \in \mathcal{A}_R$  converging to  $\Re f$  uniformly on  $X$ . Similarly, there exists a sequence  $h_n \in \mathcal{A}_R$  converging to  $\Im f$  uniformly on  $X$ . Then  $g_n + ih_n \in \mathcal{A}$  converge to  $f$  uniformly on  $X$ . □

## 5.8 Operators between Lebesgue spaces: Marcinkiewicz's interpolation theorem

**5.40. Weak and strong types.** Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be measure spaces, and let  $p, q \in [1, \infty]$  (presently,  $q$  is *not* the conjugate exponent of  $p$ !) We consider operators  $T$  defined on  $L^p(\mu)$  such that  $Tf$  is a  $\mathcal{B}$ -measurable function on  $Y$  and

$$|T(f + g)| \leq |Tf| + |Tg| \quad (f, g \in L^p(\mu)). \quad (1)$$

We refer to such operators  $T$  as *sublinear operators*. (This includes linear operators with range as above.) Let  $n$  be the distribution function of  $|Tf|$  for some *fixed non-zero element*  $f$  of  $L^p(\mu)$ , relative to the measure  $\nu$  on  $\mathcal{B}$ , that is (cf. Section 1.58)

$$n(y) := \nu(|Tf| > y) \quad (y > 0).$$

We say that  $T$  is of *weak type*  $(p, q)$  if there exists a constant  $0 < M < \infty$  independent of  $f$  and  $y$  such that

$$n(y)^{1/q} \leq M \|f\|_p / y \quad (2)$$

in case  $q < \infty$ , and

$$\|Tf\|_\infty \leq M \|f\|_p \quad (3)$$

in case  $q = \infty$ . We say that  $T$  is of strong type  $(p, q)$  if there exists  $M$  (as above) such that

$$\|Tf\|_q \leq M \|f\|_p. \quad (4)$$

The concepts of weak and strong types  $(p, \infty)$  coincide by definition, while in general strong type  $(p, q)$  implies weak type  $(p, q)$ , because by (2) in Section 1.58 (for  $q < \infty$ )

$$n(y)^{1/q} \leq \|Tf\|_q/y \leq M \|f\|_p/y. \quad (5)$$

The infimum of all  $M$  for which (2) or (4) is satisfied is called the weak or strong  $(p, q)$ -norm of  $T$ , respectively. By (5), the weak  $(p, q)$ -norm of  $T$  is no larger than its strong  $(p, q)$ -norm.

A linear operator  $T$  is of strong type  $(p, q)$  iff  $T \in B(L^p(\nu), L^q(\nu))$ , and in that case, the strong  $(p, q)$ -norm is the corresponding operator norm of  $T$ .

In the sequel, we consider only the case  $q < \infty$ . If  $T$  is of weak type  $(p, q)$ , it follows from (2) that  $n$  is finite and  $n(\infty) = 0$ , and consequently, by Theorem 1.59,

$$\|Tf\|_q^q = q \int_0^\infty y^{q-1} n(y) dy, \quad (6)$$

where the two sides could be finite or infinite.

Let  $u > 0$ , and consider the decomposition  $f = f_u + f'_u$  as in Section 1.61. Since  $f_u$  and  $f'_u$  are both in  $L^p(\mu)$  (for  $f \in L^p(\mu)$ ), the given sublinear operator  $T$  is defined on them, and

$$|Tf| \leq |Tf_u| + |Tf'_u|. \quad (7)$$

Let  $n_u$  and  $n'_u$  denote the distribution functions of  $|Tf_u|$  and  $|Tf'_u|$ , respectively. Since by (7)

$$[|Tf| > y] \subset [|Tf_u| > y/2] \cup [|Tf'_u| > y/2] \quad (y > 0),$$

we have

$$n(y) \leq n_u(y/2) + n'_u(y/2). \quad (8)$$

We now assume that  $T$  is of weak types  $(r, s)$  and  $(a, b)$ , with  $r \leq s$  and  $a \leq b$ , and  $s \neq b$ . We consider the case  $r \neq a$  (without loss of generality,  $r > a$ ) and  $s > b$ . Denote the respective weak norms of  $T$  by  $M$  and  $N$ , and

$$s/r := \sigma (\geq 1), \quad b/a := \beta (\geq 1).$$

Let  $a < p < r$  and  $f \in L^p(\mu)$ . By Section 1.61,  $f_u \in L^r(\mu)$  and  $f'_u \in L^a(\mu)$ . Since  $T$  is of weak type  $(r, s)$  with weak  $(r, s)$ -norm  $M$ , we have

$$n_u(y) \leq M^s y^{-s} \|f_u\|_r^s. \quad (9)$$

Since  $T$  is of weak type  $(a, b)$  with weak  $(a, b)$ -norm  $N$ , we have

$$n'_u(y) \leq N^b y^{-b} \|f'_u\|_a^b. \quad (10)$$

By (8)–(10),

$$n(y) \leq (2M)^s y^{-s} \|f_u\|_r^s + (2N)^b y^{-b} \|f'_u\|_a^b. \quad (11)$$

Let  $b < q < s$ . By Theorem 1.59 and (11),

$$\begin{aligned} (1/q) \|Tf\|_q^q &= \int_0^\infty y^{q-1} n(y) dy \\ &\leq (2M)^s \int_0^\infty y^{q-s-1} \|f_u\|_r^s dy + (2N)^b \int_0^\infty y^{q-b-1} \|f'_u\|_a^b dy. \end{aligned} \quad (12)$$

Since  $a < p < r$ , we may apply Formulae (8) and (9) of Section 1.61; we then conclude from (12) that

$$\begin{aligned} (1/q) \|Tf\|_q^q &\leq (2M)^s r^\sigma \int_0^\infty y^{q-s-1} \left( \int_0^u v^{r-1} m(v) dv \right)^\sigma dy \\ &\quad + (2N)^b a^\beta \int_0^\infty y^{q-b-1} \left( \int_u^\infty (v-u)^{a-1} m(v) dv \right)^\beta dy. \end{aligned} \quad (13)$$

In this formula, we may also take  $u$  dependent monotonically on  $y$  (integrability is then clear). Denote the two integrals in (13) by  $\Phi$  and  $\Psi$ . Since  $\sigma, \beta \geq 1$ , it follows from Corollary 5.8 and Theorem 4.6 (applied to the spaces  $L^\sigma(\mathbb{R}^+, \mathcal{M}, y^{q-s-1} dy)$  and  $L^\beta(\mathbb{R}^+, \mathcal{M}, y^{q-b-1} dy)$  respectively, where  $\mathcal{M}$  is the Lebesgue  $\sigma$ -algebra over  $\mathbb{R}^+$ ) that

$$\Phi^{1/\sigma} = \sup \int_0^\infty y^{q-s-1} \left( \int_0^u v^{r-1} m(v) dv \right) g(y) dy, \quad (14)$$

where the supremum is taken over all measurable functions  $g \geq 0$  on  $\mathbb{R}^+$  such that  $\int_0^\infty y^{q-s-1} g^{\sigma'}(y) dy \leq 1$  ( $\sigma'$  denotes here the conjugate exponent of  $\sigma$ ).

Similarly

$$\Psi^{1/\beta} = \sup \int_0^\infty y^{q-b-1} \left( \int_u^\infty (v-u)^{a-1} m(v) dv \right) h(y) dy, \quad (15)$$

where the supremum is taken over all measurable function  $h \geq 0$  on  $\mathbb{R}^+$  such that  $\int_0^\infty y^{q-b-1} h^{\beta'}(y) dy \leq 1$  ( $\beta'$  is the conjugate exponent of  $\beta$ ).

We now choose  $u = (y/c)^k$ , where  $c, k$  are positive parameters to be determined later. By Tonelli's theorem, the integral in (14) is equal to

$$\int_0^\infty v^{r-1} m(v) \left( \int_{cv^{1/k}}^\infty y^{q-s-1} g(y) dy \right) dv.$$

By Holder's inequality on the measure space  $(\mathbb{R}^+, \mathcal{M}, y^{q-s-1} dy)$ , the inner integral is

$$\begin{aligned} &\leq \left( \int_{cv^{1/k}}^\infty y^{q-s-1} dy \right)^{1/\sigma} \left( \int_0^\infty y^{q-s-1} g^{\sigma'}(y) dy \right)^{1/\sigma'} \\ &\leq \left( \frac{y^{q-s}}{q-s} \Big|_{cv^{1/k}}^\infty \right)^{1/\sigma} = (s-q)^{-1/\sigma} c^{(q-s)/\sigma} v^{(q-s)/k\sigma}. \end{aligned}$$

Consequently

$$\Phi \leq (s - q)^{-1} c^{q-s} \left( \int_0^\infty v^{r-1+(q-s)/k\sigma} m(v) dv \right)^\sigma. \quad (16)$$

Similarly, the integral in (15) is equal to

$$\begin{aligned} & \int_0^\infty \left( \int_0^{cv^{1/k}} y^{q-b-1} h(y) [v - (y/c)^k]^{a-1} dy \right) m(v) dv \\ & \leq \int_0^\infty v^{a-1} m(v) \left( \int_0^{cv^{1/k}} y^{q-b-1} h(y) dy \right) dv \\ & \leq \int_0^\infty v^{a-1} m(v) \left( \int_0^{cv^{1/k}} y^{q-b-1} dy \right)^{1/\beta} \left( \int_0^{cv^{1/k}} y^{q-b-1} h^{\beta'}(y) dy \right)^{1/\beta'} dv \\ & \leq \int_0^\infty v^{a-1} m(v) \left( \frac{y^{q-b}}{q-b} \Big|_0^{cv^{1/k}} \right)^{1/\beta} dv \\ & = (q-b)^{-1/\beta} c^{(q-b)/\beta} \int_0^\infty v^{a-1+(q-b)/k\beta} m(v) dv. \end{aligned}$$

Therefore

$$\Psi \leq (q-b)^{-1} c^{q-b} \left( \int_0^\infty v^{a-1+(q-b)/k\beta} m(v) dv \right)^\beta. \quad (17)$$

Since  $b < q < s$ , the integrals in (16) and (17) contain the terms  $v^{\kappa-1}$  and  $v^{\lambda-1}$  respectively, with  $\kappa := r + (q-s)/k\sigma < r$  and  $\lambda := a + (q-b)/k\beta > a$ . Recall that we also have  $a < p < r$ . If we can choose the parameter  $k$  so that  $\kappa = \lambda = p$ , then by Corollary 1.60 both integrals will be equal to  $(1/p) \|f\|_p^p$ . Since  $\beta = b/a$  and  $\sigma = s/r$ , the unique solutions for  $k$  of the equations  $\kappa = p$  and  $\lambda = p$  are

$$k = (a/b) \frac{q-b}{p-a} \quad \text{and} \quad k = (r/s) \frac{s-q}{r-p}, \quad (18)$$

respectively, so that the above choice of  $k$  is possible iff the two expressions in (18) coincide. Multiplying both expressions by  $p/q$ , this condition on  $p, q$  can be rearranged as

$$\frac{(1/b) - (1/q)}{(1/a) - (1/p)} = \frac{(1/q) - (1/s)}{(1/p) - (1/r)},$$

or equivalently

$$\frac{(1/b) - (1/q)}{(1/q) - (1/s)} = \frac{(1/a) - (1/p)}{(1/p) - (1/r)}, \quad (19)$$

that is,  $1/p$  and  $1/q$  divide the segments  $[1/r, 1/a]$  and  $[1/s, 1/b]$ , respectively, according to the same (positive, finite) ratio. Equivalently,

$$\frac{1}{p} = (1-t) \frac{1}{r} + t \frac{1}{a}; \quad \frac{1}{q} = (1-t) \frac{1}{s} + t \frac{1}{b} \quad (20)$$

for some  $t \in (0, 1)$ .



With the above choice of  $k$ , it now follows from (13), (16), and (17) that

$$\begin{aligned} (1/q)\|Tf\|_q^q &\leq (2M)^s(r/p)^\sigma(s-q)^{-1}c^{q-s}\|f\|_p^{p\sigma} \\ &\quad + (2N)^b(a/p)^\beta(q-b)^{-1}c^{q-b}\|f\|_p^{p\beta}. \end{aligned} \quad (21)$$

We now choose the parameter  $c$  in the form

$$c = (2M)^x(2N)^z\|f\|_p^w,$$

with  $x, z, w$  real parameters to be determined so that the two summands on the right-hand side of (21) contain the same powers of  $M, N$ , and  $\|f\|_p$ , respectively. This yields to the following equations for the unknown parameters  $x, z, w$ :

$$s + (q-s)x = (q-b)x; \quad b + (q-b)z = (q-s)z; \quad p\sigma + (q-s)w = p\beta + (q-b)w.$$

The unique solution is

$$x = \frac{s}{s-b}; \quad z = \frac{-b}{s-b}; \quad w = p\frac{\sigma-\beta}{s-b}.$$

With the above choice of  $c$ , the right-hand side of (21) is equal to

$$[(r/p)^\sigma(s-q)^{-1} + (a/p)^\beta(q-b)^{-1}](2M)^{s(q-b)/(s-b)}(2N)^{b(s-q)/(s-b)}\|f\|_p^{p\gamma}, \quad (22)$$

where

$$\begin{aligned} \gamma &= \beta + (q-b)\frac{\sigma-\beta}{s-b} = \beta\frac{s-q}{s-b} + \sigma\frac{q-b}{s-b} \\ &= t(q/a) + (1-t)(q/r) = q/p, \end{aligned}$$

by the relations (20) and  $\beta = b/a, \sigma = s/r$ . By (20), the exponents of  $2M$  and  $2N$  in (22) are equal to  $(1-t)q$  and  $tq$ , respectively. By (21) and (22), we conclude that

$$\|Tf\|_q \leq KM^{1-t}N^t\|f\|_p,$$

where

$$K := 2q^{1/q}[(r/p)^{s/r}(s-q)^{-1} + (a/p)^{b/a}(q-b)^{-1}]^{1/q}$$

does not depend on  $f$ . Thus  $T$  is of strong type  $(p, q)$ , with strong  $(p, q)$ -norm  $\leq KM^{1-t}N^t$ . Note that the constant  $K$  depends only on the parameters  $a, r, b, s$  and  $p, q$ ; it tends to  $\infty$  when  $q$  approaches either  $b$  or  $s$ .

A similar argument (which we shall omit) yields the same conclusion in case  $s < b$ . The result (which we proved for  $r \neq a$  and finite  $b, s$ ) is also valid when  $r = a$  and one or both exponents  $b, s$  are infinite (again, we omit the details). We formalize our conclusion in the following

**Theorem 5.41 (Marcinkiewicz's interpolation theorem).** *Let  $1 \leq a \leq b \leq \infty$  and  $1 \leq r \leq s \leq \infty$ . For  $0 < t < 1$ , let  $p, q$  be such that*

$$\frac{1}{p} = (1-t)\frac{1}{r} + t\frac{1}{a} \quad \text{and} \quad \frac{1}{q} = (1-t)\frac{1}{s} + t\frac{1}{b}. \quad (23)$$

Suppose that the sublinear operator  $T$  is of weak types  $(r, s)$  and  $(a, b)$ , with respective weak norms  $M$  and  $N$ . Then  $T$  is of strong type  $(p, q)$ , with strong  $(p, q)$ -norm  $\leq KM^{1-t}N^t$ .

The constant  $K$  depends only on the parameters  $a, b, r, s$  and  $t$ . For  $a, b, r, s$  fixed,  $K = K(t)$  is bounded for  $t$  bounded away from the end points  $0, 1$ .

**Corollary 5.42.** *Let  $a, b, r, s, t$  be as in Theorem 5.41. For any  $p, q \in [1, \infty]$ , denote the strong  $(p, q)$ -norm of the sublinear operator  $T$  by  $\|T\|_{p,q}$ . If  $T$  is of strong types  $(r, s)$  and  $(a, b)$ , then  $T$  is of strong type  $(p, q)$  whenever (23) is satisfied, and*

$$\|T\|_{p,q} \leq K \|T\|_{r,s}^{1-t} \|T\|_{a,b}^t,$$

with  $K$  as in Theorem 5.41.

In particular,

$$\begin{aligned} B(L^r(X, \mathcal{A}, \mu), L^s(Y, \mathcal{B}, \nu)) \cap B(L^a(X, \mathcal{A}, \mu), L^b(Y, \mathcal{B}, \nu)) \\ \subset B(L^p(X, \mathcal{A}, \mu), L^q(Y, \mathcal{B}, \nu)). \end{aligned}$$

## Exercises

1. A Banach space  $X$  is *separable* if it contains a countable dense subset. Prove that if  $X^*$  is separable, then  $X$  is separable (but the converse is false). Hint: let  $\{x_n^*\}$  be a sequence of unit vectors dense in the unit sphere of  $X^*$ . Pick unit vectors  $x_n$  in  $X$  such that  $|x_n^* x_n| > 1/2$ . Use Corollary 5.5 to show that  $\text{span } \{x_n\}$  is dense in  $X$ ; the same is true when the scalars are complex numbers with *rational* real and imaginary parts.
2. Consider the normed space

$$C_c^n(\mathbb{R}) := \{f \in C_c(\mathbb{R}); f^{(k)} \in C_c(\mathbb{R}), k = 1, \dots, n\},$$

with the norm

$$\|f\| = \sum_{k=0}^n \|f^{(k)}\|_u.$$

Given  $\phi \in C_c^n(\mathbb{R})$ , prove that there exist complex Borel measures  $\mu_k$  ( $k = 0, \dots, n$ ) such that

$$\phi(f) = \sum_k \int_{\mathbb{R}} f^{(k)} d\mu_k$$

for all  $f \in C_c^n(\mathbb{R})$ . Hint: consider the subspace

$$Z = \{[f, f', \dots, f^{(n)}]; f \in C_c^n(\mathbb{R})\}$$

of  $C_c \times \dots \times C_c$  ( $n+1$  times). Define  $\psi$  on  $Z$  by  $\psi([f, f', \dots, f^{(n)}]) = \phi(f)$ . (Cf. Exercise 3, Chapter 4.)

3. Let  $X$  be a Banach space, and let  $\Gamma \subset X^*$  be (norm) bounded and *weak\**-closed. Prove:
  - (a)  $\Gamma$  is *weak\**-compact.
  - (b) If  $\Gamma$  is also convex, then it is the *weak\**-closed convex hull of its extremal points.
4. Let  $X, Y$  be normed spaces and  $T \in B(X, Y)$ . Prove that  $T$  is continuous with respect to the weak topologies on  $X$  and  $Y$ , and  $T^* : Y^* \rightarrow X^*$  is continuous with respect to the *weak\**-topologies on  $Y^*$  and  $X^*$ . (Recall that the Banach adjoint  $T^*$  of  $T \in B(X)$  is defined by means of the identity  $(T^*y^*)x = y^*(Tx)$ ,  $x \in X, y^* \in Y^*$ .)
5. Let  $p, q \in [1, \infty]$  be conjugate exponents, and let  $(X, \mathcal{A}, \mu)$  be a positive measure space. Let  $g$  be a complex measurable function on  $X$  such that  $\|g\|_q \leq M$  for some constant  $M$ . Then  $\|fg\|_1 \leq M\|f\|_p$  for all  $f \in L^p(\mu)$  (by Theorems 1.26 and 1.33). *Prove the converse!*

## Uniform convexity

6. Let  $X$  be a normed space, and let  $B$  and  $S$  denote its closed unit ball and its unit sphere, respectively. We say that  $X$  is *uniformly convex* (u.c.) if for each  $\epsilon > 0$  there exists  $\delta = \delta(\epsilon) > 0$  such that  $\|x - y\| < \epsilon$  whenever  $x, y \in B$  are such that  $\|(x + y)/2\| > 1 - \delta$ . Prove:
  - (a)  $X$  is u.c. iff whenever  $x_n, y_n \in S$  are such that  $\|x_n + y_n\| \rightarrow 2$ , it follows that  $\|x_n - y_n\| \rightarrow 0$ .
  - (b) Every inner product space is u.c.
  - (c) Let  $X$  be a u.c. normed space and  $\{x_n\} \subset X$ . Then  $x_n \rightarrow x \in X$  strongly iff  $x_n \rightarrow x$  weakly and  $\|x_n\| \rightarrow \|x\|$ . (Hint: suppose  $x_n \rightarrow x$  weakly and  $\|x_n\| \rightarrow \|x\|$ . We may assume that  $x_n, x \in S$ . Pick  $x_0^* \in X^*$  such that  $x_0^*x = 1 = \|x_0^*\|$ , cf. Corollary 5.7.)
  - (d) The ‘distance theorem’ (Theorem 1.35) is valid for a u.c. Banach space  $X$ .  
*The following parts are steps in the proof of the result stated in Part (i) below.*
  - (e) Let  $X$  be a u.c. Banach space, and let  $\epsilon, \delta$  be as in the definition above. Denote by  $S^*$  and  $S^{**}$  the unit spheres of  $X^*$  and  $X^{**}$ , respectively. Given  $x_0^{**} \in S^{**}$ , there exists  $x_0^* \in S^*$  such that  $|x_0^{**}x_0^* - 1| < \delta$ . Also there exists  $x \in B$  such that  $|x_0^*x - 1| < \delta$ . Define

$$E_\delta = \{x \in B; |x_0^*x - 1| < \delta\} (\neq \emptyset).$$

Show that  $\|x - y\| < \epsilon$  for all  $x, y \in E_\delta$ .

- (f) In any normed space  $X$ , the set

$$U := \{x^{**} \in X^{**}; |x^{**}x_0^* - 1| < \delta\}$$

is a *weak\**-neighbourhood of  $x_0^{**}$ .

- (g) For any *weak\**-neighbourhood  $V$  of  $x_0^{**}$ , the *weak\**-neighbourhood  $W := V \cap U$  of  $x_0^{**}$  meets  $\kappa B$ . ( $\kappa$  denotes the canonical imbedding of  $X$  in  $X^{**}$ .) Thus  $V \cap \kappa(E_\delta) \neq \emptyset$ , and therefore  $x_0^{**}$  belongs to the *weak\**-closure of  $\kappa(E_\delta)$ . (Cf. Goldstine's theorem.)
- (h) Fix  $x \in E_\delta$ . Then  $x_0^{**} \in \kappa x + \epsilon B^{**}$ , where  $B^{**}$  denotes the (norm) closed unit ball of  $X^{**}$ . (Hint: apply Parts (e) and (g), and the fact that  $B^{**}$  is *weak\**-compact, hence *weak\**-closed.)
- (i) Conclude from Part (h) that  $d(x_0^{**}, \kappa B) = 0$ , and therefore  $x_0^{**} \in \kappa B$  since  $\kappa B$  is norm-closed in  $X^{**}$  (cf. paragraph preceding Theorem 5.9). This proves the following theorem: *uniformly convex Banach spaces are reflexive*.

7. Let  $\{\beta_n\}_{n=0}^\infty \in l^\infty$  be such that there exists a positive constant  $K$  for which

$$\left| \sum_{n=0}^N \alpha_n \beta_n \right| \leq K \max_{t \in [0, 1]} \left| \sum_{n=0}^N \alpha_n t^n \right|.$$

for all  $\alpha_0, \dots, \alpha_N \in \mathbb{C}$  and  $N = 0, 1, 2, \dots$ . Prove that there exists a unique regular complex Borel measure  $\mu$  on  $[0, 1]$  such that  $\beta_n = \int_0^1 t^n d\mu$  for all  $n = 0, 1, 2, \dots$ . Moreover  $\|\mu\| \leq K$ . Formulate and prove the converse.

8. Prove the converse of Theorem 4.4.

# 6

## Bounded operators

We recall that  $B(X, Y)$  denotes the normed space of all bounded linear mappings from the normed space  $X$  to the normed space  $Y$ . The norm on  $B(X, Y)$  is the *operator norm*

$$\|T\| := \sup_{\|x\| \leq 1} \|Tx\| = \sup_{\|x\| < 1} \|Tx\| = \sup_{\|x\|=1} \|Tx\| = \sup_{0 \neq x \in X} \frac{\|Tx\|}{\|x\|}.$$

The elements of  $B(X, Y)$  will be referred to briefly as *operators*.

Two of the basic theorems about  $B(X, Y)$ , the *Uniform Boundedness Theorem* and the *Open Mapping Theorem*, use so-called *category arguments* in their proofs. These are based on Baire's theorem about complete metric spaces.

### 6.1 Category

**Theorem 6.1 (Baire's theorem).** *Let  $X$  be a complete metric space (with metric  $d$ ), and let  $\{V_i\}$  be a sequence of open dense subsets of  $X$ . Then  $V := \bigcap_{i=1}^{\infty} V_i$  is dense in  $X$ .*

**Proof.** Let  $U$  be a non-empty open subset of  $X$ . We must show that  $U \cap V \neq \emptyset$ .

Since  $V_1$  is dense, it follows that the open set  $U \cap V_1$  is non-empty, and we may then choose a closed ball  $\bar{B}(x_1, r_1) := \{x \in X; d(x, x_1) \leq r_1\} \subset U \cap V_1$  with radius  $r_1 < 1$ . Let  $B(x_1, r_1) := \{x \in X; d(x, x_1) < r_1\}$  be the corresponding open ball. Since  $V_2$  is dense, it follows that the open set  $B(x_1, r_1) \cap V_2$  is non-empty, and we may then choose a closed ball  $\bar{B}(x_2, r_2) \subset B(x_1, r_1) \cap V_2$  with radius  $r_2 < 1/2$ . Continuing inductively, we obtain a sequence of balls  $B(x_n, r_n)$  with  $r_n < 1/n$ , such that

$$\bar{B}(x_n, r_n) \subset B(x_{n-1}, r_{n-1}) \cap V_n \quad \text{for } n = 2, 3, \dots$$

If  $i, j > n$ , we have  $x_i, x_j \in B(x_n, r_n)$ , and therefore  $d(x_i, x_j) < 2r_n < 2/n$ . This means that  $\{x_i\}$  is a Cauchy sequence. Since  $X$  is complete, the sequence

converges to some  $x \in X$ . For  $i \geq n$ ,  $x_i \in \bar{B}(x_n, r_n)$  (a closed set!), and therefore  $x \in \bar{B}(x_n, r_n) \subset U \cap V_n$  for all  $n$ , that is,  $x \in U \cap V$ .  $\square$

**Definition 6.2.** A subset  $E$  of a metric space  $X$  is *nowhere dense* if its closure  $\bar{E}$  has empty interior. A countable union of nowhere dense sets in  $X$  is called a set of (*Baire's*) *first category* in  $X$ . A subset of  $X$  which is *not* of first category in  $X$  is said to be of (*Baire's*) *second category* in  $X$ .

The family of subsets of first category is closed under countable unions. Subsets of sets of first category are of first category.

Using category terminology, Baire's theorem has the following variant form, which is the basis for the 'category arguments' mentioned above.

**Theorem 6.3 (Baire's category theorem).** *A complete metric space is of Baire's second category in itself.*

**Proof.** Suppose the complete metric space  $X$  is of Baire's first category in itself. Then

$$X = \bigcup_i E_i$$

with  $E_i$  nowhere dense ( $i = 1, 2, \dots$ ). Hence  $X = \bigcup_i \bar{E}_i$ . Taking complements, we see that

$$\bigcap_i (\bar{E}_i)^c = \emptyset.$$

Since  $E_i$  are nowhere dense, the sets in the above intersection are open dense sets. By Baire's theorem, the intersection is dense, a contradiction.  $\square$

## 6.2 The uniform boundedness theorem

**Theorem 6.4 (The uniform boundedness theorem, general version).** *Let  $X, Y$  be normed spaces, and let  $\mathcal{T} \subset B(X, Y)$ . Suppose the subspace*

$$Z := \{x \in X; \sup_{T \in \mathcal{T}} \|Tx\| < \infty\}$$

*is of Baire's second category in  $X$ . Then*

$$\sup_{T \in \mathcal{T}} \|T\| < \infty.$$

**Proof.** Denote

$$r(x) := \sup_{T \in \mathcal{T}} \|Tx\| \quad (x \in Z).$$

If  $S_Y := \overline{B_Y}(0, 1)$  is the closed unit ball in  $Y$ , then for all  $T \in \mathcal{T}$ ,  $Tx \in r(x)S_Y \subset nS_Y$  if  $x \in Z$  and  $n$  is an integer  $\geq r(x)$ . Thus  $T(x/n) \in S_Y$  for all  $T \in \mathcal{T}$ , that is,

$$x/n \in \bigcap_{T \in \mathcal{T}} T^{-1}S_Y := E$$

for  $n \geq r(x)$ . This shows that

$$Z \subset \bigcup_n nE.$$

Since  $Z$  is of Baire's second category in  $X$  and  $nE$  are closed sets (by continuity of each  $T \in \mathcal{T}$ ), there exists  $n$  such that  $nE$  has non-empty interior. However, multiplication by scalars is a homeomorphism; therefore  $E$  has non-empty interior  $E^\circ$ . Let then  $B_X(a, \delta) \subset E$ . Then for all  $T \in \mathcal{T}$ ,

$$\begin{aligned} \delta TB_X(0, 1) &= TB_X(0, \delta) = T[B_X(a, \delta) - a] = TB_X(a, \delta) - Ta \\ &\subset TE - Ta \subset S_Y - S_Y \subset 2S_Y. \end{aligned}$$

Hence  $TB_X(0, 1) \subset (2/\delta)S_Y$  for all  $T \in \mathcal{T}$ . This means that

$$\sup_{T \in \mathcal{T}} \|T\| \leq \frac{2}{\delta}.$$

□

If  $X$  is *complete*, it is of Baire's second category in itself by Theorem 6.3. This implies the non-trivial part of the following.

**Corollary 6.5 (The uniform boundedness theorem).** *Let  $X$  be a Banach space,  $Y$  a normed space, and  $\mathcal{T} \subset B(X, Y)$ . Then the following two statements are equivalent*

$$\sup_{T \in \mathcal{T}} \|Tx\| < \infty \quad \text{for all } x \in X; \tag{i}$$

$$\sup_{T \in \mathcal{T}} \|T\| < \infty. \tag{ii}$$

**Corollary 6.6.** *Let  $X$  be a Banach space,  $Y$  a normed space, and let  $\{T_n\}_{n \in \mathbb{N}} \subset B(X, Y)$  be such that*

$$\exists \lim_n T_n x := Tx \quad \text{for all } x \in X.$$

*Then  $T \in B(X, Y)$  and  $\|T\| \leq \liminf_n \|T_n\| \leq \sup_n \|T_n\| < \infty$ .*

**Proof.** The linearity of  $T$  is trivial. For each  $x \in X$ ,  $\sup_n \|T_n x\| < \infty$  (since  $\lim_n T_n x$  exists). By Corollary 6.5, it follows that  $\sup_n \|T_n\| := M < \infty$ . For all unit vectors  $x \in X$  and all  $n \in \mathbb{N}$ ,  $\|T_n x\| \leq \|T_n\|$ ; therefore

$$\|Tx\| = \lim_n \|T_n x\| \leq \liminf_n \|T_n\| \leq M,$$

that is,  $\|T\| \leq \liminf_n \|T_n\| \leq \sup_n \|T_n\| < \infty$ . □

**Corollary 6.7.** *Let  $X$  be a normed space, and  $E \subset X$ . Then the following two statements are equivalent:*

$$\sup_{x \in E} |x^* x| < \infty \quad \text{for all } x^* \in X^*. \tag{1}$$

$$\sup_{x \in E} \|x\| < \infty. \tag{2}$$

**Proof.** Let  $\mathcal{T} := \kappa E \subset (X^*)^* = B(X^*, \mathbb{C})$ , where  $\kappa$  denotes the canonical embedding of  $X$  into  $X^{**}$ . Then (1) is equivalent to

$$\sup_{\kappa x \in \mathcal{T}} |(\kappa x)x^*| < \infty \quad \text{for all } x^* \in X^*,$$

and since the *conjugate* space  $X^*$  is complete (cf. Corollary 4.5), Corollary 6.5 shows that this is equivalent to

$$\sup_{\kappa x \in \mathcal{T}} \|\kappa x\| < \infty.$$

Since  $\kappa$  is isometric, the last statement is equivalent to (2).  $\square$

Combining Corollaries 6.7 and 6.5, we obtain

**Corollary 6.8.** *Let  $X$  be a Banach space,  $Y$  a normed space, and  $\mathcal{T} \subset B(X, Y)$ . Then the following two statements are equivalent*

$$\sup_{T \in \mathcal{T}} |x^*Tx| < \infty \quad \text{for all } x \in X, x^* \in X^*. \quad (3)$$

$$\sup_{T \in \mathcal{T}} \|T\| < \infty. \quad (4)$$

### 6.3 The open mapping theorem

**Lemma 1.** *Let  $X, Y$  be normed spaces and  $T \in B(X, Y)$ . Suppose the range  $TX$  of  $T$  is of Baire's second category in  $Y$ , and let  $\epsilon > 0$  be given. Then there exists  $\delta > 0$  such that*

$$B_Y(0, \delta) \subset \overline{TB_X(0, \epsilon)}. \quad (1)$$

Moreover, one has necessarily  $\delta \leq \|T\|\epsilon$ .

The bar sign stands for the closure operation in  $Y$ .

**Proof.** We may write

$$X = \bigcup_{n=1}^{\infty} B_X(0, n\epsilon/2),$$

and therefore

$$TX = \bigcup_{n=1}^{\infty} TB_X(0, n\epsilon/2).$$

Since  $TX$  is of Baire's second category in  $Y$ , there exists  $n$  such that  $\overline{TB_X(0, n\epsilon/2)}$  has non-empty interior. Therefore,  $\overline{TB_X(0, \epsilon/2)}$  has non-empty interior (because  $T$  is homogeneous and multiplication by  $n$  is a homeomorphism of  $Y$  onto itself). Let then  $B_Y(a, \delta)$  be a ball contained in it. If  $y \in B_Y(0, \delta)$ , then

$$a, a + y \in B_Y(a, \delta) \subset \overline{TB_X(0, \epsilon/2)},$$

and therefore there exist sequences

$$\{x'_k\}, \{x''_k\} \subset B_X(0, \epsilon/2)$$



such that

$$Tx'_k \rightarrow a + y; \quad Tx''_k \rightarrow a.$$

Let  $x_k = x'_k - x''_k$ . Then  $\|x_k\| \leq \|x'_k\| + \|x''_k\| < \epsilon$  and

$$Tx_k = Tx'_k - Tx''_k \rightarrow (a + y) - a = y,$$

that is,  $\{x_k\} \subset B_X(0, \epsilon)$  and  $y \in \overline{TB_X(0, \epsilon)}$ . This proves (1).

We show finally that the relation  $\delta \leq \|T\|\epsilon$  follows necessarily from (1). Fix  $y \in Y$  with  $\|y\| = 1$  and  $0 < t < 1$ . Since  $t\delta y \in B_Y(0, \delta)$ , it follows from (1) that for each  $k \in \mathbb{N}$ , there exists  $x_k \in B_X(0, \epsilon)$  such that  $\|t\delta y - Tx_k\| < 1/k$ . Therefore

$$\begin{aligned} t\delta &= \|t\delta y\| \leq \|t\delta y - Tx_k\| + \|Tx_k\| \\ &< 1/k + \|T\|\epsilon. \end{aligned}$$

Letting  $k \rightarrow \infty$  and then  $t \rightarrow 1$ , the conclusion follows.  $\square$

**Lemma 2.** *Let  $X$  be a Banach space,  $Y$  a normed space, and  $T \in B(X, Y)$ . Suppose  $TX$  is of Baire's second category in  $Y$ , and let  $\epsilon > 0$  be given. Then there exists  $\delta > 0$  such that*

$$B_Y(0, \delta) \subset TB_X(0, \epsilon). \quad (2)$$

Comparing the lemmas, we observe that the payoff for the added completeness hypothesis is the stronger conclusion (2) (instead of (1)).

**Proof.** We apply Lemma 1 with  $\epsilon_n = \epsilon/2^{n+1}$ ,  $n = 0, 1, 2, \dots$ . We then obtain  $\delta_n > 0$  such that

$$B_Y(0, \delta_n) \subset \overline{TB_X(0, \epsilon_n)} \quad (n = 0, 1, 2, \dots). \quad (3)$$

We shall show that (2) is satisfied with  $\delta := \delta_0$ .

Let  $y \in B_Y(0, \delta)$ .

By (3) with  $n = 0$ , there exists  $x_0 \in B_X(0, \epsilon_0)$  such that

$$\|y - Tx_0\| < \delta_1,$$

that is,  $y - Tx_0 \in B_Y(0, \delta_1)$ .

By (3) with  $n = 1$ , there exists  $x_1 \in B_X(0, \epsilon_1)$  such that

$$\|(y - Tx_0) - Tx_1\| < \delta_2.$$

Proceeding inductively, we obtain a sequence  $\{x_n; n = 0, 1, 2, \dots\}$  such that (for  $n = 0, 1, 2, \dots$ )

$$x_n \in B_X(0, \epsilon_n) \quad (4)$$

and

$$\|y - T(x_0 + \dots + x_n)\| < \delta_{n+1}. \quad (5)$$

Write

$$s_n = x_0 + \cdots + x_n.$$

Then for non-negative integers  $n > m$

$$\begin{aligned} \|s_n - s_m\| &= \|x_{m+1} + \cdots + x_n\| \leq \|x_{m+1}\| + \cdots + \|x_n\| \\ &< \frac{\epsilon}{2^{m+1}} + \cdots + \frac{\epsilon}{2^{n+1}} < \frac{\epsilon}{2^{m+1}}, \end{aligned} \quad (6)$$

so that  $\{s_n\}$  is a Cauchy sequence in  $X$ .

Since  $X$  is complete,  $s := \lim s_n$  exists in  $X$ . By continuity of  $T$  and of the norm, the left-hand side of (5) converges to  $\|y - Ts\|$  as  $n \rightarrow \infty$ . The right-hand side of (5) converges to 0 (cf. Lemma 1). Therefore  $y = Ts$ . However by (6) with  $m = 0$ ,  $\|s_n - s_0\| < \epsilon/2$  for all  $n$ ; hence  $\|s - x_0\| \leq \epsilon/2$ , and by (4)  $\|s\| \leq \|x_0\| + \|s - x_0\| < \epsilon$ . This shows that  $y \in TB_X(0, \epsilon)$ .  $\square$

**Theorem 6.9 (The open mapping theorem).** *Let  $X$  be a Banach space, and  $T \in B(X, Y)$  for some normed space  $Y$ . Suppose  $TX$  is of Baire's second category in  $Y$ . Then  $T$  is an open mapping.*

**Proof.** Let  $V$  be a non-empty open subset of  $X$ , and let  $y \in TV$ . Let then  $x \in V$  be such that  $y = Tx$ . Since  $V$  is open, there exists  $\epsilon > 0$  such that  $B_X(x, \epsilon) \subset V$ . Let  $\delta$  correspond to  $\epsilon$  as in Lemma 2. Then

$$\begin{aligned} B_Y(y, \delta) &= y + B_Y(0, \delta) \subset Tx + TB_X(0, \epsilon) = T[x + B_X(0, \epsilon)] \\ &= TB_X(x, \epsilon) \subset TV. \end{aligned}$$

This shows that  $TV$  is an open set in  $Y$ .  $\square$

**Corollary 6.10.** *Let  $X, Y$  be Banach spaces, and let  $T \in B(X, Y)$  be onto. Then  $T$  is an open map.*

**Proof.** Since  $T$  is onto, its range  $TX = Y$  is a Banach space, and is therefore of Baire's second category in  $Y$  by Theorem 6.3. The result then follows from Theorem 6.9.  $\square$

**Corollary 6.11 (Continuity of the inverse).** *Let  $X, Y$  be Banach spaces, and let  $T \in B(X, Y)$  be one-to-one and onto. Then  $T^{-1} \in B(Y, X)$ .*

**Proof.** By Corollary 6.10,  $T$  is a (linear) bijective continuous open map, that is, a (linear) homeomorphism. This means in particular that the inverse map is continuous.  $\square$

**Corollary 6.12.** *Suppose the vector space  $X$  is a Banach space under two norms  $\|\cdot\|_k, k = 1, 2$ . If there exists a constant  $M > 0$  such that  $\|x\|_2 \leq M \|x\|_1$  for all  $x \in X$ , then there exists a constant  $N > 0$  such that  $\|x\|_1 \leq N \|x\|_2$  for all  $x \in X$ .*

Norms satisfying inequalities of the form

$$\frac{1}{N} \|x\|_1 \leq \|x\|_2 \leq M \|x\|_1 \quad (x \in X)$$

for suitable constants  $M, N > 0$  are said to be *equivalent*. They induce the same metric topology on  $X$ .

**Proof.** Let  $T$  be the identity map from the Banach space  $(X, \|\cdot\|_1)$  to the Banach space  $(X, \|\cdot\|_2)$ . Then  $T$  is bounded (by the hypothesis on the norms), and clearly one-to-one and onto. The result then follows from Corollary 6.11.  $\square$

## 6.4 Graphs

For the next corollary, we consider the cartesian product  $X \times Y$  of two normed spaces, as a normed space with the usual operations and with the norm

$$\|[x, y]\| = \|x\| + \|y\| \quad ([x, y] \in X \times Y).$$

Clearly the sequence  $\{[x_n, y_n]\}$  is Cauchy in  $X \times Y$  if and only if both sequences  $\{x_n\}$  and  $\{y_n\}$  are Cauchy in  $X$  and  $Y$ , respectively, and it converges to  $[x, y]$  in  $X \times Y$  if and only if both  $x_n \rightarrow x$  in  $X$  and  $y_n \rightarrow y$  in  $Y$ . Therefore,  $X \times Y$  is complete if and only if both  $X$  and  $Y$  are complete.

Let  $T$  be a linear map with domain  $D(T) \subset X$  and range in  $Y$ . The domain is a subspace of  $X$ . The *graph* of  $T$  is the subspace of  $X \times Y$  defined by

$$\Gamma(T) := \{[x, Tx]; x \in D(T)\}.$$

If  $\Gamma(T)$  is a *closed* subspace of  $X \times Y$ , we say that  $T$  is a *closed operator*. Clearly  $T$  is closed if and only if whenever  $\{x_n\} \subset D(T)$  is such that  $x_n \rightarrow x$  and  $Tx_n \rightarrow y$ , then  $x \in D(T)$  and  $Tx = y$ .

**Corollary 6.13 (The closed graph theorem).** *Let  $X, Y$  be Banach spaces and let  $T$  be a closed operator with  $D(T) = X$  and range in  $Y$ . Then  $T \in B(X, Y)$ .*

**Proof.** Let  $P_X$  and  $P_Y$  be the projections of  $X \times Y$  onto  $X$  and  $Y$ , respectively, *restricted to the closed subspace*  $\Gamma(T)$  (which is a Banach space, as a closed subspace of the Banach space  $X \times Y$ ). They are continuous, and  $P_X$  is one-to-one and onto. By Corollary 6.11,  $P_X^{-1}$  is continuous, and therefore the composition  $P_Y \circ P_X^{-1}$  is continuous. However

$$P_Y \circ P_X^{-1}x = P_Y[x, Tx] = Tx \quad (x \in X),$$

that is,  $T$  is continuous.  $\square$

**Remark 6.14.** The proof of Lemma 2 used the completeness of  $X$  to get the convergence of the sequence  $\{s_n\}$ , which is the sequence of partial sums of the series  $\sum x_k$ . The point of the argument is that, if  $X$  is complete, then the convergence of a series  $\sum x_k$  in  $X$  follows from its *absolute convergence* (that is, the convergence of the series  $\sum \|x_k\|$ ). This property actually characterizes completeness of normed spaces.

**Theorem 6.15.** *Let  $X$  be a normed space. Then  $X$  is complete if and only if absolute convergence implies convergence (of series in  $X$ ).*

**Proof.** Suppose  $X$  is complete. If  $\sum \|x_k\|$  converges and  $s_n$  denote the partial sums of  $\sum x_k$ , then for  $n > m$

$$\|s_n - s_m\| = \left\| \sum_{k=m+1}^n x_k \right\| \leq \sum_{k=m+1}^n \|x_k\| \rightarrow 0$$

as  $m \rightarrow \infty$ , and therefore  $\{s_n\}$  converges in  $X$ .

Conversely, suppose absolute convergence implies convergence of series in  $X$ . Let  $\{x_n\}$  be a Cauchy sequence in  $X$ . There exists a subsequence  $\{x_{n_k}\}$  such that

$$\|x_{n_{k+1}} - x_{n_k}\| < \frac{1}{2^k}$$

(cf. proof of Lemma 1.30). The series

$$x_{n_1} + \sum_{k=1}^{\infty} (x_{n_{k+1}} - x_{n_k})$$

converges absolutely, and therefore converges in  $X$ . Its  $(p-1)$ -th partial sum is  $x_{n_p}$ , and so the Cauchy sequence  $\{x_n\}$  has the convergent subsequence  $\{x_{n_p}\}$ ; it follows that  $\{x_n\}$  itself converges.  $\square$

## 6.5 Quotient space

If  $M$  is a closed subspace of the Banach space  $X$ , the vector space  $X/M$  is a normed space for the *quotient norm*

$$\|[x]\| := \text{dist}\{0, [x]\} := \inf_{y \in [x]} \|y\|,$$

where  $[x] := x + M$ . The properties of the norm are easily verified; the assumption that  $M$  is *closed* is needed for the implication  $\|[x]\| = 0$  implies  $[x] = [0]$ . For later use, we prove the following.

**Theorem 6.16.** *Let  $M$  be a closed subspace of the Banach space  $X$ . Then  $X/M$  is a Banach space.*

**Proof.** Let  $\{[x_n]\}$  be a Cauchy sequence in  $X/M$ . It has a subsequence  $\{[x'_n]\}$  such that

$$\|[x'_{n+1}] - [x'_n]\| < \frac{1}{2^{n+1}}$$

(cf. proof of Lemma 1.30). Pick  $y_1 \in [x'_1]$  arbitrarily. Since

$$\inf_{y \in [x'_2]} \|y - y_1\| = \|[x'_2] - [x'_1]\| < 1/4,$$

there exists  $y_2 \in [x'_2]$  such that  $\|y_2 - y_1\| < 1/2$ . Assuming we found  $y_k \in [x'_k]$ ,  $k = 1, \dots, n$ , such that  $\|y_{k+1} - y_k\| < 1/2^k$  for  $k \leq n-1$ , since

$$\inf_{y \in [x'_{n+1}]} \|y - y_n\| = \|[x'_{n+1}] - [x'_n]\| < \frac{1}{2^{n+1}},$$

there exists  $y_{n+1} \in [x'_{n+1}]$  such that

$$\|y_{n+1} - y_n\| < \frac{2}{2^{n+1}} = 1/2^n.$$

The series

$$y_1 + \sum_{n=1}^{\infty} (y_{n+1} - y_n)$$

converges absolutely, and therefore converges in  $X$ , since  $X$  is complete (cf. Theorem 6.15). Its partial sums  $y_n$  converge therefore to some  $y \in X$ . Then

$$\|[x'_n] - [y]\| = \|[y_n] - [y]\| = \|[y_n - y]\| \leq \|y_n - y\| \rightarrow 0.$$

Since the *Cauchy sequence*  $\{[x_n]\}$  has the *convergent subsequence*  $\{[x'_n]\}$ , it follows that it converges as well.  $\square$

**Corollary 6.17.** *The natural map  $\pi : x \rightarrow [x]$  of the Banach space  $X$  onto the Banach space  $X/M$  (for a given closed subspace  $M$  of  $X$ ) is an open mapping.*

**Proof.** The map  $\pi$  is a norm-decreasing, hence continuous, linear map of the Banach space  $X$  onto the Banach space  $X/M$ ; therefore,  $\pi$  is an open map by Corollary 6.10.  $\square$

## 6.6 Operator topologies

The norm topology on  $B(X, Y)$  is also called the *uniform operator topology*. This terminology is motivated by the fact that a sequence  $\{T_n\} \subset B(X, Y)$  converges in the norm topology of  $B(X, Y)$  iff it converges (strongly) pointwise, uniformly on every bounded subset of  $X$  (that is, the sequence  $\{T_n x\}$  converges strongly in  $Y$ , uniformly in  $x$  on any bounded subset of  $X$ ). Indeed, if  $\|T_n - T\| \rightarrow 0$ , then for any bounded set  $Q \subset X$ ,

$$\sup_{x \in Q} \|T_n x - T x\| \leq \|T_n - T\| \sup_{x \in Q} \|x\| \rightarrow 0,$$

so that  $T_n x \rightarrow T x$  strongly in  $Y$ , uniformly on  $Q$ . Conversely, if  $T_n x$  converge strongly in  $Y$  uniformly for  $x$  in bounded subsets of  $X$ , this is true in particular for  $x$  in the unit ball  $S = S_X$ . Hence as  $n, m \rightarrow \infty$ ,

$$\|T_n - T_m\| := \sup_{x \in S} \|(T_n - T_m)x\| = \sup_{x \in S} \|T_n x - T_m x\| \rightarrow 0.$$

If  $Y$  is complete,  $B(X, Y)$  is complete by Theorem 4.4, and therefore  $T_n$  converge in the norm topology of  $B(X, Y)$ .

We shall consider two additional topologies on  $B(X, Y)$ , weaker than the uniform operator topology (u.o.t.). A net  $\{T_j; j \in J\}$  converges to  $T$  in the *strong operator topology* (s.o.t.) of  $B(X, Y)$  if  $T_j x \rightarrow Tx$  strongly in  $Y$ , for each  $x \in X$  (this is *strong pointwise convergence* of the functions  $T_j$ !). Since the uniformity requirement has been dropped, this convergence is clearly weaker than convergence in the u.o.t. If one requires that  $T_j x$  converge *weakly* to  $Tx$  (rather than strongly!), for each  $x \in X$ , one gets a still weaker convergence concept, called *convergence in the weak operator topology* (w.o.t.).

The s.o.t. and the w.o.t. may be defined by giving bases as follows.

**Definition 6.18.**

1. A base for the strong operator topology on  $B(X, Y)$  consists of all the sets of the form

$$N(T, F, \epsilon) := \{S \in B(X, Y); \|(S - T)x\| < \epsilon, x \in F\},$$

where  $T \in B(X, Y)$ ,  $F \subset X$  is *finite*, and  $\epsilon > 0$ .

2. A base for the weak operator topology on  $B(X, Y)$  consists of all sets of the form

$$N(T, F, \Lambda, \epsilon) := \{S \in B(X, Y); |y^*(S - T)x| < \epsilon, x \in F, y^* \in \Lambda\},$$

where  $T \in B(X, Y)$ ,  $F \subset X$  and  $\Lambda \subset Y^*$  are finite sets, and  $\epsilon > 0$ .

The sets  $N$  are referred to as *basic neighbourhoods of  $T$*  in the s.o.t. (w.o.t., respectively). It is clear that net convergence in these topologies is precisely as described above.

Since the bases in Definition 6.18 consist of convex sets, it is clear that  $B(X, Y)$  is a locally convex topological vector space (t.v.s.) for each of the above topologies. We denote by  $B(X, Y)_{s.o.}$  and  $B(X, Y)_{w.o.}$  the t.v.s.  $B(X, Y)$  with the s.o.t. and the w.o.t., respectively.

**Theorem 6.19.** *Let  $X, Y$  be normed spaces. Then*

$$B(X, Y)_{s.o.}^* = B(X, Y)_{w.o.}^*.$$

*Moreover, the general form of an element  $g$  of this (common) dual is*

$$g(T) = \sum_k y_k^* T x_k \quad (T \in B(X, Y)),$$

*where the sum is finite,  $x_k \in X$ , and  $y_k^* \in Y^*$ .*

**Proof.** Let  $g \in B(X, Y)_{s.o.}^*$ . Since  $g(0) = 0$ , strong-operator continuity of  $g$  at zero implies the existence of  $\epsilon > 0$  and of a finite set  $F = \{x_1, \dots, x_n\}$ , such that

$|g(T)| < 1$  for all  $T \in N(0, F, \epsilon)$ . Thus, the inequalities

$$\|Tx_k\| < \epsilon \quad (k = 1, \dots, n) \quad (1)$$

imply  $|g(T)| < 1$ .

Consider the normed space  $Y^n$  with the norm  $\|[y_1, \dots, y_n]\| := \sum_k \|y_k\|$ . One verifies easily that  $(Y^n)^*$  is isomorphic to  $(Y^*)^n$ : given  $\Gamma \in (Y^n)^*$ , there exists a unique vector  $[y_1^*, \dots, y_n^*] \in (Y^*)^n$  such that

$$\Gamma([y_1, \dots, y_n]) = \sum_k y_k^* y_k \quad (2)$$

for all  $[y_1, \dots, y_n] \in Y^n$ .

With  $x_1, \dots, x_n$  as in (1), define the linear map

$$\Phi : B(X, Y) \rightarrow Y^n$$

by

$$\Phi(T) = [Tx_1, \dots, Tx_n] \quad (T \in B(X, Y)).$$

On the range of  $\Phi$  (a subspace of  $Y^n$ !), define  $\Gamma$  by

$$\Gamma(\Phi(T)) = g(T) \quad (T \in B(X, Y)).$$

If  $T, S \in B(X, Y)$  are such that  $\Phi(T) = \Phi(S)$ , then  $\Phi(m(T - S)) = 0$ , so that  $m(T - S)$  satisfy (1) for all  $m \in \mathbb{N}$ . Hence  $m|g(T - S)| = |g(m(T - S))| < 1$  for all  $m$ , and therefore  $g(T) = g(S)$ . This shows that  $\Gamma$  is well defined. It is clearly a linear functional on  $\text{range}(\Phi)$ . If  $\|\Phi(T)\| < 1$ , then  $\|(\epsilon T)x_k\| < \epsilon$  for all  $k$ , hence  $|g(\epsilon T)| < 1$ , that is,  $|\Gamma(\Phi(T))| (= |g(T)|) < 1/\epsilon$ . This shows that  $\Gamma$  is bounded, with norm  $\leq 1/\epsilon$ . By the Hahn–Banach theorem,  $\Gamma$  has an extension as an element  $\tilde{\Gamma} \in (Y^n)^*$ . As observed above, it follows that there exist  $y_1^*, \dots, y_n^* \in Y^*$  such that  $\tilde{\Gamma}([y_1, \dots, y_n]) = \sum_k y_k^* y_k$ . In particular,

$$g(T) = \Gamma([Tx_1, \dots, Tx_n]) = \sum_k y_k^* T x_k \quad (3)$$

for all  $T \in B(X, Y)$ .

In particular, this representation shows that  $g$  is continuous with respect to the w.o.t. Since (linear) functionals continuous with respect to the w.o.t. are trivially continuous with respect to the s.o.t., the theorem follows.  $\square$

**Corollary 6.20.** *A convex subset of  $B(X, Y)$  has the same closure in the w.o.t. and in the s.o.t.*

**Proof.** Let  $K \subset B(X, Y)$  be convex (non-empty, without loss of generality), and denote by  $K_s$  and  $K_w$  its closures with respect to the s.o.t. and the w.o.t., respectively. Since the w.o.t. is weaker than the s.o.t., we clearly have  $K_s \subset K_w$ . Suppose there exists  $T \in K_w$  such that  $T \notin K_s$ . By Corollary 5.21, there exists  $f \in B(X, Y)_{s.o.}^*$  such that

$$\Re f(T) < \inf_{S \in K_s} \Re f(S). \quad (4)$$

By Theorem 6.19,  $f \in B(X, Y)_{w.o.}^*$ . Since  $T \in K_w$  and  $K \subset K_s$ , it follows that  $\inf_{S \in K_s} \Re f(S) \leq \inf_{S \in K} \Re f(S) \leq \Re f(T)$ , a contradiction.  $\square$

## Exercises

1. Let  $X$  be a Banach space,  $Y$  a normed space, and  $T \in B(X, Y)$ . Prove that if  $TX \neq Y$ , then  $TX$  is of Baire's first category in  $Y$ .
2. Let  $X, Y$  be normed spaces, and  $T \in B(X, Y)$ . Prove that

$$\|T\| = \sup\{|y^*Tx|; x \in X, y^* \in Y^*, \|x\| = \|y^*\| = 1\}.$$

3. Let  $(S, \mathcal{A}, \mu)$  be a positive measure space, and let  $p, q \in [1, \infty]$  be conjugate exponents. Let  $T : L^p(\mu) \rightarrow L^q(\mu)$ . Prove that

$$\|T\| = \sup \left\{ \left\| \int_S (Tf)g \, d\mu \right\|; f \in L^p(\mu), g \in L^q(\mu), \|f\|_p = \|g\|_q = 1 \right\}.$$

(In case  $p = 1$  or  $p = \infty$ , assume that the measure space is  $\sigma$ -finite.)

4. Let  $X$  be a Banach space,  $\{Y_\alpha; \alpha \in I\}$  a family of normed spaces, and  $T_\alpha \in B(X, Y_\alpha)$ , ( $\alpha \in I$ ). Define

$$Z = \left\{ x \in X; \sup_{\alpha \in I} \|T_\alpha x\| = \infty \right\}.$$

Prove that  $Z$  is either empty or a dense  $G_\delta$  in  $X$ .

5. Let  $X, Y$  be Banach spaces, and let  $T : D(T) \subset X \rightarrow Y$  be a closed operator with range  $R(T)$  of the second category in  $Y$ . Prove:

- (a)  $R(T) = Y$ .
- (b) There exists a constant  $c > 0$  such that, for each  $y \in Y$ , there exists  $x \in D(T)$  such that  $y = Tx$  and  $\|x\| \leq c\|y\|$ .
- (c) If  $T$  is one-to-one, then it is bounded.  
(Hint: adapt the *proof* of Theorem 6.9.)

6. Let  $m$  denote Lebesgue measure on the interval  $[0, 1]$ , and let  $1 \leq p < r \leq \infty$ . Prove that the identity map of  $L^r(m)$  into  $L^p(m)$  is norm decreasing with range of Baire's first category in  $L^p(m)$ .

7. Let  $X$  be a Banach space, and let  $T : D(T) \subset X \rightarrow X$  be a linear operator. Suppose there exists  $\alpha \in \mathbb{C}$  such that  $(\alpha I - T)^{-1} \in B(X)$ . Let  $p(\lambda) = \sum c_k \lambda^k$  be any polynomial (over  $\mathbb{C}$ ) of degree  $n \geq 1$ . Prove that the operator  $p(T) := \sum c_k T^k$  (with domain  $D(T^n)$ ) is closed. (Hint: induction on  $n$ . Write  $p(\lambda) = (\lambda - \alpha)q(\lambda) + r$ , where the constant  $r$  may be assumed to be zero, without loss of generality, and  $q$  is a polynomial of degree  $n - 1$ .)

8. Let  $X, Y$  be Banach spaces. The operator  $T \in B(X, Y)$  is *compact* if the set  $TB_X$  is conditionally compact in  $Y$  (where  $B_X$  denotes here the



closed unit ball of  $X$ ). Let  $K(X, Y)$  be the set of all compact operators in  $B(X, Y)$ . Prove:

- (a)  $K(X, Y)$  is a (norm-)closed subspace of  $B(X, Y)$ .
- (b) If  $Z$  is a Banach space, then

$$K(X, Y)B(Z, X) \subset K(Z, Y) \quad \text{and} \quad B(Y, Z)K(X, Y) \subset K(X, Z).$$

In particular,  $K(X) := K(X, X)$  is a closed two-sided ideal in  $B(X)$ .

- (c)  $T \in B(X, Y)$  is a *finite range operator* if its range  $TX$  is finite dimensional. Prove that every finite range operator is compact.

## Adjoint

- 9. Let  $X, Y$  be Banach spaces, and let  $T : X \rightarrow Y$  be a linear operator with domain  $D(T) \subset X$  and range  $R(T)$ . If  $T$  is one-to-one, the inverse map  $T^{-1}$  is a linear operator with domain  $R(T)$  and range  $D(T)$ .

If  $D(T)$  is *dense* in  $X$ , the (Banach) adjoint  $T^*$  of  $T$  is defined as follows:

$$D(T^*) = \{y^* \in Y^*; y^* \circ T \text{ is continuous on } D(T)\}.$$

Since  $D(T)$  is dense in  $X$ , it follows that for each  $y^* \in D(T^*)$  there exists a *unique* extension  $x^* \in X^*$  of  $y^* \circ T$  (cf. Exercise 1, Chapter 4); we set  $x^* = T^*y^*$ . Thus,  $T^*$  is uniquely defined on  $D(T^*)$  by the relation

$$(T^*y^*)x = y^*(Tx) \quad (x \in D(T)).$$

Prove:

- (a)  $T^*$  is closed. If  $T$  is closed,  $D(T^*)$  is *weak*<sup>\*</sup>-dense in  $Y^*$ , and if  $Y$  is reflexive,  $D(T^*)$  is strongly dense in  $Y^*$ .
- (b) If  $T \in B(X, Y)$ , then  $T^* \in B(Y^*, X^*)$ , and  $\|T^*\| = \|T\|$ . If  $S, T \in B(X, Y)$ , then  $(\alpha S + \beta T)^* = \alpha S^* + \beta T^*$  for all  $\alpha, \beta \in \mathbb{C}$ . If  $T \in B(X, Y)$  and  $S \in B(Y, Z)$ , then  $(ST)^* = T^*S^*$ .
- (c) If  $T \in B(X, Y)$ , then  $T^{**} := (T^*)^* \in B(X^{**}, Y^{**})$ ,  $T^{**}|_X = T$ , and  $\|T^{**}\| = \|T\|$ . In particular, if  $X$  is reflexive, then  $T^{**} = T$  (note that  $\kappa X$  is identified with  $X$ ).
- (d) If  $T \in B(X, Y)$ , then  $T^*$  is continuous with respect to the *weak*<sup>\*</sup>-topologies on  $Y^*$  and  $X^*$  (cf. Exercise 4, Chapter 5). Conversely, if  $S \in B(Y^*, X^*)$  is continuous with respect to the *weak*<sup>\*</sup>-topologies on  $Y^*$  and  $X^*$ , then  $S = T^*$  for some  $T \in B(X, Y)$ . Hint: given  $x \in X$ , consider the functional  $\phi_x(y^*) = (Sy^*)x$  on  $Y^*$ .
- (e)  $\overline{R(T)} = \bigcap \{\ker(y^*); y^* \in \ker(T^*)\}$ . In particular,  $T^*$  is one-to-one iff  $R(T)$  is dense in  $Y$ .

- (f) Let  $x^* \in X^*$  and  $M > 0$  be given. Then there exists  $y^* \in D(T^*)$  with  $\|y^*\| \leq M$  such that  $x^* = T^*y^*$  if and only if

$$|x^*x| \leq M\|Tx\| \quad (x \in D(T)).$$

In particular,  $x^* \in R(T^*)$  if and only if

$$\sup_{x \in D(T), Tx \neq 0} \frac{|x^*x|}{\|Tx\|} < \infty.$$

(Hint: Hahn–Banach.)

- (g) Let  $T \in B(X, Y)$  and let  $S^*$  be the (norm-)closed unit ball of  $Y^*$ . Then  $T^*S^*$  is *weak\**-compact.
- (h) Let  $T \in B(X, Y)$  have closed range  $TX$ . Suppose  $x^* \in X^*$  vanishes on  $\ker(T)$ . Show that the map  $\phi : TX \rightarrow \mathbb{C}$  defined by  $\phi(Tx) = x^*x$  is a well-defined continuous linear functional, and therefore there exists  $y^* \in Y^*$  such that  $\phi = y^*|_{TX}$ . (Hint: apply Corollary 6.10 to  $T \in B(X, TX)$  to conclude that there exists  $r > 0$  such that  $\{y \in TX; \|y\| < r\} \subset TB_X(0, 1)$ , and deduce that  $\|\phi\| \leq (1/r)\|x^*\|$ .)
- (k) With  $T$  as in Part (h), prove that

$$T^*Y^* = \{x^* \in X^*; \ker(T) \subset \ker(x^*)\}.$$

In particular,  $T^*$  has (norm-)closed range in  $X^*$ .

10. Let  $X$  be a Banach space, and let  $T$  be a one-to-one linear operator with domain and range dense in  $X$ . Prove that  $(T^*)^{-1} = (T^{-1})^*$ , and  $T^{-1}$  is bounded (on its domain) iff  $(T^*)^{-1} \in B(X^*)$ .
11. Let  $T : D(T) \subset X \rightarrow X$  have dense domain in the Banach space  $X$ . Prove:
- If the range  $R(T^*)$  of  $T^*$  is *weak\**-dense in  $X^*$ , then  $T$  is one-to-one.
  - $T^{-1}$  exists and is bounded (on its domain) iff  $R(T^*) = X^*$ .
12. Let  $X$  be a Banach space, and  $T \in B(X)$ . We say that  $T$  is *bounded below* if

$$\inf_{0 \neq x \in X} \frac{\|Tx\|}{\|x\|} > 0.$$

Prove:

- If  $T$  is bounded below, then it is one-to-one and has closed range.
- $T$  is non-singular (that is, invertible in  $B(X)$ ) if and only if it is bounded below and  $T^*$  is one-to-one.

## Hilbert adjoint

13. Let  $X$  be a Hilbert space, and  $T : D(T) \subset X \rightarrow X$  a linear operator with dense domain. The *Hilbert adjoint*  $T^*$  of  $T$  is defined in a way analogous to that of Exercise 9, through the Riesz representation:

$$D(T^*) := \{y \in X; x \mapsto (Tx, y) \text{ is continuous on } D(T)\}.$$

Since  $D(T)$  is dense, given  $y \in D(T^*)$ , there exists a unique vector in  $X$ , which we denote by  $T^*y$ , such that

$$(Tx, y) = (x, T^*y) \quad (x \in D(T)).$$

Prove:

- (a) If  $T \in B(X)$ , then  $T^* \in B(X)$ ,  $\|T^*\| = \|T\|$ ,  $T^{**} = T$ , and  $(\alpha T)^* = \bar{\alpha}T^*$  for all  $\alpha \in \mathbb{C}$ . Also  $I^* = I$ .
  - (b) If  $S, T \in B(X)$ , then  $(S + T)^* = S^* + T^*$  and  $(ST)^* = T^*S^*$ .
  - (c)  $T \in B(X)$  is called a *normal* operator if  $T^*T = TT^*$ . Prove that  $T$  is normal iff
 
$$(T^*x, T^*y) = (Tx, Ty) \quad (x, y \in X) \quad (1)$$
  - (d) If  $T \in B(X)$  is normal, then  $\|T^*x\| = \|Tx\|$  and  $\|T^*Tx\| = \|T^2x\|$  for all  $x \in X$ . Conclude that  $\|T^*T\| = \|T^2\|$  and  $\|T^2\| = \|T\|^2$ . (Hint: apply (1).)
14. Let  $X$  be a Hilbert space, and  $T : D(T) \subset X \rightarrow X$  be a linear operator.  $T$  is *symmetric* if  $(Tx, y) = (x, Ty)$  for all  $x, y \in D(T)$ . Prove that if  $T$  is symmetric and everywhere defined, then  $T \in B(X)$  and  $T = T^*$ . (Hint: Corollary 6.13.)
15. Let  $X$  be a Hilbert space, and  $B : X \times X \rightarrow \mathbb{C}$  be a sesquilinear form such that

$$|B(x, y)| \leq M\|x\| \|y\| \quad \text{and} \quad B(x, x) \geq m\|x\|^2$$

for all  $x, y \in X$ , for some constants  $M < \infty$  and  $m > 0$ . Prove that there exists a unique non-singular  $T \in B(X)$  such that  $B(x, y) = (x, Ty)$  for all  $x, y \in X$ . Moreover,

$$\|T\| \leq M \quad \text{and} \quad \|T^{-1}\| \leq 1/m.$$

(This is the *Lax–Milgram theorem*.) Hint: apply Theorem 1.37 to get  $T$ ; show that  $R(T)$  is closed and dense (cf. Theorem 1.36), and apply Corollary 6.11.

16. Let  $X, Y$  be normed spaces, and  $T : X \rightarrow Y$  be linear. Prove that  $T$  is an open map iff  $T\bar{B}_X(0, 1)$  contains  $\bar{B}_Y(0, r)$  for some  $r > 0$ . When this is the case,  $T$  is *onto*.

17. Let  $X$  be a Banach space,  $Y$  a normed space, and  $T \in B(X, Y)$ . Suppose the closure of  $T\bar{B}_X(0, 1)$  contains some ball  $\bar{B}_Y(0, r)$ . Prove that  $T$  is open. (Hint: adapt the *proof* of Lemma 2 in the proof of Theorem 6.9, and use Exercise 16.)
18. Let  $X$  be a Banach space, and let  $P \in B(X)$  be such that  $P^2 = P$ . Such an operator is called a *projection*. Verify:
- (a)  $I - P$  is a projection (called *the complementary projection*).
  - (b) The ranges  $PX$  and  $(I - P)X$  are *closed* subspaces such that  $X = PX \oplus (I - P)X$ . Moreover  $PX = \ker(I - P) = \{x; Px = x\}$  and  $(I - P)X = \ker P$ .
  - (c) Conversely, if  $Y, Z$  are closed subspaces of  $X$  such that  $X = Y \oplus Z$  ('complementary subspaces'), and  $P : X \rightarrow Y$  is defined by  $P(y+z) = y$  for all  $y \in Y, z \in Z$ , then  $P$  is a projection with  $PX = Y$ .
  - (d) If  $Y, Z$  are closed subspaces of  $X$  such that  $Y \cap Z = \{0\}$ , then  $Y + Z$  is closed iff there exists a positive constant  $c$  such that  $\|y\| \leq c\|y+z\|$  for all  $y \in Y$  and  $z \in Z$ . (Hint: Corollary 6.13.)
19. Let  $X, Y$  be Banach spaces, and let  $\{T_n\}_{n \in \mathbb{N}} \subset B(X, Y)$  be Cauchy in the s.o.t. (that is,  $\{T_n x\}$  is Cauchy for each  $x \in X$ ). Prove that  $\{T_n\}$  is convergent in  $B(X, Y)$  in the s.o.t.
20. Let  $X, Y$  be Banach spaces and  $T \in B(X, Y)$ . Prove that  $T$  is one-to-one with closed range iff there exists a positive constant  $c$  such that  $\|Tx\| \geq c\|x\|$  for all  $x \in X$ . In that case,  $T^{-1} \in B(TX, X)$ .
21. Let  $X$  be a Banach space, and let  $C \in B(X)$  be a *contraction*, that is,  $\|C\| \leq 1$ . Prove:
- (a)  $e^{t(C-I)}$  (defined by means of the usual series) is a contraction for all  $t \geq 0$ .
  - (b)  $\|C^m x - x\| \leq m\|Cx - x\|$  for all  $m \in \mathbb{N}$  and  $x \in X$ .
  - (c) Let  $Q_n := e^{n(C-I)} - C^n$  ( $n \in \mathbb{N}$ ). Then

$$\|Q_n x\| \leq e^{-n} \sum_{k=0}^{\infty} (n^k/k!) \|C^{k-n} x - x\| \quad (2)$$

for all  $n \in \mathbb{N}$  and  $x \in X$ . (Hint: note that  $C^n x = e^{-n} \sum_k (n^k/k!) C^n x$ ; break the ensuing series for  $Q_n x$  into series over  $k \leq n$  and over  $k > n$ ).

- (d)  $\|Q_n x\| \leq \sum_{k \geq 0} e^{-n} (n^k/k!) |k - n| \|Cx - x\|$ .
- (e)  $\|Q_n x\| \leq \sqrt{n} \|(C - I)x\|$  for all  $n \in \mathbb{N}$  and  $x \in X$ . Hint: consider the *Poisson probability measure*  $\mu$  (with 'parameter'  $n$ ) on  $\mathbb{P}(\mathbb{N})$ , defined

by  $\mu(\{k\}) = e^{-n} n^k / k!$ ; apply Schwarz's inequality in  $L^2(\mu)$  and Part (d) to get the inequality

$$\|Q_n x\| \leq \|k - n\|_{L^2(\mu)} \|Cx - x\| = \sqrt{n} \|Cx - x\|. \quad (3)$$

- (f) Let  $F : [0, \infty) \rightarrow B(X)$  be contraction-valued. For  $t > 0$  fixed, set  $A_n := (n/t)[F(t/n) - I]$ ,  $n \in \mathbb{N}$ . Suppose  $\sup_n \|A_n x\| < \infty$  for all  $x$  in a dense subspace  $D$  of  $X$ . Then

$$\lim_{n \rightarrow \infty} \|e^{tA_n} x - F(t/n)^n x\| = 0 \quad (4)$$

for all  $t > 0$  and  $x \in X$ . Hint: by Part (a),  $\|e^{tA_n}\| \leq 1$ , and therefore  $\|e^{tA_n} - F(t/n)^n\| \leq 2$ . By Part (e) with  $C = F(t/n)$ , the limit in (4) is 0 for all  $x \in D$ .

# Banach algebras

If  $X$  is a Banach space, the Banach space  $B(X)$  (cf. Notation 4.3) is also an *algebra* under the composition of operators as multiplication. The operator norm (cf. Definition 4.1) clearly satisfies the relations

$$\|ST\| \leq \|S\|\|T\| \quad (S, T \in B(X))$$

and

$$\|I\| = 1,$$

where  $I$  denotes the *identity operator*, defined by  $Ix = x$  for all  $x \in X$ .

If the dimension of  $X$  is at least 2, it is immediate that  $B(X)$  is not commutative. On the other hand, if  $X$  is a compact Hausdorff space, the Banach space  $C(X)$  of all complex continuous functions on  $X$  with pointwise operations and the supremum norm  $\|f\| = \sup_X |f|$  is a *commutative* algebra, and again  $\|fg\| \leq \|f\|\|g\|$  for all  $f, g \in C(X)$  and  $\|1\| = 1$ , where 1, the *unit* of the algebra, is the function with the constant value 1 on  $X$ .

## 7.1 Basics

This section is an introduction to the theory of abstract Banach algebras, of which  $B(X)$  and  $C(X)$  are two important examples.

**Definition 7.1.** A (*unital, complex*) *Banach algebra* is an (associative) algebra  $\mathcal{A}$  over  $\mathbb{C}$  with a unit  $e$ , which is a Banach space (as a vector space over  $\mathbb{C}$ ) under a norm that satisfies the relations:

- (1)  $\|xy\| \leq \|x\|\|y\|$  for all  $x, y \in \mathcal{A}$ , and
- (2)  $\|e\| = 1$ .

If we omit the completeness requirement in Definition 7.1,  $\mathcal{A}$  is called a *normed algebra*.

Note that the submultiplicativity of the norm implies the boundedness (i.e. the continuity) of the linear map of left multiplication by  $a$ ,  $L_a : x \rightarrow ax$  (for any given  $a \in \mathcal{A}$ ), and clearly  $\|L_a\| = \|a\|$ . The same is true for the right multiplication map  $R_a : x \rightarrow xa$ . Actually, multiplication is continuous as a map from  $\mathcal{A}^2$  to  $\mathcal{A}$ , since

$$\|xy - x'y'\| \leq \|x\|\|y - y'\| + \|x - x'\|\|y'\| \rightarrow 0$$

as  $[x', y'] \rightarrow [x, y]$  in  $\mathcal{A}^2$ .

**Definition 7.2.** Let  $\mathcal{A}$  be a Banach algebra, and  $a \in \mathcal{A}$ . We say that  $a$  is *regular* (or *non-singular*) if it is invertible in  $\mathcal{A}$ , that is, if there exists  $b \in \mathcal{A}$  such that  $ab = ba = e$ . If  $a$  is not regular, we say that it is *singular*.

If  $a$  is regular, the element  $b$  in Definition 7.2 is uniquely determined (if also  $b'$  satisfies the requirement, then  $b' = b'e = b'(ab) = (b'a)b = eb = b$ ), and is called *the inverse of  $a$* , denoted  $a^{-1}$ . Thus  $aa^{-1} = a^{-1}a = e$ . In particular,  $a \neq 0$  (otherwise  $1 = \|e\| = \|aa^{-1}\| = \|0\| = 0$ ).

We denote by  $G(\mathcal{A})$  the set of all regular elements of  $\mathcal{A}$ . It is a group under the multiplication of  $\mathcal{A}$ , and the map  $x \rightarrow x^{-1}$  is an anti-automorphism of  $G(\mathcal{A})$ . Topologically, we have

**Theorem 7.3.** Let  $\mathcal{A}$  be a Banach algebra, and let  $G(\mathcal{A})$  be the group of regular elements of  $\mathcal{A}$ . Then  $G(\mathcal{A})$  is open in  $\mathcal{A}$ , and the map  $x \rightarrow x^{-1}$  is a homeomorphism of  $G(\mathcal{A})$  onto itself.

**Proof.** Let  $y \in G := G(\mathcal{A})$  and  $\delta := 1/\|y^{-1}\|$  (note that  $\|y^{-1}\| \neq 0$ , since  $y^{-1} \in G$ ). We show that the ball  $B(y, \delta)$  is contained in  $G$  (so that  $G$  is indeed open).

Let  $x \in B(y, \delta)$ , and set  $a := y^{-1}x$ . We have

$$\|e - a\| = \|y^{-1}(y - x)\| \leq \|y^{-1}\|\|y - x\| < \|y^{-1}\|\delta = 1. \quad (1)$$

Therefore, the geometric series  $\sum_n \|e - a\|^n$  converges. By submultiplicativity of the norm, the series  $\sum_n \|(e - a)^n\|$  converges as well, and since  $\mathcal{A}$  is complete, it follows (cf. Theorem 6.15) that the series

$$\sum_{n=0}^{\infty} (e - a)^n$$

converges in  $\mathcal{A}$  to some element  $z \in \mathcal{A}$  ( $v^0 = e$  by definition, for any  $v \in \mathcal{A}$ ).

By continuity of  $L_a$  and  $R_a$  (with  $a = y^{-1}x$ ),

$$az = \sum_n a(e - a)^n = \sum_n [e - (e - a)](e - a)^n = \sum_n [(e - a)^n - (e - a)^{n+1}] = e,$$

and similarly  $za = e$ . Hence  $a \in G$  and  $a^{-1} = z$ . Since  $x = ya$ , also  $x \in G$ , as wanted.

Furthermore (for  $x \in B(y, \delta)$ !)

$$x^{-1} = a^{-1}y^{-1} = zy^{-1}, \quad (2)$$

and therefore by (1)

$$\begin{aligned} \|x^{-1} - y^{-1}\| &= \|(z - e)y^{-1}\| = \left\| \sum_{n=1}^{\infty} (e - a)^n y^{-1} \right\| \\ &\leq \sum_{n=1}^{\infty} \|e - a\|^n \|y^{-1}\| = (1/\delta) \frac{\|e - a\|}{1 - \|e - a\|} \\ &\leq (1/\delta^2) \frac{\|x - y\|}{1 - (\|x - y\|/\delta)} \rightarrow 0 \end{aligned}$$

as  $x \rightarrow y$ . This proves the continuity of the map  $x \rightarrow x^{-1}$  at  $y \in G$ . Since this map is its own inverse (on  $G$ ), it is a homeomorphism.  $\square$

**Remark 7.4.** If we take in the preceding proof  $y = e$  (so that  $\delta = 1$  and  $a = x$ ), we obtain in particular that  $B(e, 1) \subset G$  and

$$x^{-1} = \sum_{n=0}^{\infty} (e - x)^n \quad (x \in B(e, 1)). \quad (3)$$

Since  $B(e, 1) = e - B(0, 1)$ , this is equivalent to

$$(e - u)^{-1} = \sum_{n=0}^{\infty} u^n \quad (u \in B(0, 1)). \quad (4)$$

Relation (4) is the abstract version of the elementary geometric series summation formula.

For  $x \in \mathcal{A}$  arbitrary and  $\lambda$  complex with modulus  $> \|x\|$ , since  $u := x/\lambda \in B(0, 1)$ , we then have  $e - x/\lambda \in G$ , and

$$(e - x/\lambda)^{-1} = \sum_{n=0}^{\infty} (x/\lambda)^n.$$

Therefore  $\lambda e - x = \lambda(e - x/\lambda) \in G$  and

$$(\lambda e - x)^{-1} = \sum_{n=0}^{\infty} \frac{x^n}{\lambda^{n+1}} \quad (5)$$

(for all complex  $\lambda$  with modulus  $> \|x\|$ ).

**Definition 7.5.** The *resolvent set* of  $x \in \mathcal{A}$  is the set

$$\rho(x) := \{\lambda \in \mathbb{C}; \lambda e - x \in G\} = f^{-1}(G),$$

where  $f : \mathbb{C} \rightarrow \mathcal{A}$  is the continuous function  $f(\lambda) := \lambda e - x$ .



The complement of  $\rho(x)$  in  $\mathbb{C}$  is called the *spectrum* of  $x$ , denoted  $\sigma(x)$ . Thus  $\lambda \in \sigma(x)$  iff  $\lambda e - x$  is *singular*. The *spectral radius* of  $x$ , denoted  $r(x)$  is defined by

$$r(x) = \sup\{|\lambda|; \lambda \in \sigma(x)\}.$$

By Theorem 7.3,  $\rho(x)$  is *open*, as the inverse image of the open set  $G$  by the continuous function  $f$  above. Therefore,  $\sigma(x)$  is a *closed* subset of  $\mathbb{C}$ .

The *resolvent* of  $x$ , denoted  $R(\cdot; x)$ , is the function from  $\rho(x)$  to  $G$  defined by

$$R(\lambda; x) = (\lambda e - x)^{-1} \quad (\lambda \in \rho(x)).$$

The series expansion (5) of the resolvent, valid for  $|\lambda| > \|x\|$ , is called the *Neumann expansion*.

Note the trivial but useful identity

$$xR(\lambda; x) = \lambda R(\lambda; x) - e \quad (\lambda \in \rho(x)).$$

If  $x, y \in \mathcal{A}$  and  $1 \in \rho(xy)$ , then

$$(e - yx)[e + yR(1; xy)x] = (e - yx) + y(e - xy)R(1; xy)x = e,$$

and

$$[e + yR(1; xy)x](e - yx) = (e - yx) + yR(1; xy)(e - xy)x = e.$$

Therefore  $1 \in \rho(yx)$  and

$$R(1; yx) = e + yR(1; xy)x.$$

Next, for any  $\lambda \neq 0$ , write  $\lambda e - xy = \lambda[e - (x/\lambda)y]$ . If  $\lambda \in \rho(xy)$ , then  $1 \in \rho((x/\lambda)y)$ ; hence  $1 \in \rho(y(x/\lambda))$ , and therefore  $\lambda \in \rho(yx)$ . By symmetry, this proves that

$$\sigma(xy) \cup \{0\} = \sigma(yx) \cup \{0\}.$$

Hence

$$r(yx) = r(xy).$$

With  $f$  as above restricted to  $\rho(x)$ ,  $R(\lambda; x) = f(\lambda)^{-1}$ ; by Theorem 7.3,  $R(\cdot; x)$  is therefore *continuous* on the open set  $\rho(x)$ .

By Remark 7.4,

$$\{\lambda \in \mathbb{C}; |\lambda| > \|x\|\} \subset \rho(x) \tag{6}$$

and

$$R(\lambda; x) = \sum_{n=0}^{\infty} \frac{x^n}{\lambda^{n+1}} \quad (|\lambda| > \|x\|). \tag{7}$$

Thus

$$\sigma(x) \subset \Delta(0, \|x\|) := \{\lambda \in \mathbb{C}; |\lambda| \leq \|x\|\} \tag{8}$$

and so

$$r(x) \leq \|x\|. \tag{9}$$

The spectrum of  $x$  is closed and bounded (since it is contained in  $\Delta(0, \|x\|)$ ). Thus  $\sigma(x)$  is a *compact* subset of the plane.

**Theorem 7.6.** *Let  $\mathcal{A}$  be a Banach algebra and  $x \in \mathcal{A}$ . Then  $\sigma(x)$  is a non-empty compact set, and  $R(\cdot; x)$  is an analytic function on  $\rho(x)$  that vanishes at  $\infty$ .*

**Proof.** We observed already that  $\sigma(x)$  is compact. By continuity of the map  $y \rightarrow y^{-1}$ ,  $(e - \lambda^{-1}x)^{-1} \rightarrow e^{-1} = e$  as  $\lambda \rightarrow \infty$ , and therefore

$$\lim_{\lambda \rightarrow \infty} R(\lambda; x) = \lim_{\lambda \rightarrow \infty} \lambda^{-1}(e - \lambda^{-1}x)^{-1} = 0.$$

For  $\lambda \in \rho(x)$ , since  $\lambda e - x$  and  $x$  commute, also the inverse  $R(\lambda; x)$  commutes with  $x$ . If also  $\mu \in \rho(x)$ , writing  $R(\cdot) := R(\cdot; x)$ , we have

$$R(\mu) = (\lambda e - x)R(\lambda)R(\mu) = \lambda R(\lambda)R(\mu) - xR(\lambda)R(\mu)$$

and

$$R(\lambda) = R(\lambda)(\mu e - x)R(\mu) = \mu R(\lambda)R(\mu) - xR(\lambda)R(\mu).$$

Subtracting, we obtain the so-called *resolvent identity*

$$R(\mu) - R(\lambda) = (\lambda - \mu)R(\lambda)R(\mu). \quad (10)$$

For  $\mu \neq \lambda$  in  $\rho(x)$ , rewrite (10) as

$$\frac{R(\mu) - R(\lambda)}{\mu - \lambda} = -R(\lambda)R(\mu).$$

Since  $R(\cdot)$  is continuous on  $\rho(x)$ , we have

$$\exists \lim_{\mu \rightarrow \lambda} \frac{R(\mu) - R(\lambda)}{\mu - \lambda} = -R(\lambda)^2.$$

This shows that  $R(\cdot)$  is analytic on  $\rho(x)$ , and  $R'(\cdot) = -R(\cdot)^2$ .

For any  $x^* \in \mathcal{A}^*$ , it follows that  $x^*R(\cdot)$  is a complex analytic function in  $\rho(x)$ . If  $\sigma(x)$  is empty,  $x^*R(\cdot)$  is entire and vanishes at  $\infty$ . By Liouville's theorem,  $x^*R(\cdot)$  is identically 0, for all  $x^* \in \mathcal{A}^*$ . Therefore,  $R(\cdot) = 0$ , which is absurd since  $R(\cdot)$  has values in  $G(\mathcal{A})$ . This shows that  $\sigma(x) \neq \emptyset$ .  $\square$

**Corollary 7.7 (The Gelfand–Mazur theorem).** *A (complex unital) Banach algebra which is a division algebra is isomorphic and isometric to the complex field.*

**Proof.** Suppose the Banach algebra  $\mathcal{A}$  is a division algebra. If  $x \in \mathcal{A}$ ,  $\sigma(x) \neq \emptyset$  (by Theorem 7.6); pick then  $\lambda \in \sigma(x)$ . Since  $\lambda e - x$  is singular, and  $\mathcal{A}$  is a division algebra, we must have  $\lambda e - x = 0$ . Hence  $x = \lambda e$  and therefore  $\mathcal{A} = \mathbb{C}e$ .  $\square$

If  $p(\lambda) = \sum \alpha_k \lambda^k$  is a polynomial with complex coefficients, and  $x \in \mathcal{A}$ , we denote as usual  $p(x) := \sum \alpha_k x^k$  (where  $x^0 := e$ ). The map  $p \rightarrow p(x)$  is an algebra homomorphism  $\tau$  of the algebra of polynomials (over  $\mathbb{C}$ ) into  $\mathcal{A}$ , that sends 1 to  $e$  and  $\lambda$  to  $x$ .

**Theorem 7.8 (The spectral mapping theorem).** *For any polynomial  $p$  (over  $\mathbb{C}$ ) and any  $x \in \mathcal{A}$*

$$\sigma(p(x)) = p(\sigma(x)).$$

**Proof.** Let  $\mu = p(\lambda_0)$ . Then  $\lambda_0$  is a root of the polynomial  $\mu - p$ , and therefore

$$\mu - p(\lambda) = (\lambda - \lambda_0)q(\lambda)$$

for some polynomial  $q$  over  $\mathbb{C}$ . Applying the homomorphism  $\tau$ , we get

$$\mu e - p(x) = (x - \lambda_0 e)q(x) = q(x)(x - \lambda_0 e).$$

If  $\mu \in \rho(p(x))$ , it follows that

$$(x - \lambda_0 e)(q(x)R(\mu; p(x))) = (R(\mu; p(x))q(x))(x - \lambda_0 e) = e,$$

so that  $\lambda_0 \in \rho(x)$ . Therefore, if  $\lambda_0 \in \sigma(x)$ , it follows that  $\mu := p(\lambda_0) \in \sigma(p(x))$ . This shows that

$$p(\sigma(x)) \subset \sigma(p(x)).$$

On the other hand, factor the polynomial  $\mu - p$  into linear factors

$$\mu - p(\lambda) = \alpha \prod_{k=1}^n (\lambda - \lambda_k).$$

Note that  $\mu = p(\lambda_k)$  for all  $k = 1, \dots, n$ . Applying the homomorphism  $\tau$ , we get

$$\mu e - p(x) = \alpha \prod_{k=1}^n (x - \lambda_k e).$$

If  $\lambda_k \in \rho(x)$  for all  $k$ , then the product above is in  $G(\mathcal{A})$ , and therefore  $\mu \in \rho(p(x))$ . Consequently, if  $\mu \in \sigma(p(x))$ , there exists  $k \in \{1, \dots, n\}$  such that  $\lambda_k \in \sigma(x)$ , and therefore  $\mu = p(\lambda_k) \in p(\sigma(x))$ . This shows that  $\sigma(p(x)) \subset p(\sigma(x))$ .  $\square$

**Theorem 7.9 (The Beurling–Gelfand spectral radius formula).** *For any element  $x$  of a Banach algebra  $\mathcal{A}$ ,*

$$\exists \lim_n \|x^n\|^{1/n} = r(x).$$

**Proof.** By Theorem 7.8 with the polynomial  $p(\lambda) = \lambda^n$  ( $n \in \mathbb{N}$ ),

$$\sigma(x^n) = \sigma(x)^n := \{\lambda^n; \lambda \in \sigma(x)\}.$$

Hence by (8) applied to  $x^n$ ,  $|\lambda^n| \leq \|x^n\|$  for all  $n$  and  $\lambda \in \sigma(x)$ . Thus  $|\lambda| \leq \|x^n\|^{1/n}$  for all  $n$ , and therefore

$$|\lambda| \leq \liminf_n \|x^n\|^{1/n}$$

for all  $\lambda \in \sigma(x)$ . Taking the supremum over all such  $\lambda$ , we obtain

$$r(x) \leq \liminf_n \|x^n\|^{1/n}. \quad (11)$$

For each  $x^* \in \mathcal{A}^*$ , the complex function  $x^*R(\cdot)$  is analytic in  $\rho(x)$ , and since  $\sigma(x)$  is contained in the closed disc around 0 with radius  $r(x)$ ,  $\rho(x)$  contains the open ‘annulus’  $r(x) < |\lambda| < \infty$ . By Laurent’s theorem,  $x^*R(\cdot)$  has a unique Laurent series expansion in this annulus. In the possibly smaller annulus  $\|x\| < |\lambda| < \infty$  (cf. (9)), this function has the expansion (cf. (7))

$$x^*R(\lambda) = \sum_{n=0}^{\infty} \frac{x^*(x^n)}{\lambda^{n+1}}.$$

This is a Laurent expansion; by uniqueness, this is *the* Laurent expansion of  $x^*R(\cdot)$  in the full annulus  $r(x) < |\lambda| < \infty$ . The convergence of the series implies in particular that

$$\sup_n |x^*(\frac{x^n}{\lambda^{n+1}})| < \infty \quad (|\lambda| > r(x))$$

for all  $x^* \in \mathcal{A}^*$ . By Corollary 6.7, it follows that (whenever  $|\lambda| > r(x)$ )

$$\sup_n \left\| \frac{x^n}{\lambda^{n+1}} \right\| := M_\lambda < \infty.$$

Hence, for all  $n \in \mathbb{N}$  and  $|\lambda| > r(x)$ ,

$$\|x^n\| \leq M_\lambda |\lambda|^{n+1},$$

so that

$$\limsup_n \|x^n\|^{1/n} \leq |\lambda|,$$

and therefore

$$\limsup_n \|x^n\|^{1/n} \leq r(x). \quad (12)$$

The conclusion of the theorem follows from (11) and (12).  $\square$

**Definition 7.10.** The element  $x$  (of a Banach algebra) is said to be *quasi-nilpotent* if  $\lim \|x^n\|^{1/n} = 0$ .

By Theorem 7.9, the element  $x$  is quasi-nilpotent if and only if  $r(x) = 0$ , that is, iff  $\sigma(x) = \{0\}$ .

In particular, *nilpotent* elements ( $x^n = 0$  for some  $n$ ) are quasi-nilpotent.

We consider now the *boundary* points of the open set  $G(\mathcal{A})$ .

**Theorem 7.11.** *Let  $x$  be a boundary point of  $G(\mathcal{A})$ . Then  $x$  is a (two-sided) topological divisor of zero, that is, there exists sequences of unit vectors  $\{x_n\}$  and  $\{x'_n\}$  such that  $x_n x \rightarrow 0$  and  $x x'_n \rightarrow 0$ .*

**Proof.** Let  $x \in \partial G$  ( $:=$  the boundary of  $G := G(\mathcal{A})$ ). Since  $G$  is open, there exists a sequence  $\{y_n\} \subset G$  such that  $y_n \rightarrow x$  and  $x \notin G$ .

If  $\{\|y_n^{-1}\|\}$  is bounded (say by  $0 < M < \infty$ ), and  $n$  is so large that  $\|x - y_n\| < 1/M$ , then

$$\|y_n^{-1}x - e\| = \|y_n^{-1}(x - y_n)\| \leq \|y_n^{-1}\| \|x - y_n\| < 1,$$

and therefore  $z := y_n^{-1}x \in G$  by Remark 7.4. Hence  $x = y_n z \in G$ , contradiction. Thus  $\{\|y_n^{-1}\|\}$  is unbounded, and has therefore a subsequence  $\{\|y_{n_k}^{-1}\|\}$  diverging to infinity. Define

$$x_k := \frac{y_{n_k}^{-1}}{\|y_{n_k}^{-1}\|} \quad (k \in \mathbb{N}).$$

Then  $\|x_k\| = 1$  and

$$\|x_k x\| = \|x_k y_{n_k} + x_k(x - y_{n_k})\| \leq \frac{1}{\|y_{n_k}^{-1}\|} + \|x_k\| \|x - y_{n_k}\| \rightarrow 0,$$

and similarly  $xx_k \rightarrow 0$ . □

**Theorem 7.12.** *Let  $\mathcal{B}$  be a Banach subalgebra of the Banach algebra  $\mathcal{A}$ . If  $x \in \mathcal{B}$ , denote the spectrum of  $x$  as an element of  $\mathcal{B}$  by  $\sigma_{\mathcal{B}}(x)$ . Then*

- (1)  $\sigma(x) \subset \sigma_{\mathcal{B}}(x)$  and
- (2)  $\partial\sigma_{\mathcal{B}}(x) \subset \partial\sigma(x)$ .

**Proof.** The first inclusion is trivial, since  $G(\mathcal{B}) \subset G(\mathcal{A})$ , so that  $\rho_{\mathcal{B}}(x) \subset \rho(x)$ .

Let  $\lambda \in \partial\sigma_{\mathcal{B}}(x)$ . Equivalently,  $\lambda e - x \in \partial G(\mathcal{B})$ , and therefore, by Theorem 7.11,  $\lambda e - x$  is a topological divisor of zero in  $\mathcal{B}$ , hence in  $\mathcal{A}$ . In particular,  $\lambda e - x \notin G(\mathcal{A})$ , that is,  $\lambda \in \sigma(x)$ . This shows that  $\partial\sigma_{\mathcal{B}}(x) \subset \sigma(x)$ . Since  $\rho_{\mathcal{B}}(x) \subset \rho(x)$ , we obtain (using (1)):

$$\begin{aligned} \partial\sigma_{\mathcal{B}}(x) &= \overline{\rho_{\mathcal{B}}(x)} \cap \sigma_{\mathcal{B}}(x) \subset \overline{[\rho(x) \cap \sigma_{\mathcal{B}}(x)]} \cap \sigma(x) \\ &= \overline{\rho(x)} \cap \sigma(x) = \partial\sigma(x). \end{aligned}$$

□

**Corollary 7.13.** *Let  $\mathcal{B}$  be a Banach subalgebra of the Banach algebra  $\mathcal{A}$ , and  $x \in \mathcal{B}$ . Then  $\sigma_{\mathcal{B}}(x) = \sigma(x)$  if either  $\sigma_{\mathcal{B}}(x)$  is nowhere dense or  $\rho(x)$  is connected.*

**Proof.** If  $\sigma_{\mathcal{B}}(x)$  is nowhere dense, Theorem 7.12 implies that

$$\sigma_{\mathcal{B}}(x) = \partial\sigma_{\mathcal{B}}(x) \subset \partial\sigma(x) \subset \sigma(x) \subset \sigma_{\mathcal{B}}(x),$$

and the conclusion follows.

If  $\rho(x)$  is connected and  $\sigma(x)$  is a *proper* subset of  $\sigma_{\mathcal{B}}(x)$ , there exists  $\lambda \in \sigma_{\mathcal{B}}(x) \cap \rho(x)$ , and it can be connected with the point at  $\infty$  by a continuous curve lying in  $\rho(x)$ . Since  $\sigma_{\mathcal{B}}(x)$  is compact, the curve meets  $\partial\sigma_{\mathcal{B}}(x)$  at some point, and therefore  $\partial\sigma_{\mathcal{B}}(x) \cap \rho(x) \neq \emptyset$ , contradicting Statement 2. of Theorem 7.12. □

If  $M$  is an (two sided,  $\neq \mathcal{A}$ ) ideal in  $\mathcal{A}$ , the quotient space  $\mathcal{A}/M$  is an algebra. If  $M$  is *closed*, the quotient norm on the Banach space  $\mathcal{A}/M$  (cf. Theorem 6.16) satisfies the requirements 1. and 2. of Definition 7.1, that is,  $\mathcal{A}/M$  is a Banach algebra. We shall discuss the case of *commutative* Banach algebras with more detail.

## 7.2 Commutative Banach algebras

Let  $\mathcal{A}$  be a (complex, unital) *commutative* Banach algebra.

1. Let  $x \in \mathcal{A}$ . Then  $x \in G(\mathcal{A})$  if and only if  $x\mathcal{A} = \mathcal{A}$ . Equivalently,  $x$  is singular if and only if  $x\mathcal{A} \neq \mathcal{A}$ , that is, iff  $x$  is contained in an ideal  $(x\mathcal{A})$ . Ideals are contained therefore in the *closed* set  $G(\mathcal{A})^c$ , and it follows that *the closure of an ideal is an ideal* (recall that we reserve the word ‘ideal’ to  $\mathcal{A}$ -invariant subspaces *not equal to*  $\mathcal{A}$ ).

2. A *maximal* ideal  $M$  (in  $\mathcal{A}$ ) is an ideal in  $\mathcal{A}$  with the property that if  $N$  is an ideal in  $\mathcal{A}$  containing  $M$ , then  $N = M$ . Since the closure of  $M$  is an ideal containing  $M$ , it follows that *maximal ideals are closed*. In particular  $\mathcal{A}/M$  is a Banach algebra, and *is also a field* (by a well-known elementary algebraic characterization of maximal ideals). By the Gelfand–Mazur theorem (Theorem 7.7),  $\mathcal{A}/M$  is isomorphic (and isometric) to  $\mathbb{C}$ . Composing the natural homomorphism  $\mathcal{A} \rightarrow \mathcal{A}/M$  with this isomorphism, we obtain a (norm-decreasing) homomorphism  $\phi_M$  of  $\mathcal{A}$  onto  $\mathbb{C}$ , *whose kernel is*  $M$ . Thus, for any  $x \in \mathcal{A}$ ,  $\phi_M(x)$  is the unique scalar  $\lambda$  such that  $x + M = \lambda e + M$ . Equivalently,  $\phi_M(x)$  is uniquely determined by the relation  $\phi_M(x)e - x \in M$ .

Let  $\Phi$  denote the set of all homomorphisms of  $\mathcal{A}$  onto  $\mathbb{C}$ . Note that  $\phi \in \Phi$  iff  $\phi$  is a homomorphism of  $\mathcal{A}$  *into*  $\mathbb{C}$  such that  $\phi(e) = 1$  (equivalently, iff  $\phi$  is a *non-zero* homomorphism of  $\mathcal{A}$  into  $\mathbb{C}$ ).

The mapping  $M \rightarrow \phi_M$  described above is a mapping of the set  $\mathcal{M}$  of all maximal ideals into  $\Phi$ .

On the other hand, if  $\phi \in \Phi$ , and  $M := \ker \phi$ , then (by Noether’s ‘first homomorphism theorem’)  $\mathcal{A}/M$  is isomorphic to  $\mathbb{C}$ , and is therefore a field. By the algebraic characterization of maximal ideals mentioned before, it follows that  $M$  is a maximal ideal. We have  $\ker \phi_M = M = \ker \phi$ . For any  $x \in \mathcal{A}$ ,  $x - \phi(x)e \in \ker \phi = \ker \phi_M$ , hence  $0 = \phi_M(x - \phi(x)e) = \phi_M(x) - \phi(x)$ . This shows that  $\phi = \phi_M$ , that is, the mapping  $M \rightarrow \phi_M$  is *onto*. It is clearly one-to-one, because if  $M, N \in \mathcal{M}$  are such that  $\phi_M = \phi_N$ , then  $M = \ker \phi_M = \ker \phi_N = N$ . We conclude that the mapping  $M \rightarrow \phi_M$  is a bijection of  $\mathcal{M}$  onto  $\Phi$ , with the inverse mapping  $\phi \rightarrow \ker \phi$ .

3. If  $J$  is an ideal, then  $e \notin J$ . The set  $\mathcal{U}$  of all ideals containing  $J$  is partially ordered by inclusion, and every totally ordered subset  $\mathcal{U}_0$  has the upper bound  $\bigcup \mathcal{U}_0$  (which is an *ideal* because the identity does *not* belong to it) in  $\mathcal{U}$ . By Zorn’s lemma,  $\mathcal{U}$  has a maximal element, which is clearly a maximal ideal containing  $J$ . Thus, *every ideal is contained in a maximal ideal*. Together with 1., this shows that *an element  $x$  is singular iff it is contained in a maximal ideal  $M$* . By the bijection established above between  $\mathcal{M}$  and  $\Phi$ , this means that  $x$  is singular iff  $\phi(x) = 0$  for some  $\phi \in \Phi$ . Therefore, for any  $x \in \mathcal{A}$ ,  $\lambda \in \sigma(x)$  iff  $\phi(\lambda e - x) = 0$  for some  $\phi \in \Phi$ , that is, iff  $\lambda = \phi(x)$  for some  $\phi$ . Thus

$$\sigma(x) = \{\phi(x); \phi \in \Phi\}. \quad (1)$$

Therefore

$$\sup_{\phi \in \Phi} |\phi(x)| = \sup_{\lambda \in \sigma(x)} |\lambda| := r(x). \quad (2)$$

By (9) in Section 7.1, it follows in particular that  $|\phi(x)| \leq \|x\|$ , so that the homomorphism  $\phi$  is necessarily continuous, with norm  $\leq 1$ . Actually, since  $\|\phi\| \geq |\phi(e)| = 1$ , we have  $\|\phi\| = 1$  (for all  $\phi \in \Phi$ ).

We have then  $\Phi \subset S^*$ , where  $S^*$  is the strongly closed unit ball of  $\mathcal{A}^*$ . By Theorem 5.24 (Alaoglu's theorem),  $S^*$  is compact in the *weak\** topology. If  $\phi_\alpha$  is a net in  $\Phi$  converging *weak\** to  $h \in S^*$ , then

$$h(xy) = \lim_{\alpha} \phi_{\alpha}(xy) = \lim_{\alpha} \phi_{\alpha}(x)\phi_{\alpha}(y) = h(x)h(y)$$

and  $h(e) = \lim_{\alpha} \phi_{\alpha}(e) = 1$ , so that  $h \in \Phi$ . This shows that  $\Phi$  is a *closed* subset of the compact space  $S^*$  (with the *weak\** topology), hence  $\Phi$  is compact (in this topology). If  $\phi, \psi \in \Phi$  are distinct, then there exists  $x_0 \in \mathcal{A}$  such that  $\epsilon := |\phi(x_0) - \psi(x_0)| > 0$ . Then  $N(\phi; x_0; \epsilon/2) \cap \Phi$  and  $N(\psi; x_0; \epsilon/2) \cap \Phi$  are disjoint neighbourhoods of  $\phi$  and  $\psi$  in the relative *weak\** topology on  $\Phi$  (cf. notations preceding Theorem 5.24). We conclude that  $\Phi$  with this topology (called the *Gelfand topology on  $\Phi$* ) is a *compact Hausdorff space*.

4. For any  $x \in \mathcal{A}$ , let  $\hat{x} := (\kappa x)|_{\Phi}$ , the restriction of  $\kappa x : \mathcal{A}^* \rightarrow \mathbb{C}$  to  $\Phi$  (where  $\kappa$  is the canonical embedding of  $\mathcal{A}$  in its second dual). By definition of the *weak\** topology,  $\kappa x$  is continuous on  $\mathcal{A}^*$  (with the *weak\** topology), therefore its restriction  $\hat{x}$  to  $\Phi$  (with the Gelfand topology) is continuous. The function  $\hat{x} \in C(\Phi)$  is called *the Gelfand transform of  $x$* . By definition

$$\hat{x}(\phi) = \phi(x) \quad (\phi \in \Phi), \quad (3)$$

and therefore, by (1),

$$\hat{x}(\Phi) = \sigma(x) \quad (4)$$

and

$$\|\hat{x}\|_{C(\Phi)} = r(x). \quad (5)$$

Note that the subalgebra  $\hat{\mathcal{A}} := \{\hat{x}; x \in \mathcal{A}\}$  of  $C(\Phi)$  contains  $1 = \hat{e}$  and separates the points of  $\Phi$  (if  $\phi \neq \psi$  are elements of  $\Phi$ , there exists  $x \in \mathcal{A}$  such that  $\phi(x) \neq \psi(x)$ , that is,  $\hat{x}(\phi) \neq \hat{x}(\psi)$ , by (3)).

5. It is also customary to consider  $\mathcal{M}$  with the Gelfand topology of  $\Phi$  transferred to it through the bijection  $M \rightarrow \phi_M$ . In this case the compact Hausdorff space  $\mathcal{M}$  (with this Gelfand topology) is called *the maximal ideal space of  $\mathcal{A}$* , and  $\hat{x}$  is considered as defined on  $\mathcal{M}$  through the above bijection, that is, we write  $\hat{x}(M)$  instead of  $\hat{x}(\phi_M)$ , so that

$$\hat{x}(M) = \phi_M(x) \quad (M \in \mathcal{M}). \quad (6)$$

The basic neighbourhoods for the Gelfand topology on  $\mathcal{M}$  are of the form

$$N(M_0; x_1, \dots, x_n; \epsilon) := \{M \in \mathcal{M}; |\hat{x}_k(M) - \hat{x}_k(M_0)| < \epsilon, k = 1, \dots, n\}.$$

6. The mapping  $\Gamma : x \rightarrow \hat{x}$  is clearly a *representation* of the algebra  $\mathcal{A}$  into the algebra  $C(\Phi)$  (or  $C(\mathcal{M})$ ), that is, an algebra homomorphism sending  $e$  to 1:

$$[\Gamma(x+y)](\phi) = \phi(x+y) = \phi(x) + \phi(y) = (\Gamma x + \Gamma y)(\phi)$$

for all  $\phi \in \Phi$ , etc. It is called the *Gelfand representation* of  $\mathcal{A}$ . By (5), the map  $\Gamma$  is also norm-decreasing (hence continuous), and (cf. also (3))

$$\begin{aligned}\ker \Gamma &= \{x \in \mathcal{A}; r(x) = 0\} = \{x \in \mathcal{A}; \sigma(x) = \{0\}\} \\ &= \{x \in \mathcal{A}; x \text{ is quasi-nilpotent}\} = \bigcap \mathcal{M}.\end{aligned}\quad (7)$$

Note that since  $\Gamma$  is a homomorphism, it follows from (5) that for all  $x, y \in \mathcal{A}$ ,

$$r(x + y) \leq r(x) + r(y); \quad r(xy) \leq r(x)r(y). \quad (8)$$

Since  $r(e) = 1$  trivially, it follows that  $\mathcal{A}$  is a normed algebra for the so-called *spectral norm*  $r(\cdot)$  if and only if  $r(x) = 0$  implies  $x = 0$ , that is (in view of (7)!), iff the so called *radical of  $\mathcal{A}$* ,  $\text{rad } \mathcal{A} := \ker \Gamma$ , is trivial. In that case we say that  $\mathcal{A}$  is *semi-simple*. By (7) and (3), equivalent characterizations of semi-simplicity are:

- (i) The Gelfand representation  $\Gamma$  of  $\mathcal{A}$  is injective.
- (ii)  $\mathcal{A}$  contains no non-zero quasi-nilpotent elements.
- (iii)  $\mathcal{A}$  is a normed algebra for the spectral norm.
- (iv) The maximal ideals of  $\mathcal{A}$  have trivial intersection.
- (v)  $\Phi$  separates the points of  $\mathcal{A}$ .

**Theorem 7.14.** *Let  $\mathcal{A}$  be a commutative Banach algebra. Then  $\mathcal{A}$  is semi-simple and  $\hat{\mathcal{A}}$  is closed in  $C(\Phi)$  if and only if there exists  $K > 0$  such that*

$$\|x\|^2 \leq K\|x^2\| \quad (x \in \mathcal{A}).$$

*In that case, the spectral norm is equivalent to the given norm on  $\mathcal{A}$  and  $\Gamma$  is a homeomorphism of  $\mathcal{A}$  onto  $\hat{\mathcal{A}}$ .  $\Gamma$  is isometric iff  $K = 1$  (i.e.  $\|x\|^2 = \|x^2\|$  for all  $x \in \mathcal{A}$ ).*

**Proof.** If  $\mathcal{A}$  is semi-simple and  $\hat{\mathcal{A}}$  is closed,  $\Gamma$  is a one-to-one continuous linear map of the Banach space  $\mathcal{A}$  onto the Banach space  $\hat{\mathcal{A}}$ . By Corollary 6.11,  $\Gamma$  is a homeomorphism. The continuity of  $\Gamma^{-1}$  means that there exists a constant  $K > 0$  such that  $\|x\| \leq \sqrt{K}\|\hat{x}\|_{C(\Phi)}$  for all  $x \in \mathcal{A}$ . Therefore

$$\begin{aligned}\|x\|^2 &\leq K \left( \sup_{\phi \in \Phi} |\hat{x}(\phi)| \right)^2 = K \sup_{\phi \in \Phi} |\hat{x}^2(\phi)| \\ &= K \|\hat{x}^2\|_{C(\Phi)} \leq K\|x^2\|.\end{aligned}\quad (9)$$

Conversely, if there exists  $K > 0$  such that  $\|x\|^2 \leq K\|x^2\|$  for all  $x \in \mathcal{A}$ , it follows by induction that

$$\|x\| \leq K^{(1/2)+\dots+(1/2^n)}\|x^{2^n}\|^{1/2^n}$$



for all  $n \in \mathbb{N}$  and  $x \in \mathcal{A}$ . Letting  $n \rightarrow \infty$ , it follows that

$$\|x\| \leq Kr(x) = K\|\hat{x}\|_{C(\Phi)} \quad (x \in \mathcal{A}). \quad (10)$$

Hence  $\ker \Gamma = \{0\}$ , that is,  $\mathcal{A}$  is semi-simple, and  $\Gamma$  is a homeomorphism of  $\mathcal{A}$  onto  $\hat{\mathcal{A}}$ . Since  $\mathcal{A}$  is complete, so is  $\hat{\mathcal{A}}$ , that is,  $\hat{\mathcal{A}}$  is closed in  $C(\Phi)$ .

If  $K = 1$  (i.e. if  $\|x\|^2 = \|x^2\|$  for all  $x \in \mathcal{A}$ ), it follows from (10) that  $\|x\| = r(x) = \|\hat{x}\|_{C(\Phi)}$  and  $\Gamma$  is isometric. Conversely, if  $\Gamma$  is isometric, it follows from (9) (with  $K = 1$  and equality throughout) that  $\|x\|^2 = \|x^2\|$  for all  $x$ .  $\square$

## 7.3 Involution

Let  $\mathcal{A}$  be a *semi-simple* commutative Banach algebra. Since  $\Phi$  is a compact Hausdorff space (with the Gelfand topology), and  $\hat{\mathcal{A}}$  is a *separating* subalgebra of  $C(\Phi)$  containing 1, it follows from the Stone–Weierstrass theorem (Theorem 5.39) that  $\hat{\mathcal{A}}$  is *dense* in  $C(\Phi)$  if it is *selfadjoint*. In that case, if  $J : f \rightarrow \bar{f}$  is the conjugation conjugate automorphism of  $C(\Phi)$ , define  $C : \mathcal{A} \rightarrow \mathcal{A}$  by

$$C = \Gamma^{-1}J\Gamma. \quad (1)$$

Since  $J\hat{\mathcal{A}} \subset \hat{\mathcal{A}}$  and  $\Gamma$  maps  $\mathcal{A}$  bijectively onto  $\hat{\mathcal{A}}$  (when  $\mathcal{A}$  is semi-simple),  $C$  is well defined. As a composition of two isomorphisms and the conjugate isomorphism  $J$ ,  $C$  is a conjugate isomorphism of  $\mathcal{A}$  onto itself such that  $C^2 = I$  (because  $J^2 = I$ , where  $I$  denotes the identity operator in the relevant space). Such a map  $C$  is called an *involution*. In the non-commutative case, multiplicativity of the involution is replaced by *anti-multiplicativity*:

$$C(xy) = C(y)C(x).$$

It is also customary to denote  $Cx = x^*$  whenever  $C$  is an involution on  $\mathcal{A}$  (not to be confused with elements of the conjugate space!). An algebra with an involution is then called a *\*-algebra*.

If  $\mathcal{A}$  and  $\mathcal{B}$  are \*-algebras, a \*-homomorphism (or isomorphism)  $f : \mathcal{A} \rightarrow \mathcal{B}$  is a homomorphism (or isomorphism) such that  $f(x^*) = f(x)^*$  for all  $x \in \mathcal{A}$ .

An element  $x$  in a \*-algebra is *normal* if it commutes with its *adjoint*  $x^*$ . Special normal elements are the *selfadjoint* ( $x^* = x$ ) and the *unitary* ( $x^* = x^{-1}$ ) elements. The identity is necessarily selfadjoint (and unitary), because

$$e^* = ee^* = e^{**}e^* = (ee^*)^* = e^{**} = e.$$

Every element  $x$  can be uniquely written as  $x = a + ib$  with  $a, b \in \mathcal{A}$  selfadjoint: we have  $a = \Re x := (x + x^*)/2$ ,  $b = \Im x := (x - x^*)/2i$ , and  $x^* = a - ib$ . Clearly,  $x$  is normal iff  $a, b$  commute.

The ‘canonical involution’  $C$  defined by (1) on a *semi-simple commutative* Banach algebra is *uniquely determined* by the natural relation  $\Gamma C = J\Gamma$ , that is, by the relation

$$\hat{x}^* = \overline{\hat{x}} \quad (x \in \mathcal{A}), \quad (2)$$

which is equivalent to the property that  $\hat{a}$  is real whenever  $a$  is selfadjoint.

In case the Gelfand representation  $\Gamma$  is *isometric*, Relation (2) implies the norm-identity:

$$\|x^*x\| = \|x\|^2 \quad (x \in \mathcal{A}). \quad (3)$$

Indeed

$$\begin{aligned} \|x^*x\| &= \|\Gamma(x^*x)\|_{C(\Phi)} = \|\bar{\hat{x}}\hat{x}\|_{C(\Phi)} = \|\hat{x}\|^2_{C(\Phi)} \\ &= \|\hat{x}\|^2_{C(\Phi)} = \|x\|^2. \end{aligned}$$

A Banach algebra with an involution satisfying the norm-identity (3) is called a *B\*-algebra*. If  $X$  is any compact Hausdorff space,  $C(X)$  is a (commutative) *B\*-algebra* for the involution  $J$ . The Gelfand–Naimark theorem (Theorem 7.16) establishes that this is a universal model (up to *B\*-algebras isomorphism*) for *commutative B\*-algebras*.

Note that in *any B\*-algebra*, the norm-identity (3) implies that the involution is *isometric*:

$$\begin{aligned} \text{Since } \|x\|^2 = \|x^*x\| \leq \|x^*\| \|x\|, \text{ we have } \|x\| \leq \|x^*\|, \text{ hence } \|x^*\| \leq \\ \|x^{**}\| = \|x\|, \text{ and therefore } \|x^*\| = \|x\| \end{aligned}$$

It follows in particular that  $\|\Re x\| = \|(x + x^*)/2\| \leq \|x\|$ , and similarly  $\|\Im x\| \leq \|x\|$ .

We prove now the following converse to some of the preceding remarks.

**Lemma 7.15.** *Let  $\mathcal{A}$  be a commutative *B\*-algebra*. Then it is semi-simple, and its involution coincides with the canonical involution.*

**Proof.** By the norm-identity (3) successively applied to  $x$ ,  $x^*x$ , and  $x^2$ , we have

$$\|x\|^4 = \|x^*x\|^2 = \|(x^*x)^*(x^*x)\| = \|(x^2)^*x^2\| = \|x^2\|^2.$$

Thus  $\|x\|^2 = \|x^2\|$ , and Theorem 7.14 implies that  $\Gamma$  is isometric. In particular  $\mathcal{A}$  is semi-simple, so that the canonical involution  $C$  is well-defined and uniquely determined by the relation  $\Gamma C = J\Gamma$ . The conclusion of the lemma will follow if we prove that the given involution satisfies (2), or equivalently, if we show that  $\hat{a}$  is *real* whenever  $a \in \mathcal{A}$  is selfadjoint (with respect to the given involution).

Suppose then that  $a \in \mathcal{A}$  is selfadjoint, but  $\beta := \Im \hat{a}(\phi) \neq 0$  for some  $\phi \in \Phi$ . Let  $\alpha := \Re \hat{a}(\phi)$  and  $b := (1/\beta)(a - \alpha e)$ . Then  $b$  is selfadjoint, and  $\hat{b}(\phi) = i$ . For any *real*  $\lambda$ , since  $\Gamma$  is isometric,

$$\begin{aligned} (1 + \lambda)^2 &= |(1 + \lambda)i|^2 = |(\Gamma(b + i\lambda e))(\phi)|^2 \\ &\leq \|\Gamma(b + i\lambda e)\|_{C(\Phi)}^2 = \|b + i\lambda e\|^2 = \|(b + i\lambda e)^*(b + i\lambda e)\| \\ &= \|(b - i\lambda e)(b + i\lambda e)\| = \|b^2 + \lambda^2 e\| \leq \|b^2\| + \lambda^2. \end{aligned}$$

Therefore  $2\lambda < \|b^2\|$ , which is absurd since  $\lambda$  is arbitrary.  $\square$

Putting together all the ingredients accumulated above, we obtain the following important result.

**Theorem 7.16 (The Gelfand–Naimark theorem).** *Let  $\mathcal{A}$  be a commutative  $B^*$ -algebra. Then the Gelfand representation  $\Gamma$  is an isometric  $*$ -isomorphism of  $\mathcal{A}$  onto  $C(\Phi)$ .*

**Proof.** It was observed in the preceding proof that  $\Gamma$  is an isometry of  $\mathcal{A}$  onto  $\hat{\mathcal{A}}$ . It follows in particular that  $\hat{\mathcal{A}}$  is a closed subalgebra of  $C(\Phi)$ . By Lemma 7.15,  $\hat{\mathcal{A}}$  is selfadjoint, and coincides therefore with  $C(\Phi)$ , by the Stone–Weierstrass theorem (Corollary 5.35). Since the involution on  $C(\Phi)$  is  $J$ , it follows from Lemma 7.15 that  $\Gamma$  is a  $*$ -isomorphism.  $\square$

Let  $\mathcal{A}$  be any (not necessarily commutative!)  $B^*$ -algebra, and let  $x \in \mathcal{A}$  be *selfadjoint*. Denote by  $[x]$  the closure in  $\mathcal{A}$  of the set  $\{p(x); p \text{ complex polynomial}\}$ . Then  $[x]$  is clearly the  $B^*$ -subalgebra of  $\mathcal{A}$  generated by  $x$ , and it is commutative. Let  $\Gamma : y \rightarrow \hat{y}$  be the Gelfand representation of  $[x]$ . Since it is a  $*$ -isomorphism and  $x$  is selfadjoint,  $\hat{x}$  is *real*. By (4) in Section 7.2, this means that  $\sigma_{[x]}(x)$  is real. In particular, it is nowhere dense in  $\mathbb{C}$ , and therefore, by Corollary 7.13,  $\sigma(x) = \sigma_{[x]}(x)$  is real. In this argument, we could replace  $\mathcal{A}$  by any  $B^*$ -subalgebra  $\mathcal{B}$  of  $\mathcal{A}$  that contains  $x$ . Hence  $\sigma_{\mathcal{B}}(x) = \sigma_{[x]}(x) = \sigma(x)$ . This proves the following

**Lemma 7.17.** *Let  $\mathcal{B}$  be a  $B^*$ -subalgebra of the  $B^*$ -algebra  $\mathcal{A}$ , and let  $x \in \mathcal{B}$  be selfadjoint. Then  $\sigma(x) = \sigma_{\mathcal{B}}(x) \subset \mathbb{R}$ .*

For arbitrary elements of the subalgebra  $\mathcal{B}$  we still have

**Theorem 7.18.** *Let  $\mathcal{B}$  be a  $B^*$ -subalgebra of the  $B^*$ -algebra  $\mathcal{A}$ . Then  $G(\mathcal{B}) = G(\mathcal{A}) \cap \mathcal{B}$  and  $\sigma_{\mathcal{B}}(x) = \sigma(x)$  for all  $x \in \mathcal{B}$ .*

**Proof.** Since  $G(\mathcal{B}) \subset G(\mathcal{A}) \cap \mathcal{B}$  trivially, we must show that if  $x \in \mathcal{B}$  has an inverse  $x^{-1} \in \mathcal{A}$ , then  $x^{-1} \in \mathcal{B}$ . The element  $x^*x \in \mathcal{B}$  is selfadjoint, and has clearly the inverse  $x^{-1}(x^{-1})^*$  in  $\mathcal{A}$ :

$$[x^{-1}(x^{-1})^*][x^*x] = x^{-1}(xx^{-1})^*x = x^{-1}x = e,$$

and similarly for multiplication in reversed order. Thus  $0 \notin \sigma(x^*x) = \sigma_{\mathcal{B}}(x^*x)$  by Lemma 7.17. Hence  $x^*x \in G(\mathcal{B})$ , and therefore the inverse  $x^{-1}(x^{-1})^*$  belongs to  $G(\mathcal{B})$ . Consequently  $x^{-1} = [x^{-1}(x^{-1})^*]x^* \in \mathcal{B}$ , as wanted.

It now follows that  $\rho_{\mathcal{B}}(x) = \rho(x)$ , hence  $\sigma_{\mathcal{B}}(x) = \sigma(x)$ , for all  $x \in \mathcal{B}$ .  $\square$

## 7.4 Normal elements

**Terminology 7.19.** If  $x$  is a *normal* element of the arbitrary  $B^*$ -algebra  $\mathcal{A}$ , we still denote by  $[x]$  the  $B^*$ -subalgebra generated by  $x$ , that is, the closure in  $\mathcal{A}$  of all complex polynomials in  $x$  and  $x^*$ ,  $\sum \alpha_{kj}x^k(x^*)^j$  (finite sums, with  $\alpha_{kj} \in \mathbb{C}$ ). Since  $x$  is normal, it is clear that  $[x]$  is a *commutative*  $B^*$ -algebra. By Theorem 7.16, the Gelfand representation  $\Gamma$  is an isometric  $*$ -isomorphism of  $[x]$  onto  $C(\Phi)$ , where  $\Phi$  denotes the space of all non-zero complex homomorphisms of  $[x]$  (with the Gelfand topology).

If  $\phi, \psi \in \Phi$  are such that  $\hat{x}(\phi) = \hat{x}(\psi)$ , then  $\hat{x}^*(\phi) = \overline{\hat{x}(\phi)} = \overline{\hat{x}(\psi)} = \hat{x}^*(\psi)$ , and therefore  $\hat{y}(\phi) = \hat{y}(\psi)$  for all  $y \in [x]$ . Since  $\{\hat{y}; y \in [x]\}$  separates the points of  $\Phi$ , it follows that  $\phi = \psi$ , so that  $\hat{x} : \Phi \rightarrow \sigma(x)$  (cf. (4) in Section 7.2) is a continuous bijective map. Since both  $\Phi$  and  $\sigma(x)$  are compact Hausdorff spaces, the map  $\hat{x}$  is a homeomorphism. It induces the isometric  $*$ -isomorphism

$$\Xi : f \in C(\sigma(x)) \rightarrow f \circ \hat{x} \in C(\Phi).$$

The composition  $\tau = \Gamma^{-1} \circ \Xi$  is an isometric  $*$ -isomorphism of  $C(\sigma(x))$  and  $[x]$ , that carries the function  $f_1(\lambda) = \lambda$  onto  $\Gamma^{-1}(\hat{x}) = x$ . This isometric  $*$ -isomorphism is called the  $C(\sigma(x))$ -operational calculus for the normal element  $x$  of  $\mathcal{A}$ .

The  $C(\sigma(x))$ -operational calculus for  $x$  is uniquely determined by the weaker property:  $\tau : C(\sigma(x)) \rightarrow \mathcal{A}$  is a continuous  $*$ -representation (that is, a  $*$ -algebra homomorphism sending identity to identity) that sends  $f_1$  onto  $x$ . Indeed,  $\tau$  sends any polynomial  $\sum \alpha_{kj} \lambda^k (\bar{\lambda})^j$  onto  $\sum \alpha_{kj} x^k (x^*)^j$ , and by continuity,  $\tau$  is then uniquely determined on  $C(\sigma(x))$ , since the above polynomials are dense in  $C(\sigma(x))$  by the Stone–Weierstrass theorem (cf. Theorem 5.39).

It is customary to write  $f(x)$  instead of  $\tau(f)$ . Note that  $f(x)$  is a normal element of  $\mathcal{A}$ , for each  $f \in C(\sigma(x))$ . Its adjoint is  $\bar{f}(x)$ .

### Remarks.

- (1) Since  $\tau$  is *onto*  $[x]$ , we have

$$[x] = \{f(x); f \in C(\sigma(x))\}.$$

This means that  $f(x)$  is the limit in  $\mathcal{A}$  of a sequence of polynomials in  $x$  and  $x^*$ .

- (2) Since  $\tau$  is *isometric*, we have

$$\|f(x)\| = \|f\|_{C(\sigma(x))}$$

for all  $f \in C(\sigma(x))$ . Taking in particular  $f = f_1$ , this shows that

$$\|x\| = \|f_1\|_{C(\sigma(x))} = \sup_{\lambda \in \sigma(x)} |\lambda| := r(x).$$

Thus, the spectral radius of a normal element coincides with its norm. Obvious consequences of this fact are that a normal quasi-nilpotent element is necessarily zero, and that the spectrum of a normal element  $x$  contains a complex number with modulus equal to  $\|x\|$ . In particular, if  $x$  is selfadjoint,  $\sigma(x)$  is contained in the closed interval  $[-\|x\|, \|x\|]$  (cf. (8) following Definition 7.5 and Lemma 7.17), and either  $\|x\|$  or  $-\|x\|$  (or both) belong to  $\sigma(x)$ .

**Theorem 7.20 (The spectral mapping and composition theorems).** *Let  $x$  be a normal element of the  $B^*$ -algebra  $\mathcal{A}$ , and let  $f \mapsto f(x)$  be its  $C(\sigma(x))$ -operational calculus. Then for all  $f \in C(\sigma(x))$ ,  $\sigma(f(x)) = f(\sigma(x))$ ,*

and furthermore, for all  $g \in C(\sigma(f(x)))$  (so that necessarily  $g \circ f \in C(\sigma(x))$ ), the identity  $g(f(x)) = (g \circ f)(x)$  is valid.

If  $x, y \in \mathcal{A}$  are normal, and  $f \in C(\sigma(x) \cup \sigma(y))$  has a continuous inverse, then  $f(x) = f(y)$  implies  $x = y$ .

**Proof.** Since  $\tau$  is an isomorphism of  $C(\sigma(x))$  and  $[x]$  that sends 1 to  $e$ , it follows that  $\mu e - f(x) = \tau(\mu - f)$  is singular in  $[x]$  iff  $\mu - f$  is singular in  $C(\sigma(x))$ , that is, iff there exists  $\lambda \in \sigma(x)$  such that  $\mu = f(\lambda)$ . Since  $\sigma(f(x)) = \sigma_{[x]}(f(x))$  by Theorem 7.18, we conclude that  $\mu \in \sigma(f(x))$  iff  $\mu \in f(\sigma(x))$ .

The maps  $g \rightarrow g \circ f$  and  $g \circ f \rightarrow (g \circ f)(x)$  are isometric  $*$ -isomorphisms of  $C(\sigma(f(x))) = C(f(\sigma(x)))$  onto  $C(\sigma(x))$  and of  $C(\sigma(x))$  into  $\mathcal{A}$ , respectively. Their composition  $g \rightarrow (g \circ f)(x)$  is an isometric  $*$ -isomorphism of  $C(\sigma(f(x)))$  into  $\mathcal{A}$ , that carries 1 onto  $e$  and  $f_1$  onto  $f(x)$ . By the uniqueness of the operational calculus for the normal element  $f(x)$ , we have  $(g \circ f)(x) = g(f(x))$  for all  $g \in C(\sigma(f(x)))$ .

The last statement of the theorem follows by taking  $g = f^{-1}$  in the last formula applied to both  $x$  and  $y$ :

$$x = f_1(x) = (g \circ f)(x) = g(f(x)) = g(f(y)) = (g \circ f)(y) = f_1(y) = y.$$

□

## 7.5 General $B^*$ -algebras

A standard example of a generally non-commutative  $B^*$ -algebra is the Banach algebra  $B(X)$  of all bounded linear operators on a Hilbert space  $X$ . The involution is the Hilbert adjoint operation  $T \rightarrow T^*$ . Given  $y \in X$ , the map  $x \in X \rightarrow (Tx, y)$  is a continuous linear functional on  $X$ . By the ‘Little’ Riesz Representation theorem (Theorem 1.37), there exists a unique vector (depending on  $T$  and  $y$ , hence denoted  $T^*y$ ) such that  $(Tx, y) = (x, T^*y)$  for all  $x, y \in X$ . The uniqueness implies that  $T^* : X \rightarrow X$  is linear, and (with the suprema below taken over all unit vectors  $x, y$ )

$$\|T^*\| = \sup |(x, T^*y)| = \sup |(Tx, y)| = \|T\| < \infty.$$

Thus,  $T^* \in B(X)$ , and an easy calculation shows that the map  $T \rightarrow T^*$  is an (isometric) involution on  $B(X)$ . Moreover

$$\begin{aligned} \|T\|^2 &= \|T^*\| \|T\| \geq \|T^*T\| = \sup_{x,y} |(T^*Tx, y)| \\ &= \sup_{x,y} |(Tx, Ty)| \geq \sup_x \|Tx\|^2 = \|T\|^2. \end{aligned}$$

Therefore,  $\|T^*T\| = \|T\|^2$ , and  $B(X)$  is indeed a  $B^*$ -algebra. Any closed selfadjoint subalgebra of  $B(X)$  containing the identity (that is, any  $B^*$ -subalgebra of  $B(X)$ ) is likewise an example of a generally non-commutative  $B^*$ -algebra. The Gelfand–Naimark theorem (Theorem 7.29) establishes that this example is (up to  $B^*$ -algebra isomorphism) the most general example of a  $B^*$ -algebra.

We begin with some preliminaries.

Let  $\mathcal{A}$  be a  $B^*$ -algebra. An element  $x \in \mathcal{A}$  is *positive* if it is selfadjoint and  $\sigma(x) \subset \mathbb{R}^+ := [0, \infty)$ .

Since the operational calculus for a normal element  $x$  is a  $*$ -isomorphism of  $C(\sigma(x))$  and  $[x]$ , the element  $f(x)$  is selfadjoint iff  $f$  is *real* on  $\sigma(x)$ , and by Theorem 7.20, it is positive iff  $f(\sigma(x)) \subset \mathbb{R}^+$ , that is, iff  $f \geq 0$  on  $\sigma(x)$ . In particular, for  $x$  selfadjoint, decompose the real function  $f_1(\lambda) = \lambda$  ( $\lambda \in \mathbb{R}$ ) as  $f_1 = f_1^+ - f_1^-$ , so that  $x = x^+ - x^-$ , where  $x^+ := f_1^+(x)$  and  $x^- := f_1^-(x)$  are both positive elements (since  $f_1^+ \geq 0$  on  $\sigma(x)$ , etc. . .) and  $x^+x^- = x^-x^+ = 0$  (since  $f_1^+f_1^- = 0$  on the spectrum of  $x$ ). We call  $x^+$  and  $x^-$  the *positive part* and the *negative part* of  $x$  respectively.

Denote by  $\mathcal{A}^+$  the set of all positive elements of  $\mathcal{A}$ .

If  $x \in \mathcal{A}^+$ ,  $\sigma(x) \subset [0, \|x\|]$  and  $\|x\| \in \sigma(x)$  (cf. last observation in Section 7.19).

For any real scalar  $\alpha \geq \|x\|$ ,  $\alpha e - x$  is selfadjoint, and by Theorem 7.20,

$$\sigma(\alpha e - x) = \alpha - \sigma(x) \subset \alpha - [0, \|x\|] \subset [\alpha - \|x\|, \alpha] \subset [0, \alpha].$$

Therefore

$$\|\alpha e - x\| = r(\alpha e - x) \leq \alpha. \quad (1)$$

Conversely, if  $x$  is selfadjoint and (1) is satisfied, then

$$\alpha - \sigma(x) = \sigma(\alpha e - x) \subset [-\alpha, \alpha],$$

hence

$$\sigma(x) \subset \alpha + [-\alpha, \alpha] = [0, 2\alpha],$$

and therefore  $x \in \mathcal{A}^+$ . This proves the following

**Lemma 7.21.** *Let  $x \in \mathcal{A}$  be selfadjoint and fix  $\alpha \geq \|x\|$ . Then  $x$  is positive iff  $\|\alpha e - x\| \leq \alpha$ .*

**Theorem 7.22.**  $\mathcal{A}^+$  is a closed positive cone in  $\mathcal{A}$  (i.e. a closed subset of  $\mathcal{A}$ , closed under addition and multiplication by non-negative scalars, such that  $\mathcal{A}^+ \cap (-\mathcal{A}^+) = \{0\}$ ).

**Proof.** Let  $x_n \in \mathcal{A}^+$ ,  $x_n \rightarrow x$ . Then  $x$  is selfadjoint (because  $x_n$  are selfadjoint and the involution is continuous). By Lemma 7.21 with  $\alpha_n := \|x_n\|$  ( $\rightarrow \|x\| := \alpha$ ),

$$\|\alpha e - x\| = \lim_n \|\alpha_n e - x_n\| \leq \lim_n \alpha_n = \alpha,$$

hence  $x \in \mathcal{A}^+$  (by the same lemma).

Let  $x_n \in \mathcal{A}^+$ ,  $\alpha_n := \|x_n\|$ ,  $n = 1, 2$ ,  $x := x_1 + x_2$ , and  $\alpha = \alpha_1 + \alpha_2$  ( $\geq \|x\|$ ). Again by Lemma 7.21,

$$\|\alpha e - x\| = \|(\alpha_1 e - x_1) + (\alpha_2 e - x_2)\| \leq \|\alpha_1 e - x_1\| + \|\alpha_2 e - x_2\| \leq \alpha_1 + \alpha_2 = \alpha,$$

hence  $x \in \mathcal{A}^+$ .

If  $x \in \mathcal{A}^+$  and  $\alpha \geq 0$ , then  $\alpha x$  is selfadjoint and  $\sigma(\alpha x) = \alpha \sigma(x) \subset \mathbb{R}^+$ , so that  $\alpha x \in \mathcal{A}^+$ .

Finally, if  $x \in \mathcal{A}^+ \cap (-\mathcal{A}^+)$ , then  $\sigma(x) \subset \mathbb{R}^+ \cap (-\mathbb{R}^+) = \{0\}$ , so that  $x$  is both selfadjoint and quasi-nilpotent, hence  $x = 0$  (see remarks preceding Theorem 7.20).  $\square$

Let  $f : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  be the positive square root function. Since it belongs to  $C(\sigma(x))$  for any  $x \in \mathcal{A}^+$ , it ‘operates’ on each element  $x \in \mathcal{A}^+$  through the  $C(\sigma(x))$ -operational calculus. The element  $f(x)$  is positive (since  $f \geq 0$  on  $\sigma(x)$ ), and  $f(x)^2 = x$  (since the operational calculus is a homomorphism). It is called *the positive square root of  $x$* , denoted  $x^{1/2}$ . Note that  $x^{1/2} \in [x]$ , which means that it is the limit of polynomials in  $x$ ,  $p_n(x)$ , where  $p_n \rightarrow f$  uniformly on  $\sigma(x)$ . Suppose also  $y \in \mathcal{A}^+$  satisfies  $y^2 = x$ . The polynomials  $q_n(\lambda) = p_n(\lambda^2)$  converge uniformly to  $f(\lambda^2) = \lambda$  on  $\sigma(y)$  (since  $\lambda^2 \in \sigma(x) = \sigma(y)^2$  when  $\lambda \in \sigma(y)$ ). Therefore (by continuity of the operational calculus)  $q_n(y) \rightarrow y$ . But  $q_n(y) = p_n(y^2) = p_n(x) \rightarrow x^{1/2}$ . Hence  $y = x^{1/2}$ , which means that the positive square root is *unique*.

The representation  $x = y^2$  (with  $y \in \mathcal{A}^+$ ) of the positive element  $x$  shows in particular that  $x = y^*y$  (since  $y$  is selfadjoint). This last property characterizes positive elements:

**Theorem 7.23.**

- (i) *The element  $x \in \mathcal{A}$  is positive if and only if  $x = y^*y$  for some  $y \in \mathcal{A}$ .*
- (ii) *If  $x$  is positive, then  $z^*xz$  is positive for all  $z \in \mathcal{A}$ .*
- (iii) *If  $\mathcal{A}$  is a  $B^*$ -subalgebra of  $B(X)$  for some Hilbert space  $X$ , then  $T \in \mathcal{A}$  is positive if and only if  $(Tx, x) \geq 0$  for all  $x \in X$ .*

**Proof.**

(i) The preceding remarks show that we need only to prove that  $x := y^*y$  is positive (for any  $y \in \mathcal{A}$ ). Since it is trivially selfadjoint, we decompose it as  $x = x^+ - x^-$ , and we need only to show that  $x^- = 0$ . Let  $z = yx^-$ . Then since  $x^+x^- = 0$ ,

$$z^*z = x^-y^*yx^- = x^-xx^- = x^-(x^+ - x^-)x^- = -(x^-)^3. \quad (2)$$

But  $(x^-)^3$  is positive; therefore

$$-z^*z \in \mathcal{A}^+. \quad (3)$$

Write  $z = a + ib$  with  $a, b$  selfadjoint elements of  $\mathcal{A}$ . Then  $a^2, b^2 \in \mathcal{A}^+$ , and therefore, by Theorem 7.22,

$$z^*z + zz^* = 2a^2 + 2b^2 \in \mathcal{A}^+. \quad (4)$$

By the remarks following Definition 7.5 and (3)

$$\sigma(-zz^*) \subset \sigma(-z^*z) \cup \{0\} \subset \mathbb{R}^+.$$

Thus,  $-zz^* \in \mathcal{A}^+$ , and so by (4) (cf. Theorem 7.22)

$$z^*z = (z^*z + zz^*) + (-zz^*) \in \mathcal{A}^+.$$

Together with (3), this shows that  $z^*z \in \mathcal{A}^+ \cap (-\mathcal{A}^+)$ , hence  $z^*z = 0$  by Theorem 7.22 (and therefore  $z = 0$  because  $\|z\|^2 = \|z^*z\| = 0$ ). By (2), we conclude that  $x^- = 0$  (because  $x^-$  is both selfadjoint and nilpotent), as wanted.

(ii) Write the positive element  $x$  in the form  $x = y^*y$  with  $y \in \mathcal{A}$  (by (i)). Then (again by (i))  $z^*xz = z^*(y^*y)z = (yz)^*(yz) \in \mathcal{A}^+$ .

(iii) If  $T$  is positive, write  $T = S^*S$  for some  $S \in \mathcal{A}$  (by (i)). Then  $(Tx, x) = (Sx, Sx) \geq 0$  for all  $x \in X$ .

Conversely, if  $(Tx, x) \in \mathbb{R}$  for all  $x \in X$ , then  $(T^*x, x) = (x, Tx) = \overline{(Tx, x)} = (Tx, x)$  for all  $x$ , and by polarization (cf. identity (11) following Definition 1.34)  $(T^*x, y) = (Tx, y)$  for all  $x, y \in X$ , hence  $T^* = T$ .

For any  $\delta > 0$ , we have

$$\|(-\delta I - T)x\|^2 = \|\delta x + Tx\|^2 = \delta^2\|x\|^2 + 2\Re[\delta(x, Tx)] + \|Tx\|^2 \geq \delta^2\|x\|^2,$$

because  $(x, Tx) \geq 0$ . Therefore

$$\|(-\delta I - T)x\| \geq \delta\|x\| \quad (x \in X).$$

This implies that  $T_\delta := -\delta I - T$  is injective (trivially) and has closed range  $:= Y$  (indeed, if  $T_\delta x_n \rightarrow y$ , then

$$\|x_n - x_m\| \leq \delta^{-1}\|T_\delta(x_n - x_m)\| \rightarrow 0;$$

hence  $\exists \lim x_n := x$ , and  $y := \lim_n T_\delta x_n = T_\delta x \in Y$ ).

If  $z \in Y^\perp$ , then for all  $x \in X$ , since  $T_\delta$  is selfadjoint,

$$(x, T_\delta z) = (T_\delta x, z) = 0.$$

Hence  $T_\delta z = 0$ , and therefore  $z = 0$  since  $T_\delta$  is injective. Consequently  $Y = X$ , by Theorem 1.36. This shows that  $T_\delta$  is bijective, and therefore  $T_\delta^{-1} \in B(X)$ , by Corollary 6.11. Thus,  $-\delta \in \rho_{B(X)}(T) = \rho_{\mathcal{A}}(T)$ , by Theorem 7.18. However, since  $T$  is selfadjoint,  $\sigma_{\mathcal{A}}(T) \subset \mathbb{R}$  by Lemma 7.17. Therefore,  $\sigma_{\mathcal{A}}(T) \subset \mathbb{R}^+$ .  $\square$

**Definition 7.24.** Let  $\mathcal{A}$  be a  $B^*$ -algebra, and let  $\mathcal{A}_s$  denote the set of all selfadjoint elements of  $\mathcal{A}$ . A linear functional on  $\mathcal{A}$  is *hermitian* (*positive*) if it is *real-valued* on  $\mathcal{A}_s$  (*non-negative* on  $\mathcal{A}^+$ , respectively).

Note that since  $e \in \mathcal{A}^+$  ( $\sigma(e) = \{1\} \subset \mathbb{R}^+$ ), we have  $\phi(e) \geq 0$  for any positive linear functional  $\phi$ . In particular, if  $\phi$  is *normalized*, that is,  $\phi(e) = 1$ , we call it a *state*. The set of all states on  $\mathcal{A}$  will be denoted by  $\mathcal{S} = \mathcal{S}(\mathcal{A})$ . It will play in the non-commutative case a role as crucial as the role that  $\Phi$  played in the commutative Gelfand–Naimark theorem.

Clearly, the linear functional  $\phi$  is hermitian iff  $\phi(x^*) = \overline{\phi(x)}$  for all  $x \in \mathcal{A}$  (this relation evidently implies that  $\phi(x)$  is real for  $x$  selfadjoint; on the other hand, if  $\phi$  is hermitian, write  $x = a + ib$  with  $a, b \in \mathcal{A}_s$ ; then  $x^* = a - ib$ , and



therefore  $\phi(x^*) = \phi(a) - i\phi(b)$  is the conjugate of  $\phi(x) = \phi(a) + i\phi(b)$ , since  $\phi(a), \phi(b) \in \mathbb{R}$ .

Note that  $\Re\phi(x) = \phi(a) = \phi(\Re x)$ .

If  $\phi$  is *positive*, it is necessarily hermitian (write any selfadjoint  $x$  as  $x^+ - x^-$ ; since  $\phi(x^+), \phi(x^-) \in \mathbb{R}^+$ , we have  $\phi(x) = \phi(x^+) - \phi(x^-) \in \mathbb{R}$ ).

If  $x \in \mathcal{A}_s$ , the element  $\|x\|e - x$  is positive (it is selfadjoint and  $\sigma(\|x\|e - x) = \|x\| - \sigma(x) \subset \|x\| - [-\|x\|, \|x\|] = [0, 2\|x\|] \subset \mathbb{R}^+$ ). Therefore, for any positive linear functional  $\phi$ ,  $\phi(\|x\|e - x) \geq 0$ , that is,  $\phi(x) \leq \phi(e)\|x\|$ . Replacing  $x$  by  $-x$ , we get also  $-\phi(x) = \phi(-x) \leq \phi(e)\| -x \| = \phi(e)\|x\|$ . Therefore,  $|\phi(x)| \leq \phi(e)\|x\|$  for all  $x \in \mathcal{A}_s$ .

Next, for  $x \in \mathcal{A}$  arbitrary, write the complex number  $\phi(x)$  in its polar form  $|\phi(x)|e^{i\theta}$ . Then

$$\begin{aligned} |\phi(x)| &= e^{-i\theta}\phi(x) = \phi(e^{-i\theta}x) \\ &= \Re\phi(e^{-i\theta}x) = \phi(\Re[e^{-i\theta}x]) \leq \phi(e)\|\Re[e^{-i\theta}x]\| \leq \phi(e)\|x\|. \end{aligned}$$

This shows that  $\phi$  is bounded with norm  $\leq \phi(e)$ . On the other hand,  $\phi(e) \leq \|\phi\|\|e\| = \|\phi\|$ . Therefore  $\|\phi\| = \phi(e)$  (in particular, states satisfy  $\|\phi\| = \phi(e) = 1$ ). Conversely, we show below that a bounded linear functional  $\phi$  such that  $\|\phi\| = \phi(e)$  is positive.

**Theorem 7.25.** *A linear functional  $\phi$  on the  $B^*$ -algebra  $\mathcal{A}$  is positive if and only if it is bounded with norm equal to  $\phi(e)$ .*

**Proof.** It remains to prove that if  $\phi$  is a bounded linear functional with norm equal to  $\phi(e)$ , then it is positive. This is trivial if  $\|\phi\| = 0$  (the zero functional is positive!), so we may assume that  $\|\phi\| = \phi(e) = 1$  (replace  $\phi$  by  $\psi := \phi/\|\phi\|$ ; if we prove that  $\psi$  is positive, the same is true for  $\phi = \|\phi\|\psi$ ).

It suffices to prove that  $\phi(x) \geq 0$  for *unit* vectors  $x \in \mathcal{A}^+$ . Write  $\phi(x) = \alpha + i\beta$  with  $\alpha, \beta$  real. If  $\beta \neq 0$ , define  $y = \beta^{-1}(x - \alpha e)$ . Then  $y$  is selfadjoint and  $\phi(y) = i$ . For all  $n \in \mathbb{N}$ ,

$$\begin{aligned} |(1+n)i|^2 &= |\phi(y + ine)|^2 \leq \|y + ine\|^2 = \|(y + ine)^*(y + ine)\| \\ &= \|(y - ine)(y + ine)\| = \|y^2 + n^2e\| \leq \|y\|^2 + n^2, \end{aligned}$$

and we get the absurd statement  $1 + 2n \leq \|y\|^2$  for all  $n \in \mathbb{N}$ . Therefore,  $\beta = 0$  and  $\phi(x) = \alpha$ .

Since  $\sigma(x) \subset [0, \|x\|] = [0, 1]$ , we have  $\sigma(e - x/2) = 1 - (1/2)\sigma(x) \subset 1 - (1/2)[0, 1] = [1/2, 1]$ , and therefore  $\|e - x/2\| = r(e - x/2) \leq 1$ . Now

$$1 - \alpha/2 \leq |1 - \alpha/2| = |\phi(e - x/2)| \leq \|e - x/2\| \leq 1,$$

hence  $\alpha \geq 0$ . □

It is now clear that the set of states  $\mathcal{S} = \mathcal{S}(\mathcal{A})$  is not empty. Indeed, consider the linear functional  $\phi_0$  on  $\mathbb{C}e$  defined by  $\phi_0(\lambda e) = \lambda$ . Then clearly  $\|\phi_0\| = 1 = \phi_0(e)$ . By the Hahn–Banach theorem,  $\phi_0$  extends to a bounded linear functional  $\phi$  on  $\mathcal{A}$  satisfying  $\|\phi\| = 1 = \phi(e)$ . By Theorem 7.25,  $\phi \in \mathcal{S}$  (actually, any state on  $\mathcal{A}$  is an extension of  $\phi_0$ ).

## 7.6 The Gelfand–Naimark–Segal construction

Let  $\mathcal{A}$  be any  $B^*$ -algebra. Given a state  $s \in \mathcal{S} := \mathcal{S}(\mathcal{A})$ , we set

$$(x, y)_s := s(y^*x) \quad (x, y \in \mathcal{A}).$$

By Theorem 7.23, the form  $(\cdot, \cdot)_s$  is a semi-inner product (s.i.p.). The induced semi-norm

$$\|\cdot\|_s := (x, x)^{1/2} = s(x^*x)^{1/2} \quad (x \in \mathcal{A})$$

is ‘normalized’, that is,  $\|e\|_s = 1$  (because  $s(e) = 1$ ), and continuous on  $\mathcal{A}$  (by continuity of  $s$ , of the involution, and the multiplication).

Let

$$J_s := \{x \in \mathcal{A}; \|x\|_s = 0\} = \|\cdot\|_s^{-1}(\{0\}).$$

The properties of the semi-norm imply that  $J_s$  is a closed subspace of  $\mathcal{A}$ .

By the Cauchy–Schwarz inequality for the semi inner product, if  $x \in J_s$ , then  $(x, y)_s = (y, x)_s = 0$  for all  $y \in \mathcal{A}$ .

This implies that  $J_s$  is left  $\mathcal{A}$ -invariant (i.e. a left ideal), because for all  $x \in J_s$  and  $y \in \mathcal{A}$

$$\|yx\|_s^2 := s((yx)^*(yx)) = s((y^*yx)^*x) = (x, y^*yx)_s = 0.$$

**Lemma 7.26.** *Let  $s \in \mathcal{S}(\mathcal{A})$ . Then for all  $x, y \in \mathcal{A}$ ,*

$$\|xy\|_s \leq \|x\|_s \|y\|_s.$$

**Proof.** By Theorem 7.23,  $x^*x \in \mathcal{A}^+$ , and since  $\|x^*x\| = \|x\|^2$ , it follows that  $\sigma(x^*x) \subset [0, \|x\|^2]$ . Therefore, the selfadjoint element  $\|x\|^2 e - x^*x$  has spectrum contained in  $\|x\|^2 - [0, \|x\|^2] = [0, \|x\|^2]$ , and is therefore positive. By the second statement in Theorem 7.23, it follows that  $y^*(\|x\|^2 e - x^*x)y$  is positive, that is,  $\|x\|^2 y^*y - (xy)^*(xy) \in \mathcal{A}^+$ . Hence  $s(\|x\|^2 y^*y - (xy)^*(xy)) \geq 0$ , that is,  $\|x\|^2 \|y\|_s^2 - \|xy\|_s^2 \geq 0$ .  $\square$

The s.i.p.  $(\cdot, \cdot)_s$  induces an inner product (same notation) on the quotient space  $\mathcal{A}/J_s$ :

$$(x + J_s, y + J_s)_s := (x, y)_s \quad (x, y \in \mathcal{A}).$$

This definition is independent on the cosets representatives, because if  $x + J_s = x' + J_s$  and  $y + J_s = y' + J_s$ , then  $x - x', y - y' \in J_s$ , and therefore (dropping the subscript  $s$ )

$$(x', y') = (x' - x, y') + (x, y) + (x, y' - y) = (x, y)$$

(the first and third summands vanish, because one of the factors of the s.i.p. is in  $J_s$ ).

If  $\|x + J_s\|_s^2 := (x + J_s, x + J_s)_s = 0$ , then  $\|x\|_s^2 = (x, x)_s = 0$ , that is,  $x \in J_s$ , hence  $x + J_s$  is the zero coset  $J_s$ . This means that  $\|\cdot\|_s$  is a norm on  $\mathcal{A}/J_s$ . Let  $X_s$

be the completion of  $\mathcal{A}/J_s$  with respect to this norm. Then  $X_s$  is a Hilbert space (its inner product is the unique continuous extension of  $(\cdot, \cdot)_s$  from the dense subspace  $\mathcal{A}/J_s \times \mathcal{A}/J_s$  to  $X_s \times X_s$ ; the extension is also denoted by  $(\cdot, \cdot)_s$ ).

For each  $x \in \mathcal{A}$ , consider the map

$$L_x := L_x^s : \mathcal{A}/J_s \rightarrow \mathcal{A}/J_s$$

defined by

$$L_x(y + J_s) = xy + J_s \quad (y \in \mathcal{A}).$$

It is well defined, because if  $y, y'$  represent the same coset, then  $\|y - y'\|_s = 0$ , and therefore, by Lemma 7.26,  $\|x(y - y')\|_s = 0$ , which means that  $xy$  and  $xy'$  represent the same coset.

The map  $L_x$  is clearly linear. It is also *bounded*, with operator norm (on the normed space  $\mathcal{A}/J_s$ )  $\|L_x\| \leq \|x\|$ : indeed, by Lemma 7.26, for all  $y \in \mathcal{A}$ ,

$$\|L_x(y + J_s)\|_s = \|xy + J_s\|_s = \|xy\|_s \leq \|x\|\|y\|_s = \|x\|\|y + J_s\|_s.$$

Therefore,  $L_x$  extends uniquely by continuity to a bounded operator on  $X_s$  (also denoted  $L_x$ ), with operator norm  $\|L_x\| \leq \|x\|$ .

Since  $L_e$  is the identity operator on  $\mathcal{A}/J_s$ , then  $L_e = I$ , the identity operator on  $X_s$ . Routine calculation shows that  $x \rightarrow L_x$  is an algebra homomorphism of  $\mathcal{A}$  into  $B(\mathcal{A}/J_s)$ , and a continuity argument implies that it is a homomorphism of  $\mathcal{A}$  into  $B(X_s)$ .

For all  $x, y, z \in \mathcal{A}$ , we have (dropping the index  $s$ )

$$\begin{aligned} (L_x(y + J), z + J) &= (xy + J, z + J) = (xy, z) = s(z^*xy) = s((x^*z)^*y) \\ &= (y, x^*z) = (y + J, x^*z + J) = (y + J, L_{x^*}(z + J)). \end{aligned}$$

By continuity, we obtain the identity

$$(L_x u, v) = (u, L_{x^*} v) \quad (u, v \in X_s),$$

that is

$$(L_x)^* = L_{x^*} \quad (x \in \mathcal{A}).$$

We conclude that  $L : x \rightarrow L_x$  is a (norm-decreasing)  $*$ -homomorphism of  $\mathcal{A}$  into  $B(X_s)$  that sends  $e$  onto  $I$  (such a homomorphism is called a *representation of  $\mathcal{A}$  on the Hilbert space  $X_s$* ). The construction of the ‘canonical’ representation  $L$  is referred to as the *Gelfand–Naimark–Segal (GNS) construction*; accordingly,  $L := L^s : x \rightarrow L_x$  will be called the GNS representation (associated with the given state  $s$  on  $\mathcal{A}$ ).

Consider the unit vector  $v_s := e + J_s \in X_s$  (it is a unit vector because  $\|v_s\|_s = \|e\|_s = s(e^*e)^{1/2} = 1$ ). By definition of  $X_s$ , the set

$$\{L_x v_s; x \in \mathcal{A}\} = \{x + J_s; x \in \mathcal{A}\} = \mathcal{A}/J_s,$$

is dense in  $X_s$ . We express this fact by saying that *the representation  $L : x \rightarrow L_x$  is cyclic, with cyclic vector  $v_s$* .

Note also the identity (dropping the index  $s$ )

$$s(x) = s(e^*x) = (x, e) = (x + J, e + J) = (L_x(e + J), e + J) = (L_x v, v). \quad (1)$$

Thus, the state  $s$  is realized through the representation  $L$  as the composition  $s_v \circ L$ , where  $s_v$  is the so-called *vector state* on  $B(X_s)$  defined (through the unit vector  $v$ ) by

$$s_v(T) = (Tv, v) \quad (T \in B(X_s)).$$

The GNS representation is ‘universal’ in a certain sense which we proceed to specify.

Suppose  $\Lambda : x \rightarrow \Lambda_x$  is *any* cyclic representation of  $\mathcal{A}$  on a Hilbert space  $Z$ , with unit cyclic vector  $u \in Z$  such that  $s = s_u \circ \Lambda$ . Then by (1), for all  $x \in \mathcal{A}$ ,

$$\begin{aligned} \|\Lambda_x u\|_Z^2 &= (\Lambda_x u, \Lambda_x u)_Z = (\Lambda_{x^*x} u, u)_Z \\ &= (s_u \circ \Lambda)(x^*x) = s(x^*x) = (L_{x^*x} v, v)_s = \|L_x v\|_s^2. \end{aligned} \quad (2)$$

Since  $L$  and  $\Lambda$  are *cyclic* representations on  $X_s$  and  $Z$  with respective cyclic vectors  $v$  and  $u$ , the subspaces  $\tilde{X}_s := \{L_x v; x \in \mathcal{A}\}$  and  $\tilde{Z} := \{\Lambda_x u; x \in \mathcal{A}\}$  are dense in  $X_s$  and  $Z$ , respectively. Define  $U : \tilde{X}_s \rightarrow \tilde{Z}$  by

$$UL_x v = \Lambda_x u \quad (x \in \mathcal{A}).$$

It follows from (2) that  $U$  is a linear isometry of  $\tilde{X}_s$  onto  $\tilde{Z}$ . It extends uniquely by continuity as a linear isometry of  $X_s$  onto  $Z$ . Thus,  $U$  is a Hilbert space isomorphism of  $X_s$  onto  $Z$ , that carries  $v$  onto  $u$  (because  $Uv = UIv = UL_e v := \Lambda_e u = Iu = u$ , where we use the notation  $I$  for the identity operator in both Hilbert spaces). We have

$$\begin{aligned} (UL_x)(L_y v) &= UL_{xy} v = \Lambda_{xy} u = \Lambda_x \Lambda_y u \\ &= \Lambda_x (UL_y v) = (\Lambda_x U)(L_y v). \end{aligned}$$

Thus,  $UL_x = \Lambda_x U$  on the dense subspace  $\tilde{X}_s$  of  $X_s$ ; by continuity of the operators, it follows that  $UL_x = \Lambda_x U$ , that is,  $\Lambda_x = UL_x U^{-1}$  for all  $x \in \mathcal{A}$  (one says that the representations  $\Lambda$  and  $L$  are *unitarily equivalent*, through the *unitary equivalence*  $U : X_s \rightarrow Z$ ). This concludes the proof of the following

**Theorem 7.27 (The Gelfand–Naimark–Segal–theorem).** *Let  $s$  be a state of the  $B^*$ -algebra  $\mathcal{A}$ . Then the associated GNS representation  $L := L^s$  is a cyclic norm-decreasing representation of  $\mathcal{A}$  on the Hilbert space  $X = X_s$ , with a unit cyclic vector  $v = v_s$  such that  $s = s_v \circ L$ . If  $\Lambda$  is any cyclic representation of  $\mathcal{A}$  on a Hilbert space  $Z$  with unit cyclic vector  $u$  such that  $s = s_u \circ \Lambda$ , then  $\Lambda$  is unitarily equivalent to  $L$  under a unitary equivalence  $U : X \rightarrow Z$  such that  $Uv = u$ .*

**Lemma 7.28.** *Let  $\mathcal{A}$  be a  $B^*$ -algebra, and let  $\mathcal{S}$  be the set of all states of  $\mathcal{A}$ . Then, for each  $x \in \mathcal{A}$ ,*

- (i)  $\sigma(x) \subset \{s(x); s \in \mathcal{S}\}$ ;
- (ii) *if  $x$  is normal, then  $\|x\| = \max_{s \in \mathcal{S}} |s(x)|$ ; for  $x$  arbitrary,  $\|x\| = \max_{s \in \mathcal{S}} \|x\|_s$ , where  $\|x\|_s^2 := s(x^*x)$ ;*
- (iii) *if  $s(x) = 0$  for all  $s \in \mathcal{S}$ , then  $x = 0$ ;*
- (iv) *if  $s(x) \in \mathbb{R}$  for all  $s \in \mathcal{S}$ , then  $x$  is selfadjoint;*
- (v) *if  $s(x) \in \mathbb{R}^+$  for all  $s \in \mathcal{S}$ , then  $x \in \mathcal{A}^+$ .*

**Proof.** (i) Let  $\lambda \in \sigma(x)$ . Then for any  $\alpha, \beta \in \mathbb{C}$ ,  $\alpha\lambda + \beta \in \sigma(\alpha x + \beta e)$ , and therefore

$$|\alpha\lambda + \beta| \leq \|\alpha x + \beta e\|. \quad (3)$$

Define  $s_0 : Z := \mathbb{C}x + \mathbb{C}e \rightarrow \mathbb{C}$  by

$$s_0(\alpha x + \beta e) = \alpha\lambda + \beta.$$

If  $\alpha x + \beta e = \alpha'x + \beta'e$ , then by (3)

$$\begin{aligned} |(\alpha\lambda + \beta) - (\alpha'\lambda + \beta')| &= |(\alpha - \alpha')\lambda + (\beta - \beta')| \\ &\leq \|(\alpha - \alpha')x + (\beta - \beta')e\| = 0. \end{aligned}$$

Therefore  $s_0$  is well defined. It is clearly linear and bounded, with norm  $\leq 1$  by (3). Since  $s_0(e) = 1$ , we have  $\|s_0\| = 1$ . By the Hahn–Banach theorem,  $s_0$  has an extension  $s$  as a bounded linear functional on  $\mathcal{A}$  with norm  $\|s\| = \|s_0\| = 1$ . Since also  $s(e) = s_0(e) = 1$ , it follows from Theorem 7.25 that  $s \in \mathcal{S}$ , and  $\lambda = s_0(x) = s(x)$ .

(ii) Since  $|s(x)| \leq \|x\|$ , we have  $\sup_{s \in \mathcal{S}} |s(x)| \leq \|x\|$  for any  $x$ . When  $x$  is normal, we have  $r(x) = \|x\|$ , and therefore there exists  $\lambda_1 \in \sigma(x)$  such that  $|\lambda_1| = \|x\|$ . By (i),  $\lambda_1 = s_1(x)$  for some  $s_1 \in \mathcal{S}$ . This shows that the above supremum is a maximum, attained at  $s_1$ , and is equal to  $\|x\|$ .

For  $x$  arbitrary, we apply the preceding identity to the selfadjoint (hence normal!) element  $x^*x$ :

$$\|x\|^2 = \|x^*x\| = \max_{s \in \mathcal{S}} |s(x^*x)| := \max_{s \in \mathcal{S}} \|x\|_s^2.$$

(iii) Suppose  $s(x) = 0$  for all  $s \in \mathcal{S}$ . Write  $x = a + ib$  with  $a, b \in \mathcal{A}$  selfadjoint. Since  $s$  is real on selfadjoint elements (being a positive linear functional, hence hermitian!), the relation  $0 = s(x) = s(a) + is(b)$  implies that  $s(a) = s(b) = 0$  for all  $s \in \mathcal{S}$ . By (ii), it follows that  $a = b = 0$ , hence  $x = 0$ .

(iv) If  $s(x) \in \mathbb{R}$  for all  $s$ , then (with notation as in (iii))  $s(b) = 0$  for all  $s$ , and therefore  $b = 0$  by (iii). Hence  $x = a$  is selfadjoint.

(v) If  $s(x) \in \mathbb{R}^+$  for all  $s$ , then  $x$  is selfadjoint by (iv), and  $\sigma(x) \subset \mathbb{R}^+$  by (i); hence  $x \in \mathcal{A}^+$ .  $\square$

Let  $X$  be the vector space (under pointwise operations) of all  $f \in \prod_{s \in \mathcal{S}} X_s$  such that  $\mathcal{S}(f) := \{s \in \mathcal{S}; f(s) \neq 0\}$  is at most countable, and

$$\|f\|^2 := \sum_{s \in \mathcal{S}} \|f(s)\|_s^2 < \infty.$$

By the Cauchy–Schwarz inequality in  $\mathbb{C}^n$ , if  $f, g \in X$ , the series

$$(f, g) := \sum_{s \in \mathcal{S}} (f(s), g(s))_s$$

converges absolutely (hence unconditionally), and defines an inner product on  $X$  with induced norm  $\|\cdot\|$ . Let  $\{f_n\}$  be a Cauchy sequence with respect to this norm. For all  $s \in \mathcal{S}$ , the inequality  $\|f\| \geq \|f(s)\|_s$  ( $f \in X$ ) implies that  $\{f_n(s)\}$  is a Cauchy sequence in the Hilbert space  $X_s$ . Let  $f(s) := \lim_n f_n(s) \in X_s$ . Then  $f \in \prod_{s \in \mathcal{S}} X_s$  and  $\mathcal{S}(f) \subset \bigcup_n \mathcal{S}(f_n)$  is at most countable. Given  $\epsilon > 0$ , let  $n_0 \in \mathbb{N}$  be such that  $\|f_n - f_m\| < \epsilon$  for all  $n, m > n_0$ . By Fatou's lemma for the counting measure on  $\mathcal{S}$ , for  $n, m > n_0$ ,

$$\begin{aligned} \|f_n - f\|^2 &= \sum_{s \in \mathcal{S}} \|f_n(s) - f(s)\|_s^2 = \sum_{s \in \mathcal{S}} \liminf_m \|f_n(s) - f_m(s)\|_s^2 \\ &\leq \liminf_m \sum_{s \in \mathcal{S}} \|(f_n - f_m)(s)\|_s^2 = \liminf_m \|f_n - f_m\|^2 \leq \epsilon^2. \end{aligned}$$

This shows that  $f = f_n - (f_n - f) \in X$  and  $f_n \rightarrow f$  in the  $X$ -norm, so that  $X$  is a Hilbert space. It is usually called *the direct sum of the Hilbert spaces  $X_s$*  and is denoted

$$X = \sum_{s \in \mathcal{S}} \oplus X_s.$$

The elements of  $X$  are usually denoted by  $\sum_{s \in \mathcal{S}} \oplus x_s$  (rather than the functional notation  $f$ ). We shall keep up with the preceding notation for simplicity of symbols.

Given  $x \in \mathcal{A}$ , consider the operators  $L_x^s \in B(X_s)$  defined above (the fixed superscript  $s$ , often omitted above, will be allowed now to vary over  $\mathcal{S}$ !). We define a new map  $L_x : X \rightarrow X$  by

$$(L_x f)(s) = L_x^s f(s) \quad (s \in \mathcal{S}).$$

Clearly  $L_x f \in \prod_{s \in \mathcal{S}} X_s$  and  $\mathcal{S}(L_x f) \subset \mathcal{S}(f)$  is at most countable. Also, since  $L^s$  is norm-decreasing,

$$\|L_x f\|^2 = \sum_{s \in \mathcal{S}} \|L_x^s f(s)\|_s^2 \leq \sum_{s \in \mathcal{S}} \|x\|^2 \|f(s)\|_s^2 = \|x\|^2 \|f\|^2 < \infty.$$

Therefore  $L_x f \in X$ , and  $L_x$  is a bounded linear operator on  $X$  with operator norm  $\leq \|x\|$ . The usual notation for the operator  $L_x$  is  $\sum_{s \in \mathcal{S}} \oplus L_x^s$ . Actually, we have  $\|L_x\| = \|x\|$ . Indeed, for each  $s \in \mathcal{S}$ , consider the function  $f_s \in X$  defined by

$$f_s(s) = v_s; \quad f_s(t) = 0 \quad (t \in \mathcal{S}, t \neq s).$$

Then  $\|f_s\| = \|v_s\|_s = 1$  and

$$\|L_x f_s\| = \|L_x^s v_s\|_s = \|x + J_s\|_s = \|x\|_s.$$

Hence  $\|L_x\| \geq \|x\|_s$  for all  $s \in \mathcal{S}$ , and therefore

$$\|L_x\| \geq \sup_{s \in \mathcal{S}} \|x\|_s = \|x\|,$$

by Lemma 7.28 (ii). Together with the preceding inequality, we obtain  $\|L_x\| = \|x\|$  for all  $x \in \mathcal{A}$ .

An easy calculation shows that the map  $L : x \rightarrow L_x$  of  $\mathcal{A}$  into  $B(X)$  is an algebra homomorphism that sends  $e$  onto the identity operator  $I$ . Also  $L_{x^*} = (L_x)^*$  because for all  $f, g \in X$ ,

$$\begin{aligned} (L_{x^*} f, g) &:= \sum_{s \in \mathcal{S}} \left( L_{x^*}^s f(s), g(s) \right)_s = \sum_s \left( (L_x^s)^* f(s), g(s) \right)_s \\ &= \sum_s \left( f(s), L_x^s g(s) \right)_s = (f, L_x g). \end{aligned}$$

Thus,  $L$  is an *isometric  $*$ -isomorphism of the  $B^*$ -algebra  $\mathcal{A}$  onto the  $B^*$ -subalgebra  $L\mathcal{A}$  of  $B(X)$* . The usual notation for  $L$  is  $\sum_{s \in \mathcal{S}} \oplus L^s$ ; it is called *the direct sum of the representations  $L^s$ , ( $s \in \mathcal{S}$ )*. This particular representation of  $\mathcal{A}$  is usually referred to as the *universal representation* of  $\mathcal{A}$ . It is *faithful* (that is, injective), since it is isometric. Our construction proves the following

**Theorem 7.29 (The Gelfand–Naimark theorem).** *Any  $B^*$ -algebra is isometrically  $*$ -isomorphic to a  $B^*$ -subalgebra of  $B(X)$  for some Hilbert space  $X$ .*

A special isometric  $*$ -isomorphism  $L : \mathcal{A} \rightarrow L\mathcal{A} \subset B(X)$  (called the ‘universal representation’) of the  $B^*$ -algebra  $\mathcal{A}$  is the direct sum representation  $L = \sum_{s \in \mathcal{S}} \oplus L^s$  on  $X = \sum_{s \in \mathcal{S}} \oplus X_s$  of the GNS representations  $\{L^s; s \in \mathcal{S} := \mathcal{S}(\mathcal{A})\}$ .

## Exercises

1. A general Banach algebra  $\mathcal{A}$  is not required to possess an identity. Consider then the cartesian product Banach space  $\mathcal{A}_e := \mathcal{A} \times \mathbb{C}$  with the norm  $\|[x, \lambda]\| = \|x\| + |\lambda|$  and the multiplication

$$[x, \lambda][y, \mu] = [xy + \lambda y + \mu x, \lambda\mu].$$

Prove that  $\mathcal{A}_e$  is a unital Banach algebra with the identity  $e := [0, 1]$ , commutative if  $\mathcal{A}$  is commutative, and the map  $x \in \mathcal{A} \rightarrow [x, 0] \in \mathcal{A}_e$  is an isometric isomorphism of  $\mathcal{A}$  onto a maximal two-sided ideal (identified with  $\mathcal{A}$ ) in  $\mathcal{A}_e$ . (With this identification, we have  $\mathcal{A}_e = \mathcal{A} + \mathbb{C}e$ .)

If  $\phi$  is a homomorphism of the commutative Banach algebra  $\mathcal{A}$  into  $\mathbb{C}$ , it extends uniquely to a homomorphism (also denoted by  $\phi$ ) of  $\mathcal{A}_e$  into  $\mathbb{C}$  by the identity  $\phi([x, \lambda]) = \phi(x) + \lambda$ . Conclude that  $\|\phi\| = 1$ .

2. The requirement  $\|xy\| \leq \|x\|\|y\|$  in the definition of a Banach algebra implies the joint continuity of multiplication. Prove:
- (a) If  $\mathcal{A}$  is a Banach space and also an algebra for which multiplication is *separately* continuous, then multiplication is jointly continuous. Hint: consider the bounded operators  $L_x : y \rightarrow xy$  and  $R_y : x \rightarrow xy$  on  $\mathcal{A}$  and use the uniform boundedness theorem.
- (b) (Notation as in Part (a)) The norm  $|x| := \|L_x\|$  (where the norm on the right is the  $B(\mathcal{A})$ -norm) is equivalent to the given norm on  $\mathcal{A}$ , and satisfies the submultiplicativity requirement  $|xy| \leq |x||y|$ . If  $\mathcal{A}$  has an identity  $e$ , then  $|e| = 1$ .
3. Let  $\mathcal{A}$  be a unital complex Banach algebra. If  $F \subset \mathcal{A}$  consists of commuting elements, denote by  $\mathcal{C}_F$  the maximal commutative Banach subalgebra of  $\mathcal{A}$  containing  $F$ . Prove:
- (a)  $\sigma_{\mathcal{C}_F}(a) = \sigma(a)$  for all  $a \in \mathcal{C}_F$ .
- (b) If  $a, b \in \mathcal{A}$  commute, then

$$\sigma(a+b) \subset \sigma(a) + \sigma(b) \text{ and } \sigma(ab) \subset \sigma(a)\sigma(b).$$

Conclude that

$$r(a+b) \leq r(a) + r(b) \text{ and } r(ab) \leq r(a)r(b).$$

- (c) For all  $a \in \mathcal{A}$  and  $\lambda \in \rho(a)$ ,

$$r(R(\lambda; a)) = \frac{1}{d(\lambda, \sigma(a))}.$$

Hint: use the Gelfand representation of  $\mathcal{C}_{\{a,b\}}$ .

4. Let  $\mathcal{A}$  be a commutative (unital) Banach algebra (over  $\mathbb{C}$ ). A set  $E \subset \mathcal{A}$  *generates*  $\mathcal{A}$  if the minimal closed subalgebra of  $\mathcal{A}$  containing  $E$  and the identity  $e$  coincides with  $\mathcal{A}$ . In that case, prove that the maximal ideal space of  $\mathcal{A}$  is homeomorphic to a closed subset of the cartesian product  $\prod_{a \in E} \sigma(a)$ .
5. Let  $\mathcal{A}$  be a unital (complex) Banach algebra, and let  $G$  be the group of regular elements of  $\mathcal{A}$ . Suppose  $\{a_n\} \subset G$  has the following properties:
- (i)  $a_n \rightarrow a$  and  $a_n a = a a_n$  for all  $n$ ;
- (ii) the sequence  $\{r(a_n^{-1})\}$  is bounded.

Prove that  $a \in G$ . Hint: observe that  $r(e - a_n^{-1}a) \leq r(a_n^{-1})r(a_n - a) \rightarrow 0$ , hence  $1 - \sigma(a_n^{-1}a) = \sigma(e - a_n^{-1}a) \subset B(0, 1/2)$  for  $n$  large enough, and therefore  $0 \notin \sigma(a_n^{-1}a)$ .



6. Let  $\mathcal{A}$  be a unital (complex) Banach algebra,  $a \in \mathcal{A}$ , and  $\lambda \in \rho(a)$ . Prove that  $\lambda \in \rho(b)$  for all  $b \in \mathcal{A}$  for which the series  $s(\lambda) := \sum_n [(b-a)R(\lambda; a)]^n$  converges in  $\mathcal{A}$ , and for such  $b$ ,  $R(\lambda; b) = R(\lambda; a)s(\lambda)$ .
7. Let  $\mathcal{A}$  and  $a$  be as in Exercise 6, and let  $V$  be an open set in  $\mathbb{C}$  such that  $\sigma(a) \subset V$ . Prove that there exists  $\delta > 0$  such that  $\sigma(b) \subset V$  for all  $b$  in the ball  $B(a, \delta)$ . Hint: if  $M$  is a bound for  $R(\cdot; a)$  on the complement of  $V$  in the Riemann sphere, take  $\delta = 1/M$  and apply Exercise 6.
8. Let  $\phi$  be a non-zero linear functional on the Banach algebra  $\mathcal{A}$ . Trivially, if  $\phi$  is *multiplicative*, then  $\phi(e) = 1$  and  $\phi \neq 0$  on  $G(\mathcal{A})$ . The following steps provide a proof of the *converse*. Suppose  $\phi(e) = 1$  and  $\phi \neq 0$  on  $G(\mathcal{A})$ . Denote  $N = \ker \phi$  (note that  $N \cap G(\mathcal{A}) = \emptyset$ ). Prove:
- $d(e, N) = 1$ . Hint: if  $\|e - x\| < 1$ , then  $x \in G(\mathcal{A})$ , hence  $x \notin N$ .
  - $\phi \in \mathcal{A}^*$  and has norm 1. Hint: if  $a \notin N$ ,  $a_1 := e - \phi(a)^{-1}a \in N$ , hence  $d(e, a_1) \geq 1$  by Part a.
  - Fix  $a \in N$  with norm 1, and let  $f(\lambda) := \phi(\exp(\lambda a))$  (where the exponential is defined by means of the usual power series, converging absolutely in  $\mathcal{A}$  for all  $\lambda \in \mathbb{C}$ ). Then  $f$  is an entire function with no zeros such that  $f(0) = 1$ ,  $f'(0) = 0$ , and  $|f(\lambda)| \leq e^{|\lambda|}$ .
  - (*This is a result about entire functions.*) If  $f$  has the properties listed in Part (c), then  $f = 1$  identically. *Sketch of proof:* since  $f$  has no zeros, it can be represented as  $f = e^g$  with  $g$  entire; necessarily  $g(0) = g'(0) = 0$ , so that  $g(\lambda) = \lambda^2 h(\lambda)$  with  $h$  entire, and  $\Re g(\lambda) \leq |\lambda|$ . For any  $r > 0$ , verify that  $|2r - g| \geq |g|$  in the disc  $|\lambda| \leq r$  and  $|2r - g| > 0$  in the disc  $|\lambda| < 2r$ . Therefore  $F(\lambda) := [r^2 h(\lambda)]/[2r - g(\lambda)]$  is analytic in  $|\lambda| < 2r$ , and  $|F| \leq 1$  on the circle  $|\lambda| = r$ , hence in the disc  $|\lambda| \leq r$  by the maximum modulus principle. Thus

$$\frac{|h|}{|2 - g/r|} \leq 1/r \quad (|\lambda| < r).$$

Given  $\lambda$ , let  $r \rightarrow \infty$  to conclude that  $h = 0$ .

- If  $a \in N$ , then  $a^2 \in N$ . Hint: apply Parts (c) and (d) and look at the coefficient of  $\lambda^2$  in the series for  $f$ .
- $\phi(x^2) = \phi(x)^2$  for all  $x \in \mathcal{A}$ . (Represent  $x = x_1 + \phi(x)e$  with  $x_1 \in N$  and apply Part (e).) In particular,  $x \in N$  iff  $x^2 \in N$ .
- If either  $x$  or  $y$  belong to  $N$ , then (i)  $xy + yx \in N$ ; (ii)  $(xy)^2 + (yx)^2 \in N$ ; and (iii)  $xy - yx \in N$ . (For (i), apply Part (f) to  $x + y$ ; for (ii), apply (i) to  $yxy$  instead of  $y$ , when  $x \in N$ ; for (iii), write  $(xy - yx)^2 = 2[(xy)^2 + (yx)^2] - (xy + yx)^2$  and use Part (f).) Conclude that  $N$  is a two-sided ideal in  $\mathcal{A}$  and  $\phi$  is multiplicative (use the representation  $x = x_1 + \phi(x)e$  with  $x_1 \in N$ ).

9. Let  $\mathcal{A}$  be a (unital complex) Banach algebra. For  $a, b \in \mathcal{A}$ , denote  $C(a, b) = L_a - R_b$  (cf. Section 7.1), and consider the series

$$b_L(\lambda) = \sum_{j=0}^{\infty} (-1)^j R(\lambda; a)^{j+1} [C(a, b)^j e];$$

$$b_R(\lambda) = \sum_{j=0}^{\infty} [C(b, a)^j e] R(\lambda; a)^{j+1}$$

for  $\lambda \in \rho(a)$ . Prove that if  $b_L(\lambda)$  ( $b_R(\lambda)$ ) converges in  $\mathcal{A}$  for some  $\lambda \in \rho(a)$ , then its sum is a left inverse (right inverse, respectively) for  $\lambda e - b$ . In particular, if  $\lambda \in \rho(a)$  is such that both series converge in  $\mathcal{A}$ , then  $\lambda \in \rho(b)$  and  $R(\lambda; b) = b_L(\lambda) = b_R(\lambda)$ .

10. (Notation as in Exercise 9.) Set

$$r(a, b) = \limsup_n \|C(a, b)^n e\|^{1/n},$$

and consider the compact subsets of  $\mathbb{C}$

$$\sigma_L(a, b) = \{\lambda \in \mathbb{C}; d(\lambda, \sigma(a)) \leq r(a, b)\};$$

$$\sigma_R(a, b) = \{\lambda \in \mathbb{C}; d(\lambda, \sigma(a)) \leq r(b, a)\};$$

$$\sigma(a, b) = \sigma_L(a, b) \cup \sigma_R(a, b).$$

Prove that the series  $b_L(\lambda)$  ( $b_R(\lambda)$ ) converge absolutely and uniformly on compact subsets of  $\sigma_L(a, b)^c$  ( $\sigma_R(a, b)^c$ , respectively). In particular,  $\sigma(b) \subset \sigma(a, b)$ , and  $R(\cdot; b) = b_L = b_R$  on  $\sigma(a, b)^c$ .

11. (Notation as in Exercise 10.) Set

$$d(a, b) = \max\{r(a, b), r(b, a)\},$$

so that trivially

$$\sigma(a, b) = \{\lambda; d(\lambda, \sigma(a)) \leq d(a, b)\}$$

and  $\sigma(a, b) = \sigma(a)$  iff  $d(a, b) = 0$ . In this case, it follows from Exercise 10 (and symmetry) that  $\sigma(b) = \sigma(a)$  (for this reason, elements  $a, b$  such that  $d(a, b) = 0$  are said to be *spectrally equivalent*).

12. Let  $D$  be a *derivation* on  $\mathcal{A}$ , that is, a linear map  $D : \mathcal{A} \rightarrow \mathcal{A}$  such that  $D(ab) = (Da)b + a(Db)$  for all  $a, b \in \mathcal{A}$ . (Example: given  $s \in \mathcal{A}$ , the map  $D_s := L_s - R_s$  is a derivation; it is called an *inner derivation*.) Prove:

- (a) If  $D$  is a derivation on  $\mathcal{A}$  and  $Dv$  commutes with  $v$  for some  $v$ , then  $Df(v) = f'(v)Dv$  for all polynomials  $f$ .

- (b) Let  $s \in \mathcal{A}$ . The element  $v \in \mathcal{A}$  is  $s$ -Volterra if  $D_s v = v^2$ . (Example: in  $\mathcal{A} = B(L^p([0, 1]))$ , take  $S : f(t) \rightarrow tf(t)$  and  $V : f(t) \rightarrow \int_0^t f(u)du$ , the so-called *classical Volterra operator*.) Prove: (i)  $D_s v^n = nv^{n+1}$ ; (ii)  $v^{n+1} = D_s^n v/n!$ ; and (iii)  $C(s + \alpha v, s + \beta v)^n e = (-1)^n n! \binom{\beta - \alpha}{n} v^n$ , for all  $n \in \mathbb{N}$  and  $\alpha, \beta \in \mathbb{C}$ .
- (c) If  $v \in \mathcal{A}$  is  $s$ -Volterra, then (i)  $\|v^n\|^{1/n} = O(1/n)$ . In particular,  $v$  is quasi-nilpotent. (ii)  $r(s + \alpha v, s + \beta v) = 0$  if  $\beta - \alpha \in \mathbb{N} \cup \{0\}$ , and  $= \limsup(n! \|v^n\|)^{1/n}$  otherwise. (iii)  $d(s + \alpha v, s + \beta v) = \limsup(n! \|v^n\|)^{1/n}$  if  $\alpha \neq \beta$ . (iv) For  $\alpha \neq \beta$ ,  $s + \alpha v$  and  $s + \beta v$  are spectrally equivalent iff  $\|v^n\|^{1/n} = o(1/n)$ . (v)  $d(s + \alpha v, s + \beta v) \leq \text{diam } \sigma(s)$ . (vi)  $d(S + \alpha V, S + \beta V) = 1$  when  $\alpha \neq \beta$  (cf. Part (b) for notation). In particular,  $S + \alpha V$  and  $S + \beta V$  are spectrally equivalent iff  $\alpha = \beta$  (however, they all have the same spectrum, but do not try to prove this here!) Note that if  $\beta - \alpha \in \mathbb{N}$ , then  $r(S + \alpha V, S + \beta V) = 0$  while  $r(S + \beta V, S + \alpha V) = 1$ .
- (d) If  $v$  is  $s$ -Volterra, then

$$R(\lambda; v) = \lambda^{-1} e + \lambda^{-2} \exp(s/\lambda) v \exp(-s/\lambda) \quad (\lambda \neq 0).$$

- (e) If  $v$  is  $s$ -Volterra, then for all  $\alpha, \lambda \in \mathbb{C}$

$$\exp[\lambda(s + \alpha v)] = \exp(\lambda s)(e + \lambda v)^\alpha = (e - \lambda v)^{-\alpha} \exp(\lambda v),$$

where the binomials are defined by means of the usual series (note that  $v$  is quasi-nilpotent, so that the binomial series converge for all complex  $\lambda$ ).

- (f) If  $v$  is  $s$ -Volterra and  $\rho(s)$  is connected, then  $\sigma(s + kv) \subset \sigma(s)$  for all  $k \in \mathbb{Z}$ . For all  $\lambda \in \rho(s)$ ,

$$\begin{aligned} R(\lambda; s + kv) &= \sum_{j=0}^k \binom{k}{j} j! R(\lambda; s)^{j+1} v^j \quad (k \geq 0); \\ &= \sum_{j=0}^{|k|} (-1)^j \binom{|k|}{j} j! v^j R(\lambda; s)^{j+1} \quad (k < 0). \end{aligned}$$

(Apply Exercise 9.) If  $\rho(s)$  and  $\rho(s + kv)$  are both connected for some integer  $k$ , then  $\sigma(s + kv) = \sigma(s)$ . In particular, if  $\sigma(s) \subset \mathbb{R}$ , then  $\sigma(s + kv) = \sigma(s)$  for all  $k \in \mathbb{Z}$ .

13. Let  $\mathcal{A}$  be a (unital, complex) Banach algebra, and let  $a, b, c \in \mathcal{A}$  be such that  $C(a, b)c = 0$  (i.e.  $ac = cb$ ). Prove:

- (a)  $C(e^a, e^b)c = 0$  (i.e.  $e^a c = c e^b$ , where the exponential function  $e^a$  is defined by the usual absolutely convergent series; the base of the exponential should not be confused with the identity of  $\mathcal{A}$ !)

- (b) If  $\mathcal{A}$  is a  $B^*$ -algebra, then  $e^{x-x^*}$  is unitary, for any  $x \in \mathcal{A}$ . (In particular,  $\|e^{x-x^*}\| = 1$ .)
- (c) If  $\mathcal{A}$  is a  $B^*$ -algebra and  $a, b$  are normal elements (such that  $ac = cb$ , as before!), then

$$e^{a^*} c e^{-b^*} = e^{a^*-a} c e^{b-b^*};$$

$$\|e^{a^*} c e^{-b^*}\| \leq \|c\|.$$

- (d) For  $a, b, c$  as in Part (c), define

$$f(\lambda) = e^{\lambda a^*} c e^{-\lambda b^*} \quad (\lambda \in \mathbb{C}).$$

Prove that  $\|f(\lambda)\| \leq \|c\|$  for all  $\lambda \in \mathbb{C}$ , and conclude that  $f(\lambda) = c$  for all  $\lambda$  (i.e.  $e^{\lambda a^*} c = c e^{\lambda b^*}$  for all  $\lambda \in \mathbb{C}$ ).

- (e) If  $\mathcal{A}$  is a  $B^*$ -algebra, and  $a, b$  are normal elements of  $\mathcal{A}$  such that  $ac = cb$  for some  $c \in \mathcal{A}$ , then  $a^*c = cb^*$ . (Consider the coefficient of  $\lambda$  in the last identity in Part (d).)

In particular, if  $c$  commutes with a normal element  $a$ , it commutes also with its adjoint; this is *Fuglede's theorem*.

14. Consider  $L^1(\mathbb{R})$  (with respect to Lebesgue measure) *with convolution as multiplication*. Prove that  $L^1(\mathbb{R})$  is a commutative Banach algebra with no identity, and the Fourier transform  $F$  is a contractive (i.e. norm-decreasing) homomorphism of  $L^1(\mathbb{R})$  into  $C_0(\mathbb{R})$ . (Cf. Exercise 7, Chapter 2.)
15. Let  $\phi$  be a non-zero homomorphism of the Banach algebra  $L^1 = L^1(\mathbb{R})$  into  $\mathbb{C}$  (cf. Exercises 14 and 1). Prove:

- (a) There exists a unique  $h \in L^\infty = L^\infty(\mathbb{R})$  such that  $\phi(f) = \int f h dx$  for all  $f \in L^1$ , and  $\|h\|_\infty = 1$ . Moreover

$$\phi(f_y)\phi(g) = \phi(f)\phi(g_y) \quad (f, g \in L^1; y \in \mathbb{R}),$$

where  $f_y(x) = f(x - y)$ .

- (b) For any  $f \in L^1$  such that  $\phi(f) \neq 0$ ,
- (i)  $h(y) = \phi(f_y)/\phi(f)$  a.e. (in particular,  $h$  may be chosen to be continuous).
  - (ii)  $\phi(f_y) \neq 0$  for all  $y \in \mathbb{R}$ .
  - (iii)  $|h(y)| = 1$  for all  $y \in \mathbb{R}$ .
  - (iv)  $h(x + y) = h(x)h(y)$  for all  $x, y \in \mathbb{R}$  and  $h(0) = 1$ .

Conclude that  $h(y) = e^{-ity}$  for some  $t \in \mathbb{R}$  (for all  $y$ ) and that  $\phi(f) = (Ff)(t)$ , where  $F$  is the Fourier transform.

Conversely, each  $t \in \mathbb{R}$  determines the homomorphism  $\phi_t(f) = (Ff)(t)$ . Conclude that the map  $t \rightarrow \phi_t$  is a homeomorphism of  $\mathbb{R}$  onto the Gelfand space  $\Phi$  of  $L^1$  (that is, the space of non-zero complex homomorphisms of  $L^1$  with the Gelfand topology). (Hint: the Gelfand topology is Hausdorff and is weaker than the metric topology on  $\mathbb{R}$ .)

16. Let  $\mathcal{A}, \mathcal{B}$  be commutative Banach algebras,  $\mathcal{B}$  semi-simple. Let  $\tau : \mathcal{A} \rightarrow \mathcal{B}$  be an algebra homomorphism. Prove that  $\tau$  is continuous. Hint: for each  $\phi \in \Phi(\mathcal{B})$  (the Gelfand space of  $\mathcal{B}$ ),  $\phi \circ \tau \in \Phi(\mathcal{A})$ . Use the closed graph theorem.
17. Let  $X$  be a compact Hausdorff space, and let  $\mathcal{A} = C(X)$ . Prove that the Gelfand space  $\Phi$  of  $\mathcal{A}$  (terminology as in Exercise 15) is homeomorphic to  $X$ . Hint: consider the map  $t \in X \rightarrow \phi_t \in \Phi$ , where  $\phi_t(f) = f(t)$ , ( $f \in C(X)$ ) (this is the so-called ‘evaluation at  $t$ ’ homomorphism). If  $\exists \phi \in \Phi$  such that  $\phi \neq \phi_t$  for all  $t \in X$  and  $M = \ker \phi$ , then for each  $t \in X$  there exists  $f_t \in M$  such that  $f_t(t) \neq 0$ . Use continuity of the functions and compactness of  $X$  to get a finite set  $\{f_{t_j}\} \subset M$  such that  $h := \sum |f_{t_j}|^2 > 0$  on  $X$ , hence  $h \in G(\mathcal{A})$ ; however,  $h \in M$ , contradiction. Thus  $t \rightarrow \phi_t$  is onto  $\Phi$ , and one-to-one (by Urysohn’s lemma). Identifying  $\Phi$  with  $X$  through this map, observe that the Gelfand topology is weaker than the given topology on  $X$  and is Hausdorff.
18. Let  $U$  be the open unit disc in  $\mathbb{C}$ . For  $n \in \mathbb{N}$ , let  $\mathcal{A} = \mathcal{A}(U^n)$  denote the Banach algebra of all complex functions analytic in  $U^n := U \times \cdots \times U$  and continuous on the closure  $\overline{U^n}$  of  $U^n$  in  $\mathbb{C}^n$ , with pointwise operations and supremum norm  $\|f\|_u := \sup\{|f(z)|; z \in \overline{U^n}\}$ . Let  $\Phi$  be the Gelfand space of  $\mathcal{A}$  (terminology as in Exercise 15). Given  $f \in \mathcal{A}$  and  $0 < r < 1$ , denote  $f_r(z) = f(rz)$  and  $Z(f) = \{z \in \overline{U^n}; f(z) = 0\}$ . Prove:
  - (a)  $f_r$  is the sum of an absolutely and uniformly convergent power series in  $\overline{U^n}$ . Conclude that the polynomials (in  $n$  variables) are dense in  $\mathcal{A}$ .
  - (b) Each  $\phi \in \Phi$  is an ‘evaluation homomorphism’  $\phi_w$  for some  $w \in \overline{U^n}$ , where  $\phi_w(f) = f(w)$ . Hint: consider the polynomials  $p_j(z) = z_j$  (where  $z = (z_1, \dots, z_n)$ ). Then  $w := (\phi(p_1), \dots, \phi(p_n)) \in \overline{U^n}$  and  $\phi(p_j) = p_j(w)$ . Hence  $\phi(p) = p(w)$  for all polynomials  $p$ . Apply Part (a) to conclude that  $\phi = \phi_w$ . The map  $w \rightarrow \phi_w$  is the wanted homeomorphism of  $\overline{U^n}$  onto  $\Phi$ .
  - (c) Given  $f_1, \dots, f_m \in \mathcal{A}$  such that  $\bigcap_{k=1}^m Z(f_k) = \emptyset$ , there exist  $g_1, \dots, g_m \in \mathcal{A}$  such that  $\sum_k f_k g_k = 1$  (on  $\overline{U^n}$ ). Hint: otherwise, the ideal  $J$  generated by  $f_1, \dots, f_m$  is proper, and therefore there exists  $\phi \in \Phi$  vanishing on  $J$ . Apply Part (b) to reach a contradiction.
19. Let  $\mathcal{A}$  be a (unital, complex) Banach algebra such that

$$K := \sup_{0 \neq a \in \mathcal{A}} \frac{\|a\|^2}{\|a^2\|} < \infty.$$

Prove:

- (a)  $\|a\| \leq K r(a)$  for all  $a \in \mathcal{A}$ .
  - (b)  $\|p(a)\| \leq K \|p\|_{C(\sigma(a))}$  for all  $p \in \mathcal{P}$ , where  $\mathcal{P}$  denotes the algebra of all polynomials of one complex variable over  $\mathbb{C}$ .
  - (c) If  $a \in \mathcal{A}$  has the property that  $\mathcal{P}$  is dense in  $C(\sigma(a))$ , then there exists a continuous algebra homomorphism (with norm  $\leq K$ )  $\tau : C(\sigma(a)) \rightarrow \mathcal{A}$  such that  $\tau(p) = p(a)$  for all  $p \in \mathcal{P}$ .
20. Let  $K \subset \mathbb{C}$  be compact  $\neq \emptyset$ , and let  $C(K)$  be the corresponding Banach algebra of continuous functions with the supremum norm  $\|f\|_K := \sup_K |f|$ . Denote  $\mathcal{P}_1 := \{p \in \mathcal{P}; \|p\|_K \leq 1\}$  (cf. Exercise 19 b).

Let  $X$  be a Banach space, and  $T \in B(X)$ . For  $x \in X$ , denote

$$\|x\|_T := \sup_{p \in \mathcal{P}_1} \|p(T)x\|;$$

$$Z_T := \{x \in X; \|x\|_T < \infty\}.$$

Prove:

- (a)  $Z_T$  is a Banach space for the norm  $\|\cdot\|_T$  (which is greater than the given norm on  $X$ ).
- (b)  $\|p(T)\|_{B(Z_T)} \leq 1$ .
- (c) If the compact set  $K$  is such that  $\mathcal{P}$  is dense in  $C(K)$ , there exists a contractive algebra homomorphism  $\tau : C(K) \rightarrow B(Z_T)$  such that  $\tau(p) = p(T)$  for all  $p \in \mathcal{P}$ .

# 8

## Hilbert spaces

### 8.1 Orthonormal sets

We recall that the vectors  $x, y$  in an inner product space  $X$  are (mutually) *orthogonal* if  $(x, y) = 0$  (cf. Theorem 1.35). The set  $A \subset X$  is orthogonal if any two distinct vectors of  $A$  are orthogonal; it is *orthonormal* if it is orthogonal and all vectors in  $A$  are unit vectors, that is, if

$$(a, b) = \delta_{a,b} \quad (a, b \in A)$$

where  $\delta_{a,b}$  is *Kronecker's delta*, which equals zero for  $a \neq b$  and equals one for  $a = b$ .

A classical example is the set  $A = \{e^{int}; n \in \mathbb{N}\}$  in the Hilbert space  $X = L^2([0, 2\pi])$  with the normalized Lebesgue measure  $dt/2\pi$ .

**Lemma 8.1 (Pythagores' theorem).** *Let  $\{x_k; k = 1, \dots, n\}$ , be an orthogonal subset of the inner product space  $X$ . Then*

$$\left\| \sum_{k=1}^n x_k \right\|^2 = \sum_{k=1}^n \|x_k\|^2.$$

**Proof.** By 'sesqui-linearity' of the inner product, the left-hand side equals

$$\left( \sum_k x_k, \sum_j x_j \right) = \sum_{k,j} (x_k, x_j).$$

Since  $(x_k, x_j) = 0$  for  $k \neq j$ , the last sum equals  $\sum_k (x_k, x_k) = \sum_k \|x_k\|^2$ .  $\square$

If  $A := \{a_1, \dots, a_n\} \subset X$  is *orthonormal* and  $\{\lambda_1, \dots, \lambda_n\} \subset \mathbb{C}$ , then (taking  $x_k = \lambda_k a_k$  in Lemma 8.1),

$$\left\| \sum_{k=1}^n \lambda_k a_k \right\|^2 = \sum_{k=1}^n |\lambda_k|^2. \quad (1)$$

**Theorem 8.2.** Let  $\{a_k; k = 1, 2, \dots\}$  be an orthonormal sequence in the Hilbert space  $X$ , and let  $\Lambda := \{\lambda_k; k = 1, 2, \dots\}$  be a complex sequence. Then

- (a) The series  $\sum_k \lambda_k a_k$  converges in  $X$  iff  $\|\Lambda\|_2^2 := \sum_k |\lambda_k|^2 < \infty$ .  
 (b) In this case, the above series converges unconditionally in  $X$ , and  $\|\sum_k \lambda_k a_k\| = \|\Lambda\|_2$ .

**Proof.** By (1) applied to the orthonormal set  $\{a_{m+1}, \dots, a_n\}$  and the set of scalars  $\{\lambda_{m+1}, \dots, \lambda_n\}$  with  $n > m \geq 0$ ,

$$\left\| \sum_{k=m+1}^n \lambda_k a_k \right\|^2 = \sum_{k=m+1}^n |\lambda_k|^2. \quad (2)$$

This means that the series  $\sum \lambda_k a_k$  satisfies Cauchy's condition iff the series  $\sum |\lambda_k|^2$  satisfies Cauchy's condition. Since  $X$  is complete, this is equivalent to Statement (a) of the theorem.

Suppose now that  $\|\Lambda\|_2 < \infty$ , and let then  $s \in X$  denote the sum of the series  $\sum_k \lambda_k a_k$ . Taking  $m = 0$  and letting  $n \rightarrow \infty$  in (2), we obtain  $\|s\| = \|\Lambda\|_2$ .

Since  $(\cdot, x)$  is a continuous linear functional (for any fixed  $x \in X$ ), we have

$$(s, x) = \sum_{k=1}^{\infty} \lambda_k (a_k, x). \quad (3)$$

If  $\pi : \mathbb{N} \rightarrow \mathbb{N}$  is any permutation of  $\mathbb{N}$ , the series  $\sum_k |\lambda_{\pi(k)}|^2$  converges to  $\|\Lambda\|_2^2$  by a well-known property of positive series. Therefore, by what we already proved, the series  $\sum_k \lambda_{\pi(k)} a_{\pi(k)}$  converges in  $X$ ; denoting its sum by  $t$ , we also have  $\|t\| = \|\Lambda\|_2$ , and by (3), for any  $x \in X$ ,

$$(t, x) = \sum_k \lambda_{\pi(k)} (a_{\pi(k)}, x).$$

Choose  $x = a_j$  for  $j \in \mathbb{N}$  fixed. By orthonormality, we get  $(t, a_j) = \lambda_j$ , and therefore, by (3) with  $x = t$ ,

$$(t, s) = \overline{(s, t)} = \overline{\sum_k \lambda_k (a_k, t)} = \sum_k \overline{\lambda_k} (t, a_k) = \sum_k |\lambda_k|^2 = \|\Lambda\|_2^2.$$

Hence  $\|t - s\|^2 = \|t\|^2 - 2\Re(t, s) + \|s\|^2 = 0$ , and  $t = s$ .  $\square$

**Lemma 8.3 (Bessel's inequality).** Let  $\{a_1, \dots, a_n\}$  be an orthonormal set in the inner product space  $X$ . Then for all  $x \in X$ ,

$$\sum_k |(x, a_k)|^2 \leq \|x\|^2.$$

**Proof.** Given  $x \in X$ , denote  $y := \sum_k (x, a_k) a_k$ . Then by (1)

$$(y, x - y) = (y, x) - (y, y) = \sum_k (x, a_k)(a_k, x) - \|y\|^2 = \sum_k |(x, a_k)|^2 - \|y\|^2 = 0.$$



Therefore, by Lemma 8.1,

$$\|x\|^2 = \|(x - y) + y\|^2 = \|x - y\|^2 + \|y\|^2 \geq \|y\|^2 = \sum_k |(x, a_k)|^2.$$

□

**Corollary 8.4.** *Let  $\{a_k; k = 1, 2, \dots\}$  be an orthonormal sequence in the Hilbert space  $X$ . Then for any  $x \in X$ , the series  $\sum_k (x, a_k)a_k$  converges unconditionally in  $X$  to an element  $Px$ , and  $P \in B(X)$  has the following properties:*

- (a)  $\|P\| = 1$ ;
- (b)  $P^2 = P$ ;
- (c) the ranges  $PX$  and  $(I - P)X$  are orthogonal
- (d)  $P^* = P$ .

**Proof.** Let  $x \in X$ . By Bessel's inequality (Lemma 8.3), the partial sums of the positive series  $\sum_k |(x, a_k)|^2$  are bounded by  $\|x\|^2$ ; the series therefore converges, and consequently  $\sum_k (x, a_k)a_k$  converges unconditionally to an element  $Px \in X$ , by Theorem 8.2. The linearity of  $P$  is trivial, and by Theorem 8.2,  $\|Px\|^2 = \sum_k |(x, a_k)|^2 \leq \|x\|^2$ , so that  $P \in B(X)$  and  $\|P\| \leq 1$ . By (3)

$$(Px, a_j) = \sum_{k=1}^{\infty} (x, a_k)(a_k, a_j) = (x, a_j). \quad (4)$$

Therefore

$$P^2x = P(Px) = \sum_j (Px, a_j)a_j = \sum_j (x, a_j)a_j = Px.$$

This proves Property (b)

By Property (b),  $\|P\| = \|P^2\| \leq \|P\|^2$ . Since  $P \neq 0$  (e.g.  $Pa_j = a_j \neq 0$  for all  $j \in \mathbb{N}$ ), it follows that  $\|P\| \geq 1$ , and therefore  $\|P\| = 1$ , by our previous inequality. This proves Property (a). By (3) with the proper choices of scalars and vectors, we have for all  $x, y \in X$

$$(Py, x) = \sum_k (y, a_k)(a_k, x) = \sum_k \overline{(x, a_k)}(a_k, y) = \overline{(Px, y)} = (y, Px).$$

This proves Property (d), and therefore by Property (b),

$$(Px, (I - P)y) = (x, P(I - P)y) = (x, (P - P^2)y) = 0,$$

which verifies Property (c). □

## 8.2 Projections

### Terminology 8.5.

- (a) Any  $P \in B(X)$  satisfying Property (b) in Corollary 8.4 is called a *projection* ( $X$  could be any Banach space!). If  $P$  is a projection, so is  $I - P$  (because  $(I - P)^2 = I - 2P + P = I - P$ );  $I - P$  is called the *complementary* projection of  $P$ . Note that the complementary projections  $P$  and  $I - P$  commute and have product equal to zero.
- (b) For any projection  $P$  on a Banach space  $X$ ,

$$PX = \{x \in X; Px = x\} = \ker(I - P) \quad (1)$$

(if  $x = Py$  for some  $y \in X$ , then  $Px = P^2y = Py = x$ ). Since  $I - P$  is continuous, it follows from (1) that the range  $PX$  is closed. The closed subspaces  $PX$  and  $(I - P)X$  have trivial intersection (if  $x$  is in the intersection, then  $x = Px$ , and  $x = (I - P)x = x - Px = 0$ ), and their sum is  $X$  (every  $x$  can be written as  $Px + (I - P)x$ ). This means that  $X$  is the *direct sum of the closed subspaces*  $PX$  and  $(I - P)X$ .

A closed subspace  $M \subset X$  is *T-invariant* (for a given  $T \in B(X)$ ) if  $TM \subset M$ . By (1), the closed subspace  $PX$  is  $T$ -invariant iff  $P(TPx) = TPx$  for all  $x \in X$ , that is, iff  $TP = PTP$ . Applying this to the complementary projection  $I - P$ , we conclude that  $(I - P)X$  is  $T$ -invariant iff  $T(I - P) = (I - P)T(I - P)$ , that is (expand and cancel!), iff  $PT = PTP$ . Therefore *both* complementary subspaces  $PX$  and  $(I - P)X$  are  $T$ -invariant iff  $P$  *commutes* with  $T$ . One says in this case that  $PX$  is a *reducing subspace* for  $T$  (or that  $P$  *reduces*  $T$ ).

- (c) When  $X$  is a Hilbert space and a projection  $P$  in  $X$  satisfies Property (c) in Corollary 8.4, it is called an *orthogonal projection*. In that case the direct sum decomposition  $X = PX \oplus (I - P)X$  is an *orthogonal decomposition*. Conversely, if  $Y$  is any closed subspace of  $X$ , we may use the orthogonal decomposition  $X = Y \oplus Y^\perp$  (Theorem 1.36) to define an orthogonal projection  $P$  in  $X$  with range equal to  $Y$ : given any  $x \in X$ , it has the unique orthogonal decomposition  $x = y + z$  with  $y \in Y$  and  $z \in Y^\perp$ ; define  $Px = y$ . It is easy to verify that  $P$  is the wanted (selfadjoint!) projection; it is called *the orthogonal projection onto*  $Y$ . Given  $T \in B(X)$ ,  $Y$  is a reducing subspace for  $T$  iff  $P$  commutes with  $T$  (by Point (b) above); since  $P$  is selfadjoint, the relations  $PT = TP$  and  $T^*P = PT^*$  are equivalent (since one follows from the other by taking adjoints). Thus  $Y$  reduces  $T$  iff it reduces  $T^*$ . If  $Y$  is invariant for both  $T$  and  $T^*$ , then (cf. Point (b))  $TP = PTP$  and  $T^*P = PT^*P$ ; taking adjoints in the second relation, we get  $PT = PTP$ , hence  $TP = PT$ . As observed before, this last relation implies in particular that  $Y$  is invariant for both  $T$  and  $T^*$ . Thus, if  $Y$  is a closed subspace of the Hilbert space  $X$  and  $P$  is the corresponding orthogonal projection, then for any  $T \in B(X)$ , the following propositions

are equivalent:

- (i)  $Y$  reduces  $T$ ;
  - (ii)  $Y$  is invariant for both  $T$  and  $T^*$ ;
  - (iii)  $P$  commutes with  $T$ .
- (d) In the proof of Corollary 8.4, we deduced Property (c) from Property (d). Conversely, if  $P$  is an orthogonal projection, then it is selfadjoint. Indeed, for any  $x, y \in X$ ,  $(Px, (I - P)y) = 0$ , that is,  $(Px, y) = (Px, Py)$ . Interchanging the roles of  $x$  and  $y$  and taking complex adjoints, we get  $(x, Py) = (Px, Py)$ , and therefore  $(Px, y) = (x, Py)$ . Thus, *a projection in Hilbert space is orthogonal if and only if it is selfadjoint.*

We consider now an arbitrary orthonormal set  $A$  in the Hilbert space  $X$ .

**Lemma 8.6.** *Let  $X$  be an inner product space and  $A \subset X$  be orthonormal. Let  $\delta > 0$  and  $x \in X$  be given. Then the set*

$$A_\delta(x) := \{a \in A; |(x, a)| > \delta\}$$

*is finite (it has at most  $\lceil \|x\|^2/\delta^2 \rceil$  elements).*

**Proof.** We may assume that  $A_\delta(x) \neq \emptyset$  (otherwise there is nothing to prove). Let then  $a_1, \dots, a_n$  be  $n \geq 1$  distinct elements in  $A_\delta(x)$ . By Bessel's inequality,

$$n\delta^2 \leq \sum_{k=1}^n |(x, a_k)|^2 \leq \|x\|^2,$$

so that  $n \leq \|x\|^2/\delta^2$ , and the conclusion follows.  $\square$

**Theorem 8.7.** *Let  $X$  be an inner product space, and  $A \subset X$  be orthonormal. Then for any given  $x \in X$ , the set*

$$A(x) := \{a \in A; (x, a) \neq 0\}$$

*is at most countable.*

**Proof.** Since

$$A(x) = \bigcup_{m=1}^{\infty} A_{1/m}(x),$$

this is an immediate consequence of Lemma 8.6.  $\square$

**Notation 8.8.** Let  $A$  be any orthonormal subset in the Hilbert space  $X$ . Given  $x \in X$ , write  $A(x)$  as a (finite or infinite) sequence  $A(x) = \{a_k\}$ . Let  $Px$  be defined as before with respect to the orthonormal sequence  $\{a_k\}$ . Since the convergence is unconditional, the definition is *independent* of the particular representation of  $A(x)$  as a sequence, and one may use the following notation that ignores the sequential representation:

$$Px = \sum_{a \in A(x)} (x, a)a.$$

Since  $(x, a) = 0$  for all  $a \in A$  not in  $A(x)$ , one may add to the sum above the zero terms  $(x, a)a$  for all such vectors  $a$ , that is,

$$Px = \sum_{a \in A} (x, a)a.$$

By Corollary 8.4,  $P$  is an orthogonal projection.

**Lemma 8.9.** *The ranges of the orthogonal projections  $P$  and  $I - P$  are  $\overline{\text{span}(A)}$  (the closed span of  $A$ , i.e. the closure of the linear span of  $A$ ) and  $A^\perp$ , respectively.*

**Proof.** Since  $Pb = b$  for any  $b \in A$ , we have  $A \subset PX$ , and since  $PX$  is a closed subspace, it follows that  $\overline{\text{span}(A)} \subset PX$ . On the other hand, given  $x \in X$ , represent  $A(x) = \{a_k\}$ ; then  $Px = \sum_k (x, a_k)a_k = \lim_n \sum_{k=1}^n (x, a_k)a_k \in \overline{\text{span}(A)}$ , and the first statement of the lemma follows.

By uniqueness of the orthogonal decomposition, we have

$$(I - P)X = \left(\overline{\text{span}(A)}\right)^\perp. \quad (2)$$

Clearly,  $x \in A^\perp$  iff  $A \subset \ker(\cdot, x)$ . Since  $(\cdot, x)$  is a continuous linear functional, its kernel is a closed subspace, and therefore the last inclusion is equivalent to  $\overline{\text{span}(A)} \subset \ker(\cdot, x)$ , that is, to  $x \in \left(\overline{\text{span}(A)}\right)^\perp$ . This shows that the set on the right of (2) is equal to  $A^\perp$ .  $\square$

### 8.3 Orthonormal bases

**Theorem 8.10.** *Let  $A$  be an orthonormal set in the Hilbert space  $X$ , and let  $P$  be the associated projection. Then the following statements are equivalent:*

- (1)  $A^\perp = \{0\}$ .
- (2) If  $A \subset B \subset X$  and  $B$  is orthonormal, then  $A = B$ .
- (3)  $X = \overline{\text{span}(A)}$ .
- (4)  $P = I$ .
- (5) Every  $x \in X$  has the representation

$$x = \sum_{a \in A} (x, a)a.$$

- (6) For every  $x, y \in X$ , one has

$$(x, y) = \sum_{a \in A} (x, a)\overline{(y, a)},$$

where the series, which has at most countably many non-zero terms, converges absolutely.

(7) For every  $x \in X$ , one has

$$\|x\|^2 = \sum_{a \in A} |(x, a)|^2.$$

**Proof.**

- (1) *implies* (2). Suppose  $A \subset B$  with  $B$  orthonormal. If  $B \neq A$ , pick  $b \in B$ ,  $b \notin A$ . For any  $a \in A$ , the vectors  $a, b$  are distinct (unit) vectors in the orthonormal set  $B$ , and therefore  $(b, a) = 0$ . Hence  $b \in A^\perp$ , and therefore  $b = 0$  by (1), contradicting the fact that  $b$  is a unit vector. Hence  $B = A$ .
- (2) *implies* (3). If (3) does not hold, then by the orthogonal decomposition theorem and Lemma 8.9, the subspace  $A^\perp$  is non-trivial, and contains therefore a unit vector  $b$ . Then  $b \notin A$ , so that the orthonormal set  $B := A \cup \{b\}$  contains  $A$  properly, contradicting (2).
- (3) *implies* (4). By (3) and Lemma 8.9, the range of  $P$  is  $X$ , hence  $Px = x$  for all  $x \in X$  (cf. remark (b) in Terminology 8.5), that is,  $P = I$ .
- (4) *implies* (5). By (4), for all  $x \in X$ ,  $x = Ix = Px = \sum_{a \in A} (x, a)a$ .
- (5) *implies* (6). Given  $x, y \in A$  and representing  $A(x) = \{a_k\}$ , we have by the Cauchy-Schwarz inequality in the Hilbert space  $\mathbb{C}^n$  (for any  $n \in \mathbb{N}$ ):

$$\sum_{k=1}^n |(x, a_k) \overline{(y, a_k)}| \leq \left( \sum_{k=1}^n |(x, a_k)|^2 \right)^{1/2} \left( \sum_{k=1}^n |(y, a_k)|^2 \right)^{1/2} \leq \|x\| \|y\|,$$

where we used Bessel's inequality for the last step.

Since the partial sums of the positive series  $\sum_k |(x, a_k) \overline{(y, a_k)}|$  are bounded, the series  $\sum_k (x, a_k) \overline{(y, a_k)}$  converges absolutely, hence unconditionally, and may be written without specifying the particular ordering of  $A(x)$ ; after adding zero terms, we finally write it in the form  $\sum_{a \in A} (x, a) \overline{(y, a)}$ . Now since  $(\cdot, y)$  is a continuous linear functional, this sum is equal to

$$\sum_k (x, a_k) (a_k, y) = \left( \sum_k (x, a_k) a_k, y \right) = (x, y),$$

where we used (5) for the last step.

- (6) *implies* (7). Take  $x = y$  in (6).
- (7) *implies* (1). If  $x \in A^\perp$ ,  $(x, a) = 0$  for all  $a \in A$ , and therefore  $\|x\| = 0$  by (7). □

**Terminology 8.11.** Let  $A$  be an orthonormal set in  $X$ . If  $A$  has Property (1), it is called a *complete orthonormal set*. If it has Property (2), it is called a *maximal orthonormal set*. If it has Property (3), one says that  $A$  *spans*  $X$ . We express Property (5) by saying that *every*  $x \in X$  *has a generalized Fourier expansion with respect to*  $A$ . Properties (6) and (7) are the *Parseval* and *Bessel* identities for  $A$ , respectively.

If  $\mathcal{A}$  has any (and therefore all) of these seven properties, it is called an *orthonormal basis* or a *Hilbert basis* for  $X$ .

**Example** (1) Let  $\Gamma$  denote the unit circle in  $\mathbb{C}$ , and let  $\mathcal{A}$  be the subalgebra of  $C(\Gamma)$  consisting of the restrictions to  $\Gamma$  of all the functions in  $\text{span}\{z^k; k \in \mathbb{Z}\}$ . Since  $\bar{z} = z^{-1}$  on  $\Gamma$ ,  $\mathcal{A}$  is selfadjoint. The function  $z \in \mathcal{A}$  assumes distinct values at distinct points of  $\Gamma$ , so that  $\mathcal{A}$  is separating, and contains  $1 = z^0$ . By the Stone–Weierstrass theorem (Theorem 5.39),  $\mathcal{A}$  is dense in  $C(\Gamma)$ .

Given  $f \in L^2(\Gamma)$  (with the arc-length measure) and  $\epsilon > 0$ , let  $g \in C(\Gamma)$  be such that  $\|f - g\|_2 < \epsilon/2$  (by density of  $C(\Gamma)$  in  $L^2(\Gamma)$ ). Next let  $p \in \mathcal{A}$  be such that  $\|g - p\|_{C(\Gamma)} < \epsilon/2\sqrt{2\pi}$  (by density of  $\mathcal{A}$  in  $C(\Gamma)$ ). Then  $\|g - p\|_2 < \epsilon/2$ , and therefore  $\|f - p\|_2 < \epsilon$ . This shows that  $\mathcal{A}$  is dense in  $L^2(\Gamma)$ . Equivalently, writing  $z = e^{ix}$  with  $x \in [-\pi, \pi]$ , we proved that the span of the (obviously) orthonormal sequence

$$\left\{ \frac{e^{ikx}}{\sqrt{2\pi}}; k \in \mathbb{Z} \right\} \quad (*)$$

in the Hilbert space  $L^2(-\pi, \pi)$  is dense, that is, the sequence  $(*)$  is an *orthonormal basis* for  $L^2(-\pi, \pi)$ . In particular, every  $f \in L^2(-\pi, \pi)$  has the unique so-called  $L^2(-\pi, \pi)$ -convergent *Fourier expansion*

$$f = \sum_{k \in \mathbb{Z}} c_k e^{ikx},$$

with

$$c_k = (1/2\pi) \int_{-\pi}^{\pi} f(x) e^{-ikx} dx.$$

By Bessel's identity,

$$\sum_{k \in \mathbb{Z}} |c_k|^2 = (1/2\pi) \|f\|_2^2. \quad (**)$$

Take for example  $f = I_{(0, \pi)}$ . A simple calculation shows that  $c_0 = 1/2$  and  $c_k = (1 - e^{-ik\pi})/2\pi ik$  for  $k \neq 0$ . Therefore, for  $k \neq 0$ ,  $|c_k|^2 = (1/4\pi^2 k^2)(2 - 2\cos k\pi)$  vanishes for  $k$  even and equals  $1/\pi^2 k^2$  for  $k$  odd. Substituting in  $(**)$ , we get  $1/4 + (1/\pi^2) \sum_{k \text{ odd}} (1/k^2) = 1/2$ , hence

$$\sum_{k \text{ odd}} (1/k^2) = \pi^2/4.$$

If  $a := \sum_{k=1}^{\infty} (1/k^2)$ , then

$$\sum_{k \text{ even} \geq 2} (1/k^2) = \sum_{j=1}^{\infty} (1/4j^2) = a/4,$$

and therefore

$$a = \sum_{k \text{ odd} \geq 1} (1/k^2) + a/4 = \pi^2/8 + a/4.$$

Solving for  $a$ , we get  $a = \pi^2/6$ .

(2) Let

$$f_k(x) = \frac{e^{ikx}}{\sqrt{2\pi}} I_{(-\pi, \pi)}(x) \quad (k \in \mathbb{Z}, x \in \mathbb{R}).$$

The sequence  $\{f_k\}$  is clearly orthonormal in  $L^2(\mathbb{R})$ . If  $f$  belongs to the closure of  $\text{span}\{f_k\}$  and  $\epsilon > 0$  is given, there exists  $h$  in the span such that  $\|f - h\|_2^2 < \epsilon$ . Since  $h = 0$  on  $(-\pi, \pi)^c$ , we have

$$\int_{(-\pi, \pi)^c} |f|^2 dx = \int_{(-\pi, \pi)^c} |f - h|^2 dx < \epsilon,$$

and the arbitrariness of epsilon implies that the integral on the left-hand side vanishes. Hence  $f = 0$  a.e. outside  $(-\pi, \pi)$  (since  $f$  represents an equivalence class of functions, we may say that  $f = 0$  identically outside  $(-\pi, \pi)$ ). On the other hand, the density of the span of  $\{e^{ikx}\}$  in  $L^2(-\pi, \pi)$  means that every  $f \in L^2(\mathbb{R})$  vanishing outside  $(-\pi, \pi)$  is in the closure of  $\text{span}\{f_k\}$  in  $L^2(\mathbb{R})$ . Thus  $\{f_k\}$  is an orthonormal sequence in  $L^2(\mathbb{R})$  which is *not* an orthonormal basis for the space; the closure of its span consists precisely of all  $f \in L^2(\mathbb{R})$  vanishing outside  $(-\pi, \pi)$ .

**Theorem 8.12 (Existence of orthonormal bases).** *Every (non-trivial) Hilbert space has an orthonormal base.*

*More specifically, given any orthonormal set  $A_0$  in the Hilbert space  $X$ , it can be completed to an orthonormal base  $A$ .*

**Proof.** Since a non-trivial Hilbert space contains a unit vector  $a_0$ , the first statement of the theorem follows from the second with the orthonormal set  $A_0 = \{a_0\}$ .

To prove the second statement, consider the family  $\mathcal{A}$  of all orthonormal sets in  $X$  containing the given set  $A_0$ . It is non-empty (since  $A_0 \in \mathcal{A}$ ) and partially ordered by inclusion. If  $\mathcal{A}' \subset \mathcal{A}$  is totally ordered, it is clear that  $\bigcup \mathcal{A}'$  is an *orthonormal set* (because of the total order!) containing  $A_0$ , that is, it is an element of  $\mathcal{A}$ , and is an upper bound for all the sets in  $\mathcal{A}'$ . Therefore, by Zorn's lemma, the family  $\mathcal{A}$  contains a maximal element  $A$ ;  $A$  is a maximal orthonormal set (hence an orthonormal base, cf. Section 8.11) containing  $A_0$ .  $\square$

## 8.4 Hilbert dimension

**Theorem 8.13 (Equi-cardinality of orthonormal bases).** *All orthonormal bases of a given Hilbert space have the same cardinality.*

**Proof.** Note first that an orthonormal set  $A$  is necessarily linearly independent. Indeed, if the finite linear combination  $x := \sum_{k=1}^n \lambda_k a_k$  of vectors  $a_k \in A$  vanishes, then  $\lambda_j = (x, a_j) = (0, a_j) = 0$  for all  $j = 1, \dots, n$ .

We consider two orthonormal bases  $A, B$  of the Hilbert space  $X$ .

*Case 1.* At least one of the bases is finite.

We may assume that  $A$  is finite, say  $A = \{a_1, \dots, a_n\}$ . By Property (5) of the orthonormal base  $A$ , the vectors  $a_1, \dots, a_n$  span  $X$  (in the algebraic sense!), and are linearly independent by the above remark. This means that  $\{a_1, \dots, a_n\}$  is a base (in the algebraic sense) for the vector space  $X$ , and therefore the algebraic dimension of  $X$  is  $n$ . Since  $B \subset X$  is linearly independent (by the above remark), its cardinality  $|B|$  is at most  $n$ , that is,  $|B| \leq |A|$ . In particular,  $B$  is finite, and the preceding conclusion applies with  $B$  and  $A$  interchanged, that is,  $|A| \leq |B|$ , and the equality  $|A| = |B|$  follows.

*Case 2.* Both bases are infinite.

With notation as in Theorem 8.7, we claim that

$$A = \bigcup_{b \in B} A(b).$$

Indeed, suppose some  $a \in A$  is *not* in the above union. Then  $a \notin A(b)$  for all  $b \in B$ , that is,  $(a, b) = 0$  for all  $b \in B$ . Hence  $a \in B^\perp = \{0\}$  (by Property (1) of the orthonormal base  $B$ ), but this is absurd since  $a$  is a unit vector.

By Theorem 8.7, each set  $A(b)$  is at most countable; therefore the union above has cardinality  $\leq \aleph_0 \times |B|$ . Also  $\aleph_0 \leq |B|$  (since  $B$  is infinite). Hence  $|A| \leq |B|^2 = |B|$  (the last equation is a well-known property of *infinite* cardinalities). By symmetry of the present case with respect to  $A$  and  $B$ , we also have  $|B| \leq |A|$ , and the equality of the cardinalities follows.  $\square$

**Definition 8.14.** The cardinality of any (hence of all) orthonormal bases of the Hilbert space  $X$  is called the *Hilbert dimension of  $X$* , denoted  $\dim_H X$ .

## 8.5 Isomorphism of Hilbert spaces

Two Hilbert spaces  $X$  and  $Y$  are *isomorphic* if there exists an algebraic isomorphism  $V : X \rightarrow Y$  that preserves the inner product:  $(Vp, Vq) = (p, q)$  for all  $p, q \in X$  (the same notation is used for the inner product in both spaces).

The map  $V$  is necessarily isometric (since it is linear and norm-preserving). Conversely, by the polarization identity (cf. (11) following Definition 1.34), any bijective isometric linear map between Hilbert spaces is an isomorphism (of Hilbert spaces). Such an isomorphism is also called a *unitary equivalence*; accordingly, isomorphic Hilbert spaces are said to be unitarily equivalent.

The isomorphism relation between Hilbert spaces is clearly an equivalence relation. Each equivalence class is completely determined by the Hilbert dimension:

**Theorem 8.15.** *Two Hilbert spaces are isomorphic if and only if they have the same Hilbert dimension.*

**Proof.** If the Hilbert spaces  $X, Y$  have the same Hilbert dimension, and  $A, B$  are orthonormal bases for  $X$  and  $Y$ , respectively, then since  $|A| = |B|$ , we can



choose an index set  $J$  to index the elements of *both*  $A$  and  $B$ :

$$A = \{a_j; j \in J\}; \quad B = \{b_j; j \in J\}.$$

By Property (5) of orthonormal bases, there is a unique continuous linear map  $V : X \rightarrow Y$  such that  $Va_j = b_j$  for all  $j \in J$  (namely  $Vx = \sum_{j \in J} (x, a_j)b_j$  for all  $x \in X$ ). There is also a unique continuous linear map  $W : Y \rightarrow X$  such that  $Wb_j = a_j$  for all  $j \in J$ . Clearly  $VW = WV = I$ , where  $I$  denotes the identity map in both spaces. Therefore  $V$  is bijective, and by Parseval's identity

$$(Vx, Vy) = \left( \sum_{j \in J} (x, a_j)b_j, \sum_{j \in J} (y, a_j)b_j \right) = \sum_{j \in J} (x, a_j)\overline{(y, a_j)} = (x, y)$$

for all  $x, y \in X$ . Thus  $V$  is a Hilbert space isomorphism of  $X$  onto  $Y$ .

Conversely, suppose the Hilbert spaces  $X$  and  $Y$  are isomorphic, and let then  $V : X \rightarrow Y$  be an isomorphism (of Hilbert spaces). If  $A$  is an orthonormal base for  $X$ , then  $VA$  is an orthonormal base for  $Y$ . Indeed,  $VA$  is orthonormal, because

$$(Vs, Vt) = (s, t) = \delta_{s,t} = \delta_{Vs, Vt}$$

for all  $s, t \in A$  (the last equality follows from the injectiveness of  $V$ ). In order to show completeness, let  $y \in (VA)^\perp$  (in  $Y$ ), and write  $y = Vx$  for some  $x \in X$  (since  $V$  is onto). Then for all  $a \in A$

$$(x, a) = (Vx, Va) = (y, Va) = 0,$$

that is,  $x \in A^\perp = \{0\}$  (by Property (1) of the orthonormal base  $A$ ). Hence  $y = 0$ , and we conclude that  $VA$  is indeed an orthonormal base for  $Y$  (cf. Section 8.11).

Now by Theorem 8.13 and the fact that  $V : A \rightarrow VA$  is bijective,  $\dim_H Y = |VA| = |A| = \dim_H X$ .  $\square$

## 8.6 Canonical model

Given any cardinality  $\gamma$ , choose a set  $J$  with this cardinality. Consider the vector space  $l^2(J)$  of all functions  $f : J \rightarrow \mathbb{C}$  with  $J(f) := \{j \in J; f(j) \neq 0\}$  at most countable and

$$\|f\|_2 := \left( \sum_{j \in J} |f(j)|^2 \right)^{1/2} < \infty.$$

The inner product associated with the norm  $\|\cdot\|_2$  is defined by the absolutely convergent series

$$(f, g) := \sum_{j \in J} f(j)\overline{g(j)} \quad (f, g \in l^2(J)).$$

The space  $l^2(J)$  is the Hilbert direct sum of copies  $\mathbb{C}_j$  ( $j \in J$ ) of the Hilbert space  $\mathbb{C}$  (cf. section following Lemma 7.28). In particular,  $l^2(J)$  is a Hilbert space (it is

actually the  $L^2$  space of the measure space  $(J, \mathbb{P}(J), \mu)$ , with  $\mu$  the *counting measure*, that is equal to one on singletons). For each  $j \in J$ , let  $a_j \in l^2(J)$  be defined by  $a_j(i) = \delta_{i,j}$  ( $i, j \in J$ ). Then  $A := \{a_j; j \in J\}$  is clearly orthonormal, and  $(f, a_j) = f(j)$  for all  $f \in l^2(J)$  and  $j \in J$ . In particular  $A^\perp = \{0\}$ , so that  $A$  is an orthonormal base for  $l^2(J)$ . Since  $|A| = |J| = \gamma$ , the space  $l^2(J)$  has Hilbert dimension  $\gamma$ . By Theorem 8.15, every Hilbert space with Hilbert dimension  $\gamma$  is isomorphic to the ‘canonical model’  $l^2(J)$ .

## 8.7 Commutants

Let  $X$  be a Hilbert space, and let  $B(X)$  be the  $B^*$ -algebra of all bounded operators on  $X$ . For any subset  $\mathcal{A} \subset B(X)$ , the *commutant*  $\mathcal{A}'$  of  $\mathcal{A}$  is the set

$$\mathcal{A}' := \{T \in B(X); TA = AT \text{ for all } A \in \mathcal{A}\}.$$

(For a singleton  $\{T\}$ , we write  $T'$  instead of  $\{T\}'$ .)

The commutant  $\mathcal{A}'$  is closed in the w.o.t. Indeed, if the net  $\{T_j; j \in J\} \subset \mathcal{A}'$  converges to the operator  $T$  in the w.o.t. (that is, by the Riesz Representation theorem,  $(T_j x, y) \rightarrow (T x, y)$  for all  $x, y \in X$ ), then for all  $x, y \in X$  and  $A \in \mathcal{A}$ ,

$$\begin{aligned} (TAx, y) &= \lim_j (T_j A x, y) = \lim_j (A T_j x, y) \\ &= \lim_j (T_j x, A^* y) = (T x, A^* y) = (ATx, y), \end{aligned}$$

hence  $TA = AT$  for all  $A \in \mathcal{A}$ , that is,  $T \in \mathcal{A}'$ .

It follows that the *second commutant*

$$\mathcal{A}'' := (\mathcal{A}')'$$

is a w.o.-closed set (to be read: *weak operator closed*, i.e., a set closed in the w.o.t.) that contains  $\mathcal{A}$  (trivially). In particular, the relation  $\mathcal{A} = \mathcal{A}''$  implies that  $\mathcal{A}$  is w.o.-closed. We show below that the converse is also true in case  $\mathcal{A}$  is a *\*-subalgebra* of  $B(X)$ .

**Theorem 8.16 (Von Neumann’s double commutant theorem).** *Let  $\mathcal{A}$  be a selfadjoint subalgebra of  $B(X)$ . Then  $\mathcal{A}$  is w.o.-closed if and only if  $\mathcal{A} = \mathcal{A}''$ .*

**Proof.** By the comments preceding the statement of the theorem, we must only show that if  $T \in \mathcal{A}''$ , then  $T \in \mathcal{A}_w$  (where  $\mathcal{A}_w$  denotes the w.o.-closure of  $\mathcal{A}$ ). However, since  $\mathcal{A}$  is convex, its w.o.-closure coincides with its s.o.-closure  $\mathcal{A}_s$ , by Corollary 6.20. We must then show that any strong operator basic neighbourhood  $N(T, F, \epsilon)$  of  $T$  meets  $\mathcal{A}$ .

Recall that  $F$  is an arbitrary finite subset of  $X$ , say  $F = \{x_1, \dots, x_n\}$ .

Consider first the special case when  $n = 1$ . Let then  $M := \overline{\mathcal{A}x_1}$  (the norm-closure of the  $\mathcal{A}$ -cycle generated by  $x_1$  in  $X$ ), and let  $P$  be the orthogonal projection onto the closed subspace  $M$  of  $X$ . Since  $\mathcal{A}$  is stable under multiplication,  $M$  is  $A$ -invariant for all  $A \in \mathcal{A}$ . Therefore, for each  $A \in \mathcal{A}$ ,  $M$  is invariant

for both  $A$  and  $A^*$  (because  $A^* \in \mathcal{A}$ , by selfadjointness of  $\mathcal{A}$ ). By Section 8.5, Point 3,  $PA = AP$  for all  $A \in \mathcal{A}$ , that is,  $P \in \mathcal{A}'$ . Therefore  $TP = PT$  (since  $T \in \mathcal{A}''$ ). In particular,  $M$  is  $T$ -invariant, and so  $Tx_1 \in M$ . Therefore there exists  $A \in \mathcal{A}$  such that  $\|Tx_1 - Ax_1\| < \epsilon$ , that is,  $A \in N(T, \{x_1\}, \epsilon)$ , as wanted.

The case of arbitrary  $n$  is reduced to the case  $n = 1$  by looking at the space  $X^n$ , direct sum of  $n$  copies of  $X$ . The algebra  $B(X^n)$  is isomorphic to the matrix algebra  $M_n(B(X))$  (of  $n \times n$ -matrices with coefficients in  $B(X)$ ). In particular, two operators on  $X^n$  commute iff their associated matrices commute.

If  $T \in B(X)$ , let  $T^{(n)}$  be the operator on  $X^n$  with matrix  $\text{diag}(T, \dots, T)$  (the diagonal matrix with  $T$  on its diagonal). For any  $\underline{x} := [x_1, \dots, x_n] \in X^n$ ,

$$\begin{aligned}\|T^{(n)}\underline{x}\|^2 &= \|[Tx_1, \dots, Tx_n]\|^2 \\ &= \sum_{k=1}^n \|Tx_k\|^2 \leq \|T\|^2 \sum_k \|x_k\|^2 = \|T\|^2 \|\underline{x}\|^2.\end{aligned}$$

Hence  $\|T^{(n)}\| \leq \|T\|$ . On the other hand, for any unit vector  $x \in X$ ,

$$\|T^{(n)}\| \geq \|T^{(n)}[x, 0, \dots, 0]\| = \|[Tx, 0, \dots, 0]\| = \|Tx\|,$$

and therefore  $\|T^{(n)}\| \geq \|T\|$ . Hence  $\|T^{(n)}\| = \|T\|$ .

A simple calculation shows that  $(T^{(n)})^* = (T^*)^{(n)}$ . It follows that  $\mathcal{A}^{(n)} := \{A^{(n)}; A \in \mathcal{A}\}$  is a selfadjoint subalgebra of  $B(X^n)$ .

One verifies easily that  $T \in (\mathcal{A}^{(n)})'$  iff its associated matrix has entries  $T_{ij}$  in  $\mathcal{A}'$ .

Let  $T \in \mathcal{A}''$  and consider as above the s.o.-neighbourhood  $N(T, \{x_1, \dots, x_n\}, \epsilon)$ . Then  $T^{(n)} \in (\mathcal{A}^{(n)})''$ , and therefore, by the case  $n = 1$  proved above, the s.o.-neighbourhood  $N(T^{(n)}, \underline{x}, \epsilon)$  in  $B(X^n)$  meets  $\mathcal{A}^{(n)}$ , that is, there exists  $A \in \mathcal{A}$  such that

$$\|(T^{(n)} - A^{(n)})\underline{x}\| < \epsilon.$$

Hence, for all  $k = 1, \dots, n$ ,

$$\|(T - A)x_k\| \leq \left( \sum_k \|(T - A)x_k\|^2 \right)^{1/2} = \|(T - A)^{(n)}\underline{x}\| < \epsilon,$$

that is,  $A \in N(T, \{x_1, \dots, x_n\}, \epsilon)$ . □

## Exercises

### The trigonometric Hilbert basis of $L^2$

1. Let  $m$  denote the normalized Lebesgue measure on  $[-\pi, \pi]$ . (The integral of  $f \in L^1(m)$  over this interval will be denoted by  $\int f dm$ .) Let

$e_k(x) = e^{ikx}$  ( $k \in \mathbb{Z}$ ), and denote

$$s_n = \sum_{|k| \leq n} e_k \quad (n = 0, 1, 2, \dots)$$

$$\sigma_n = (1/n) \sum_{j=0}^{n-1} s_j \quad (n \in \mathbb{N}).$$

Note that  $\sigma_1 = s_0 = e_0 = 1$ ,  $\{e_k; k \in \mathbb{Z}\}$  is orthonormal in the Hilbert space  $L^2(m)$ , and

$$\int s_n dm = (s_n, e_0) = \sum_{|k| \leq n} (e_k, e_0) = 1$$

$$\int \sigma_n dm = (1/n) \sum_{j=0}^{n-1} \int s_j dm = 1.$$

Prove

(a)

$$s_n(x) = \frac{\cos(nx) - \cos(n+1)x}{1 - \cos x} = \frac{\sin(n+1/2)x}{\sin(x/2)}.$$

(b)

$$\sigma_n(x) = (1/n) \frac{1 - \cos(nx)}{1 - \cos x} = (1/n) \left( \frac{\sin(nx/2)}{\sin(x/2)} \right)^2.$$

- (c)  $\{\sigma_n\}$  is an ‘approximate identity’ in the sense of Exercise 6, Chapter 4. Consequently  $\sigma_n * f \rightarrow f$  uniformly on  $[-\pi, \pi]$  if  $f \in C([-\pi, \pi])$  and  $f(\pi) = f(-\pi)$ , and in  $L^p$ -metric if  $f \in L^p := L^p(m)$  (for each  $p \in [1, \infty)$ ). (Cf. Exercise 6, Chapter 4.)
- (d) Consider the orthogonal projections  $P$  and  $P_n$  associated with the orthonormal sets  $\{e_k; k \in \mathbb{Z}\}$  and  $\{e_k; |k| \leq n\}$  respectively, and denote  $Q_n = (1/n) \sum_{j=0}^{n-1} P_j$  for  $n \in \mathbb{N}$ .

*Terminology:*  $Pf := \sum_{k \in \mathbb{Z}} (f, e_k) e_k$  is called the (formal) Fourier series of  $f$  for any integrable  $f$  (it converges in  $L^2$  if  $f \in L^2$ );  $(f, e_k)$  is the  $k$ th Fourier coefficient of  $f$ ;  $P_n f$  is the  $n$ th partial sum of the Fourier series for  $f$ ;  $Q_n f$  is the  $n$ th Cesaro mean of the Fourier series of  $f$ .

Observe that  $P_n f = s_n * f$  and  $Q_n f = \sigma_n * f$  for any integrable function  $f$ . Consequently  $Q_n f \rightarrow f$  uniformly in  $[-\pi, \pi]$  if  $f \in C_T := \{f \in C([-\pi, \pi]); f(\pi) = f(-\pi)\}$ , and in  $L^p$ -norm if  $f \in L^p$  (for each  $p \in [1, \infty)$ ). If  $f \in L^\infty := L^\infty(-\pi, \pi)$ ,  $Q_n f \rightarrow f$  in the weak\*-topology on  $L^\infty$ . (Cf. Exercise 6, Chapter 4.)

- (e)  $\{e_k; k \in \mathbb{Z}\}$  is a Hilbert basis for  $L^2(m)$ . (Note that  $Q_n f \in \text{span}\{e_k\}$  and use Part (d).)

## Fourier coefficients

2. (Notation as in Exercise 1.) Given  $k \in \mathbb{Z} \rightarrow c_k \in \mathbb{C}$ , denote  $g_n = \sum_{|k| \leq n} c_k e_k$  ( $n = 0, 1, 2, \dots$ ) and  $G_n = (1/n) \sum_{j=0}^{n-1} g_j$  ( $n \in \mathbb{N}$ ). Note that  $(g_n, e_m) = c_m$  for  $n \geq |m|$  and  $= 0$  for  $n < |m|$ , and consequently  $(G_n, e_k) = (1 - |k|/n)c_k$  for  $n > |k|$  and  $= 0$  for  $n \leq |k|$ .

- (a) Let  $p \in (1, \infty]$ . Prove that if

$$M := \sup_n \|G_n\|_p < \infty,$$

then there exists  $f \in L^p$  such that  $c_k = (f, e_k)$  for all  $k \in \mathbb{Z}$ , and conversely. (Hint: the ball  $\bar{B}(0, M)$  in  $L^p$  is *weak\**-compact, cf. Theorems 5.24 and 4.6. For the converse, see Exercise 1.)

- (b) If  $\{G_n\}$  converges in  $L^1$ -norm, then there exists  $f \in L^1$  such that  $c_k = (f, e_k)$  for all  $k \in \mathbb{Z}$ , and conversely.
- (c) If  $\{G_n\}$  converges uniformly in  $[-\pi, \pi]$ , then there exists  $f \in C_T$  such that  $c_k = (f, e_k)$  for all  $k \in \mathbb{Z}$ , and conversely.
- (d) If  $\sup_n \|G_n\|_1 < \infty$ , there exists a complex Borel measure  $\mu$  on  $[-\pi, \pi]$  with  $\mu(\{\pi\}) = \mu(\{-\pi\})$  (briefly,  $\mu \in M_T$ ) such that  $c_k = \int e_k d\mu$  for all  $k$ , and conversely. (Hint: Consider the measures  $d\mu_n = G_n dm$ , and apply Theorems 4.9 and 5.24.)
- (e) If  $G_n \geq 0$  for all  $n \in \mathbb{N}$ , there exists a finite *positive* Borel measure  $\mu$  as in Part (d), and conversely.

## Poisson integrals

3. (Notation as in Exercise 1.) Let  $D$  be the open unit disc in  $\mathbb{C}$ .
- (a) Verify that  $(e_1 + z)/(e_1 - z) = 1 + 2 \sum_{k \in \mathbb{N}} e_{-k} z^k$  for all  $z \in D$ , where the series converges absolutely and uniformly in  $z$  in any compact subset of  $D$ . Conclude that for any complex Borel measure  $\mu$  on  $[-\pi, \pi]$ ,

$$g(z) := \int \frac{e_1 + z}{e_1 - z} d\mu = \mu([-\pi, \pi]) + 2 \sum_{k \in \mathbb{N}} c_k z^k, \quad (1)$$

where  $c_k = \int e_{-k} d\mu$  and integration is over  $[-\pi, \pi]$ . In particular,  $g$  is analytic in  $D$ , and if  $\mu$  is *real*,  $\Re g(z) = \int \Re((e_1 + z)/(e_1 - z)) d\mu$  is (real) harmonic in  $D$ . Verify that the ‘kernel’ in the last integral has the form  $P_r(\theta - t)$ , where  $z = re^{i\theta}$  and

$$P_r(\theta) := \frac{1 - r^2}{1 - 2r \cos \theta + r^2}$$

is the classical *Poisson kernel* (for the disc). Thus

$$(\Re g)(r e^{i\theta}) = (P_r * \mu)(\theta) := \int P_r(\theta - t) d\mu(t). \quad (2)$$

- (b) Let  $\mu$  be a complex Borel measure on  $[-\pi, \pi]$ . Then  $P_r * \mu$  is a complex harmonic function in  $D$  (as a function of  $z = r e^{i\theta}$ ). (This is true in particular for  $P_r * f$ , for any  $f \in L^1(m)$ .)
- (c) Verify that  $\{P_r; 0 < r < 1\}$  is an ‘approximate identity’ for  $L^1$  in the sense of Exercise 6, Chapter 4 (with the continuous parameter  $r$  instead of the discrete  $n$ ). Consequently, as  $r \rightarrow 1$ ,
- (i) if  $f \in L^p$  for some  $p < \infty$ , then  $P_r * f \rightarrow f$  in  $L^p$ -norm;
  - (ii) if  $f \in C_T$ , then  $P_r * f \rightarrow f$  uniformly in  $[-\pi, \pi]$ ;
  - (iii) if  $f \in L^\infty$ , then  $P_r * f \rightarrow f$  in the *weak\**-topology on  $L^\infty$ ;
  - (iv) if  $\mu \in M_T$ , then  $(P_r * \mu)dm \rightarrow d\mu$  in the *weak\**-topology.
- (d) For  $\mu \in M_T$ , denote  $F(t) = \mu([-\pi, t])$  and verify the identity

$$(P_r * \mu)(\theta) = r \int K_r(t) \frac{F(\theta + t) - F(\theta - t)}{2 \sin t} dt, \quad (3)$$

where

$$K_r(t) = \frac{(1 - r^2) \sin^2 t}{(1 - 2r \cos t + r^2)^2}. \quad (4)$$

Verify that  $\{K_r; 0 < r < 1\}$  is an approximate identity for  $L^1(m)$  in the sense of Exercise 6, Chapter 4. (Hint: integration by parts.)

- (e) Let  $G_\theta(t)$  denote the function integrated against  $K_r(t)$  in (3). If  $F$  is differentiable at the point  $\theta$ ,  $G_\theta(\cdot)$  is continuous at 0 and  $G_\theta(0) = F'(\theta)$ . Conclude from Part (d) that  $P_r * \mu \rightarrow 2\pi F' (= 2\pi D\mu = d\mu/dm)$  as  $r \rightarrow 1$  at all points  $\theta$  where  $F$  is differentiable, that is,  $m$ -almost everywhere in  $[-\pi, \pi]$ . (Cf. Theorem 3.28 with  $k = 1$  and Exercise 4e, Chapter 3; note that here  $m$  is *normalized* Lebesgue measure on  $[-\pi, \pi]$ .) This is the ‘radial limit’ version of *Fatou’s theorem* on ‘Poisson integrals’. (The same conclusion is true with ‘non-tangential convergence’ of  $re^{i\theta}$  to points of the unit circle.)
- (f) State and prove the analogue of Exercise 2 for the representation of harmonic functions in  $D$  as Poisson integrals.
4. *Poisson integrals in the right half-plane.* Let  $\mathbb{C}^+$  denote the right half-plane, and

$$P_x(y) := \pi^{-1} \frac{x}{x^2 + y^2} \quad (x > 0; y \in \mathbb{R}).$$

(This is the so-called *Poisson kernel of the right half-plane*.) Prove:

- (a)  $\{P_x; x > 0\}$  is an approximate identity for  $L^1(\mathbb{R})$  (as  $x \rightarrow 0+$ ). (Cf. Exercise 6, Chapter 4.) Consequently, as  $x \rightarrow 0+$ ,  $P_x * f \rightarrow f$  uniformly on  $\mathbb{R}$  if  $f \in C_c(\mathbb{R})$ , and in  $L^p$ -norm if  $f \in L^p(\mathbb{R})$  ( $1 \leq p < \infty$ ).

- (b)  $(P_x * f)(y)$  is a harmonic function of  $(x, y)$  in  $\mathbb{C}^+$ .
- (c) If  $f \in L^p(\mathbb{R})$ , then for each  $\delta > 0$ ,  $(P_x * f)(y) \rightarrow 0$  uniformly for  $x \geq \delta$  as  $x^2 + y^2 \rightarrow \infty$ . (Hint: use Holder's inequality with the probability measure  $(1/\pi)P_x(y-t)dt$  for  $x, y$  fixed.)
- (d) If  $f \in L^1(dt/(1+t^2))$ , then  $P_x * f \rightarrow f$  as  $x \rightarrow 0+$  pointwise a.e. on  $\mathbb{R}$ . (Hint: immitate the argument in Parts (d)–(e) of the preceding exercise, or transform the disc onto the half-plane and use Fatou's theorem for the disc.)
5. Let  $\mu$  be a complex Borel measure on  $[-\pi, \pi]$ . Show that

$$\lim_{n \rightarrow \infty} \int e_{-n} d\mu = 0 \quad (5)$$

if and only if

$$\lim_{n \rightarrow \infty} \int e_{-n} d|\mu| = 0. \quad (6)$$

(Hint: (5) for the measure  $\mu$  implies (5) for the measure  $d\nu = h d\mu$  for any trigonometric polynomial  $h$ ; use a density argument and the relation  $d|\mu| = h d\mu$  for an appropriate  $h$ .)

## Divergence of Fourier series

6. (Notation as in Exercise 1.) Consider the partial sums  $P_n f$  of the Fourier series of  $f \in C_T$ . Let

$$\phi_n(f) := (P_n f)(0) = (s_n * f)(0) \quad (f \in C_T). \quad (7)$$

Prove:

- (a) For each  $n \in \mathbb{N}$ ,  $\phi_n$  is a bounded linear functional on  $C_T$  with norm  $\|s_n\|_1$  (the  $L^1(m)$ -norm of  $s_n$ ). Hint:  $\|\phi_n\| \leq \|s_n\|_1$  trivially. Consider real functions  $f_j \in C_T$  such that  $\|f_j\|_\infty \leq 1$  and  $f_j \rightarrow \operatorname{sgn} s_n$  a.e., cf. Exercise 9, Chapter 3. Then  $\phi_n(f_j) \rightarrow \|s_n\|_1$ .
- (b)  $\lim_n \|s_n\|_1 = \infty$ . (Use the fact that the *Dirichlet integral*  $\int_0^\infty ((\sin t)/t)dt$  does *not* converge *absolutely*.)
- (c) The subspace

$$Z := \{f \in C_T; \sup_n |\phi_n(f)| < \infty\}$$

is of Baire's first category in the Banach space  $C_T$ . Conclude that the subspace of  $C_T$  consisting of all  $f \in C_T$  with convergent Fourier series at 0 is of Baire's first category in  $C_T$ . (Hint: assume  $Z$  is of Baire's second category in  $C_T$  and apply Theorem 6.4 and Parts (a) and (b).)

## Fourier coefficients of $L^1$ functions

7. (Notation as in Exercise 1.)

(a) If  $f \in L^1 := L^1(m)$ , prove that

$$\lim_{|k| \rightarrow \infty} (f, e_k) = 0. \quad (8)$$

Hint: if  $f = e_n$  for some  $n \in \mathbb{Z}$ ,  $(f, e_k) = 0$  for all  $k$  such that  $|k| > |n|$ , and (8) is trivial. Hence (8) is true for  $f \in \text{span}\{e_n; n \in \mathbb{Z}\}$ , and the general case follows by density of this span in  $L^1$ , cf. Exercise 1, Part (d).

(b) Consider  $\mathbb{Z}$  with the discrete topology (this is a locally compact Hausdorff space!), and let  $c_0 := C_0(\mathbb{Z})$ ; cf. Definition 3.23. Consider the map

$$F : f \in L^1 \rightarrow \{(f, e_k)\} \in c_0.$$

(Cf. Part a) Then  $F \in B(L^1, c_0)$  is one-to-one. Hint: if  $(f, e_k) = 0$  for all  $k \in \mathbb{Z}$ , then  $(f, g) = 0$  for all  $g \in \text{span}\{e_k\}$ , hence for all  $g \in C_T$  by Exercise 1(d), hence for  $g = I_E$  for any measurable subset  $E$  of  $[-\pi, \pi]$ ; Cf. Exercise 9, Chapter 3, and Proposition 1.22.

(c) Prove that the range of  $F$  is of Baire's first category in the Banach space  $c_0$  (in particular,  $F$  is *not onto*). Hint: if the range of  $F$  is of Baire's second category in  $c_0$ ,  $F^{-1}$  with domain range  $F$  is continuous, by Theorem 6.9. Therefore there exists  $c > 0$  such that  $\|Ff\|_u \geq c\|f\|_1$  for all  $f \in L^1$ . Get a contradiction by choosing  $f = s_n$ ; cf. Exercise 6(b).

## Miscellaneous

8. Let  $\{a_n\}$  and  $\{b_n\}$  be two orthonormal sequences in the Hilbert space  $X$  such that

$$\sum_n \|b_n - a_n\|^2 < 1.$$

Prove that  $\{a_n\}$  is a Hilbert basis for  $X$  iff this is true for  $\{b_n\}$ .

9. Let  $\{a_k\}$  be a Hilbert basis for the Hilbert space  $X$ . Define  $T \in B(X)$  by  $Ta_k = a_{k+1}$ ,  $k \in \mathbb{N}$ . Prove:

(a)  $T$  is isometric.

(b)  $T^n \rightarrow 0$  in the weak operator topology.

10. If  $\{a_n\}$  is an orthonormal (infinite) sequence in an inner product space, then  $a_n \rightarrow 0$  weakly (however  $\{a_n\}$  has no strongly convergent subsequence, because  $\|a_n - a_m\|^2 = 2$ ; this shows in particular that the



closed unit ball of an infinite dimensional Hilbert space is not strongly compact.)

11. Let  $\{x_\alpha\}$  be a net in the inner product space  $X$ . Then  $x_\alpha \rightarrow x \in X$  strongly iff  $x_\alpha \rightarrow x$  weakly and  $\|x_\alpha\| \rightarrow \|x\|$ .
12. Let  $A$  be an orthonormal basis for the Hilbert space  $X$ . Prove:
  - (a) If  $f : A \rightarrow X$  is any map such that  $(f(a), a) = 0$  for all  $a \in A$ , then it does not necessarily follow that  $f = 0$  (the zero map on  $A$ ).
  - (b) If  $T \in B(X)$  is such that  $(Tx, x) = 0$  for all  $x \in X$ , then  $T = 0$  (the zero operator).
  - (c) If  $S, T \in B(X)$  are such that  $(Tx, x) = (Sx, x)$  for all  $x \in X$ , then  $S = T$ .
13. Let  $X$  be a Hilbert space, and let  $\mathcal{N}$  be the set of normal operators in  $B(X)$ . Prove that the adjoint operation  $T \rightarrow T^*$  is continuous on  $\mathcal{N}$  in the s.o.t.
14. Let  $X$  be a Hilbert space, and  $T \in B(X)$ . Denote by  $P(T)$  and  $Q(T)$  the orthogonal projections onto the closed subspaces  $\ker T$  and  $\overline{TX}$ , respectively. Prove:
  - (a) The complementary orthogonal projections of  $P(T)$  and  $Q(T)$  are  $Q(T^*)$  and  $P(T^*)$ , respectively.
  - (b)  $P(T^*T) = P(T)$  and  $Q(T^*T) = Q(T^*)$ .
15. For any (non-empty) set  $A$ , denote by  $\mathbb{B}(A)$  the  $B^*$ -algebra of all bounded complex functions on  $A$  with pointwise operations, the involution  $f \rightarrow \bar{f}$  (complex conjugation), and the supremum norm  $\|f\|_u = \sup_A |f|$ .

Let  $A$  be an orthonormal basis of the Hilbert space  $X$ . For each  $f \in \mathbb{B}(A)$  and  $x \in X$ , let

$$T_f x = \sum_{a \in A} f(a)(x, a)a.$$

Prove:

- (a) The map  $f \rightarrow T_f$  is an isometric  $*$ -isomorphism of the  $B^*$ -algebra  $\mathbb{B}(A)$  into  $B(X)$ . (In particular,  $T_f$  is a normal operator.)
  - (b)  $T_f$  is selfadjoint (positive, unitary) iff  $f$  is real-valued ( $f \geq 0$ ,  $|f| = 1$ , respectively).
16. Let  $(S, \mathcal{A}, \mu)$  be a  $\sigma$ -finite positive measure space. Consider  $L^\infty(\mu)$  as a  $B^*$ -algebra with pointwise multiplication and complex conjugation as involution. Let  $p \in [1, \infty)$ . For each  $f \in L^\infty(\mu)$  define

$$T_f g = fg \quad (g \in L^p(\mu)).$$

Prove:

- (a) The map  $f \rightarrow T_f$  is an isometric isomorphism of  $L^\infty(\mu)$  into  $B(L^p(\mu))$ ; in case  $p = 2$ , the map is an isometric  $*$ -isomorphism of  $L^\infty(\mu)$  onto a commutative  $B^*$ -algebra of (normal) operators on  $L^2(\mu)$ .
  - (b) (Case  $p = 2$ .)  $T_f$  is selfadjoint (positive, unitary) iff  $f$  is real-valued ( $f \geq 0$ ,  $|f| = 1$ , respectively) almost everywhere.
17. Let  $X$  be a Hilbert space. Show that multiplication in  $B(X)$  is *not* (jointly) continuous in the w.o.t., even on the norm-closed unit ball  $B(X)_1$  of  $B(X)$  in the relative w.o.t.; however it is continuous on  $B(X)_1$  in the relative s.o.t.
18. (Notation as in Exercise 17.) Prove that  $B(X)_1$  is compact in the w.o.t. but *not* in the s.o.t.
19. Let  $X, Y$  be Hilbert spaces, and  $T \in B(X, Y)$ . Imitate the definition of the Hilbert adjoint of an operator in  $B(X)$  to define the (Hilbert) adjoint  $T^* \in B(Y, X)$ . Observe that  $T^*T$  is a positive operator in  $B(X)$ . Let  $\{a_n\}_{n \in \mathbb{N}}$  be an orthonormal basis for  $X$  and let  $Vx = \{(x, a_n)\}$ . We know that  $V$  is a Hilbert space isomorphism of  $X$  onto  $l^2$ . What are  $V^*$ ,  $V^{-1}$ , and  $V^*V$ ?
20. Let  $\{a_n\}_{n \in \mathbb{N}}$  be an orthonormal basis for the Hilbert space  $X$ , and let  $Q \in B(X)$  be invertible in  $B(X)$ . Let  $b_n = Qa_n$ ,  $n \in \mathbb{N}$ . Prove that there exist positive constants  $A, B$  such that

$$A \sum |\lambda_k|^2 \leq \left\| \sum \lambda_k b_k \right\|^2 \leq B \sum |\lambda_k|^2$$

for all finite sets of scalars  $\lambda_k$ .

21. Let  $X$  be a Hilbert space. A sequence  $\{a_n\} \subset X$  is *upper (lower) Bessel* if there exists a positive constant  $B$  (resp.  $A$ ) such that

$$\sum |(x, a_n)|^2 \leq B \|x\|^2 \quad (\geq A \|x\|^2, \text{ resp.})$$

for all  $x \in X$ . The sequence is *two-sided Bessel* if it is both upper and lower Bessel (e.g., an orthonormal sequence is upper Bessel with  $B = 1$ , and is two-sided Bessel with  $A = B = 1$  iff it is an orthonormal basis for  $X$ ).

- (a) Let  $\{a_n\}$  be an upper Bessel sequence in  $X$ , and define  $V$  as in Exercise 19. Then  $V \in B(X, l^2)$  and  $\|V\| \leq B^{1/2}$ . On the other hand, for any  $\{\lambda_n\} \in l^2$ , the series  $\sum \lambda_n a_n$  converges in  $X$  and its sum equals  $V^*\{\lambda_n\}$ . The operator  $S := V^*V \in B(X)$  is a positive operator with norm  $\leq B$ .
- (b) If  $\{a_n\}$  is two-sided Bessel,  $S - AI$  is positive, and therefore  $\sigma(S) \subset [A, B]$ . In particular,  $S$  is *onto*. Conclude that every  $x \in X$  can be represented as  $x = \sum (x, S^{-1}a_n)a_n$  (convergent in  $X$ ).

# Integral representation

## 9.1 Spectral measure on a Banach subspace

Let  $X$  be a Banach space. A *Banach subspace*  $Z$  of  $X$  is a subspace of  $X$  in the algebraic sense, which is a Banach space for a norm  $\|\cdot\|_Z$  *larger than or equal to the given norm*  $\|\cdot\|$  of  $X$ . Clearly, if  $Z$  and  $X$  are Banach subspaces of each other, then they coincide as Banach spaces (with equality of norms).

Let  $K$  be a compact subset of the complex plane  $\mathbb{C}$ , and let  $C(K)$  be the Banach algebra of all complex continuous functions on  $K$  with the supremum norm  $\|f\| := \sup_K |f|$ .

Let  $T \in B(X)$ , and suppose  $Z$  is a  $T$ -invariant Banach subspace of  $X$ , such that  $T|_Z$ , the restriction of  $T$  to  $Z$ , belongs to  $B(Z)$ .

A *(contractive)  $C(K)$ -operational calculus for  $T$  on  $Z$*  is a *(contractive)  $C(K)$ -operational calculus for  $T|_Z$*  in  $B(Z)$ , that is, a (norm-decreasing) continuous algebra-homomorphism  $\tau : C(K) \rightarrow B(Z)$  such that  $\tau(f_0) = I|_Z$  and  $\tau(f_1) = T|_Z$ , where  $f_k(\lambda) = \lambda^k, k = 0, 1$ . When such  $\tau$  exists, we say that  $T$  is of *(contractive) class  $C(K)$  on  $Z$*  (or that  $T|_Z$  is of (contractive) class  $C(K)$ ).

If the complex number  $\beta$  is not in  $K$ , the function  $g_\beta(\lambda) = (\beta - \lambda)^{-1}$  belongs to  $C(K)$  and  $(\beta - \lambda)g_\beta(\lambda) = 1$  on  $K$ . Since  $\tau$  is an algebra homomorphism of  $C(K)$  into  $B(Z)$ , it follows that  $(\beta I - T)|_Z$  has the inverse  $\tau(g_\beta)$  in  $B(Z)$ . In particular,  $\sigma_{B(Z)}(T|_Z) \subset K$ .

Let  $f$  be any rational function with poles off  $K$ . Write

$$f(\lambda) = \alpha \prod_{k,j} (\alpha_k - \lambda)(\beta_j - \lambda)^{-1},$$

where  $\alpha \in \mathbb{C}$ ,  $\alpha_k$  are the zeroes of  $f$  and  $\beta_j$  are its poles (reduced decomposition). Since  $\tau$  is an algebra homomorphism,  $\tau(f)$  is uniquely determined as

$$\tau(f) = \alpha \prod_{k,j} (\alpha_k I - T)|_Z (\beta_j I - T)|_Z^{-1}. \quad (0)$$

If  $K$  has planar Lebesgue measure zero, the rational functions with poles off  $K$  are dense in  $C(K)$ , by the Hartogs–Rosenthal theorem (cf. Exercise 2). The continuity of  $\tau$  implies then that the  $C(K)$ -operational calculus for  $T$  on  $Z$  is *unique* (when it exists). The operator  $\tau(f) \in B(Z)$  is usually denoted by  $f(T|_Z)$ , for  $f \in C(K)$ .

A *spectral measure on  $Z$*  is a map  $E$  of the Borel algebra  $\mathcal{B}(\mathbb{C})$  of  $\mathbb{C}$  into  $B(Z)$ , such that:

- (1) For each  $x \in Z$  and  $x^* \in X^*$ ,  $x^*E(\cdot)x$  is a regular complex Borel measure;
- (2)  $E(\mathbb{C}) = I|_Z$ , and  $E(\delta \cap \epsilon) = E(\delta)E(\epsilon)$  for all  $\delta, \epsilon \in \mathcal{B}(\mathbb{C})$ .

The spectral measure  $E$  is *contractive* if  $\|E(\delta)\|_{B(Z)} \leq 1$  for all  $\delta \in \mathcal{B}(\mathbb{C})$ .

By Property (2),  $E(\delta)$  is a projection in  $Z$ ; therefore a contractive spectral measure satisfies  $\|E(\delta)\|_{B(Z)} = 1$  whenever  $E(\delta) \neq 0$ . The closed subspaces  $E(\delta)Z$  of  $Z$  satisfy the relation

$$E(\delta)Z \cap E(\epsilon)Z = E(\delta \cap \epsilon)Z$$

for all  $\delta, \epsilon \in \mathcal{B}(\mathbb{C})$ . In particular,  $E(\delta)Z \cap E(\epsilon)Z = \{0\}$  if  $\delta \cap \epsilon = \emptyset$ . Therefore, for any partition  $\{\delta_k; k \in \mathbb{N}\}$  of  $\mathbb{C}$ ,  $Z$  has the direct sum decomposition  $Z = \sum_k \oplus E(\delta_k)Z$ . (cf. Property (2) of  $E$ ). This is equivalent to the decomposition  $I|_Z = \sum_k E(\delta_k)$ , with projections  $E(\delta_k) \in B(Z)$  such that  $E(\delta_k)E(\delta_j) = 0$  for  $k \neq j$ . For that reason,  $E$  is also called a *resolution of the identity on  $Z$* .

Property (1) of  $E$  together with Theorem 1.43 and Corollary 6.7 imply that  $E(\cdot)x$  is a  $Z$ -valued additive set function and

$$M_x := \sup_{\delta \in \mathcal{B}(\mathbb{C})} \|E(\delta)x\| < \infty \quad (1)$$

for each  $x \in Z$ .

## 9.2 Integration

If  $\mu$  is a *real* Borel measure on  $\mathbb{C}$  (to fix the ideas), and  $\{\delta_k\}$  is a partition of  $\mathbb{C}$ , let  $J = \{k; \mu(\delta_k) > 0\}$ . Then

$$\begin{aligned} \sum_k |\mu(\delta_k)| &= \sum_{k \in J} \mu(\delta_k) - \sum_{k \in \mathbb{N} - J} \mu(\delta_k) \\ &= \mu\left(\bigcup_{k \in J} \delta_k\right) - \mu\left(\bigcup_{k \in \mathbb{N} - J} \delta_k\right) \leq 2M, \end{aligned}$$

where  $M := \sup_{\delta \in \mathcal{B}(\mathbb{C})} |\mu(\delta)| (\leq \|\mu\|)$ , the total variation norm of  $\mu$ .

If  $\mu$  is a complex Borel measure, apply the above inequality to the real Borel measures  $\Re\mu$  and  $\Im\mu$  to conclude that  $\sum_k |\mu(\delta_k)| \leq 4M$  for all partitions, hence  $\|\mu\| \leq 4M$ .

By (1) of Section 9.1, for each  $x^* \in X^*$  and  $x \in Z$ ,

$$\sup_{\delta \in \mathcal{B}(\mathbb{C})} |x^*E(\delta)x| \leq M_x \|x^*\|.$$

Therefore

$$\|x^*E(\cdot)x\| \leq 4M_x\|x^*\|. \quad (1)$$

The Banach algebra of all bounded complex Borel functions on  $\mathbb{C}$  or  $\mathbb{R}$  is denoted by  $\mathbb{B}(\mathbb{C})$  or  $\mathbb{B}(\mathbb{R})$ , respectively (briefly,  $\mathbb{B}$ );  $\mathbb{B}_0(\mathbb{C})$  and  $\mathbb{B}_0(\mathbb{R})$  (briefly  $\mathbb{B}_0$ ) are the respective dense subalgebras of simple Borel functions. The norm on  $\mathbb{B}$  is the supremum norm (denoted  $\|\cdot\|$ ).

Integration with respect to the vector measure  $E(\cdot)x$  is defined on simple Borel functions as in Definition 1.12. It follows from (1) that for  $f \in \mathbb{B}_0$ ,  $x \in Z$ , and  $x^* \in X^*$ ,

$$\left| x^* \int_{\mathbb{C}} f dE(\cdot)x \right| = \left| \int_{\mathbb{C}} f dx^*E(\cdot)x \right| \leq \|f\| \|x^*E(\cdot)x\| \leq 4M_x \|f\| \|x^*\|.$$

Therefore

$$\left\| \int_{\mathbb{C}} f dE(\cdot)x \right\| \leq 4M_x \|f\|,$$

and we conclude that the map  $f \rightarrow \int_{\mathbb{C}} f dE(\cdot)x$  is a continuous linear map from  $\mathbb{B}_0$  to  $X$  with norm  $\leq 4M_x$ . It extends uniquely (by density of  $\mathbb{B}_0$  in  $\mathbb{B}$ ) to a continuous linear map (same notation!) of  $\mathbb{B}$  into  $X$ , with the same norm ( $\leq 4M_x$ ).

It follows clearly from the above definition that the vector  $\int_{\mathbb{C}} f dE(\cdot)x$  (belonging to  $X$ , and not necessarily in  $Z$ !) satisfies the relation

$$x^* \int_{\mathbb{C}} f dE(\cdot)x = \int_{\mathbb{C}} f dx^*E(\cdot)x \quad (2)$$

for all  $f \in \mathbb{B}(\mathbb{C})$ ,  $x \in Z$ , and  $x^* \in X^*$ .

As for scalar measures, the *support* of  $E$ ,  $\text{supp } E$ , is the complement in  $\mathbb{C}$  of the union of all open set  $\delta$  such that  $E(\delta) = 0$ . The support of each complex measure  $x^*E(\cdot)x$  is then contained in the support of  $E$ . One has

$$\int_{\mathbb{C}} f dE(\cdot)x = \int_{\text{supp } E} f dE(\cdot)x$$

for all  $f \in \mathbb{B}$  and  $x \in Z$  (where the right-hand side is defined as usual as the integral over  $\mathbb{C}$  of  $f\chi_{\text{supp } E}$ , and  $\chi_V$  denotes in *this chapter* the indicator of  $V \subset \mathbb{C}$ ) (cf. (2) of Section 3.5). The right-hand side of the last equation can be used to extend the definition of the integral to complex Borel function *that are only bounded on*  $\text{supp } E$ .

In case  $Z = X$ , it follows from (1) of Section 9.1 and the uniform boundedness theorem that  $M := \sup_{\delta \in \mathcal{B}(\mathbb{C})} \|E(\delta)\| < \infty$ ,  $M_x \leq M\|x\|$ , and (1) takes the form  $\|x^*E(\cdot)x\| \leq 4M\|x\|\|x^*\|$  for all  $x \in X$  and  $x^* \in X^*$ . It then follows that the *spectral integral*  $\int_{\mathbb{C}} f dE$ , defined by

$$\left( \int_{\mathbb{C}} f dE \right) x := \int_{\mathbb{C}} f dE(\cdot)x \quad (f \in \mathbb{B}) \quad (3)$$

belongs to  $B(X)$  and has operator norm  $\leq 4M\|f\|$ .

If  $X$  is a *Hilbert space* and  $Z = X$ , one may use the Riesz representation for  $X^*$  to express Property (1) of  $E$  and Relation (2) in the following form:

*For each  $x, y \in X$ ,  $(E(\cdot)x, y)$  is a regular complex measure.*

*For each  $x, y \in X$  and  $f \in \mathbb{B}(\mathbb{C})$ ,*

$$\left( \int_{\mathbb{C}} f dE(\cdot)x, y \right) = \int_{\mathbb{C}} f d(E(\cdot)x, y). \quad (4)$$

In the Hilbert space context, it is particularly significant to consider *selfadjoint* spectral measures on  $X$ ,  $E(\cdot)$ , that is,  $E(\delta)^* = E(\delta)$  for all  $\delta \in \mathcal{B}(\mathbb{C})$ . The operators  $E(\delta)$  are then *orthogonal projections*, and any partition  $\{\delta_k; k \in \mathbb{N}\}$  gives an *orthogonal decomposition*  $X = \sum_k \oplus E(\delta_k)X$  into mutually orthogonal closed subspaces (by Property (2) of  $E(\cdot)$ ). Equivalently, the identity operator is the sum of the *mutually orthogonal projections*  $E(\delta_k)$  (the adjective ‘orthogonal’ is transferred from the subspace to the corresponding projection). For this reason,  $E$  is also called a (selfadjoint) *resolution of the identity*.

### 9.3 Case $Z = X$

The relationship between  $C(K)$ -operational calculi and spectral measures is especially simple in case  $Z = X$ .

#### Theorem 9.1.

- (1) *Let  $E$  be a spectral measure on the Banach space  $X$ , supported by the compact subset  $K$  of  $\mathbb{C}$ , and let  $\tau(f)$  be the associated spectral integral, for  $f \in \mathbb{B} := \mathbb{B}(\mathbb{C})$ . Then  $\tau : \mathbb{B} \rightarrow B(X)$  is a continuous representation of  $\mathbb{B}$  on  $X$  with norm  $\leq 4M$ . The restriction of  $\tau$  to (Borel) functions continuous on  $K$  defines a  $C(K)$ -operational calculus for  $T := \tau(f_1) = \int_{\mathbb{C}} \lambda dE(\lambda)$ .*

*If  $X$  is a Hilbert space and the spectral measure  $E$  is selfadjoint, then  $\tau$  is a norm-decreasing  $*$ -representation of  $\mathbb{B}$ , and  $\tau|_{C(K)}$  is a contractive  $C(K)$ -operational calculus on  $X$  for  $T$  (sending adjoints to adjoints).*

- (2) *Conversely, Let  $\tau$  be a  $C(K)$ -operational calculus for a given operator  $T$  on the reflexive Banach space  $X$ . Then there exists a unique spectral measure  $E$  commuting with  $T$  with support in  $K$ , such that  $\tau(f) = \int_{\mathbb{C}} f dE$  for all  $f \in C(K)$  (and the spectral integral on the right-hand side extends  $\tau$  to a continuous representation of  $\mathbb{B}$  on  $X$ ).*

*If  $X$  is a Hilbert space and  $\tau$  sends adjoints to adjoints, then  $E$  is a contractive selfadjoint spectral measure on  $X$ .*

#### Proof.

(1) A calculation shows that Property (2) of  $E$  implies the multiplicativity of  $\tau$  on  $\mathbb{B}_0$ . Since  $\tau : \mathbb{B} \rightarrow B(X)$  was shown to be linear and continuous (with norm at most  $4M$ ), it follows from the density of  $\mathbb{B}_0$  in  $\mathbb{B}$  that  $\tau$  is a continuous algebra homomorphism of  $\mathbb{B}$  into  $B(X)$  and  $\tau(1) = E(\mathbb{C}) = I$ .

If  $X$  is a Hilbert space and  $E$  is selfadjoint, then  $\tau$  restricted to  $\mathbb{B}_0$  sends adjoints to adjoints, because if  $f = \sum \alpha_k \chi_{\delta_k}$ , then  $\tau(\bar{f}) = \sum \bar{\alpha}_k E(\delta_k) =$

$[\sum \alpha_k E(\delta_k)]^* = \tau(f)^*$ . By continuity of  $\tau$  and of the involutions, and by density of  $\mathbb{B}_0$  in  $\mathbb{B}$ ,  $\tau(\bar{f}) = \tau(f)^*$  for all  $f \in \mathbb{B}$ .

Let  $\{\delta_k\}$  be a partition of  $\mathbb{C}$ . For any  $x \in X$ , the sequence  $\{E(\delta_k)x\}$  is orthogonal with sum equal to  $x$ . Therefore

$$\sum_k \|E(\delta_k)x\|^2 = \|x\|^2. \quad (1)$$

By Schwarz's inequality for  $X$  and for  $l^2$ , we have for all  $x, y \in X$  (since  $E(\delta_k)$  are selfadjoint projections):

$$\begin{aligned} \sum_k |(E(\delta_k)x, y)| &= \sum_k |(E(\delta_k)x, E(\delta_k)y)| \leq \sum_k \|E(\delta_k)x\| \|E(\delta_k)y\| \\ &\leq \left( \sum_k \|E(\delta_k)x\|^2 \right)^{1/2} \left( \sum_k \|E(\delta_k)y\|^2 \right)^{1/2} = \|x\| \|y\|. \end{aligned}$$

Hence

$$\|(E(\cdot)x, y)\| \leq \|x\| \|y\| \quad (x, y \in X). \quad (2)$$

Therefore

$$|(\tau(f)x, y)| = \left| \int_{\mathbb{C}} f d(E(\cdot)x, y) \right| \leq \|f\| \|x\| \|y\|,$$

that is,  $\|\tau(f)\| \leq \|f\|$  for all  $f \in \mathbb{B}$ .

(2) Let  $\tau$  be a  $C(K)$ -operational calculus on  $X$  for  $T \in B(X)$ . For each  $x \in X$  and  $x^* \in X^*$ ,  $x^*\tau(\cdot)x$  is a continuous linear functional on  $C(K)$  with norm  $\leq \|\tau\| \|x\| \|y\|$  (where  $\|\tau\|$  denotes the norm of the bounded linear map  $\tau : C(K) \rightarrow B(X)$ ). By the Riesz representation theorem, there exists a unique regular complex Borel measure  $\mu = \mu(\cdot; x, x^*)$  on  $\mathcal{B}(\mathbb{C})$ , with support in  $K$ , such that

$$x^*\tau(f)x = \int_K f d\mu(\cdot; x, x^*) \quad (3)$$

and

$$\|\mu(\cdot; x, x^*)\| \leq \|\tau\| \|x\| \|x^*\| \quad (4)$$

for all  $f \in C(K)$ ,  $x \in X$ , and  $x^* \in X^*$ .

For each fixed  $\delta \in \mathcal{B}(\mathbb{C})$  and  $x \in X$ , it follows from the uniqueness of the Riesz representation, the linearity of the left-hand side of (3) with respect to  $x^*$ , and (4), that the map  $\mu(\delta; x, \cdot)$  is a continuous linear functional on  $X^*$ , with norm  $\leq \|\tau\| \|x\|$ . Since  $X$  is assumed reflexive, there exists a unique vector in  $X$ , which we denote  $E(\delta)x$ , such that

$$\mu(\delta; x, x^*) = x^*E(\delta)x.$$

A routine calculation using the uniqueness properties mentioned above and the linearity of the left-hand side of (3) with respect to  $x$ , shows that  $E(\delta) : X \rightarrow X$  is linear. By (4),  $E(\delta)$  is bounded with operator norm  $\leq \|\tau\|$ . By definition,  $E$

satisfies Property (1) of spectral measures on  $X$ , and has support in  $K$ . Therefore, the integral  $\int_{\mathbb{C}} f dE$  makes sense (see above construction) for any Borel function  $f : \mathbb{C} \rightarrow \mathbb{C}$  bounded on  $K$ , and (by (2) of Section 9.2 and (3))

$$\tau(f) = \int_{\mathbb{C}} f dE \quad (f \in C(K)). \quad (5)$$

For all  $f, g \in C(K)$  and  $x \in X$ ,

$$\begin{aligned} \int_K f dE(\cdot) \tau(g)x &= \tau(f) \tau(g)x = \tau(g) \tau(f)x = \int_K f d\tau(g)E(\cdot)x \\ &= \tau(fg)x = \int_K fg dE(\cdot)x. \end{aligned}$$

The uniqueness of the Riesz representation implies that  $E(\cdot)\tau(g) = \tau(g)E(\cdot)$  (in particular,  $E$  commutes with  $\tau(f_1) = T$ ) and  $dE(\cdot)\tau(g)x = g dE(\cdot)x$ . Therefore, for all  $\delta \in \mathcal{B}(\mathbb{C})$  and  $g \in C(K)$ ,

$$\begin{aligned} \int_K g dE(\cdot)E(\delta)x &= \tau(g)E(\delta)x = E(\delta)\tau(g)x \\ &= \int_K \chi_{\delta} dE(\cdot)\tau(g)x = \int_K \chi_{\delta} g dE(\cdot)x. \end{aligned} \quad (6)$$

By uniqueness of the Riesz representation, we get  $dE(\cdot)E(\delta)x = \chi_{\delta} dE(\cdot)x$ . Thus, for all  $\epsilon, \delta \in \mathcal{B}(\mathbb{C})$  and  $x \in X$ ,

$$\begin{aligned} E(\epsilon)E(\delta)x &= \int_K \chi_{\epsilon} dE(\cdot)E(\delta)x = \int_K \chi_{\epsilon}\chi_{\delta} dE(\cdot)x \\ &= \int_K \chi_{\epsilon \cap \delta} dE(\cdot)x = E(\epsilon \cap \delta)x. \end{aligned}$$

Taking  $f = f_0 (= 1)$  in (5), we get  $E(\mathbb{C}) = \tau(f_0) = I$ , so that  $E$  satisfies Property (2) of spectral measures. Relation (5) provides the wanted relation between the operational calculus and the spectral measure  $E$ . The uniqueness of  $E$  follows from Property (1) of spectral measures and the uniqueness of the Riesz representation. Finally, if  $X$  is a Hilbert space and  $\tau$  sends adjoints to adjoints, then by (4) of Section 9.2

$$\begin{aligned} \int_K f d(E(\cdot)x, y) &= (\tau(f)x, y) = (x, \tau(f)^*y) \\ &= (x, \tau(\bar{f})y) = \overline{(\tau(\bar{f})y, x)} = \overline{\int_K \bar{f} d(E(\cdot)y, x)} \\ &= \int_K f d(\overline{E(\cdot)y}, x) = \int_K f d(x, E(\cdot)y) \end{aligned}$$

for all  $f \in C(K)$  and  $x, y \in X$ . Therefore (by uniqueness of the Riesz representation)  $(E(\delta)x, y) = (x, E(\delta)y)$  for all  $\delta$ , that is,  $E$  is a *selfadjoint* spectral measure. By (2), it is necessarily contractive.  $\square$



**Terminology 9.2.** Given a spectral measure  $E$  on  $X$  with compact support, the bounded operator  $T := \int_{\mathbb{C}} \lambda dE(\lambda)$  is the associated *scalar operator*. By Theorem 9.1,  $T$  is of class  $C(K)$  on  $X$  for any compact set  $K$  containing the support of  $E$ , and  $\tau : f \rightarrow \int_{\mathbb{C}} f dE$  ( $f \in C(K)$ ) is a  $C(K)$ -operational calculus for  $T$ . Conversely, if  $X$  is reflexive, then any operator of class  $C(K)$ , for a given compact set  $K$ , is a scalar operator (the scalar operator associated with the spectral measure  $E$  in Theorem 9.1, Part 2).

If  $E$  and  $F$  are spectral measures on  $X$  with support in the compact set  $K$ , and their associated scalar operators coincide, it follows from Theorem 9.1 (Part 1) that their associated spectral integrals coincide for all rational functions with poles off  $K$ . In case  $K$  has planar Lebesgue measure zero, these rational functions are dense in  $C(K)$  (by the Hartogs–Rosenthal theorem), and the continuity of the spectral integrals on  $C(K)$  (cf. Theorem 9.1) implies that they coincide on  $C(K)$ . It then follows that  $E = F$ , by the uniqueness of the Riesz representation. The uniqueness of the spectral measure associated with a scalar operator can be proved without the ‘planar measure zero’ condition on  $K$ , but this will not be done here. The unique spectral measure with compact support associated with the scalar operator  $T$  is called the *resolution of the identity for  $T$* .

For each  $\delta \in \mathcal{B}(\mathbb{C})$ , the projection  $E(\delta)$  commutes with  $T$  (cf. Theorem 9.1, Part 2), and therefore the closed subspace  $E(\delta)X$  reduces  $T$ . If  $\mu$  is a complex number not in the closure  $\bar{\delta}$  of  $\delta$ , the function  $h(\lambda) := \chi_{\delta}(\lambda)/(\mu - \lambda)$  belongs to  $\mathbb{B}$  ( $\|h\| \leq 1/\text{dist}(\mu, \delta)$ ) and  $(\mu - \lambda)h(\lambda) = \chi_{\delta}(\lambda)$ . Applying the  $\mathbb{B}$ -operational calculus, we get  $(\mu I - T)\tau(h) = E(\delta)$ . Restricting to  $E(\delta)X$ , this means that  $\mu \in \rho(T|_{E(\delta)X})$ . Hence

$$\sigma(T|_{E(\delta)X}) \subset \bar{\delta} \quad (\delta \in \mathcal{B}(\mathbb{C})). \quad (7)$$

**Remark 9.3.** A bounded operator  $T$  for which there exists a spectral measure  $E$  on  $X$ , commuting with  $T$  and with support in  $\sigma(T)$ , such that (7) is satisfied, is called a *spectral operator*. It turns out that  $E$  is uniquely determined; it is called the resolution of the identity for  $T$ , as before. If  $S$  is the scalar operator associated with  $E$ , it can be proved that  $T$  has the (unique) *Jordan decomposition*  $T = S + N$  with  $N$  *quasi-nilpotent commuting with  $S$* . Conversely, any operator with such a Jordan decomposition is spectral. The Jordan canonical form for complex matrices establishes the fact that *every* linear operator on  $\mathbb{C}^n$  is spectral (for any finite  $n$ ).

Combining 7.19 and Theorem 9.1, we obtain

## 9.4 The spectral theorem for normal operators

**Theorem 9.4.** *Let  $T$  be a normal operator on the Hilbert space  $X$ , and let  $\tau : f \rightarrow f(T)$  be its  $C(\sigma(T))$ -operational calculus (cf. Terminology 7.19). Then there exists a unique selfadjoint spectral measure  $E$  on  $X$ , commuting with  $T$  and with*

support in  $\sigma(T)$ , such that

$$f(T) = \int_{\sigma(T)} f dE$$

for all  $f \in C(\sigma(T))$ . The spectral integral above extends the  $C(\sigma(T))$ -operational calculus to a norm-decreasing  $*$ -representation  $\tau: f \rightarrow f(T)$  of  $\mathbb{B}(\mathbb{C})$  on  $X$ .

**Remark 9.5.** We can use Terminology 9.2 to restate Theorem 9.4 in the form: normal operators are scalar, and their resolutions of the identity are selfadjoint, with support in the spectrum. The converse is also true, since the map  $\tau$  associated with a selfadjoint spectral measure with compact support is a  $*$ -representation. If we consider spectral measures that are not necessarily selfadjoint, it can be shown that a bounded operator  $T$  in Hilbert space is scalar *if and only if it is similar to a normal operator*, that is, iff there exists a non-singular  $Q \in B(X)$  such that  $QTQ^{-1}$  is normal.

**Theorem 9.6.** Let  $T$  be a normal operator on the Hilbert space  $X$ , and let  $E$  be its resolution of the identity. Then

- (1) For each  $x \in X$ , the measure  $(E(\cdot)x, x) = \|E(\cdot)x\|^2$  is a positive regular Borel measure bounded by  $\|x\|^2$ , and

$$\|f(T)x\|^2 = \int_{\sigma(T)} |f|^2 d(E(\cdot)x, x)$$

for all Borel functions  $f$  bounded on  $\sigma(T)$ .

- (2) If  $\{\delta_k\}$  is a sequence of mutually disjoint Borel subsets of  $\mathbb{C}$  with union  $\delta$ , then for all  $x \in X$ ,

$$E(\delta)x = \sum_k E(\delta_k)x,$$

where the series converges strongly in  $X$  (this is ‘strong  $\sigma$ -additivity’ of  $E$ ).

- (3) If  $\{f_n\}$  is a sequence of Borel functions, uniformly bounded on  $\sigma(T)$ , such that  $f_n \rightarrow f$  pointwise on  $\sigma(T)$ , then  $f_n(T) \rightarrow f(T)$  in the s.o.t.

- (4)  $\text{supp } E = \sigma(T)$ .

**Proof.** (1) Since  $E(\delta)$  is a self-adjoint projection for each  $\delta \in \mathcal{B}(\mathbb{C})$ , we have

$$(E(\delta)x, x) = (E(\delta)^2x, x) = (E(\delta)x, E(\delta)x) = \|E(\delta)x\|^2 \leq \|x\|^2,$$

and the stated properties of the measure follow from Property (1) of spectral measures in Hilbert space.

If  $f$  is a Borel function bounded on  $\sigma(T)$ , we have for all  $x \in X$

$$\begin{aligned} \|f(T)x\|^2 &= (f(T)x, f(T)x) = (f(T)^* f(T)x, x) = ((\bar{f}f)(T)x, x) \\ &= \int_{\sigma(T)} |f|^2 d\|E(\cdot)x\|^2. \end{aligned}$$

(2) The sequence  $\{E(\delta_k)x\}$  is orthogonal. Therefore, for each  $n \in \mathbb{N}$ ,

$$\sum_{k=1}^n \|E(\delta_k)x\|^2 = \left\| \sum_{k=1}^n E(\delta_k)x \right\|^2 = \|E(\delta)x\|^2 \leq \|x\|^2.$$

This shows that the series  $\sum_k \|E(\delta_k)x\|^2$  converges. Therefore,

$$\left\| \sum_{k=m}^n E(\delta_k)x \right\|^2 = \sum_{k=m}^n \|E(\delta_k)x\|^2 \rightarrow 0$$

when  $m, n \rightarrow \infty$ . By completeness of  $X$ , it follows that  $\sum_k E(\delta_k)x$  converges strongly in  $X$ . Hence for all  $y \in X$ , Property (1) of spectral measures gives

$$(E(\delta)x, y) = \sum_k (E(\delta_k)x, y) = \left( \sum_k E(\delta_k)x, y \right),$$

and Part (2) follows.

(3) For all  $x \in X$ , we have by Part (1)

$$\|f_n(T)x - f(T)x\|^2 = \|(f_n - f)(T)x\|^2 = \int_{\sigma(T)} |f_n - f|^2 d\|E(\cdot)x\|^2,$$

and Part (3) then follows from Lebesgue's dominated convergence theorem for the *finite* positive measure  $\|E(\cdot)x\|^2$ .

(4) We already know that  $\text{supp } E \subset \sigma(T)$ . So we need to prove that  $(\text{supp } E)^c \subset \rho(T)$ . Let  $\mu \in (\text{supp } E)^c$ . By definition of the support, there exists an open neighbourhood  $\delta$  of  $\mu$  such that  $E(\delta) = 0$ . Then  $r := \text{dist}(\mu, \delta^c) > 0$ . Let  $g(\lambda) := \chi_{\delta^c}(\lambda)/(\mu - \lambda)$ . Then  $g \in \mathbb{B}$  (in fact,  $\|g\| = 1/r < \infty$ ), and since  $(\mu - \lambda)g(\lambda) = \chi_{\delta^c}(\lambda)$ , the operational calculus for  $T$  implies that  $(\mu I - T)g(T) = g(T)(\mu I - T) = E(\mathbb{C}) - E(\delta) = I$ . Hence  $\mu \in \rho(T)$ .  $\square$

Note that the proof of Part (4) of Theorem 9.6 is valid for any scalar operator in Banach space.

## 9.5 Parts of the spectrum

**Definition 9.7.** Let  $X$  be a Banach space, and  $T \in B(X)$ .

- (1) The set of all  $\lambda \in \mathbb{C}$  such that  $\lambda I - T$  is not injective is called the *point spectrum* of  $T$ , and is denoted by  $\sigma_p(T)$ ; any  $\lambda \in \sigma_p(T)$  is called an *eigenvalue* of  $T$ , and the non-zero subspace  $\ker(\lambda I - T)$  is the corresponding *eigenspace*. The non-zero vectors  $x$  in the eigenspace are the *eigenvectors* of  $T$  corresponding to the eigenvalue  $\lambda$  (briefly, the  $\lambda$ -eigenvectors of  $T$ ): they are the non-trivial solutions of the equation  $Tx = \lambda x$ .
- (2) The set of all  $\lambda \in \mathbb{C}$  such that  $\lambda I - T$  is injective, and  $\lambda I - T$  has range dense but not equal to  $X$ , is called the *continuous spectrum* of  $T$ , and is denoted by  $\sigma_c(T)$ .

- (3) The set of all  $\lambda \in \mathbb{C}$  such that  $\lambda I - T$  is injective and  $\lambda I - T$  has range not dense in  $X$  is called the *residual spectrum* of  $T$ , and is denoted by  $\sigma_r(T)$ .

Clearly, the three sets defined above are mutually disjoint subsets of  $\sigma(T)$ . If  $\lambda$  is *not* in their union, then  $\lambda I - T$  is bijective, hence invertible in  $B(X)$  (cf. Corollary 6.11). This shows that

$$\sigma(T) = \sigma_p(T) \cup \sigma_c(T) \cup \sigma_r(T).$$

**Theorem 9.8.** *Let  $X$  be a Hilbert space and  $T \in B(X)$ . Then:*

- (1) *if  $\lambda \in \sigma_r(T)$ , then  $\bar{\lambda} \in \sigma_p(T^*)$ ;*
- (2)  *$\lambda \in \sigma_p(T) \cup \sigma_r(T)$  iff  $\bar{\lambda} \in \sigma_p(T^*) \cup \sigma_r(T^*)$ .*
- (3) *If  $T$  is normal, then  $\lambda \in \sigma_p(T)$  iff  $\bar{\lambda} \in \sigma_p(T^*)$ , and in that case the  $\lambda$ -eigenspace of  $T$  coincides with the  $\bar{\lambda}$ -eigenspace of  $T^*$ .*
- (4) *If  $T$  is normal, then  $\sigma_r(T) = \emptyset$ .*
- (5) *If  $T$  is normal and  $E$  is its resolution of the identity, then*

$$\sigma_p(T) = \{\mu \in \mathbb{C}; E(\{\mu\}) \neq 0\}$$

*and the range of  $E(\{\mu\})$  coincides with the  $\mu$ -eigenspace of  $T$ , for each  $\mu \in \sigma_p(T)$ .*

**Proof.** (1) Let  $\lambda \in \sigma_r(T)$ . Since the range of  $\lambda I - T$  is not dense in  $X$ , the orthogonal decomposition theorem (Theorem 1.36) implies the existence of  $y \neq 0$  orthogonal to this range. Hence for all  $x \in X$

$$(x, [\bar{\lambda}I - T^*]y) = ((\lambda I - T)x, y) = 0,$$

and therefore  $y \in \ker(\bar{\lambda}I - T^*)$  and  $\bar{\lambda} \in \sigma_p(T^*)$ .

(2) Let  $\lambda \in \sigma_p(T)$  and let then  $x$  be an eigenvector of  $T$  corresponding to  $\lambda$ . Then for all  $y \in X$ ,  $(x, (\bar{\lambda}I - T^*)y) = ((\lambda I - T)x, y) = 0$ , which implies that the range of  $\bar{\lambda}I - T^*$  is not dense in  $X$  (because  $x \neq 0$  is orthogonal to it). Therefore  $\bar{\lambda}$  belongs to  $\sigma_r(T^*)$  or to  $\sigma_p(T^*)$  if  $\bar{\lambda}I - T^*$  is injective or not, respectively. Together with Part (1), this shows that if  $\lambda \in \sigma_p(T) \cup \sigma_r(T)$ , then  $\bar{\lambda} \in \sigma_p(T^*) \cup \sigma_r(T^*)$ . Applying this to  $\bar{\lambda}$  and  $T^*$ , we get the reverse implication (because  $T^{**} = T$ ).

- (3) For any normal operator  $S$ ,  $\|S^*x\| = \|Sx\|$  (for all  $x$ ) because

$$\|S^*x\|^2 = (S^*x, S^*x) = (SS^*x, x) = (S^*Sx, x) = (Sx, Sx) = \|Sx\|^2.$$

If  $T$  is normal, so is  $S := \lambda I - T$ , and therefore  $\|(\bar{\lambda}I - T^*)x\| = \|(\lambda I - T)x\|$  (for all  $x \in X$  and  $\lambda \in \mathbb{C}$ ). This implies Part (3).

(4) If  $\lambda \in \sigma_r(T)$ , then Part (1) implies that  $\bar{\lambda} \in \sigma_p(T^*)$ , and therefore, by Part (3),  $\lambda \in \sigma_p(T)$ , contradiction. Hence  $\sigma_r(T) = \emptyset$ .

(5) If  $E(\{\mu\}) \neq 0$ , let  $x \neq 0$  be in the range of this projection. Then  $E(\cdot)x = E(\cdot)E(\{\mu\})x = E(\cdot \cap \{\mu\})x$  is the point mass measure at  $\mu$  (with total mass 1)

multiplied by  $x$ . Hence

$$Tx = \int_{\sigma(T)} \lambda dE(\lambda)x = \mu x,$$

so that  $\mu \in \sigma_p(T)$  and each  $x \neq 0$  in  $E(\{\mu\})X$  is a  $\mu$ -eigenvector for  $T$ .

On the other hand, if  $\mu \in \sigma_p(T)$  and  $x$  is a  $\mu$ -eigenvector for  $T$ , let  $\delta_n = \{\lambda \in \mathbb{C}; |\lambda - \mu| > 1/n\}$  ( $n = 1, 2, \dots$ ) and  $f_n(\lambda) := \chi_{\delta_n}(\lambda)/(\mu - \lambda)$ . Then  $f_n \in \mathbb{B}$  (it is clearly Borel, and bounded by  $n$ ), and therefore  $f_n(T) \in B(X)$ . Since  $(\mu - \lambda)f_n(\lambda) = \chi_{\delta_n}(\lambda)$ , we have

$$E(\delta_n)x = f_n(T)(\mu I - T)x = f_n(T)0 = 0.$$

Hence, by  $\sigma$ -subadditivity of the positive measure  $\|E(\cdot)x\|^2$ ,

$$\|E(\{\lambda; \lambda \neq \mu\})x\|^2 = \|E\left(\bigcup_n \delta_n\right)x\|^2 \leq \sum_n \|E(\delta_n)x\|^2 = 0.$$

Therefore  $E(\{\lambda; \lambda \neq \mu\})x = 0$  and so  $E(\{\mu\})x = Ix = x$ , that is, the non-zero vector  $x$  belongs to the range of the projection  $E(\{\mu\})$ .  $\square$

## 9.6 Spectral representation

**Construction 9.9.** The spectral theorem for normal operators has an interesting interpretation through isomorphisms of Hilbert spaces. Let  $T$  be a (bounded) normal operator on the Hilbert space  $X$ , and let  $E$  be its resolution of the identity. Given  $x \in X$ , define the *cycle of  $x$*  (relative to  $T$ ) as the closed linear span of the vectors  $T^n(T^*)^m x$ ,  $n, m = 0, 1, 2, \dots$ . This is clearly a reducing subspace for  $T$ . It follows from the Stone-Weierstrass theorem that the cycle of  $x$ ,  $[x]$ , coincides with the closure (in  $X$ ) of the subspace

$$[x]_0 := \{g(T)x; g \in C(\sigma(T))\}.$$

Define  $V_0 : [x]_0 \rightarrow C(\sigma(T))$  by

$$V_0 g(T)x = g.$$

By Theorem 9.6, Part (1), the map  $V_0$  is a *well-defined* linear isometry of  $[x]_0$  onto the subspace  $C(\sigma(T))$  of  $L^2(\mu)$ , where  $\mu = \mu_x := \|E(\cdot)x\|^2$  is a regular finite positive Borel measure with support in  $\sigma(T)$ . By Corollary 3.21,  $C(\sigma(T))$  is dense in  $L^2(\mu)$ , and therefore  $V_0$  extends uniquely as a linear isometry  $V$  of the cycle  $[x]$  onto  $L^2(\mu)$ . If  $g$  is a Borel function bounded on  $\sigma(T)$ , we may apply Lusin's theorem (Theorem 3.20) to get a sequence  $g_n \in C(\sigma(T))$ , uniformly bounded on the spectrum by  $\sup_{\sigma(T)} |g|$ , such that  $g_n \rightarrow g$  pointwise  $\mu$ -almost everywhere on  $\sigma(T)$ . It follows from the proof of Part (3) in Theorem 9.6 that  $g(T)x = \lim_n g_n(T)x$ , hence  $g(T)x \in [x]$ . Also  $g_n \rightarrow g$  in  $L^2(\mu)$  by dominated convergence, and therefore

$$Vg(T)x = V \lim_n g_n(T)x = \lim_n V_0 g_n(T)x = \lim_n g_n = g,$$

where the last two limits are in the space  $L^2(\mu)$  and equalities are between  $L^2(\mu)$ -elements.

For each  $y = g(T)x \in [x]_0$  and each Borel function  $f$  bounded on  $\sigma(T)$ , the function  $fg$  (restricted to  $\sigma(T)$ ) is a bounded Borel function on the spectrum, hence  $(fg)(T)x$  makes sense and equals  $f(T)g(T)x = f(T)y$ . Applying  $V$ , we get

$$Vf(T)y = V(fg)(T)x = fg = fVg(T)x = fVy$$

(in  $L^2(\mu)$ ). Let  $M_f$  denote the *multiplication by  $f$*  operator in  $L^2(\mu)$ , defined by  $M_f g = fg$ . Since  $Vf(T) = M_f V$  on the dense subspace  $[x]_0$  of  $[x]$ , and both operators are continuous, we conclude that the last relation is valid on  $[x]$ . We express this by saying that *the isomorphism  $V$  intertwines  $f(T)|_{[x]}$  and  $M_f$*  (for all Borel functions  $f$  bounded on  $\sigma(T)$ ). Equivalently,  $f(T)|_{[x]} = V^{-1}M_f V$ , that is, restricted to the cycle  $[x]$ , the operators  $f(T)$  are unitarily equivalent (through  $V$ !) to the multiplication operators  $M_f$  on  $L^2(\mu)$  with  $\mu := \|E(\cdot)x\|^2$ . This is particularly interesting when  $T$  possess a *cyclic vector*, that is, a vector  $x$  such that  $[x] = X$ . Using such a cyclic vector in the above construction, we conclude that the abstract operator  $f(T)$  is unitarily equivalent to the concrete operator  $M_f$  acting in the concrete Hilbert space  $L^2(\mu)$ , through the Hilbert isomorphism  $V : X \rightarrow L^2(\mu)$ , for each Borel function  $f$  bounded on the spectrum; in particular, taking  $f(\lambda) = f_1(\lambda) := \lambda$ , we have  $T = V^{-1}MV$ , where  $M := M_{f_1}$ .

The construction is generalized to an arbitrary normal operator  $T$  by considering a maximal family of non-zero mutually orthogonal cycles  $\{[x_j], j \in J\}$  for  $T$  ( $J$  denotes some index set). Such a family exists by Zorn's lemma.

Let  $\mu_j := \|E(\cdot)x_j\|^2$  and let  $V_j : [x_j] \rightarrow L^2(\mu_j)$  be the Hilbert isomorphism constructed above, for each  $j \in J$ . If  $\sum_j \oplus [x_j] \neq X$ , pick  $x \neq 0$  orthogonal to the orthogonal sum; then the non-zero cycle  $[x]$  is orthogonal to all  $[x_j]$  (since the cycles are reducing subspaces for  $T$ ), contradicting the maximality of the family above. Hence  $X = \sum_j \oplus [x_j]$ . Consider the operator

$$V := \sum_{j \in J} \oplus V_j$$

operating on  $X = \sum_j \oplus [x_j]$ . Recall that if  $y = \sum_j \oplus y_j$  is any vector in  $X$ , then

$$Vy := \sum_j \oplus V_j y_j.$$

Then  $V$  is a linear isometry of  $X$  onto  $\sum_j \oplus L^2(\mu_j)$ :

$$\|Vy\|^2 = \sum_j \|V_j y_j\|^2 = \sum_j \|y_j\|^2 = \|y\|^2.$$

Since each cycle reduces  $f(T)$  for all Borel function  $f$  bounded on  $\sigma(T)$ , we have

$$Vf(T)y = \sum_j \oplus V_j f(T)y_j = \sum_j \oplus fV_j y_j = M_f Vy,$$

where  $M_f$  is now defined as the orthogonal sum  $M_f := \sum_j \oplus M_f^j$ , with  $M_f^j$  equal to the multiplication by  $f$  operator on the space  $L^2(\mu_j)$ . The Hilbert isomorphism  $V$  is usually referred to as a *spectral representation of  $X$*  (relative to  $T$ ). Formally

**Theorem 9.10.** *Let  $T$  be a bounded normal operator on the Hilbert space  $X$ , and let  $E$  be its resolution of the identity. Then there exists an isomorphism  $V$  of  $X$  onto  $\sum_{j \in J} \oplus L^2(\mu_j)$  with  $\mu_j := \|E(\cdot)x_j\|^2$  for suitable non-zero mutually orthogonal vectors  $x_j$ , such that  $f(T) = V^{-1}M_fV$  for all Borel functions  $f$  bounded on  $\sigma(T)$ . The operator  $M_f$  acts on  $\sum \oplus L^2(\mu_j)$  by  $M_f \sum_j \oplus g_j := \sum \oplus fg_j$ . In case  $T$  has a cyclic vector  $x$ , then there exists an isomorphism  $V$  of  $X$  onto  $L^2(\mu)$  with  $\mu := \|E(\cdot)x\|^2$ , such that  $f(T) = V^{-1}M_fV$  for all  $f$  as above; here  $M_f$  is the ordinary multiplication by  $f$  operator on  $L^2(\mu)$ .*

## 9.7 Renorming method

In the following,  $X$  denotes a given Banach space, and  $B(X)$  is the Banach algebra of all bounded linear operators on  $X$ . The identity operator is denoted by  $I$ . If  $\mathcal{A} \subset B(X)$ , the *commutant*  $\mathcal{A}'$  of  $\mathcal{A}$  consists of all  $S \in B(X)$  that commute with every  $T \in \mathcal{A}$ .

Given an operator  $T$ , we shall construct a *maximal Banach subspace*  $Z$  of  $X$  such that  $T$  has a contractive  $C(K)$ -operational calculus on  $Z$ , where  $K$  is an adequate compact subset of  $\mathbb{C}$  containing the spectrum of  $T$ . In case  $X$  is reflexive, this construction will associate with  $T$  a contractive spectral measure on  $Z$  such that  $f(T|_Z)$  is the corresponding spectral integral for each  $f \in C(K)$ . This ‘maximal’ spectral integral representation is a generalization of the spectral theorem for normal operators in Hilbert space. The construction is based on the following

**Theorem 9.11 (Renorming theorem).** *Let  $\mathcal{A} \subset B(X)$  be such that its strong closure contains  $I$ . Let*

$$\|x\|_{\mathcal{A}} := \sup_{T \in \mathcal{A}} \|Tx\| \quad (x \in X),$$

and

$$Z = Z(\mathcal{A}) := \{x \in X; \|x\|_{\mathcal{A}} < \infty\}.$$

Then:

- (i)  $Z$  with the norm  $\|\cdot\|_{\mathcal{A}}$  is a Banach subspace of  $X$ ;
- (ii) For any  $S \in \mathcal{A}'$ ,  $SZ \subset Z$  and  $S|_Z \in B(Z)$  with  $\|S|_Z\|_{B(Z)} \leq \|S\|$ ;
- (iii) If  $\mathcal{A}$  is a multiplicative semigroup (for operator multiplication), then  $Z$  is  $\mathcal{A}$ -invariant, and  $\mathcal{A}|_Z := \{T|_Z; T \in \mathcal{A}\}$  is contained in the closed unit ball  $B_1(Z)$  of  $B(Z)$ . Moreover,  $Z$  is maximal with this property, that is, if  $W$  is an  $\mathcal{A}$ -invariant Banach subspace of  $X$  such that  $\mathcal{A}|_W \subset B_1(W)$ , then  $W$  is a Banach subspace of  $Z$ .

**Proof.** Subadditivity and homogeneity of  $\|\cdot\|_{\mathcal{A}}$  are clear.

Let  $\epsilon > 0$ . The neighbourhood of  $I$

$$N := N(I, \epsilon, x) := \{T \in B(X); \|(T - I)x\| < \epsilon\}$$

(in the strong operator topology) meets  $\mathcal{A}$  for each  $x \in X$ . Fix  $x$ , and pick then  $T \in \mathcal{A} \cap N$ . Then

$$\|x\|_{\mathcal{A}} \geq \|Tx\| = \|x + (T - I)x\| \geq \|x\| - \epsilon.$$

Therefore  $\|x\|_{\mathcal{A}} \geq \|x\|$  for all  $x \in X$ , and it follows in particular that  $\|\cdot\|_{\mathcal{A}}$  is a norm on  $\mathcal{A}$ .

Let  $\{x_n\} \subset Z$  be  $\|\cdot\|_{\mathcal{A}}$ -Cauchy. In particular, it is  $\|\cdot\|_{\mathcal{A}}$ -bounded; let then  $K = \sup_n \|x_n\|_{\mathcal{A}}$ . Since  $\|\cdot\| \leq \|\cdot\|_{\mathcal{A}}$ , the sequence is  $\|\cdot\|$ -Cauchy; let  $x$  be its  $X$ -limit. Given  $\epsilon > 0$ , let  $n_0 \in \mathbb{N}$  be such that

$$\|x_n - x_m\|_{\mathcal{A}} < \epsilon \quad (n, m > n_0).$$

Then

$$\|Tx_n - Tx_m\| < \epsilon \quad (n, m > n_0; T \in \mathcal{A}).$$

Letting  $m \rightarrow \infty$ , we see that

$$\|Tx_n - Tx\| \leq \epsilon \quad (n > n_0; T \in \mathcal{A}).$$

Hence

$$\|x_n - x\|_{\mathcal{A}} \leq \epsilon \quad (n > n_0). \tag{1}$$

Fixing  $n > n_0$ , we see that

$$\|x\|_{\mathcal{A}} \leq \|x_n\|_{\mathcal{A}} + \|x - x_n\|_{\mathcal{A}} \leq K + \epsilon < \infty,$$

that is,  $x \in Z$ , and  $x_n \rightarrow x$  in  $(Z, \|\cdot\|_{\mathcal{A}})$  by (1). This proves Part (i).

If  $S \in \mathcal{A}'$ , then for all  $x \in Z$  and  $T \in \mathcal{A}$ ,

$$\|T(Sx)\| = \|S(Tx)\| \leq \|S\| \|Tx\| \leq \|S\| \|x\|_{\mathcal{A}}.$$

Therefore  $\|Sx\|_{\mathcal{A}} \leq \|S\| \|x\|_{\mathcal{A}} < \infty$ , that is,  $SZ \subset Z$  and  $S|_Z \in B(Z)$  with  $\|S|_Z\|_{B(Z)} \leq \|S\|$ .

In case  $\mathcal{A}$  is a multiplicative sub-semigroup of  $B(X)$ , we have for all  $x \in Z$  and  $T, U \in \mathcal{A}$

$$\|U(Tx)\| = \|(UT)x\| \leq \sup_{V \in \mathcal{A}} \|Vx\| = \|x\|_{\mathcal{A}}.$$

Hence  $\|Tx\|_{\mathcal{A}} \leq \|x\|_{\mathcal{A}}$ , so that  $Z$  is  $\mathcal{A}$ -invariant and  $\mathcal{A}|_Z \subset B_1(Z)$ .

Finally, if  $W$  is as stated in the theorem, and  $x \in W$ , then for all  $T \in \mathcal{A}$ ,

$$\|Tx\| \leq \|Tx\|_W \leq \|T\|_{B(W)} \|x\|_W \leq \|x\|_W,$$

and therefore  $\|x\|_{\mathcal{A}} \leq \|x\|_W$ . □



## 9.8 Semi-simplicity space

Let  $\Delta \subset \mathbb{R}$  be compact, and let  $\mathcal{P}_1(\Delta)$  denote the set of all complex polynomials  $p$  with

$$\|p\|_\Delta := \sup_{\Delta} |p| \leq 1.$$

Given an arbitrary bounded operator  $T$ , let  $\mathcal{A}$  be the (multiplicative) semigroup

$$\mathcal{A} := \{p(T); p \in \mathcal{P}_1(\Delta)\},$$

and let  $Z = Z(\mathcal{A})$ .

By Theorem 9.11, Part (iii),  $p(T)Z \subset Z$  and

$$\|p(T)|_Z\|_{B(Z)} \leq 1$$

for all  $p \in \mathcal{P}_1(\Delta)$ . This means that the polynomial operational calculus  $\tau : p \rightarrow p(T)|_Z (= p(T|_Z))$  is *norm-decreasing* as a homomorphism of  $\mathcal{P}(\Delta)$ , the subalgebra of  $C(\Delta)$  of all (complex) polynomials restricted to  $\Delta$ , into  $B(Z)$ . Since  $\mathcal{P}(\Delta)$  is dense in  $C(\Delta)$ ,  $\tau$  extends uniquely as a norm decreasing homomorphism of the algebra  $C(\Delta)$  into  $B(Z)$ , that is,  $T$  is of *contractive class*  $C(\Delta)$  on  $Z$ . The Banach subspace  $Z$  in this application of the renorming theorem is called the *semi-simplicity space* for  $T$ .

On the other hand, suppose  $W$  is a  $T$ -invariant Banach subspace of  $X$ , such that  $T$  is of contractive class  $C(\Delta)$  on  $W$ . Then for each  $p \in \mathcal{P}_1(\Delta)$  and  $w \in W$ ,

$$\|p(T)w\| \leq \|p(T)|_W\|_{B(W)} \|w\|_W \leq \|w\|_W,$$

and therefore  $w \in Z$  and  $\|w\|_{\mathcal{A}} \leq \|w\|_W$ . This shows that  $W$  is a Banach subspace of  $Z$ , and concludes the proof of the following

**Theorem 9.12.** *Let  $\Delta \subset \mathbb{R}$  be compact,  $T \in B(X)$ , and let  $Z$  be the semi-simplicity space for  $T$ , that is,  $Z = Z(\mathcal{A})$ , where  $\mathcal{A}$  is the (multiplicative) semigroup*

$$\{p(T); p \in \mathcal{P}_1(\Delta)\}.$$

*Then  $T$  is of contractive class  $C(\Delta)$  on  $Z$ .*

*Moreover,  $Z$  is maximal in the following sense: if  $W$  is a  $T$ -invariant Banach subspace of  $X$  such that  $T$  is of contractive class  $C(\Delta)$  on  $W$ , then  $W$  is a Banach subspace of  $Z$ .*

**Remark 9.13.** (1) If  $X$  is a Hilbert space and  $T$  is a bounded selfadjoint operator on  $X$ , it follows from Remark (2) in Terminology 7.19 that  $\|p(T)\| = \|p\|_{\sigma(T)} \leq 1$  for all  $p \in \mathcal{P}_1(\Delta)$  for any given compact subset  $\Delta$  of  $\mathbb{R}$  containing  $\sigma(T)$ . Therefore  $\|x\|_{\mathcal{A}} \leq \|x\|$  for all  $x \in X$ . Hence  $Z = X$  (with equality of norms) in this case.

Another way to see this is by observing that the selfadjoint operator  $T$  has a contractive  $C(\Delta)$ -operational calculus on  $X$  (see Terminology 7.19, Remark (2)). Therefore,  $X$  is a Banach subspace of  $Z$  by the maximality statement in Theorem 9.12. Hence  $Z = X$ .

(2) If  $T$  is a normal operator (with spectrum in a compact set  $\Delta \subset \mathbb{C}$ ), we may take

$$\mathcal{A} = \{p(T, T^*); p \in \mathcal{P}_1(\Delta)\},$$

where  $\mathcal{P}_1(\Delta)$  is now the set of all complex polynomials  $p$  in  $\lambda$  and  $\bar{\lambda}$  with  $\|p\|_\Delta := \sup_{\lambda \in \Delta} |p(\lambda, \bar{\lambda})| \leq 1$ .

Then  $\mathcal{A}$  is a commutative semigroup (for operator multiplication), and since polynomials in  $\lambda$  and  $\bar{\lambda}$  are dense in  $C(\Delta)$  (by the Stone-Weierstrass Theorem, Theorem 5.39), we conclude as before that  $Z := Z(\mathcal{A})$  coincides with  $X$  (with equality of norms).

(3) In the general Banach space setting, we took  $\Delta \subset \mathbb{R}$  in the construction leading to Theorem 9.12 to ensure the *density* of  $\mathcal{P}(\Delta)$  in  $C(\Delta)$ . If  $\Delta$  is *any compact* subset of  $\mathbb{C}$ , Lavrentiev's theorem (cf. Theorem 8.7 in T.W. Gamelin, *Uniform Algebras*, Prentice-Hall, Englewood Cliffs, NJ, 1969) states that the wanted density occurs if and only if  $\Delta$  is *nowhere dense and has connected complement*. Theorem 9.12 is then valid for such  $\Delta$ , with the same statement and proof.

(4) When the complement of  $\Delta$  is *not* connected, the choice of  $\mathcal{A}$  may be adapted in some cases so that Theorem 9.12 remains valid.

An important case of this kind is the unit circle:

$$\Delta = \Gamma := \{\lambda \in \mathbb{C}; |\lambda| = 1\}.$$

Suppose  $T \in B(X)$  has spectrum in  $\Gamma$ . Consider the algebra  $\mathcal{R}(\Gamma)$  of restrictions to  $\Gamma$  of all complex polynomials in  $\lambda$  and  $\lambda^{-1}$  ( $= \bar{\lambda}$  on  $\Gamma$ ). Following our previous notation, let

$$\mathcal{R}_1(\Gamma) = \left\{ p \in \mathcal{R}(\Gamma); \|p\|_\Gamma := \sup_\Gamma |p| \leq 1 \right\}.$$

Since  $\sigma(T) \subset \Gamma$ ,  $T$  is invertible in  $B(X)$ . Let  $p \in \mathcal{R}(\Gamma)$ . Writing  $p$  as the finite sum

$$p(\lambda) = \sum_k \alpha_k \lambda^k \quad (\alpha_k \in \mathbb{C}, k \in \mathbb{Z}),$$

it makes sense to define

$$p(T) = \sum_k \alpha_k T^k.$$

The map  $\tau : p \rightarrow p(T)$  is the unique algebra homomorphism of  $\mathcal{R}(\Gamma)$  into  $B(X)$  such that  $\tau(p_0) = I$  and  $\tau(p_1) = T$ , where  $p_k(\lambda) = \lambda^k$ . As before, we choose  $\mathcal{A}$  to be the (multiplicative) semigroup

$$\mathcal{A} = \{p(T); p \in \mathcal{R}_1(\Gamma)\},$$

and we call  $Z = Z(\mathcal{A})$  the *semi-simplicity space* for  $T$ .

By Theorem 9.11, Part (iii),  $Z$  is a  $T$ -invariant Banach subspace of  $X$ , and

$$\|p(T)|_Z\|_{B(Z)} \leq 1 \quad (p \in \mathcal{R}_1(\Gamma)). \quad (1)$$

By the Stone–Weierstrass theorem (Theorem 5.39),  $\mathcal{R}(\Gamma)$  is dense in  $C(\Gamma)$ . Consequently, it follows from (1) that

$$\tau_Z : p \rightarrow p(T)|_Z (= p(T|_Z))$$

extends uniquely to a norm-decreasing algebra homomorphism of  $C(\Gamma)$  into  $B(Z)$ , that is,  $T$  is of contractive class  $C(\Gamma)$  on  $Z$ .

The maximality of  $Z$  is proved word for word as in Theorem 9.12.

We restate Theorem 9.12 in the present case for future reference.

**Theorem 9.14.** *Let  $T \in B(X)$  have spectrum on the unit circle  $\Gamma$ . Let  $Z$  be its semi-simplicity space, as defined in Remark 9.13 (4). Then  $Z$  is a  $T$ -invariant Banach subspace of  $X$  such that  $T$  is of contractive class  $C(\Gamma)$  on  $Z$ .*

*Moreover,  $Z$  is maximal in the following sense: if  $W$  is a  $T$ -invariant Banach subspace of  $X$  such that  $T$  is of contractive class  $C(\Gamma)$  on  $W$ , then  $W$  is a Banach subspace of  $Z$ .*

**Remark 9.15.** With a minor change in the choice of  $\mathcal{A}$ , Theorem 9.14 generalizes (with identical statement and proof) to the case when  $\Gamma$  is an arbitrary compact subset of the plane with planar Lebesgue measure zero. In this case, let  $\mathcal{R}(\Gamma)$  denote the algebra of all restrictions to  $\Gamma$  of rational functions with poles off  $\Gamma$ . Each  $f \in \mathcal{R}(\Gamma)$  can be written as a finite product

$$f(\lambda) = \gamma \prod_{j,k} (\lambda - \alpha_j)(\lambda - \beta_k)^{-1},$$

with  $\gamma, \alpha_j \in \mathbb{C}$  and poles  $\beta_k \notin \Gamma$ . Since  $\beta_k \in \rho(T)$ , we may define

$$f(T) := \gamma \prod_{j,k} (T - \alpha_j I)(T - \beta_k I)^{-1}.$$

As before, we choose  $\mathcal{A}$  to be the (multiplicative) semigroup

$$\mathcal{A} = \{f(T); f \in \mathcal{R}_1(\Gamma)\},$$

where  $\mathcal{R}_1(\Gamma)$  is the set of all  $f \in \mathcal{R}(\Gamma)$  with  $\|f\|_\Gamma \leq 1$ .

The corresponding space  $Z = Z(\mathcal{A})$  (called as before the *semi-simplicity space for  $T$* ) is a  $T$ -invariant Banach subspace of  $X$  such that  $\|f(T)|_Z\|_{B(Z)} \leq 1$  for all  $f \in \mathcal{R}_1(T)$ . Since  $\mathcal{R}(\Gamma)$  is dense in  $C(\Gamma)$  by the Hartogs–Rosenthal theorem (cf. Exercise 2), the map  $f \rightarrow f(T)|_Z = f(T|_Z)$  has a unique extension as a norm decreasing algebra homomorphism of  $C(\Gamma)$  into  $B(Z)$ , that is,  $T$  is of contractive class  $C(\Gamma)$  on  $Z$ .

## 9.9 Resolution of the identity on $Z$

**Theorem 9.16.** *Let  $\Gamma$  be any compact subset of  $\mathbb{C}$  for which the semi-simplicity space  $Z$  was defined above, for a given bounded operator  $T$  on the Banach space*

$X$ , with spectrum in  $\Gamma$  (recall that, in the general case,  $\Gamma$  could be any compact subset of  $\mathbb{C}$  with planar Lebesgue measure zero). Then  $T$  is of contractive class  $C(\Gamma)$  on  $Z$ , and  $Z$  is maximal with this property.

Moreover, if  $X$  is reflexive, there exists a unique contractive spectral measure on  $Z$ ,  $E$ , with support in  $\Gamma$  and values in  $T''$ , such that

$$f(T|_Z)x = \int_{\Gamma} f dE(\cdot)x \quad (1)$$

for all  $x \in Z$  and  $f \in C(\Gamma)$ , where  $f \rightarrow f(T|_Z)$  denotes the (unique)  $C(\Gamma)$ -operational calculus for  $T|_Z$ .

The integral (1) extends the  $C(\Gamma)$ -operational calculus for  $T|_Z$  in  $B(Z)$  to a contractive  $\mathbb{B}(\Gamma)$ -operational calculus  $\tau : f \in \mathbb{B}(\Gamma) \rightarrow f(T|_Z) \in B(Z)$ , where  $\mathbb{B}(\Gamma)$  stands for the Banach algebra of all bounded complex Borel functions on  $\Gamma$  with the supremum norm, and  $\tau(f)$  commutes with every  $U \in B(X)$  that commutes with  $T$ .

**Proof.** The first statement of the theorem was verified in the preceding sections.

Suppose then that  $X$  is a reflexive Banach space, and let  $f \rightarrow f(T|_Z)$  denote the (unique)  $C(\Gamma)$ -operational calculus for  $T|_Z$  (in  $B(Z)$ ). For each  $x \in Z$  and  $x^* \in X^*$ , the map

$$f \in C(\Gamma) \rightarrow x^*f(T|_Z)x \in \mathbb{C}$$

is a continuous linear functional on  $C(\Gamma)$  with norm  $\leq \|x\|_Z \|x^*\|$  (where we denote the  $Z$ -norm by  $\|\cdot\|_Z$  rather than  $\|\cdot\|_A$ ). By the Riesz representation theorem, there exists a unique regular complex Borel measure  $\mu = \mu(\cdot; x, x^*)$  on  $\mathcal{B}(\mathbb{C})$ , supported by  $\Gamma$ , such that

$$x^*f(T|_Z)x = \int_{\Gamma} f d\mu(\cdot; x, x^*) \quad (2)$$

and

$$\|\mu(\cdot; x, x^*)\| \leq \|x\|_Z \|x^*\| \quad (3)$$

for all  $f \in C(\Gamma)$ ,  $x \in Z$ , and  $x^* \in X^*$ .

For each  $\delta \in \mathcal{B}(\mathbb{C})$  and  $x \in Z$ , the map  $x^* \in X^* \rightarrow \mu(\delta; x, x^*)$  is a continuous linear functional on  $X^*$  with norm  $\leq \|x\|_Z$  (by (3)). Since  $X$  is reflexive, there exists a unique vector in  $X$ , which we denote  $E(\delta)x$ , such that  $\mu(\delta; x, x^*) = x^*E(\delta)x$ . The map  $E(\delta) : Z \rightarrow X$  is clearly linear and norm-decreasing.

If  $U \in T'$ , then  $U$  commutes with  $f(T)$  for all  $f \in \mathcal{R}(\Gamma)$  (notation as in the preceding section), hence  $U \in \mathcal{A}'$ . Let  $f \in C(\Gamma)$ , and let  $f_n \in \mathcal{R}(\Gamma)$  converge to  $f$  uniformly on  $\Gamma$ . Since  $UZ \subset Z$  and  $\|U\|_{B(Z)} \leq \|U\|_{B(X)}$  (cf. Theorem 9.11, Part (ii)), we have

$$\begin{aligned} \|Uf(T|_Z) - f(T|_Z)U\|_{B(Z)} &\leq \|U[f(T|_Z) - f_n(T)]\|_{B(Z)} \\ &\quad + \|[f_n(T) - f(T|_Z)]U\|_{B(Z)} \leq 2\|U\|_{B(X)}\|f(T|_Z) - f_n(T)\|_{B(Z)} \\ &\leq 2\|U\|_{B(X)}\|f - f_n\|_{C(\Gamma)} \rightarrow 0. \end{aligned}$$

Thus,  $U$  commutes with  $f(T|_Z)$  for all  $f \in C(\Gamma)$ . Therefore, for all  $x \in Z$ ,  $x^* \in X^*$ , and  $f \in C(\Gamma)$ ,

$$\begin{aligned} \int_{\Gamma} f dx^* U E(\cdot) x &= \int_{\Gamma} f d(U^* x^*) E(\cdot) x = (U^* x^*) f(T|_Z) x \\ &= x^* U f(T|_Z) x = x^* f(T|_Z) U x = \int_{\Gamma} f dx^* E(\cdot) U x. \end{aligned}$$

By uniqueness of the Riesz representation, it follows that  $U E(\delta) = E(\delta) U$  for all  $\delta \in \mathcal{B}(\mathbb{C})$  (i.e.  $E$  has values in  $T''$ ). The last relation is true in particular for all  $U \in \mathcal{A}$ . Since  $\mathcal{A} \subset B_1(Z)$  (by Theorem 9.11, Part (iii)) and  $E(\delta) : Z \rightarrow X$  is norm-decreasing, we have for each  $x \in Z$  and  $\delta \in \mathcal{B}(\mathbb{C})$ ,

$$\begin{aligned} \|E(\delta)x\|_Z &:= \sup_{U \in \mathcal{A}} \|U E(\delta)x\| = \sup_{U \in \mathcal{A}} \|E(\delta)Ux\| \\ &\leq \sup_{U \in \mathcal{A}} \|Ux\|_Z \leq \|x\|_Z. \end{aligned}$$

Thus,  $E(\delta) \in B_1(Z)$  for all  $\delta \in \mathcal{B}(\mathbb{C})$ .

Since  $x^*[E(\cdot)x] = \mu(\cdot; x, x^*)$  is a regular countably additive complex measure on  $\mathcal{B}(\mathbb{C})$  for each  $x^* \in X^*$ ,  $E$  satisfies Condition (1) of a spectral measure on  $Z$  and has support in  $\Gamma$ . We may then rewrite (2) in the form

$$f(T|_Z)x = \int_{\Gamma} f dE(\cdot)x \quad f \in C(\Gamma), \quad x \in Z, \quad (4)$$

where the integral is defined as in Section 9.2.

Taking  $f = f_0 (= 1)$  in (4), we see that  $E(\mathbb{C}) = E(\Gamma) = I|_Z$ .

Since  $f \rightarrow f(T|_Z)$  is an algebra homomorphism of  $C(\Gamma)$  into  $B(Z)$ , we have for all  $f, g \in C(\Gamma)$  and  $x \in Z$  (whence  $g(T|_Z)x \in Z$ ):

$$\begin{aligned} \int_{\Gamma} f dE(\cdot)[g(T|_Z)x] &= f(T|_Z)[g(T|_Z)x] \\ &= (fg)(T|_Z)x = \int_{\Gamma} fg dE(\cdot)x. \end{aligned}$$

By uniqueness of the Riesz representation, it follows that

$$dE(\cdot)[g(T|_Z)x] = g dE(\cdot)x.$$

This means that for all  $\delta \in \mathcal{B}(\mathbb{C})$ ,  $g \in C(\Gamma)$ , and  $x \in Z$ ,

$$E(\delta)[g(T|_Z)x] = \int_{\Gamma} \chi_{\delta} g dE(\cdot)x, \quad (5)$$

where  $\chi_{\delta}$  denotes the characteristic function of  $\delta$ .

We observed above that  $E(\delta)$  commutes with every  $U \in B(X)$  that commutes with  $T$ . In particular,  $E(\delta)$  commutes with  $g(T)$  for all  $g \in \mathcal{R}(\Gamma)$ . If  $g \in C(\Gamma)$

and  $g_n \in \mathcal{R}(\Gamma) \rightarrow g$  uniformly on  $\Gamma$ , then since  $E(\delta) \in B_1(Z)$ , we have

$$\begin{aligned} & \|E(\delta)g(T|_Z) - g(T|_Z)E(\delta)\|_{B(Z)} \\ & \leq \|E(\delta)[g(T|_Z) - g_n(T|_Z)]\|_{B(Z)} + \|[g_n(T|_Z) - g(T|_Z)]E(\delta)\|_{B(Z)} \\ & \leq 2\|g - g_n\|_{C(\Gamma)} \rightarrow 0. \end{aligned}$$

Thus  $E(\delta)$  commutes with  $g(T|_Z)$  for all  $g \in C(\Gamma)$  and  $\delta \in \mathcal{B}(\mathbb{C})$ .

We can then rewrite (5) in the form

$$\int_{\Gamma} g \chi_{\delta} dE(\cdot)x = \int_{\Gamma} g dE(\cdot)E(\delta)x$$

for all  $x \in Z$ ,  $\delta \in \mathcal{B}(\mathbb{C})$ , and  $g \in C(\Gamma)$ .

Again, by uniqueness of the Riesz representation, it follows that

$$\chi_{\delta} dE(\cdot)x = dE(\cdot)E(\delta)x.$$

Therefore, for each  $\epsilon \in \mathcal{B}(\mathbb{C})$  and  $x \in Z$ ,

$$\begin{aligned} E(\epsilon)E(\delta)x &= \int_{\Gamma} \chi_{\epsilon} dE(\cdot)E(\delta)x = \int_{\Gamma} \chi_{\epsilon} \chi_{\delta} dE(\cdot)x \\ &= \int_{\Gamma} \chi_{\epsilon \cap \delta} dE(\cdot)x = E(\epsilon \cap \delta)x. \end{aligned}$$

We consider now the map  $f \in \mathbb{B}(\Gamma) \rightarrow \int_{\Gamma} f dE(\cdot)x \in X$ . Denote the integral by  $\tau(f)x$ . We have for all  $x \in Z$  and  $f \in \mathbb{B}(\Gamma)$

$$\begin{aligned} \|\tau(f)x\| &= \sup_{x^* \in X^*; \|x^*\|=1} \left| \int_{\Gamma} f dx^* E(\cdot)x \right| \\ &\leq \|x\|_Z \sup_{\Gamma} |f|. \end{aligned} \tag{6}$$

If  $U \in T'$ , we saw that  $UE(\delta) = E(\delta)U$  for all  $\delta \in \mathcal{B}(\mathbb{C})$ . It follows from the definition of the integral with respect to  $E(\cdot)x$  (cf. Section 9.2) that for each  $x \in Z$

$$U\tau(f)x = \tau(f)Ux.$$

In particular, this is true for all  $U \in \mathcal{A}$ . Hence by (6) and Theorem 9.11, Part (iii),

$$\begin{aligned} \|\tau(f)x\|_Z &= \sup_{U \in \mathcal{A}} \|U\tau(f)x\| = \sup_{U \in \mathcal{A}} \|\tau(f)Ux\| \\ &\leq \sup_{\Gamma} |f| \sup_{U \in \mathcal{A}} \|Ux\|_Z \leq \|f\|_{\mathbb{B}(\Gamma)} \|x\|_Z. \end{aligned}$$

Thus  $\tau(f) \in B(Z)$  for all  $f \in \mathbb{B}(\Gamma)$ . Furthermore, the map  $\tau : \mathbb{B}(\Gamma) \rightarrow B(Z)$  is linear and norm-decreasing, and clearly multiplicative on the simple Borel functions on  $\Gamma$ . A routine density argument proves the multiplicativity of  $\tau$  on  $\mathbb{B}(\Gamma)$ .

The uniqueness statement about  $E$  is an immediate consequence of the uniqueness of the Riesz representation.  $\square$

**Remark 9.17.** Let  $\Delta$  be a closed interval. The semi-simplicity space for  $T \in B(X)$  is the smallest Banach subspace  $W_0$  in the increasing scale  $\{W_m; m = 0, 1, 2, \dots\}$  of Banach subspaces defined below.

Let  $C^m(\Delta)$  denote the Banach algebra of all complex functions with continuous derivatives in  $\Delta$  up to the  $m$ th order, with pointwise operations and norm

$$\|f\|_m := \sum_{j=0}^m \sup_{\Delta} |f^{(j)}|/j!.$$

We apply Theorem 9.11, Part (iii), to the multiplicative semigroup of operators

$$\mathcal{A}_m := \{p(T); p \in \mathcal{P}, \|p\|_m \leq 1\},$$

where  $\mathcal{P}$  denotes the polynomial algebra  $\mathbb{C}[t]$  and  $m = 0, 1, 2, \dots$ .

Fix  $m \in \mathbb{N} \cup \{0\}$ . By Theorem 9.11, the Banach subspace  $W_m := Z(\mathcal{A}_m)$  is  $\mathcal{A}_m$ -invariant and  $\mathcal{A}_m|_{W_m} \subset B_1(W_m)$ , i.e.,

$$\|p(T)x\|_{W_m} \leq \|x\|_{W_m} \|p\|_m$$

for all  $p \in \mathcal{P}$  and  $x \in W_m$ . By density of  $\mathcal{P}$  in  $C^m(\Delta)$ , it follows that  $T|_{W_m}$  is of contractive class  $C^m(\Delta)$ , that is, there exists a norm-decreasing algebra homomorphism of  $C^m(\Delta)$  into  $B(W_m)$  that extends the usual polynomial operational calculus. Moreover,  $W_m$  is maximal with this property.

## 9.10 Analytic operational calculus

Let  $K \subset \mathbb{C}$  be a non-empty compact set, and denote by  $H(K)$  the (complex) algebra of all complex functions analytic in some neighbourhood of  $K$  (depending on the function), with pointwise operations. A net  $f_\alpha \subset H(K)$  converges to  $f$  if all  $f_\alpha$  are analytic in some fixed neighbourhood  $\Omega$  of  $K$ , and  $f_\alpha \rightarrow f$  pointwise uniformly on every compact subset of  $\Omega$ . When this is the case,  $f$  is analytic in  $\Omega$  (hence  $f \in H(K)$ ), and one verifies easily that the operations in  $H(K)$  are continuous relative to the above convergence concept (or, equivalently, relative to the topology associated with the above convergence concept). Thus,  $H(K)$  is a so-called *topological algebra*.

Throughout this section,  $\mathcal{A}$  will denote a fixed (complex) Banach algebra with unit  $e$ . Let  $\mathcal{F}$  be a topological algebra of complex functions defined on some subset of  $\mathbb{C}$ , with pointwise operations, such that  $f_k(\lambda) := \lambda^k \in \mathcal{F}$  for  $k = 0, 1$ . Given  $a \in \mathcal{A}$ , an  $\mathcal{F}$ -operational calculus for  $a$  is a continuous representation  $\tau : \mathcal{F} \rightarrow \mathcal{A}$  such that  $\tau(f_1) = a$  (in the present context, a *representation* is an algebra homomorphism sending the identity  $f_0$  of  $\mathcal{F}$  to the identity  $e$  of  $\mathcal{A}$ ).

**Notation 9.18.** Let  $K \subset \Omega \subset \mathbb{C}$ ,  $K$  compact and  $\Omega$  open. There exists an open set  $\Delta$  with boundary  $\Gamma$  consisting of finitely many positively oriented rectifiable Jordan curves, such that  $K \subset \Delta$  and  $\Delta \cup \Gamma \subset \Omega$ . Let  $\Gamma(K, \Omega)$  denote the family of all  $\Gamma$ s with these properties.

If  $F$  is an  $\mathcal{A}$ -valued function analytic in  $\Omega \cap K^c$  and  $\Gamma, \Gamma' \in \Gamma(K, \Omega)$ , it follows from (the vector-valued version of) Cauchy's theorem that

$$\int_{\Gamma} F(\lambda) d\lambda = \int_{\Gamma'} F(\lambda) d\lambda.$$

We apply this observation to the function  $F = f(\cdot)R(\cdot; a)$  when  $\sigma(a) \subset K$  and  $f \in H(K)$ , to conclude that the so-called *Riesz–Dunford integral*

$$\tau(f) := \frac{1}{2\pi i} \int_{\Gamma} f(\lambda) R(\lambda; a) d\lambda \quad (1)$$

(with  $\Gamma \in \Gamma(K, \Omega)$  and  $f$  analytic in the open neighborhood  $\Omega$  of  $K$ ) is a well-defined element of  $\mathcal{A}$  (independent on the choice of  $\Gamma$ !).

**Theorem 9.19.** *The element  $a \in \mathcal{A}$  has an  $H(K)$ -operational calculus iff  $\sigma(a) \subset K$ . In that case, the  $H(K)$ -operational calculus for  $a$  is unique, and is given by (1).*

**Proof.** Suppose  $\sigma(a) \subset K$ . For  $f \in H(K)$ , define  $\tau(f)$  by (1). As observed above,  $\tau : H(K) \rightarrow \mathcal{A}$  is well defined and clearly linear. Let  $f, g \in H(K)$ ; suppose both are analytic in the open neighbourhood  $\Omega$  of  $K$ . Let  $\Gamma' \in \Gamma(K, \Omega)$ , and let  $\Delta'$  be the open neighborhood of  $K$  with boundary  $\Gamma'$ . Choose  $\Gamma \in \Gamma(K, \Delta')$  (hence  $\Gamma \in \Gamma(K, \Omega)$ ). By Cauchy's integral formula,

$$\frac{1}{2\pi i} \int_{\Gamma'} \frac{g(\mu)}{\mu - \lambda} d\mu = g(\lambda) \quad (\lambda \in \Gamma), \quad (2)$$

and

$$\int_{\Gamma} \frac{f(\lambda)}{\lambda - \mu} d\lambda = 0 \quad (\mu \in \Gamma'). \quad (3)$$

Therefore, by the resolvent identity for  $R(\cdot) := R(\cdot; a)$ , Fubini's theorem, and Relations (2) and (3), we have

$$\begin{aligned} (2\pi i)^2 \tau(f) \tau(g) &= \int_{\Gamma} f(\lambda) R(\lambda) d\lambda \int_{\Gamma'} g(\mu) R(\mu) d\mu \\ &= \int_{\Gamma} \int_{\Gamma'} f(\lambda) g(\mu) \frac{R(\lambda) - R(\mu)}{\mu - \lambda} d\mu d\lambda \\ &= \int_{\Gamma} f(\lambda) R(\lambda) \int_{\Gamma'} \frac{g(\mu)}{\mu - \lambda} d\mu d\lambda + \int_{\Gamma'} g(\mu) R(\mu) \int_{\Gamma} \frac{f(\lambda)}{\lambda - \mu} d\lambda d\mu \\ &= 2\pi i \int_{\Gamma} f(\lambda) g(\lambda) R(\lambda) d\lambda = (2\pi i)^2 \tau(fg). \end{aligned}$$

This proves the multiplicativity of  $\tau$ .

Let  $\{f_{\alpha}\}$  be a net in  $H(K)$  converging to  $f$ , let  $\Omega$  be an open neighbourhood of  $K$  in which all  $f_{\alpha}$  (and  $f$ ) are analytic, and let  $\Gamma \in \Gamma(K, \Omega)$ . Since  $R(\cdot)$  is



continuous on  $\Gamma$ , we have  $M := \sup_{\Gamma} \|R(\cdot)\| < \infty$ , and therefore, denoting the (finite) length of  $\Gamma$  by  $|\Gamma|$ , we have

$$\|\tau(f_{\alpha}) - \tau(f)\| = \|\tau(f_{\alpha} - f)\| \leq \frac{M|\Gamma|}{2\pi} \sup_{\Gamma} |f_{\alpha} - f| \rightarrow 0,$$

since  $f_{\alpha} \rightarrow f$  uniformly on  $\Gamma$ . This proves the continuity of  $\tau : H(K) \rightarrow \mathcal{A}$ .

If  $f$  is analytic in the disc  $\Delta_r := \{\lambda; |\lambda| < r\}$  and  $K \subset \Delta_r$ , the (positively oriented) circle  $C_{\rho} = \{\lambda; |\lambda| = \rho\}$  is in  $\Gamma(K, \Delta_r)$  for suitable  $\rho < r$ , and for any  $a \in \mathcal{A}$  with spectrum in  $K$ , the Neumann series expansion of  $R(\lambda)$  converges uniformly on  $C_{\rho}$ . Integrating term-by-term, we get

$$\tau(f) = \sum_{n=0}^{\infty} a^n \frac{1}{2\pi i} \int_{C_{\rho}} \frac{f(\lambda)}{\lambda^{n+1}} d\lambda = \sum_{n=0}^{\infty} \frac{f^{(n)}(0)}{n!} a^n. \quad (4)$$

In particular,  $\tau(\lambda^k) = a^k$  for all  $k = 0, 1, 2, \dots$ , and we conclude that  $\tau$  is an  $H(K)$ -operational calculus for  $a$ .

If  $\tau'$  is any  $H(K)$ -operational calculus for  $a$ , then necessarily  $\tau'(\lambda^n) = a^k$  for all  $k = 0, 1, 2, \dots$ , that is,  $\tau'$  coincides with  $\tau$  on all polynomials. If  $f$  is a rational function with poles in  $K^c$ , then  $f \in H(K)$ . Writing  $f = p/q$  with polynomials  $p, q$  such that  $q \neq 0$  on  $K$  (so that  $1/q \in H(K)$ !), we have  $e = \tau'(1) = \tau'(q \cdot (1/q)) = \tau'(q)\tau'(1/q)$ , hence  $\tau'(q)$  is non-singular, with inverse  $\tau'(1/q)$ . Therefore  $\tau'(f) = \tau'(p)\tau'(1/q) = \tau'(p)\tau'(q)^{-1} = \tau(p)\tau(q)^{-1} = \tau(f)$ . By Runge's theorem (cf. Exercise 1), the rational functions with poles in  $K^c$  are dense in  $H(K)$ , and the conclusion  $\tau'(f) = \tau(f)$  for all  $f \in H(K)$  follows from the continuity of both  $\tau'$  and  $\tau$ . This proves the uniqueness of the  $H(K)$ -operational calculus for  $a$ .

Finally, suppose  $a \in \mathcal{A}$  has an  $H(K)$ -operational calculus  $\tau$ , and let  $\mu \in K^c$ . The polynomial  $q(\lambda) := \mu - \lambda$  does not vanish on  $K$ , so that (as was proved above)  $\tau(q) = (\mu e - a)$  is non-singular, and  $R(\mu; a) = \tau(1/(\mu - \lambda))$ . In particular,  $\sigma(a) \subset K$ .  $\square$

Let  $\tau$  be the  $H(\sigma(a))$ -operational calculus for  $a$ . Since  $\tau(f) = f(a)$  when  $f$  is a polynomial, it is customary to use the notation  $f(a)$  instead of  $\tau(f)$  for all  $f \in H(\sigma(a))$ .

**Theorem 9.20 (Spectral mapping theorem).** *Let  $a \in \mathcal{A}$  and  $f \in H(\sigma(a))$ . Then  $\sigma(f(a)) = f(\sigma(a))$ .*

**Proof.** Let  $\mu = f(\lambda)$  with  $\lambda \in \sigma(a)$ . Since  $\lambda$  is a zero of the analytic function  $\mu - f$ , there exists  $h \in H(\sigma(a))$  such that

$$\mu - f(\zeta) = (\lambda - \zeta)h(\zeta)$$

in a neighbourhood of  $\sigma(a)$ . Applying the  $H(\sigma(a))$ -operational calculus, we get

$$\mu e - f(a) = (\lambda e - a)h(a) = h(a)(\lambda e - a).$$

If  $\mu \in \rho(f(a))$ , and  $v = R(\mu; f(a))$ , then

$$(\lambda e - a)[h(a)v] = e \quad \text{and} \quad [vh(a)](\lambda e - a) = e,$$

that is,  $\lambda \in \rho(a)$ , contradiction. Hence  $\mu \in \sigma(f(a))$ , and we proved the inclusion  $f(\sigma(a)) \subset \sigma(f(a))$ .

If  $\mu \notin f(\sigma(a))$ , then  $\mu - f \neq 0$  on  $\sigma(a)$ , and therefore  $g := 1/(\mu - f) \in H(\sigma(a))$ . Since  $(\mu - f)g = 1$  in a neighbourhood of  $\sigma(a)$ , we have  $(\mu e - f(a))g(a) = e$ , that is,  $\mu \in \rho(f(a))$ . This proves the inclusion  $\sigma(f(a)) \subset f(\sigma(a))$ .  $\square$

**Theorem 9.21 (Composite function theorem).** *Let  $a \in \mathcal{A}$ ,  $f \in H(\sigma(a))$ , and  $g \in H(f(\sigma(a)))$ . Then  $g \circ f \in H(\sigma(a))$  and  $(g \circ f)(a) = g(f(a))$ .*

**Proof.** By Theorem 9.20,  $g(f(a))$  is well defined.

Let  $\Omega$  be an open neighbourhood of  $K := f(\sigma(a)) = \sigma(f(a))$  in which  $g$  is analytic, and let  $\Gamma \in \Gamma(K, \Omega)$ . Then

$$g(f(a)) = \frac{1}{2\pi i} \int_{\Gamma} g(\mu) R(\mu; f(a)) d\mu. \quad (5)$$

Since  $\Gamma \subset K^c$ , for each fixed  $\mu \in \Gamma$ , the function  $\mu - f$  does not vanish on  $\sigma(a)$ , and consequently  $k_{\mu} := 1/(\mu - f) \in H(\sigma(a))$ . The relation  $(\mu - f)k_{\mu} = 1$  (valid in a neighbourhood of  $\sigma(a)$ ) implies through the operational calculus for  $a$  that  $k_{\mu}(a) = R(\mu; f(a))$ , that is,

$$R(\mu; f(a)) = \frac{1}{2\pi i} \int_{\Gamma'} k_{\mu}(\lambda) R(\lambda; a) d\lambda \quad (6)$$

for a suitable  $\Gamma'$ . We now substitute (6) in (5), interchange the order of integration, and use Cauchy's integral formula:

$$\begin{aligned} (2\pi i)^2 g(f(a)) &= \int_{\Gamma} g(\mu) \int_{\Gamma'} k_{\mu}(\lambda) R(\lambda; a) d\lambda d\mu = \int_{\Gamma'} \int_{\Gamma} \frac{g(\mu)}{\mu - f(\lambda)} d\mu R(\lambda; a) d\lambda \\ &= 2\pi i \int_{\Gamma'} g(f(\lambda)) R(\lambda; a) d\lambda = (2\pi i)^2 (g \circ f)(a). \end{aligned}$$

$\square$

## 9.11 Isolated points of the spectrum

**Construction 9.22.** Let  $\mu$  be an isolated point of  $\sigma(a)$ . There exists then a function  $e_{\mu} \in H(\sigma(a))$  that equals 1 in a neighbourhood of  $\mu$  and 0 in a neighbourhood of  $\sigma_{\mu} := \sigma(a) \cap \{\mu\}^c$ . Set  $E_{\mu} = e_{\mu}(a)$ . The element  $E_{\mu}$  is independent of the choice of the function  $e_{\mu}$ , and since  $e_{\mu}^2 = e_{\mu}$  in a neighbourhood of  $\sigma(a)$ , it is an idempotent commuting with  $a$ .

Let  $\delta = \text{dist}(\mu, \sigma_{\mu})$ . By Laurent's theorem (whose classical proof applies word for word to vector-valued functions), we have for  $0 < |\lambda - \mu| < \delta$ :

$$R(\lambda; a) = \sum_{k=-\infty}^{\infty} a_k (\mu - \lambda)^k, \quad (1)$$

where

$$a_k = -\frac{1}{2\pi i} \int_{\Gamma} (\mu - \lambda)^{-k-1} R(\lambda; a) d\lambda, \quad (2)$$

and  $\Gamma$  is a positively oriented circle centred at  $\mu$  with radius  $r < \delta$ . Choosing a function  $e_{\mu}$  as above that equals 1 in a neighbourhood of the corresponding closed disc, we can add the factor  $e_{\mu}$  to the integrand in (2). For  $k \in -\mathbb{N}$ , the new integrand is analytic in a neighbourhood  $\Omega$  of  $\sigma(a)$ , and therefore, by Cauchy's theorem, the circle  $\Gamma$  may be replaced by any  $\Gamma' \in \Gamma(\sigma(a), \Omega)$ . By the multiplicativity of the analytic operational calculus, it follows that

$$a_{-k} = -(\mu e - a)^{k-1} E_{\mu} \quad (k \in \mathbb{N}). \quad (3)$$

In particular, it follows from (3) that  $a_{-k} = 0$  for all  $k \geq k_0$  iff  $a_{-k_0} = 0$ . Consequently the point  $\mu$  is a pole of order  $m$  of  $R(\cdot; a)$  iff

$$(\mu e - a)^m E_{\mu} = 0 \quad \text{and} \quad (\mu e - a)^{m-1} E_{\mu} \neq 0. \quad (4)$$

Similarly,  $R(\cdot; a)$  has a removable singularity at  $\mu$  iff  $E_{\mu} = 0$ . In this case, the relation  $(\lambda e - a)R(\lambda; a) = R(\lambda; a)(\lambda e - a) = e$  extends by continuity to the point  $\lambda = \mu$ , so that  $\mu \in \rho(a)$ , contradicting our hypothesis. Consequently  $E_{\mu} \neq 0$ .

These observations have a particular significance when  $\mathcal{A} = B(X)$  for a Banach space  $X$ . If  $\mu$  is an isolated point of the spectrum of  $T \in B(X)$  and  $E_{\mu}$  is the corresponding idempotent, the non-zero projection  $E_{\mu}$  (called the *Riesz projection at  $\mu$  for  $T$* ) commutes with  $T$ , so that its range  $X_{\mu} \neq \{0\}$  is a reducing subspace for  $T$  (cf. Terminology 8.5 (2)).

Let  $T_{\mu} := T|_{X_{\mu}}$ . If  $\zeta \neq \mu$ , the function  $h(\lambda) := e_{\mu}(\lambda)/(\zeta - \lambda)$  belongs to  $H(\sigma(T))$  for a proper choice of  $e_{\mu}$ , and  $(\zeta - \lambda)h(\lambda) = e_{\mu}$ . Applying the analytic operational calculus, we get

$$(\zeta I - T)h(T) = h(T)(\zeta I - T) = E_{\mu},$$

and therefore (since  $h(T)X_{\mu} \subset X_{\mu}$ ),

$$(\zeta I - T_{\mu})h(T)x = h(T)(\zeta I - T_{\mu})x = x \quad (x \in X_{\mu}).$$

Hence  $\zeta \in \rho(T_{\mu})$ , and consequently  $\sigma(T_{\mu}) \subset \{\mu\}$ . Since  $X_{\mu} \neq \{0\}$ , the spectrum of  $T_{\mu}$  in non-empty (cf. Theorem 7.6), and therefore

$$\sigma(T_{\mu}) = \{\mu\}. \quad (5)$$

Consider the complementary projection  $E'_{\mu} := I - E_{\mu}$ , and let  $X'_{\mu} := E'_{\mu}X$  and  $T'_{\mu} := T|_{X'_{\mu}}$ . The above argument (with  $h(\lambda) := (1 - e_{\mu})/(\zeta - \lambda)$  for a fixed  $\zeta \notin \sigma_{\mu}$ ) shows that  $\sigma(T'_{\mu}) \subset \sigma_{\mu}$ . If the inclusion is strict, pick  $\zeta \in \sigma_{\mu} \cap \rho(T'_{\mu})$ . Then  $\zeta \neq \mu$ , so that  $\exists R(\zeta; T_{\mu})$  (by (5)), and of course  $\exists R(\zeta; T'_{\mu})$ . Let

$$V := R(\zeta; T_{\mu})E_{\mu} + R(\zeta; T'_{\mu})E'_{\mu}. \quad (6)$$

Clearly  $V \in B(X)$ , and a simple calculation shows that  $(\zeta I - T)V = V(\zeta I - T) = I$ . Hence  $\zeta \in \rho(T)$ , contradicting the fact that  $\zeta \in \sigma_{\mu} \subset \sigma(T)$ . Consequently

$$\sigma(T'_{\mu}) = \sigma_{\mu}. \quad (7)$$

We also read from (6) (and the above observation) that  $V = R(\zeta; T)$  for all  $\zeta \notin \sigma(T)$ , that is,

$$R(\zeta; T_\mu) = R(\zeta; T)|_{X_\mu} \quad (\zeta \neq \mu); \quad (8)$$

$$R(\zeta; T'_\mu) = R(\zeta; T)|_{X'_\mu} \quad (\zeta \notin \sigma_\mu). \quad (9)$$

(Rather than discussing an isolated point, we could consider any closed subset  $\sigma$  of the spectrum, whose complement in the spectrum is also closed; such a set is called a *spectral set*. The above arguments and conclusions go through with very minor changes.)

By (4), the isolated point  $\mu$  of  $\sigma(T)$  is a pole of order  $m$  of  $R(\cdot; T)$  iff

$$(\mu I - T)^m X_\mu = \{0\} \quad \text{and} \quad (\mu I - T)^{m-1} X_\mu \neq \{0\}. \quad (10)$$

In this case, any non-zero vector in the latter space is an eigenvector of  $T$  for the eigenvalue  $\mu$ , that is,  $\mu \in \sigma_p(T)$ .

By (10),  $X_\mu \subset \ker(\mu I - T)^m$ . Let  $x \in \ker(\mu I - T)^m$ . Since  $E_\mu$  commutes with  $T$ ,

$$(\mu I - T'_\mu)^m (I - E_\mu)x = (I - E_\mu)(\mu I - T)^m x = 0.$$

By (7),  $\mu \in \rho(T'_\mu)$ , so that  $\mu I - T'_\mu$  is one-to-one; hence  $(I - E_\mu)x = 0$ , and therefore  $x \in X_\mu$ , and we conclude that

$$X_\mu = \ker(\mu I - T)^m. \quad (11)$$

## 9.12 Compact operators

**Definition 9.23.** Let  $X$  be a Banach space, and denote by  $S$  its closed unit ball. An operator  $T \in B(X)$  is compact if the set  $TS$  is conditionally compact.

Equivalently,  $T$  is compact iff it maps bounded sets onto conditionally compact sets.

In terms of sequences, the compactness of  $T$  is characterized by the property: if  $\{x_n\}$  is a bounded sequence, then  $\{Tx_n\}$  has a convergent subsequence.

Denote by  $K(X)$  the set of all compact operators on  $X$ .

**Proposition 9.24.**

- (i)  $K(X)$  is a closed two-sided ideal in  $B(X)$ ;
- (ii)  $K(X) = B(X)$  iff  $X$  has finite dimension.
- (iii) The restriction of a compact operator to a closed invariant subspace is compact.

**Proof.** (i)  $K(X)$  is trivially stable under linear combinations. Let  $T \in K(X)$ ,  $A \in B(X)$ , and let  $\{x_n\}$  be a bounded sequence. Let then  $\{Tx_{n_k}\}$  be

a convergent subsequence of  $\{Tx_n\}$ . Then  $\{ATx_{n_k}\}$  converges (by continuity of  $A$ ), that is,  $AT \in K(X)$ . Also  $\{Ax_n\}$  is bounded (by boundedness of  $A$ ), and therefore  $\{TAx_{n'_k}\}$  converges for some subsequence, that is,  $TA \in K(X)$ , and we conclude that  $K(X)$  is a two-sided ideal in  $B(X)$ .

Suppose  $\{T_m\} \subset K(X)$  converges in  $B(X)$  to  $T$ . Let  $\{x_n\} \subset X$  be bounded, say  $\|x_n\| < M$  for all  $n$ . By a Cantor diagonal process, we can select a subsequence  $\{x_{n_k}\}$  of  $\{x_n\}$  such that  $\{T_mx_{n_k}\}_k$  converges for all  $m$ . Given  $\epsilon > 0$ , let  $m_0 \in \mathbb{N}$  be such that  $\|T - T_m\| < \epsilon/(4M)$  for all  $m > m_0$ . Fix  $m > m_0$ , and then  $k_0 = k_0(m)$  such that  $\|T_mx_{n_k} - T_mx_{n_j}\| < \epsilon/2$  for all  $k, j > k_0$ . Then for all  $k, j > k_0$ ,

$$\begin{aligned} \|Tx_{n_k} - Tx_{n_j}\| &\leq \|(T - T_m)(x_{n_k} - x_{n_j})\| + \|T_mx_{n_k} - T_mx_{n_j}\| \\ &< [\epsilon/(4M)]2M + \epsilon/2 = \epsilon, \end{aligned}$$

and we conclude that  $\{Tx_{n_k}\}$  converges to some element  $y$ . Hence

$$\limsup_k \|Tx_{n_k} - y\| \leq \epsilon,$$

and therefore,  $Tx_{n_k} \rightarrow y$  by the arbitrariness of  $\epsilon$ .

(ii) If  $X$  has finite dimension, any linear operator  $T$  on  $X$  maps bounded sets onto bounded sets, and a bounded set in  $X$  is conditionally compact.

Conversely, if  $K(X) = B(X)$ , then, equivalently, the identity operator  $I$  is compact, and therefore the closed unit ball  $S = IS$  is compact. Hence  $X$  has finite dimension, by Theorem 5.27.

The proof of (iii) is trivial.  $\square$

**Theorem 9.25 (Schauder).**  $T \in K(X)$  iff  $T^* \in K(X^*)$ .

**Proof.** (1) Let  $T \in K(X)$ , and let  $\{x_n^*\}$  be a bounded sequence in  $X^*$ , say  $\|x_n^*\| \leq M$  for all  $n$ . Then, for all  $n$ ,  $|x_n^*x| \leq M\|T\|$  for all  $x \in \overline{TS}$  and  $|x_n^*x - x_n^*y| \leq M\|x - y\|$  for all  $x, y \in X$ , that is, the sequence of functions  $\{x_n^*\}$  is uniformly bounded and equicontinuous on the compact metric space  $\overline{TS}$ . By the Arzela–Ascoli theorem (cf. Exercise 3), there exists a subsequence  $\{x_{n_k}^*\}$  of  $\{x_n^*\}$  converging uniformly on  $\overline{TS}$ . Hence

$$\sup_{x \in S} |x_{n_k}^*(Tx) - x_{n_j}^*(Tx)| \rightarrow 0 \quad (k, j \rightarrow \infty),$$

that is,  $\|T^*x_{n_k}^* - T^*x_{n_j}^*\| \rightarrow 0$  as  $k, j \rightarrow \infty$ , and consequently  $\{T^*x_{n_k}^*\}$  converges (strongly) in  $X^*$ . This proves that  $T^*$  is compact.

(2) Let  $T^*$  be compact. By Part (1) of the proof,  $T^{**}$  is compact. Let  $\{x_n\} \subset S$ . Then  $\|\hat{x}_n\| = \|x_n\| \leq 1$ , and therefore  $T^{**}\hat{x}_{n_k}$  converges in  $X^{**}$  for some  $1 \leq n_1 < n_2 < \dots$ , that is,

$$\sup_{\|x^*\|=1} |(T^{**}\hat{x}_{n_k})x^* - (T^{**}\hat{x}_{n_j})x^*| \rightarrow 0 \quad (k, j \rightarrow \infty).$$

Equivalently,

$$\sup_{\|x^*\|=1} |x^*Tx_{n_k} - x^*Tx_{n_j}| \rightarrow 0,$$

that is,  $\|Tx_{n_k} - Tx_{n_j}\| \rightarrow 0$  as  $k, j \rightarrow \infty$ , and consequently  $T$  is compact.  $\square$

**Lemma 9.26.** *Let  $Y$  be a proper closed subspace of the Banach space  $X$ . Then  $\sup_{x \in X_1} d(x, Y) = 1$ . ( $X_1$  denotes the unit sphere of  $X$ .)*

**Proof.** Let  $1 > \epsilon > 0$ . If  $d(x, Y) = 0$  for all  $x \in X_1$ , then since  $Y$  is closed,  $X_1 \subset Y$ , and therefore  $X \subset Y$  (because  $Y$  is a subspace), contrary to the assumption that  $Y$  is a *proper* subspace of  $X$ . Thus there exists  $x_1 \in X_1$  such that  $\delta := d(x_1, Y) > 0$ . By definition of  $d(x_1, Y)$ , there exists  $y_1 \in Y$  such that  $(\delta \leq) d(x_1, y_1) < (1 + \epsilon)\delta$ . Let  $u = x_1 - y_1$  and  $x = u/\|u\|$ . Then  $x \in X_1$ ,  $\|u\| < (1 + \epsilon)\delta$ , and for all  $y \in Y$

$$(1 + \epsilon)\delta\|x - y\| \geq \|u\|\|x - y\| = \|u - \|u\|y\| = \|x_1 - (y_1 + \|u\|y)\| \geq \delta.$$

Hence  $\|x - y\| \geq 1/(1 + \epsilon) > 1 - \epsilon$  for all  $y \in Y$ , and therefore  $d(x, Y) \geq 1 - \epsilon$ . Since we have trivially  $d(x, Y) \leq 1$ , the conclusion of the lemma follows.  $\square$

**Theorem 9.27 (Riesz–Schauder).** *Let  $T$  be a compact operator on the Banach space  $X$ . Then*

- (i)  $\sigma(T)$  is at most countable. If  $\{\mu_n\}$  is a sequence of distinct non-zero points of the spectrum, then  $\mu_n \rightarrow 0$ .
- (ii) Each non-zero point  $\mu \in \sigma(T)$  is an isolated point of the spectrum, and is an eigenvalue of  $T$  and a pole of the resolvent of  $T$ . If  $m$  is the order of the pole  $\mu$ , and  $E_\mu$  is the Riesz projection for  $T$  at  $\mu$ , then its range  $E_\mu X$  equals  $\ker(\mu I - T)^m$  and is finite dimensional. In particular, the  $\mu$ -eigenspace of  $T$  is finite dimensional.

**Proof.** (1) Let  $\mu$  be a non-zero complex number, and let  $\{x_n\} \subset X$  be such that  $(\mu I - T)x_n$  converge to some  $y$ . If  $\{x_n\}$  is unbounded, say  $0 < \|x_n\| \rightarrow \infty$  without loss of generality (w.l.o.g.), consider the unit vectors  $z_n := x_n/\|x_n\|$ . Since  $T$  is compact, there exist  $1 \leq n_1 < n_2 < \dots$  such that  $Tz_{n_k} \rightarrow v \in X$ . Then

$$\mu z_{n_k} = \frac{1}{\|x_{n_k}\|}(\mu I - T)x_{n_k} + Tz_{n_k} \rightarrow 0y + v = v. \quad (*)$$

Hence

$$\mu v = \lim_k T(\mu z_{n_k}) = Tv.$$

If  $\mu \notin \sigma_p(T)$ , we must have  $v = 0$ . Then by (\*)  $|\mu| = \|\mu z_{n_k}\| \rightarrow 0$ , contradiction. Therefore (if  $\mu \notin \sigma_p(T)$ !) the sequence  $\{x_n\}$  is bounded, and has therefore a subsequence  $\{x_{n_k}\}$  such that  $\exists \lim_k Tx_{n_k} := u$ . Then as above

$$x_{n_k} = \mu^{-1}[(\mu I - T)x_{n_k} + Tx_{n_k}] \rightarrow \mu^{-1}(y + u) := x,$$

and therefore  $y = \lim_k (\mu I - T)x_{n_k} = (\mu I - T)x \in (\mu I - T)X$ . Thus  $(\mu I - T)X$  is closed. This proves that a non-zero  $\mu$  is either in  $\sigma_p(T)$  or else the range of  $\mu I - T$  is closed. In the later case, if this range is dense in  $X$ ,  $\mu I - T$  is onto (and one-to-one!), and therefore  $\mu \in \rho(T)$ . If the range is not dense in  $X$ , it is

a *proper closed subspace* of  $X$ ; by Corollary 5.4., there exists  $x^* \neq 0$  such that  $x^*(\mu I - T)x = 0$  for all  $x \in X$ , i.e.,  $(\mu I - T^*)x^* = 0$ . Thus  $\mu \in \sigma_p(T^*)$ . In conclusion, if  $\mu \in \sigma(T)$  is not zero, then  $\mu \in \sigma_p(T) \cup \sigma_p(T^*)$ .

(2) Suppose  $\mu_n, n = 1, 2, \dots$ , are distinct eigenvalues of  $T$  that do not converge to zero. By passing if necessary to a subsequence, we may assume that  $|\mu_n| \geq \epsilon$  for all  $n$ , for some positive  $\epsilon$ . Let  $x_n$  be an eigenvector of  $T$  corresponding to the eigenvalue  $\mu_n$ . Then  $\{x_n\}$  is necessarily linearly independent (an elementary linear algebra fact!). Setting  $Y_n := \text{span}\{x_1, \dots, x_n\}$ ,  $Y_{n-1}$  is therefore, a *proper closed*  $T$ -invariant subspace of the Banach space  $Y_n$ , and clearly  $(\mu_n I - T)Y_n \subset Y_{n-1}$  (for all  $n > 1$ ). By Lemma 9.26, there exists  $y_n \in Y_n$  such that  $\|y_n\| = 1$  and  $d(y_n, Y_{n-1}) > 1/2$ , for each  $n > 1$ . Set  $z_n = y_n/\mu_n$ . Since  $\|z_n\| \leq 1/\epsilon$ , there exist  $1 < n_1 < n_2 < \dots$  such that  $Tz_{n_k}$  converges. However, for  $j > k$ ,

$$\|Tz_{n_j} - Tz_{n_k}\| = \|y_{n_j} - [(\mu_{n_j} I - T)z_{n_j} + Tz_{n_k}]\| > 1/2,$$

since the vector in square brackets belongs to  $Y_{n_j-1}$ , contradiction. This proves that if  $\{\mu_n\}$  is a sequence of distinct eigenvalues of  $T$ , then  $\mu_n \rightarrow 0$ .

(3) Suppose  $\mu \in \sigma(T)$ ,  $\mu \neq 0$ , is not an isolated point of the spectrum, and let then  $\mu_n, (n \in \mathbb{N})$  be distinct non-zero points of the spectrum converging to  $\mu$ . By the conclusion of Part (1) of the proof,  $\{\mu_n\} \subset \sigma_p(T) \cup \sigma_p(T^*)$ . Since the set  $\{\mu_n\}$  is infinite, at least one of its intersections with  $\sigma_p(T)$  and  $\sigma_p(T^*)$  is infinite. This infinite intersection converges to zero, by Part (2) of the proof (since both  $T$  and  $T^*$  are compact, by Theorem 9.25). Hence  $\mu = 0$ , contradiction! This shows that the non-zero points of  $\sigma(T)$  are isolated points of the spectrum. Since  $\sigma(T)$  is compact, it then follows that it is at most countable.

(4) Let  $\mu \neq 0, \mu \in \sigma(T)$ , and let  $E_\mu$  be the Riesz projection for  $T$  at (the isolated point)  $\mu$ . As before, let  $X_\mu = E_\mu X$  and  $T_\mu = T|_{X_\mu}$ . Let  $S_\mu$  denote the closed unit ball of  $X_\mu$ . Since  $\sigma(T_\mu) = \{\mu\}$  (cf. (5) of Section 9.11), we have  $0 \in \rho(T_\mu)$ , that is,  $\exists T_\mu^{-1} \in B(X_\mu)$ , and consequently  $T_\mu^{-1}S_\mu$  is bounded. The latter's image by the compact operator  $T_\mu$  (cf. Proposition 9.24 (iii)) is then conditionally compact; this image is the closed set  $S_\mu$ , hence  $S_\mu$  is compact, and therefore  $X_\mu$  is finite dimensional (by Theorem 5.27). Since  $\sigma(\mu I - T_\mu) = \mu - \sigma(T_\mu) = \{0\}$  by (5) of Section 9.11, the operator  $\mu I - T_\mu$  on the *finite dimensional space*  $X_\mu$  is *nilpotent*, that is, there exists  $m \in \mathbb{N}$  such that  $(\mu I - T_\mu)^m = 0$  but  $(\mu I - T_\mu)^{m-1} \neq 0$ . Equivalently,

$$(\mu I - T)^m E_\mu = 0 \quad \text{and} \quad (\mu I - T)^{m-1} E_\mu \neq 0.$$

By (4) of Section 9.11,  $\mu$  is a pole of order  $m$  of  $R(\cdot; T)$ , hence an eigenvalue of  $T$  (cf. observation following (10) of Section 9.11), and  $\ker(\mu I - T)^m = X_\mu$  by (11) of Section 9.11.  $\square$

## Exercises

[The first three exercises provide the proofs of theorems used in this chapter.]

### Runge theorem

1. Let  $S^2 = \bar{\mathbb{C}}$  denote the Riemann sphere, and let  $K \subset \mathbb{C}$  be compact. Fix a point  $a_j$  in each component  $V_j$  of  $S^2 - K$ , and let  $\mathcal{R}(\{a_j\})$  denote the set of all rational functions with poles in the set  $\{a_j\}$ .

If  $\mu$  is a complex Borel measure on  $K$ , we define its *Cauchy transform*  $\tilde{\mu}$  by

$$\tilde{\mu}(z) = \int_K \frac{d\mu(w)}{w - z} \quad (z \in S^2 - K). \quad (1)$$

Prove

- (a)  $\tilde{\mu}$  is analytic in  $S^2 - K$ .
- (b) For  $a_j \neq \infty$ , let  $d_j = d(a_j, K)$  and fix  $z \in B(a_j, r) \subset V_j$  (necessarily  $r < d_j$ ). Observe that

$$\frac{1}{w - z} = \sum_{n=0}^{\infty} \frac{(z - a_j)^n}{(w - a_j)^{n+1}}, \quad (2)$$

and the series converges uniformly for  $w \in K$ .

For  $a_j = \infty$ , we have

$$\frac{1}{w - z} = - \sum_{n=0}^{\infty} \frac{w^n}{z^{n+1}} \quad (|z| > r), \quad (3)$$

and the series converges uniformly for  $w \in K$ .

- (c) If  $\int_K h d\mu = 0$  for all  $h \in \mathcal{R}(\{a_j\})$ , then  $\tilde{\mu}(z) = 0$  for all  $z \in B(a_j, r)$ , hence for all  $z \in V_j$ , for all  $j$ , and therefore  $\tilde{\mu} = 0$  on  $S^2 - K$ .
- (d) Let  $\Omega \subset \mathbb{C}$  be open such that  $K \subset \Omega$ . If  $f$  is analytic in  $\Omega$  and  $\mu$  is as in Part (c), then  $\int_K f d\mu = 0$ . (Hint: represent  $f(z) = (1/2\pi i) \int_{\Gamma} f(w)/(w - z)dw$  for all  $z \in K$ , where  $\Gamma \in \Gamma(K, \Omega)$ , cf. Notation 9.18, and use Fubini's theorem.)
- (e) Prove that  $\mathcal{R}(\{a_j\})$  is  $C(K)$ -dense in  $H(\Omega)$  (the subspace of  $C(K)$  consisting of the analytic functions in  $\Omega$  restricted to  $K$ ). Hint: Theorem 4.9, Corollary 5.3, and Part (d). The result in Part (e) is *Runge's theorem*. In particular, the rational functions with poles off  $K$  are  $C(K)$ -dense in  $H(\Omega)$ .
- (f) If  $S^2 - K$  is *connected*, the polynomials are  $C(K)$ -dense in  $H(\Omega)$ . Hint: apply Part (e) with  $a = \infty$  in the single component of  $S^2 - K$ .



## Hartogs–Rosenthal theorem

2. (Notation as in Exercise 1) Let  $m$  denote the  $\mathbb{R}^2$ -Lebesgue measure.

- (a) The integral defining the Cauchy transform  $\tilde{\mu}$  converges absolutely  $m$ -a.e. (Hint: show that

$$\int_{\mathbb{R}^2} \int_K \frac{d|\mu|(w)}{|w-z|} dx dy < \infty$$

by using Tonelli's theorem and polar coordinates.)

- (b) Let  $\mathcal{R}(K)$  denote the space of rational functions with poles off  $K$ . Then  $\int_K h d\mu = 0$  for all  $h \in \mathcal{R}(K)$  iff  $\tilde{\mu} = 0$  off  $K$ . (Hint: use Cauchy's formula and Fubini's theorem for the non-trivial implication.)
- (c) It can be shown that if  $\tilde{\mu} = 0$   $m$ -a.e., then  $\mu = 0$ . Conclude that if  $m(K) = 0$  and  $\mu$  is a complex Borel measure on  $K$  such that  $\int_K h d\mu = 0$  for all  $h \in \mathcal{R}(K)$ , then  $\mu = 0$ . Consequently, if  $m(K) = 0$ , then  $\mathcal{R}(K)$  is dense in  $C(K)$  (cf. Theorem 4.9 and Corollary 5.6). This is the Hartogs–Rosenthal theorem.

## Arzela–Ascoli theorem

3. Let  $X$  be a compact metric space. A set  $\mathcal{F} \subset C(X)$  is *equicontinuous* if for each  $\epsilon > 0$ , there exists  $\delta > 0$  such that  $|f(x) - f(y)| < \epsilon$  for all  $f \in \mathcal{F}$  and  $x, y \in X$  such that  $d(x, y) < \delta$ . The set  $\mathcal{F}$  is *equibounded* if  $\sup_{f \in \mathcal{F}} \|f\|_u < \infty$ . Prove that if  $\mathcal{F}$  is equicontinuous and equibounded, then it is relatively compact in  $C(X)$ . Sketch:  $X$  is necessarily separable. Let  $\{a_k\}$  be a countable dense set in  $X$ . Let  $\{f_n\} \subset \mathcal{F}$ .  $\{f_n(a_1)\}$  is a bounded complex sequence; therefore there is a subsequence  $\{f_{n,1}\}$  of  $\{f_n\}$  converging at  $a_1$ ;  $\{f_{n,1}(a_2)\}$  is a bounded complex sequence, and therefore there is a subsequence  $\{f_{n,2}\}$  of  $\{f_{n,1}\}$  converging at  $a_2$  (and  $a_1$ ). Continuing inductively, we get subsequences  $\{f_{n,r}\}$  such that the  $(r+1)$ -th subsequence is a subsequence of the  $r$ th subsequence, and the  $r$ th subsequence converges at the points  $a_1, \dots, a_r$ . The diagonal subsequence  $\{f_{n,n}\}$  converges at *all* the points  $a_k$ . Use the compactness of  $X$  and an  $\epsilon/3$  argument to show that  $\{f_{n,n}\}$  is Cauchy in  $C(X)$ .

## Compact normal operators

4. Let  $X$  be a Hilbert space, and  $T \in K(X)$  be normal. Prove that there exist a sequence  $\{\lambda_n\} \in c_0$  and a sequence  $\{E_n\}$  of pairwise orthogonal finite rank projections such that  $\sum_{n=1}^N \lambda_n E_n \rightarrow T$  in  $B(X)$  as  $N \rightarrow \infty$ .

## Logarithms of Banach algebra elements

5. Let  $\mathcal{A}$  be a (unital, complex) Banach algebra, and let  $x \in \mathcal{A}$ . Prove:
- (a) If 0 belongs to the unbounded component  $V$  of  $\rho(x)$ , then  $x \in \exp \mathcal{A} := \{e^a; a \in \mathcal{A}\}$  (that is,  $x$  has a logarithm in  $\mathcal{A}$ ). Hint:  $\Omega := V^c$  is a simply connected open subset of  $\mathbb{C}$  containing  $\sigma(x)$ , and the analytic function  $f_1(\lambda) = \lambda$  does not vanish on  $\Omega$ . Therefore, there exists  $g$  analytic in  $\Omega$  such that  $e^g = f_1$ .
  - (b) The group generated  $\exp \mathcal{A}$  is an open subset of  $\mathcal{A}$ .
6. Let  $\mathcal{A}$  be a (unital, complex) Banach algebra, and let  $G_e$  denote the component of  $G := G(\mathcal{A})$  containing the identity  $e$ . Prove:
- (a)  $G_e$  is open.
  - (b)  $G_e$  is a normal subgroup of  $G$ .
  - (c)  $\exp \mathcal{A} \subset G_e$ .
  - (d)  $\exp \mathcal{A} \cdots \exp \mathcal{A}$  (finite product) is an open subset of  $G_e$  (cf. Exercise 5(b)).
  - (e) Let  $H$  be the group generated by  $\exp \mathcal{A}$ . Then  $H$  is an open and closed subset of  $G_e$ . Conclude that  $H = G_e$ .
  - (f) If  $\mathcal{A}$  is commutative, then  $G_e = \exp \mathcal{A}$ .

## Non-commutative Taylor theorem

7. (Notation as in Exercise 10, Chapter 7) Let  $\mathcal{A}$  be a (unital, complex) Banach algebra, and let  $a, b \in \mathcal{A}$ . Prove the following *non-commutative Taylor theorem* for each  $f \in H(\sigma(a, b))$ :

$$\begin{aligned} f(b) &= \sum_{j=0}^{\infty} (-1)^j \frac{f^{(j)}(a)}{j!} [C(a, b)^j e] \\ &= \sum_{j=0}^{\infty} [C(a, b)^j e] \frac{f^{(j)}(a)}{j!}. \end{aligned}$$

In particular, if  $a, b$  commute,

$$f(b) = \sum \frac{f^{(j)}(a)}{j!} (b - a)^j \quad (*)$$

for all  $f \in H(\sigma(a, b))$ , where (in this special case)

$$\sigma(a, b) = \{\lambda \in \mathbb{C}; d(\lambda, \sigma(a)) \leq r(b - a)\}.$$

If  $b - a$  is quasi-nilpotent,  $(*)$  is valid for all  $f \in H(\sigma(a))$ .

## Positive operators

8. Let  $X$  be a Hilbert space. Recall that  $T \in B(X)$  is *positive* (in symbols,  $T \geq 0$ ) iff  $(Tx, x) \geq 0$  for all  $x \in X$ . Prove:
  - (a) The positive operator  $T$  is non-singular (i.e. invertible in  $B(X)$ ) iff  $T - \epsilon I \geq 0$  for some  $\epsilon > 0$  (one writes also  $T \geq \epsilon I$  to express the last relation).
  - (b) The (arbitrary) operator  $T$  is non-singular iff both  $TT^* \geq \epsilon I$  and  $T^*T \geq \epsilon I$  for some  $\epsilon > 0$ .
9. Let  $X$  be a Hilbert space,  $T \in B(X)$ . Prove:
  - (a) If  $T$  is positive, then
 
$$|(Tx, y)|^2 \leq (Tx, x)(Ty, y) \quad \text{for all } x, y \in X.$$
  - (b) Let  $\{T_k\} \subset B(X)$  be a sequence of positive operators. Then  $T_k \rightarrow 0$  in the s.o.t. iff it does so in the w.o.t.
  - (c) If  $0 \leq T_k \leq T_{k+1} \leq KI$  for all  $k$  (for some positive constant  $K$ ), then  $\{T_k\}$  converges in  $B(X)$  in the s.o.t.

## Analytic functions operate on $\hat{\mathcal{A}}$

10. Let  $\mathcal{A}$  be a complex unital commutative Banach algebra, and  $a \in \mathcal{A}$ . Let  $f \in H(\sigma(a))$ . Prove that there exists  $b \in \mathcal{A}$  such that  $\hat{b} = f \circ \hat{a}$ . ( $\hat{a}$  denotes the Gelfand transform of  $a$ .) In particular, if  $\hat{a} \neq 0$ , there exists  $b \in \mathcal{A}$  such that  $\hat{b} = 1/\hat{a}$ . (This is *Wiener's theorem*.) Hint: Use the analytic operational calculus.

## Polar decomposition

11. Let  $X$  be a Hilbert space, and let  $T \in B(X)$  be non-singular. Prove that there exist a unique pair of operators  $S, U$  such that  $S$  is non-singular and positive,  $U$  is unitary, and  $T = US$ . If  $T$  is normal, the operators  $S, U$  commute with each other and with  $T$ . Hint: assuming the result, find out how to define  $S$  and  $U|_{SX}$ ; verify that  $U$  is isometric on  $SX$ , etc.

## Cayley transform

12. Let  $X$  be a Hilbert space, and let  $T \in B(X)$  be selfadjoint. Prove:
  - (a) The operator  $V := (T + iI)(T - iI)^{-1}$  (called the *Cayley transform* of  $T$ ) is unitary and  $1 \notin \sigma(V)$ .
  - (b) Conversely, every unitary operator  $V$  such that  $1 \notin \sigma(V)$  is the Cayley transform of some selfadjoint operator  $T \in B(X)$ .

## Riemann integrals of operator functions

13. Let  $X$  be a Banach space, and let  $T(\cdot) : [a, b] \rightarrow B(X)$  be *strongly continuous* (that is, continuous with respect to the s.o.t. on  $B(X)$ ). Prove:
- (a)  $\|T(\cdot)\|$  is bounded and lower semi-continuous (l.s.c.) (cf. Exercise 6, Chapter 3).
  - (b) For each  $x \in X$ , the Riemann integral  $\int_a^b T(t)x dt$  is a well-defined element of  $X$  with norm  $\leq \int_a^b \|T(t)\| dt \|x\|$ . Therefore the operator  $\int_a^b T(t) dt$  defined by  $(\int_a^b T(t) dt)x = \int_a^b T(t)x dt$  has norm  $\leq \int_a^b \|T(t)\| dt$ . For each  $S \in B(X)$ ,  $ST(\cdot)$  and  $T(\cdot)S$  are strongly continuous on  $[a, b]$ , and  $S \int_a^b T(t) dt = \int_a^b ST(t) dt$ ;  $(\int_a^b T(t) dt)S = \int_a^b T(t)S dt$ .
  - (c)  $(\int_a^t T(s) ds)'(c) = T(c)$  (derivative in the s.o.t.).
  - (d) If  $T(\cdot) = V'(\cdot)$  (derivative in the s.o.t.) for some operator function  $V$ , then  $\int_a^b T(t) dt = V(b) - V(a)$ .
  - (e) If  $T(\cdot) : [a, \infty) \rightarrow B(X)$  is strongly continuous and  $\int_a^\infty \|T(t)\| dt < \infty$ , then  $\lim_{b \rightarrow \infty} \int_a^b T(t) dt := \int_a^\infty T(t) dt$  exists in the norm topology of  $B(X)$ , and  $\|\int_a^\infty T(t) dt\| \leq \int_a^\infty \|T(t)\| dt$ . (Note that  $\|T(\cdot)\|$  is l.s.c. by Part (a), and the integral on the right makes sense as the integral of a non-negative Borel function.)

## Semigroups of operators

14. Let  $X$  be a Banach space, and let  $T(\cdot) : [0, \infty) \rightarrow B(X)$  be such that  $T(t+s) = T(t)T(s)$  for all  $t, s \geq 0$  and  $T(0) = I$ . (Such a function is called a *semigroup of operators*.) Assume  $T(\cdot)$  is (right) continuous at 0 in the s.o.t. (briefly,  $T(\cdot)$  is a  $C_0$ -semigroup). Prove:
- (a)  $T(\cdot)$  is right continuous on  $[0, \infty)$ , in the s.o.t.
  - (b) Let  $c_n := \sup\{\|T(t)\|; 0 \leq t \leq 1/n\}$ . Then there exists  $n$  such that  $c_n < \infty$ . (Fix such an  $n$  and let  $c := c_n (\geq 1)$ .) Hint: the uniform boundedness theorem.
  - (c) With  $n$  and  $c$  as in Part (b),  $\|T(t)\| \leq Me^{at}$  on  $[0, \infty)$ , where  $M := c^n (\geq 1)$  and  $a := \log M (\geq 0)$ .
  - (d)  $T(\cdot)$  is strongly continuous on  $[0, \infty)$ .
  - (e) Let  $V(t) := \int_0^t T(s) ds$ . Then

$$T(h)V(t) = V(t+h) - V(h) \quad (h, t > 0).$$

Conclude that  $(1/h)(T(h) - I)V(t) \rightarrow T(t) - I$  in the s.o.t., as  $h \rightarrow 0+$  (i.e. the strong right derivative of  $T(\cdot)V(t)$  at 0 exists and equals  $T(t) - I$ , for each  $t > 0$ ). Hint: Exercise 13, Part (c).

- (f) Let  $\omega := \inf_{t>0} t^{-1} \log \|T(t)\|$ . Then  $\omega = \lim_{t \rightarrow \infty} t^{-1} \log \|T(t)\| (< \infty)$  (cf. Part (c)). Hint: fix  $s > 0$  and  $r > s^{-1} \log \|T(s)\|$ . Given  $t > 0$ , let  $n = [t/s]$ . Then  $t^{-1} \log \|T(t)\| < rns/t + t^{-1} \sup_{[0,s]} \log \|T(\cdot)\|$ . ( $\omega$  is called the *type* of the semigroup  $T(\cdot)$ .)
- (g) Let  $\omega$  be the type of  $T(\cdot)$ . Then the spectral radius of  $T(t)$  is  $e^{\omega t}$ , for each  $t \geq 0$ .

# Unbounded operators

## 10.1 Basics

In this chapter, we deal with (linear) operators  $T$  with domain  $D(T)$  and range  $R(T)$  in a Banach space  $X$ ;  $D(T)$  and  $R(T)$  are (linear) subspaces of  $X$ . The operators  $S, T$  are *equal* if  $D(S) = D(T)$  and  $Sx = Tx$  for all  $x$  in the (common) domain of  $S$  and  $T$ . If  $S, T$  are operators such that  $D(S) \subset D(T)$  and  $T|_{D(S)} = S$ , we say that  $T$  is an *extension* of  $S$  (notation:  $S \subset T$ ).

The algebraic operations between unbounded operators are defined with the obvious restrictions on domains. Both sum and product are associative, but the distributive laws take the form

$$AB + AC \subset A(B + C); \quad (A + B)C = AC + BC.$$

The *graph* of  $T$  is the subspace of  $X \times X$  given by

$$\Gamma(T) := \{[x, Tx]; x \in D(T)\}.$$

The operator  $T$  is *closed* if  $\Gamma(T)$  is a *closed* subspace of  $X \times X$ .

A convenient elementary criterion for  $T$  being closed is the following condition:

If  $\{x_n\} \subset D(T)$  is such that  $x_n \rightarrow x$  and  $Tx_n \rightarrow y$ , then  $x \in D(T)$  and  $Tx = y$ .

Clearly, if  $D(T)$  is closed and  $T$  is continuous on  $D(T)$ , then  $T$  is a closed operator. In particular, every  $T \in B(X)$  is closed. Conversely, if  $T$  is a closed operator with closed domain (hence a Banach space!), then  $T$  is continuous on  $D(T)$ , by the Closed Graph Theorem. Also if  $T$  is closed and continuous (on its domain), then it has a closed domain.

If  $B \in B(X)$  and  $T$  is closed, then  $T + B$  and  $TB$  (with their ‘maximal domains’  $D(T)$  and  $\{x \in X; Bx \in D(T)\}$ , respectively) are closed operators. In particular, the operators  $\lambda I - T$  and  $\lambda T$  are closed, for any  $\lambda \in \mathbb{C}$ .

If  $B \in B(X)$  is non-singular and  $T$  is closed, then  $BT$  (with domain  $D(T)$ ) is closed.

Usually, the norm taken on  $X \times X$  is  $\|[x, y]\| = \|[x, y]\|_1 := \|x\| + \|y\|$ , or in case  $X$  is a Hilbert space,  $\|[x, y]\| = \|[x, y]\|_2 := \sqrt{\|x\|^2 + \|y\|^2}$ . These norms are equivalent, since

$$\|[x, y]\|_2 \leq \|[x, y]\|_1 \leq \sqrt{2}\|[x, y]\|_2.$$

If  $X$  is a Hilbert space, the space  $X \times X$  (also denoted  $X \oplus X$ ) is a Hilbert space with the inner product

$$([x, y], [u, v]) := (x, u) + (y, v),$$

and the norm induced by this inner product is indeed  $\|[x, y]\| := \|[x, y]\|_2$ .

The *graph norm* on  $D(T)$  is defined by

$$\|x\|_T := \|[x, Tx]\| \quad (x \in D(T)).$$

We shall denote by  $[D(T)]$  the space  $D(T)$  with the graph norm. The space  $[D(T)]$  is *complete* iff  $T$  is a closed operator.

If the operator  $S$  has a closed extension  $T$ , it clearly satisfies the property

$$\text{If } \{x_n\} \subset D(S) \text{ is such that } x_n \rightarrow 0 \text{ and } Sx_n \rightarrow y, \text{ then } y = 0.$$

An operator  $S$  with this property is said to be *closable*. Conversely, if  $S$  is closable, then the  $X \times X$ -closure of its graph,  $\overline{\Gamma(S)}$ , is the graph of a (necessarily closed) operator  $\bar{S}$ , called the *closure* of  $S$ . Indeed, if  $[x, y], [x', y'] \in \overline{\Gamma(S)}$ , there exist sequences  $\{x_n, Sx_n\}$  and  $\{x'_n, Sx'_n\}$  in  $\Gamma(S)$  converging respectively to  $[x, y]$  and  $[x', y']$  in  $X \times X$ . Then  $x_n - x'_n \in D(S) \rightarrow 0$  and  $S(x_n - x'_n) \rightarrow y - y'$ . Therefore,  $y - y' = 0$  since  $S$  is closable. Consequently the map  $\bar{S} : x \rightarrow y$  is well defined, clearly linear, and by definition,

$$\Gamma(\bar{S}) = \overline{\Gamma(S)}.$$

Hence the closable operator  $S$  has the (minimal) closed extension  $\bar{S}$ .

By definition,  $D(\bar{S}) = \{x \in X; \exists \{x_n\} \subset D(S) \text{ such that } x_n \rightarrow x \text{ and } \exists \lim Sx_n\}$  and  $\bar{S}x$  is equal to the above limit for  $x \in D(\bar{S})$ .

If  $T$  is one-to-one, the inverse operator  $T^{-1}$  with domain  $R(T)$  and range  $D(T)$  has the graph

$$\Gamma(T^{-1}) = J\Gamma(T),$$

where  $J$  is the isometric automorphism of  $X \times Y$  given by  $J[x, y] = [y, x]$ . Therefore  $T$  is closed iff  $T^{-1}$  is closed. In particular, if  $T^{-1} \in B(X)$ , then  $T$  is closed.

The *resolvent set*  $\rho(T)$  of  $T$  is the set of all  $\lambda \in \mathbb{C}$  such that  $\lambda I - T$  has an inverse in  $B(X)$ ; the inverse operator is called the *resolvent* of  $T$ , and is denoted by  $R(\lambda; T)$  (or  $R(\lambda)$ , when  $T$  is understood). If  $\rho(T) \neq \emptyset$ , and  $\lambda$  is any point in  $\rho(T)$ , then  $R(\lambda)^{-1}$  (with domain  $D(T)$ ) is closed, and therefore

$T = \lambda I - R(\lambda; T)^{-1}$  is closed. On the other hand, if  $T$  is closed and  $\lambda$  is such that  $\lambda I - T$  is bijective, then  $\lambda \in \rho(T)$  (because  $(\lambda I - T)^{-1}$  is closed and everywhere defined, hence belongs to  $B(X)$ , by the Closed Graph theorem).

By definition,  $TR(\lambda) = \lambda R(\lambda) - I \in B(X)$ , while  $R(\lambda)T = (\lambda R(\lambda) - I)|_{D(T)}$ .

The complement of  $\rho(T)$  in  $\mathbb{C}$  is the *spectrum* of  $T$ ,  $\sigma(T)$ . By the preceding remark, the spectrum of the closed operator  $T$  is the disjoint union of the following sets:

the *point spectrum* of  $T$ ,  $\sigma_p(T)$ , which consists of all scalars  $\lambda$  for which  $\lambda I - T$  is not one-to-one;

the *continuous spectrum* of  $T$ ,  $\sigma_c(T)$ , which consists of all  $\lambda$  for which  $\lambda I - T$  is one-to-one but *not* onto, and its range is dense in  $X$ ;

the *residual spectrum* of  $T$ ,  $\sigma_r(T)$ , which consists of all  $\lambda$  for which  $\lambda I - T$  is one-to-one, and its range is *not dense* in  $X$ .

**Theorem 10.1.** *Let  $T$  be any (unbounded) operator. Then  $\rho(T)$  is open, and  $R(\cdot)$  is analytic on  $\rho(T)$  and satisfies the ‘resolvent identity’*

$$R(\lambda) - R(\mu) = (\mu - \lambda)R(\lambda)R(\mu) \quad (\lambda, \mu \in \rho(T)).$$

*In particular,  $R(\lambda)$  commutes with  $R(\mu)$ .*

*Moreover,  $\|R(\lambda)\| \geq 1/d(\lambda, \sigma(T))$ .*

**Proof.** We assume without loss of generality that the resolvent set is non-empty. Let then  $\lambda \in \rho(T)$ , and denote  $r = \|R(\lambda)\|^{-1}$ . We wish to prove that the disc  $B(\lambda, r)$  is contained in  $\rho(T)$ . This will imply that  $\rho(T)$  is open and  $d(\lambda, \sigma(T)) \geq r$  (i.e.  $\|R(\lambda)\| \geq 1/d(\lambda, \sigma(T))$ ).

For  $\mu \in B(\lambda, r)$ , the series

$$S(\mu) = \sum_{k=0}^{\infty} [(\lambda - \mu)R(\lambda)]^k$$

converges in  $B(X)$ , commutes with  $R(\lambda)$ , and satisfies the identity

$$(\lambda - \mu)R(\lambda)S(\mu) = S(\mu) - I.$$

For  $x \in D(T)$ ,

$$S(\mu)R(\lambda)(\mu I - T)x = S(\mu)R(\lambda)[(\lambda I - T) - (\lambda - \mu)I]x = S(\mu)x - [S(\mu) - I]x = x$$

by the above identity, and similarly, for all  $x \in X$ ,

$$(\mu I - T)R(\lambda)S(\mu)x = [(\lambda I - T) - (\lambda - \mu)I]R(\lambda)S(\mu)x = S(\mu)x - [S(\mu) - I]x = x.$$

This shows that  $\mu \in \rho(T)$  and  $R(\mu) = R(\lambda)S(\mu)$  for all  $\mu \in B(\lambda, r)$ .

In particular,  $R(\cdot)$  is analytic in  $\rho(T)$  (since it is locally the sum of a  $B(X)$ -convergent power series).



Finally, for  $\lambda, \mu \in \rho(T)$ , we have on  $X$ :

$$\begin{aligned} & (\lambda I - T)[R(\lambda) - R(\mu) - (\mu - \lambda)R(\lambda)R(\mu)] \\ &= I - [(\lambda - \mu)I + (\mu I - T)]R(\mu) - (\mu - \lambda)R(\mu) \\ &= I - (\lambda - \mu)R(\mu) - I + (\lambda - \mu)R(\mu) = 0. \end{aligned}$$

Since  $\lambda I - T$  is one-to-one, the resolvent identity follows.  $\square$

**Theorem 10.2.** *Let  $T$  be an unbounded operator in the Banach space  $X$ , with  $\rho(T) \neq \emptyset$ . Fix  $\alpha \in \rho(T)$  and let  $h(\lambda) = 1/(\alpha - \lambda)$ . Then  $h$  maps  $\sigma(T) \cup \{\infty\}$  onto  $\sigma(R(\alpha))$ .*

**Proof.** (In order to reduce the number of brackets in the following formulas, we shall write  $R_\lambda$  instead of  $R(\lambda)$ .)

Taking complements in  $\mathbb{C} \cup \{\infty\}$ , we must show that  $h$  maps  $\rho(T)$  onto  $\rho(R_\alpha) \cup \{\infty\}$ . Since  $h(\alpha) = \infty$ , we consider  $\lambda \neq \alpha$  in  $\rho(T)$ , and define

$$V := (\alpha - \lambda)[I + (\alpha - \lambda)R_\lambda].$$

Then  $V$  commutes with  $h(\lambda)I - R_\alpha$  and by the resolvent identity (cf. Theorem 10.1)

$$[h(\lambda)I - R_\alpha]V = I + (\alpha - \lambda)[R_\lambda - R_\alpha - (\alpha - \lambda)R_\alpha R_\lambda] = I.$$

This shows that  $h(\lambda) \in \rho(R_\alpha)$  and  $R(h(\lambda); R_\alpha) = V$ . Hence  $h$  maps  $\rho(T)$  into  $\rho(R_\alpha) \cup \{\infty\}$ .

Next, let  $\mu \in \rho(R_\alpha)$ . If  $\mu = 0$ ,  $T = \alpha I - (\alpha I - T) = \alpha I - R_\alpha^{-1} \in B(X)$ , contrary to our hypothesis. Hence  $\mu \neq 0$ , and let then  $\lambda = \alpha - 1/\mu$  (so that  $h(\lambda) = \mu$ ). Let

$$W := \mu R_\alpha R(\mu; R_\alpha).$$

Then  $W$  commutes with  $\lambda I - T$  and

$$\begin{aligned} (\lambda I - T)W &= [(\lambda - \alpha)I + (\alpha I - T)]W = \mu[(\lambda - \alpha)R_\alpha + I]R(\mu; R_\alpha) \\ &= (\mu I - R_\alpha)R(\mu; R_\alpha) = I. \end{aligned}$$

Thus  $\lambda \in \rho(T)$ , and we conclude that  $h$  maps  $\rho(T)$  onto  $\rho(R_\alpha) \cup \{\infty\}$ .  $\square$

## 10.2 The Hilbert adjoint

**Terminology 10.3.** Let  $T$  be an operator with *dense domain*  $D(T)$  in the *Hilbert space*  $X$ . For  $y \in X$  fixed, consider the function

$$\phi(x) = (Tx, y) \quad (x \in D(T)). \quad (1)$$

If  $\phi$  is *continuous*, it has a unique extension as a continuous linear functional on  $X$  (since  $D(T)$  is dense in  $X$ ), and there exists therefore a unique  $z \in X$  such that

$$\phi(x) = (x, z) \quad (x \in D(T)). \quad (2)$$

(Conversely, if there exists  $z \in X$  such that (2) holds, then  $\phi$  is continuous on  $D(T)$ .)

Let  $D(T^*)$  denote the subspace of all  $y$  for which  $\phi$  is continuous on  $D(T)$  (equivalently, for which  $\phi = (\cdot, z)$  for some  $z \in X$ ). Define  $T^* : D(T^*) \rightarrow X$  by  $T^*y = z$  (the map  $T^*$  is well defined, by the uniqueness of  $z$  for given  $y$ ). The defining identity for  $T^*$  is then

$$(Tx, y) = (x, T^*y) \quad (x \in D(T), y \in D(T^*)). \quad (3)$$

It follows clearly from (3) that  $T^*$  (with domain  $D(T^*)$ ) is a linear operator. It is called the *adjoint operator* of  $T$ .

By (2),  $[y, z] \in \Gamma(T^*)$  iff  $(Tx, y) = (x, z)$  for all  $x \in D(T)$ , that is, iff

$$([Tx, -x], [y, z]) = 0 \quad \text{for all } x \in D(T).$$

Consider the isometric automorphism of  $X \times X$  defined by

$$Q[x, y] = [y, -x].$$

The preceding statement means that  $[y, z] \in \Gamma(T^*)$  iff  $[y, z]$  is orthogonal to  $Q\Gamma(T)$  in  $X \times X$ . Hence

$$\Gamma(T^*) = (Q\Gamma(T))^\perp. \quad (4)$$

In particular, it follows from (4) that  $T^*$  is *closed*.

One verifies easily that if  $B \in B(X)$ , then

$$(T + B)^* = T^* + B^* \quad \text{and} \quad (BT)^* = T^*B^*.$$

It follows in particular (or directly) that  $(\lambda T)^* = [(\lambda I)T]^* = \bar{\lambda}T^*$ .

If  $T = T^*$ , the operator  $T$  is called a *selfadjoint* operator. Since  $T^*$  is closed, a selfadjoint operator is necessarily *closed and densely defined*. An everywhere defined selfadjoint operator is necessarily bounded, by the Closed Graph theorem.

The operator  $T$  is *symmetric* if

$$(Tx, y) = (x, Ty) \quad (x, y \in D(T)). \quad (5)$$

If  $T$  is densely defined (so that  $T^*$  exists), Condition (5) is equivalent to  $T \subset T^*$ . If  $T$  is everywhere defined, it is symmetric iff it is selfadjoint. Therefore, a symmetric everywhere defined operator is a bounded selfadjoint operator.

If  $T$  is one-to-one with *domain and range both dense in  $X$* , the adjoint operators  $T^*$  and  $(T^{-1})^*$  both exist. If  $T^*y = 0$  for some  $y \in D(T^*)$ , then for all  $x \in D(T)$

$$(Tx, y) = (x, T^*y) = (x, 0) = 0, \quad (6)$$

and therefore  $y = 0$  since  $R(T)$  is dense. Thus  $T^*$  is one-to-one, and  $(T^*)^{-1}$  exists. By (4)

$$\begin{aligned} \Gamma((T^{-1})^*) &= [Q\Gamma(T^{-1})]^\perp = [QJT(T)]^\perp \\ &= [-JQ\Gamma(T)]^\perp = J[Q\Gamma(T)]^\perp = J\Gamma(T^*) = \Gamma((T^*)^{-1}), \end{aligned}$$

since  $(JA)^\perp = JA^\perp$  for any  $A \subset X \times X$ . Therefore

$$(T^{-1})^* = (T^*)^{-1}. \quad (7)$$

It follows that if  $T$  is densely defined then

$$R(\lambda; T)^* = R(\bar{\lambda}; T^*) \quad (\lambda \in \rho(T)). \quad (8)$$

In particular, if  $T$  is *selfadjoint*,

$$R(\lambda, T)^* = R(\bar{\lambda}; T), \quad (9)$$

and therefore  $R(\lambda; T)$  is a *bounded normal operator* for each  $\lambda \in \rho(T)$  (cf. Theorem 10.1).

Note that (6) also shows that for any  $T$  with dense domain,  $\ker(T^*) \subset R(T)^\perp$ . On the other hand, if  $y \in R(T)^\perp$ , then  $(Tx, y) = 0$  for all  $x \in D(T)$ . In particular, the function  $x \rightarrow (Tx, y)$  is continuous on  $D(T)$ , so that  $y \in D(T^*)$ , and  $(x, T^*y) = (Tx, y) = 0$  for all  $x \in D(T)$ . Since  $D(T)$  is dense, it follows that  $T^*y = 0$ , and we conclude that

$$\ker(T^*) = R(T)^\perp. \quad (10)$$

**Theorem 10.4.** *Let  $T$  be a symmetric operator. Then for any non-real  $\lambda \in \mathbb{C}$ ,  $\lambda I - T$  is one-to-one and*

$$\|(\lambda I - T)^{-1}y\| \leq |\Im \lambda|^{-1}\|y\| \quad (y \in R(\lambda I - T)). \quad (11)$$

*If  $T$  is closed, the range  $R(\lambda I - T)$  is closed, and coincides with  $X$  if  $T$  is selfadjoint. In the latter case, every non-real  $\lambda$  is in  $\rho(T)$ ,  $R(\lambda; T)$  is a bounded normal operator, and*

$$\|R(\lambda; T)\| \leq 1/|\Im \lambda|. \quad (12)$$

**Proof.** If  $T$  is symmetric,  $(Tx, x)$  is real for all  $x \in D(T)$  (since  $\overline{(Tx, x)} = (x, Tx) = (Tx, x)$ ). Therefore  $(Tx, i\beta x)$  is pure imaginary for  $\beta \in \mathbb{R}$ . Since  $\alpha I - T$  is symmetric for any  $\alpha \in \mathbb{R}$ ,  $((\alpha I - T)x, i\beta x)$  is pure imaginary for  $\alpha, \beta \in \mathbb{R}$ . Hence, for all  $x \in D(T)$  and  $\lambda = \alpha + i\beta$ ,

$$\begin{aligned} \|(\lambda I - T)x\|^2 &= \|(\alpha I - T)x + i\beta x\|^2 \\ &= \|(\alpha I - T)x\|^2 + 2\Re(\alpha I - T)x, i\beta x) + \beta^2\|x\|^2 \\ &= \|(\alpha I - T)x\|^2 + \beta^2\|x\|^2 \geq \beta^2\|x\|^2. \end{aligned}$$

Hence

$$\|(\lambda I - T)x\| \geq |\Im \lambda| \|x\|. \quad (13)$$

If  $\lambda$  is non-real, it follows from (13) that  $\lambda I - T$  is one-to-one, and (11) holds.

If  $T$  is also closed,  $(\lambda I - T)^{-1}$  is closed and continuous on its domain  $R(\lambda I - T)$  (by (11)), and therefore this domain is closed (for non-real  $\lambda$ ).

If  $T$  is selfadjoint,

$$R(\lambda I - T)^\perp = \ker((\lambda I - T)^*) = \ker(\bar{\lambda} I - T) = \{0\}$$

since  $\bar{\lambda}$  is non-real. Therefore  $(\lambda I - T)^{-1}$  is everywhere defined, with operator norm  $\leq 1/|\Im \lambda|$ , by (11). This shows that every non-real  $\lambda$  is in  $\rho(T)$ , that is,  $\sigma(T) \subset \mathbb{R}$ .  $\square$

### 10.3 The spectral theorem for unbounded selfadjoint operators

**Theorem 10.5.** *Let  $T$  be a selfadjoint operator on the Hilbert space  $X$ . Then there exists a unique regular selfadjoint spectral measure  $E$  on  $\mathcal{B} := \mathcal{B}(\mathbb{C})$ , supported by  $\sigma(T) \subset \mathbb{R}$ , such that*

$$\begin{aligned} D(T) &= \left\{ x \in X; \int_{\sigma(T)} \lambda^2 d\|E(\lambda)x\|^2 < \infty \right\} \\ &= \left\{ x \in X; \lim_{n \rightarrow \infty} \int_{-n}^n \lambda dE(\lambda)x \text{ exists} \right\} \end{aligned} \quad (1)$$

and

$$Tx = \lim_{n \rightarrow \infty} \int_{-n}^n \lambda dE(\lambda)x \quad (x \in D(T)). \quad (2)$$

(The limits above are strong limits in  $X$ .)

**Proof.** By Theorem 10.4, every non-real  $\alpha$  (to be fixed from now on) is in  $\rho(T)$ , and  $R_\alpha := R(\alpha; T)$  is a bounded normal operator. Let  $F$  be its resolution of the identity, and define

$$E(\delta) = F(h(\delta)) \quad (\delta \in \mathcal{B}), \quad (3)$$

where  $h$  is as in Theorem 10.2.

By Theorem 9.8 (Part 5),  $F(\{0\})X = \ker R_\alpha = \{0\}$ , and therefore

$$E(\mathbb{C}) = F(\{0\}^c) = I - F(\{0\}) = I. \quad (4)$$

We conclude that  $E$  is a selfadjoint regular spectral measure from the corresponding properties of  $F$ .

By Theorem 10.2,

$$E(\sigma(T)) := F(h(\sigma(T))) = F(\sigma(R_\alpha)) - F(\{0\}) = I,$$

hence  $E$  is supported by  $\sigma(T)$  (by (4)).

Denote the sets in (1) by  $D_0$  and  $D_1$ .

If  $\delta \in \mathcal{B}$  is *bounded*, then for all  $x \in X$ ,

$$\int_{\sigma(T)} \lambda^2 d\|E(\lambda)E(\delta)x\|^2 = \int_{\delta \cap \sigma(T)} \lambda^2 d\|E(\lambda)x\|^2 < \infty, \quad (5)$$

since  $\lambda^2$  is bounded on  $\delta \cap \sigma(T)$ . Hence  $E(\delta)X \subset D_0$ . Moreover, by Theorem 9.6, the last integral in (5) equals  $\|\int_{\delta \cap \sigma(T)} \lambda dE(\lambda)x\|^2$ . For positive integers  $n > m$ ,

take  $\delta = [-n, -m] \cup [m, n]$ . Then

$$\begin{aligned} & \left\| \int_{-n}^n \lambda dE(\lambda)x - \int_{-m}^m \lambda dE(\lambda)x \right\|^2 \\ &= \int_{-n}^n \lambda^2 d\|E(\lambda)x\|^2 - \int_{-m}^m \lambda^2 d\|E(\lambda)x\|^2. \end{aligned}$$

It follows that  $D_0 = D_1$ .

Let  $x \in D(T)$ . We may then write  $x = R_\alpha y$  for a unique  $y \in X$ , and therefore

$$\begin{aligned} \int_{-n}^n \lambda dE(\lambda)x &= \int_{-n}^n \lambda dE(\lambda) \int_{\mathbb{C}} \mu dF(\mu)y \\ &= \int_{-n}^n \lambda dE(\lambda) \int_{\mathbb{R}} h(\lambda) dE(\lambda)y \\ &= \int_{-n}^n \lambda h(\lambda) dE(\lambda)y \rightarrow \int_{\mathbb{R}} \lambda h(\lambda) dE(\lambda)y. \end{aligned}$$

(The limit exists in  $X$  because  $\lambda h(\lambda)$  is *bounded*.) Thus  $x \in D_0$ , and we proved that  $D(T) \subset D_0$ .

Next, let  $x \in D_0 (= D_1)$ , and denote  $z = \lim_n \int_{-n}^n \lambda dE(\lambda)x$ . Consider the sequence  $x_n := E([-n, n])x$ . Then  $x_n \rightarrow x$  in  $X$ ,

$$x_n = R_\alpha \int_{-n}^n (\alpha - \lambda) dE(\lambda)x \in R_\alpha X = D(T), \quad (6)$$

and by (6)

$$(\alpha I - T)x_n = \int_{-n}^n (\alpha - \lambda) dE(\lambda)x \rightarrow \alpha x - z.$$

Since  $\alpha I - T$  (with domain  $D(T)$ ) is closed, it follows that  $x \in D(T)$  and  $(\alpha I - T)x = \alpha x - z$ . Hence  $D_0 \subset D(T)$  (and so  $D(T) = D_0$ ), and (2) is valid.

For each bounded  $\delta \in \mathcal{B}$ , the restriction of  $T$  to the reducing subspace  $E(\delta)X$  is the *bounded* selfadjoint operator  $\int_{\delta} \lambda dE(\lambda)$ . By the uniqueness of the resolution of the identity for bounded selfadjoint operators,  $E$  is uniquely determined on the bounded Borel sets, and therefore on all Borel sets, by Theorem 9.6 (Part 2).  $\square$

## 10.4 The operational calculus for unbounded selfadjoint operators

The unique spectral measure  $E$  of Theorem 10.5 is called the *resolution of the identity* for  $T$ .

The map  $f \rightarrow f(T) := \int_{\mathbb{R}} f dE$  of  $\mathbb{B} := \mathbb{B}(\mathbb{R})$  into  $B(X)$  is a norm-decreasing  $*$ -representation of  $\mathbb{B}$  on  $X$  (cf. Theorem 9.1). The map is extended to arbitrary complex Borel functions  $f$  on  $\mathbb{R}$  as follows. Let  $\chi_n$  be the indicator of the set  $[|f| \leq n]$ , and consider the ‘truncations’  $f_n := f\chi_n \in \mathbb{B}$ ,  $n \in \mathbb{N}$ . The operator

$f(T)$  has domain  $D(f(T))$  equal to the set of all  $x \in X$  for which the strong limit  $\lim_n f_n(T)x$  exists, and  $f(T)x$  is defined as this limit for  $x \in D(f(T))$ .

Note that if  $f$  is bounded, then  $f_n = f$  for all  $n \geq \|f\|_u$ , and therefore the new definition of  $f(T)$  coincides with the previous one for  $f \in \mathbb{B}$ . In particular,  $f(T) \in B(X)$ . For general Borel functions, we have the following

**Theorem 10.6.** *Let  $T$  be an unbounded selfadjoint operator on the Hilbert space  $X$ , and let  $E$  be its resolution of the identity. For  $f : \mathbb{R} \rightarrow \mathbb{C}$  Borel, let  $f(T)$  be defined as above. Then*

- (a)  $D(f(T)) = \{x \in X; \int_{\mathbb{R}} |f|^2 d\|E(\cdot)x\|^2 < \infty\}$ ;
- (b)  $f(T)$  is a closed densely defined operator;
- (c)  $f(T)^* = \bar{f}(T)$ .

**Proof.** (a) Let  $D$  denote the set on the right-hand side of (a).

Since  $f_n(x) = f(x)$  for all  $n \geq |f(x)|$ ,  $f_n \rightarrow f$  pointwise. If  $x \in D$ ,

$$|f_n - f_m|^2 \leq 4|f|^2 \in L^1(\|E(\cdot)x\|^2)$$

and  $|f_n - f_m|^2 \rightarrow 0$  pointwise when  $n, m \rightarrow \infty$ . Therefore, by Theorem 9.6 and Lebesgue's dominated convergence theorem,

$$\|f_n(T)x - f_m(T)x\|^2 = \|(f_n - f_m)(T)x\|^2 = \int_{\mathbb{R}} |f_n - f_m|^2 d\|E(\cdot)x\|^2 \rightarrow 0$$

as  $n, m \rightarrow \infty$ . Hence  $x \in D(f(T))$ . On the other hand, if  $x \in D(f(T))$ , we have by Fatou's lemma

$$\begin{aligned} \int_{\mathbb{R}} |f|^2 d\|E(\cdot)x\|^2 &\leq \liminf_n \int_{\mathbb{R}} |f_n|^2 d\|E(\cdot)x\|^2 \\ &= \liminf_n \|f_n(T)x\|^2 = \|f(T)x\|^2 < \infty, \end{aligned}$$

that is,  $x \in D$ , and (a) has been verified.

(b) Let  $x \in X$ , and  $\delta_n = [|f| \leq n]$ . Clearly  $\delta_{n+1}^c \subset \delta_n^c$  and  $\bigcap \delta_n^c = \emptyset$ . Since  $\|E(\cdot)x\|^2$  is a finite positive measure,

$$\lim_n \|E(\delta_n^c)x\|^2 = \left\| E\left(\bigcap \delta_n^c\right)x \right\|^2 = 0,$$

that is,

$$\lim \|x - E(\delta_n)x\| = 0 \quad (x \in X). \quad (1)$$

Now

$$\int_{\mathbb{R}} |f|^2 d\|E(\cdot)E(\delta_n)x\|^2 = \int_{\delta_n} |f|^2 d\|E(\cdot)x\|^2 \leq n^2 \|x\|^2 < \infty,$$

that is,  $E(\delta_n)x \in D(f(T))$ , by Part (a). This proves that  $D(f(T))$  is dense in  $X$ .

Fix  $x \in D(f(T))$  and  $m \in \mathbb{N}$ . Since  $E(\delta_m)$  is a bounded operator, we have by the operational calculus for bounded Borel functions and the relation  $\chi_{\delta_m} f_n = f_m$  for all  $n \geq m$ ,

$$E(\delta_m)f(T)x = \lim_n E(\delta_m)f_n(T)x = \lim_n f_m(T)x = f_m(T)x. \quad (2)$$

Similarly

$$f(T)E(\delta_m)x = f_m(T)x \quad (x \in X). \quad (3)$$

In order to show that  $f(T)$  is closed, let  $\{x_n\}$  be any sequence in  $D(f(T))$  such that  $x_n \rightarrow x$  and  $f(T)x_n \rightarrow y$ . By (2) applied to  $x_n \in D(f(T))$ ,

$$E(\delta_m)y = \lim_n E(\delta_m)f(T)x_n = \lim_n f_m(T)x_n = f_m(T)x,$$

since  $f_m(T) \in B(X)$ . Letting  $m \rightarrow \infty$ , we see that  $f_m(T)x \rightarrow y$  (by (1)). Hence  $x \in D(f(T))$ , and  $f(T)x := \lim_m f_m(T)x = y$ . This proves (b).

(c) By the operational calculus for *bounded* Borel functions,  $(f_n(T)x, y) = (x, \bar{f}_n(T)y)$  for all  $x, y \in X$ . When  $x, y \in D(f(T)) = D(\bar{f}(T))$  (cf. (a)), letting  $n \rightarrow \infty$  implies the relation  $(f(T)x, y) = (x, \bar{f}(T)y)$ . Hence  $\bar{f}(T) \subset f(T)^*$ . On the other hand, if  $y \in D(f(T)^*)$ , we have by (3) (for all  $x \in X$ )

$$\begin{aligned} (x, \bar{f}_m(T)y) &= (f_m(T)x, y) = (f(T)E(\delta_m)x, y) \\ &= (E(\delta_m)x, f(T)^*y) = (x, E(\delta_m)f(T)^*y), \end{aligned}$$

that is,

$$\bar{f}_m(T)y = E(\delta_m)f(T)^*y.$$

The right-hand side converges to  $f(T)^*y$  when  $m \rightarrow \infty$ . Hence  $y \in D(\bar{f}(T))$ , and (c) follows.  $\square$

## 10.5 The semi-simplicity space for unbounded operators in Banach space

Let  $T$  be an *unbounded* operator with *real spectrum* on the Banach space  $X$ . Its *Cayley transform*

$$V := (iI - T)(iI + T)^{-1} = -2iR(-i; T) - I$$

belongs to  $B(X)$ .

By Theorem 10.2 with  $\alpha = -i$  and the corresponding  $h$ ,

$$\sigma(R(-i; T)) = h(\sigma(T) \cup \{\infty\}),$$

where  $\infty$  denotes the point at infinity of the Riemann sphere. Therefore

$$\begin{aligned} \sigma(V) &= -2ih(\sigma(T) \cup \{\infty\}) - 1 \subset -2ih(\mathbb{R} \cup \{\infty\}) - 1 \\ &= \left\{ \frac{i - \lambda}{i + \lambda}; \lambda \in \mathbb{R} \right\} \cup \{-1\} \subset \Gamma, \end{aligned}$$

where  $\Gamma$  denotes the unit circle.

**Definition 10.7.** Let  $T$  be an unbounded operator with real spectrum, and let  $V$  be its Cayley transform. The semi-simplicity space for the unbounded

operator  $T$  is defined as the semi-simplicity space  $Z$  for the bounded operator  $V$  with spectrum in  $\Gamma$  (cf. Remark 9.13, (4)).

The function

$$\phi(s) := \frac{i - s}{i + s}$$

is a homeomorphism of  $\bar{\mathbb{R}} := \mathbb{R} \cup \{\infty\}$  onto  $\Gamma$ , with the inverse  $\phi^{-1}(\lambda) = i(1 - \lambda)/(1 + \lambda)$ .

For any  $g \in C(\bar{\mathbb{R}})$ , we have  $g \circ \phi^{-1} \in C(\Gamma)$ , and therefore, by Theorem 9.14, the operator  $(g \circ \phi^{-1})(V|_Z)$  belongs to  $B(Z)$ , with  $B(Z)$ -norm  $\leq \|g \circ \phi^{-1}\|_{C(\Gamma)} = \|g\|_{C(\bar{\mathbb{R}})}$ .

The restriction  $V|_Z$  is the Cayley transform of  $T_Z$ , which is the restriction of  $T$  to the domain

$$D(T_Z) := \{x \in D(T) \cap Z; Tx \in Z\}.$$

The operator  $T_Z$  is called *the part of  $T$  in  $Z$* .

It is therefore natural to *define*

$$g(T_Z) := (g \circ \phi^{-1})(V|_Z) \quad (g \in C(\bar{\mathbb{R}})). \quad (1)$$

The map  $\tau : g \rightarrow g(T_Z)$  is a norm-decreasing algebra homomorphism of  $C(\bar{\mathbb{R}})$  into  $B(Z)$  such that  $f_0(T_Z) = I|_Z$  and  $\phi(T_Z) = V|_Z$ . We call a map  $\tau$  with the above properties a *contractive  $C(\bar{\mathbb{R}})$ -operational calculus for  $T$  on  $Z$* ; when such  $\tau$  exists, we say that  $T$  is of *contractive class  $C(\bar{\mathbb{R}})$  on  $Z$* .

If  $W$  is a Banach subspace of  $X$  such that  $T_W$  is of contractive class  $C(\bar{\mathbb{R}})$  on  $W$ , then the map

$$f \in C(\Gamma) \rightarrow f(V|_W) := (f \circ \phi)(T_W) \in B(W)$$

is a contractive  $C(\Gamma)$ -operational calculus for  $V|_W$  in  $B(W)$  (note that  $(f_1 \circ \phi)(T_W) = \phi(T_W) = V|_W$ ). Therefore  $W$  is a Banach subspace of  $Z$ , by Theorem 9.14. We formalize the above observations as

**Theorem 10.8.** *Let  $T$  be an unbounded operator with real spectrum, and let  $Z$  be its semi-simplicity space. Then  $T$  is of contractive class  $C(\bar{\mathbb{R}})$  on  $Z$ , and  $Z$  is maximal with this property (in the sense detailed in Theorem 9.14).*

For  $X$  reflexive, we obtain a spectral integral representation for  $T_Z$ .

**Theorem 10.9.** *Let  $T$  be an unbounded operator with real spectrum on the reflexive Banach space  $X$ , and let  $Z$  be its semi-simplicity space. Then there exists a contractive spectral measure on  $Z$*

$$F : \mathcal{B}(\mathbb{R}) \rightarrow B(Z),$$



such that

- (1)  $F$  commutes with every  $U \in B(X)$  which commutes with  $T$ ;  
 (2)  $D(T_Z)$  is the set  $Z_1$  of all  $x \in Z$  such that the integral

$$\int_{\mathbb{R}} s dF(s)x := \lim_{a \rightarrow -\infty, b \rightarrow \infty} \int_a^b s dF(s)x$$

exists in  $X$  and belongs to  $Z$ ;

- (3)  $Tx = \int_{\mathbb{R}} s dF(s)x$  for all  $x \in D(T_Z)$ ;  
 (4) For all non-real  $\lambda \in \mathbb{C}$  and  $x \in Z$ ,

$$R(\lambda; T)x = \int_{\mathbb{R}} \frac{1}{\lambda - s} dF(s)x.$$

**Proof.** We apply Theorem 9.16 to the Cayley transform  $V$ . Let then  $E$  be the unique contractive spectral measure on  $Z$ , with support on the unit circle  $\Gamma$ , such that

$$f(V|_Z)x = \int_{\Gamma} f dE(\cdot)x \quad (2)$$

for all  $x \in Z$  and  $f \in C(\Gamma)$ .

If  $E(\{-1\}) \neq 0$ , each  $x \neq 0$  in  $E(\{-1\})Z$  is an eigenvector for  $V$ , corresponding to the eigenvalue  $-1$  (the argument is the same as in the proof of Theorem 9.8, Part 5, first paragraph). However, since  $V = -2iR(-i; T) - I$ , we have the relation

$$R(-i; T) = (i/2)(I + V), \quad (3)$$

from which it is evident that  $-1$  is *not* an eigenvalue of  $V$  (since  $R(-i; T)$  is one-to-one). Thus

$$E(\{-1\}) = 0. \quad (4)$$

Define

$$F(\delta) = E(\phi(\delta)) \quad (\delta \in \mathcal{B}(\mathbb{R})).$$

Then  $F$  is a contractive spectral measure on  $Z$  defined on  $\mathcal{B}(\mathbb{R})$  (note that the requirement  $F(\mathbb{R}) = I|_Z$  follows from (4):

$$F(\mathbb{R}) = E(\Gamma - \{-1\}) = E(\Gamma) = I|_Z.)$$

If  $U \in B(X)$  commutes with  $T$ , it follows that  $U$  commutes with  $V = -2iR(-i; T) - I$ , and therefore  $U$  commutes with  $E$ , hence with  $F$ . By (2)

$$f(V|_Z)x = \int_{\mathbb{R}} f \circ \phi dF(\cdot)x \quad (5)$$

for all  $x \in Z$  and  $f \in C(\Gamma)$ . By definition, the left-hand side of (5) is  $(f \circ \phi)(T_Z)x$  for  $f \in C(\Gamma)$ . We may then rewrite (5) in the form

$$g(T_Z)x = \int_{\mathbb{R}} g dF(\cdot)x \quad (x \in Z) \quad (6)$$

for all  $g \in C(\bar{\mathbb{R}})$ . Taking in particular  $g = \phi$ , we get (since  $\phi(T_Z) = V|_Z$ )

$$Vx = \int_{\mathbb{R}} \phi dF(\cdot)x \quad (x \in Z). \quad (7)$$

By (3) and (7), we have for all  $x \in Z$

$$R(-i; T)x = (i/2) \int_{\mathbb{R}} (1 + \phi) dF(\cdot)x = \int_{\mathbb{R}} \frac{1}{-i - s} dF(s)x. \quad (8)$$

Observe that

$$D(T_Z) = R(-i; T)Z. \quad (9)$$

Indeed, if  $x \in D(T_Z)$ , then  $x \in D(T) \cap Z$  and  $Tx \in Z$ , by definition. Therefore,  $z := (-iI - T)x \in Z$ , and  $x = R(-i; T)z \in R(-i; T)Z$ . On the other hand, if  $x = R(-i; T)z$  for some  $z \in Z$ , then  $x \in D(T) \cap Z$  (because  $Z$  is invariant for  $R(-i; T)$ ), and  $Tx = -ix - z \in Z$ , so that  $x \in D(T_Z)$ .

Now let  $x \in D(T_Z)$ , and write  $x = R(-i; T)z$  for a suitable  $z \in Z$  (by (9)). The spectral integral on the right-hand side of (6) defines a norm-decreasing algebra homomorphism  $\tau$  of  $\mathbb{B}(\mathbb{R})$  into  $B(Z)$ , which extends the  $C(\bar{\mathbb{R}})$ -operational calculus for  $T$  on  $Z$  (cf. Theorem 9.16). For real  $a < b$ , take  $g(s) = s\chi_{[a,b]}(s) \in \mathbb{B}(\mathbb{R})$ . By (8)

$$\begin{aligned} \int_a^b s dF(s)x &= \tau(g)\tau(1/(-i - s))z = \tau\left(\frac{g(s)}{-i - s}\right)z \\ &= \int_a^b \frac{s}{-i - s} dF(s)z \rightarrow \int_{\mathbb{R}} \frac{s}{-i - s} dF(s)z \end{aligned}$$

as  $a \rightarrow -\infty$  and  $b \rightarrow \infty$  (convergence in  $X$  of the last integral follows from the boundedness of the integrand on  $\mathbb{R}$ ). Thus, the integral  $\int_{\mathbb{R}} s dF(s)x$  exists in  $X$  (in the sense stated in the theorem). Writing  $s/(-i - s) = [-i/(-i - s)] - 1$ , the last relation and (8) show that

$$\int_{\mathbb{R}} s dF(s)x = -iR(-i; T)z - z = TR(-i; T)z = Tx \in Z. \quad (10)$$

This proves that  $D(T_Z) \subset Z_1$  and Statement 3 of the theorem is valid.

On the other hand, if  $x \in Z_1$ , consider the well-defined element of  $Z$  given by  $z := \int_{\mathbb{R}} s dF(s)x$ . Since  $R(-i; T) \in B(X)$  commutes with  $T$  (hence with  $F$ ) and  $x \in Z$ , we have by (8) and the multiplicativity of  $\tau$  on  $\mathbb{B}(\mathbb{R})$

$$\begin{aligned} R(-i; T)z &= \lim_{a \rightarrow -\infty, b \rightarrow \infty} \int_a^b sR(-i; T) dF(s)x = \lim_{a, b} \int_a^b s dF(s)R(-i; T)x \\ &= \lim_{a, b} \int_a^b \frac{s}{-i - s} dF(s)x = \int_{\mathbb{R}} \frac{s}{-i - s} dF(s)x \\ &= \int_{\mathbb{R}} \left( \frac{-i}{-i - s} - 1 \right) dF(s)x = -iR(-i; T)x - x. \end{aligned}$$

Hence  $x = -R(-i; T)(ix + z) \in R(-i; T)Z = D(T_Z)$ , and we proved that  $D(T_Z) = Z_1$ .

For any non-real  $\lambda \in \mathbb{C}$ , the function  $g_\lambda(s) := (\lambda - s)^{-1}$  belongs to  $C(\bar{\mathbb{R}})$ , so that  $g_\lambda(T_Z)$  is a well-defined operator in  $B(Z)$  and by (6)

$$g_\lambda(T_Z)x = \int_{\mathbb{R}} \frac{1}{\lambda - s} dF(s)x \quad (x \in Z). \quad (11)$$

Fix  $x \in Z$ , and let  $y := g_\lambda(T_Z)x$  ( $\in Z$ ). By the multiplicativity of  $\tau : \mathbb{B}(\mathbb{R}) \rightarrow B(Z)$  and (10),

$$\begin{aligned} \int_a^b s dF(s)y &= \int_a^b \frac{s}{\lambda - s} dF(s)x \rightarrow \int_{\mathbb{R}} \frac{s}{\lambda - s} dF(s)x \\ &= \int_{\mathbb{R}} \left( \frac{\lambda}{\lambda - s} - 1 \right) dF(s)x = \lambda y - x \in Z. \end{aligned}$$

(The limit above is the  $X$ -limit as  $a \rightarrow -\infty$  and  $b \rightarrow \infty$ , and it exists because  $s/(\lambda - s)$  is a *bounded* continuous function on  $\mathbb{R}$ .) Thus,  $y \in D(T_Z)$  and  $Ty = \lambda y - x$  (by Statements 2 and 3 of the theorem). Hence  $(\lambda I - T)y = x$ , and since  $\lambda \in \rho(T)$ , it follows that  $y = R(\lambda; T)x$ , and Statement 4 is verified.  $\square$

## 10.6 Symmetric operators in Hilbert space

In this section,  $T$  will be an unbounded *densely defined* operator on a given Hilbert space  $X$ . The adjoint operator  $T^*$  is then a well-defined *closed* operator, to which we associate the *Hilbert space*  $[D(T^*)]$  with the  $T^*$ -graph norm  $\|\cdot\|^*$  and the inner product

$$(x, y)^* := (x, y) + (T^*x, T^*y) \quad (x, y \in D(T^*)).$$

We shall also consider the continuous sesquilinear form on  $[D(T^*)]$

$$\phi(x, y) := i[(x, T^*y) - (T^*x, y)] \quad (x, y \in D(T^*)).$$

Recall that  $T$  is symmetric iff  $T \subset T^*$ . In particular, a symmetric operator  $T$  is *closable* (since it has the closed extension  $T^*$ ). If  $S$  is a symmetric extension of  $T$ , then  $T \subset S \subset S^* \subset T^*$ , so that  $S = T^*|_D$ , where  $D = D(S)$ , and  $D(T) \subset D \subset D(T^*)$ . Clearly  $\phi(x, y) = 0$  for all  $x, y \in D$ . (Call such a subspace  $D$  of  $[D(T^*)]$  a *symmetric* subspace.) By the polarization formula for the sesquilinear form  $\phi$ ,  $D$  is symmetric iff  $\phi(x, x) (= 2\Im(T^*x, x)) = 0$  on  $D$ , that is, iff  $(T^*x, x)$  is real on  $D$ . Since  $T^* \in B([D(T^*)], X)$ , the  $[D(T^*)]$ -closure  $\bar{D}$  of a symmetric subspace  $D$  is symmetric.

If  $D$  is a symmetric subspace such that  $D(T) \subset D \subset D(T^*)$ , then  $D$  is the domain of the *symmetric* extension  $S := T^*|_D$  of  $T$ . Together with the previous remarks, this shows that the symmetric extensions  $S$  of  $T$  are precisely the restrictions of  $T^*$  to symmetric subspaces of  $[D(T^*)]$ .

We verify easily that  $S$  is *closed* iff  $D$  is a *closed* (symmetric) subspace of  $[D(T^*)]$  (Suppose  $S$  is closed and  $x_n \in D \rightarrow x$  in  $[D(T^*)]$ , i.e.  $x_n \rightarrow x$  and  $Sx_n (= T^*x_n) \rightarrow T^*x$  in  $X$ . Since  $S$  is closed, it follows that  $x \in D$ , and so  $D$  is closed in  $[D(T^*)]$ . Conversely, if  $D$  is closed in  $[D(T^*)]$ ,  $x_n \rightarrow x$  and  $Sx_n \rightarrow y$  in  $X$ , then  $T^*x_n \rightarrow y$ , and since  $T^*$  is closed,  $y = T^*x$ , i.e.  $x_n \rightarrow x$  in  $[D(T^*)]$ . Hence  $x \in D$ , and  $Sx = T^*x = y$ , i.e.  $S$  is closed.)

Let  $S$  be a symmetric extension of the symmetric operator  $T$ . Since  $D(S)$  is then a symmetric subspace of  $[D(T^*)]$ , so is its  $[D(T^*)]$ -closure  $\overline{D(S)}$ ; therefore, the restriction of  $T^*$  to  $\overline{D(S)}$  is a closed symmetric extension of  $S$ , which is precisely the *closure*  $\bar{S}$  of the closable operator  $S$ . (If  $x \in D(\bar{S})$ , there exist  $x_n \in D(S) \subset D(T^*)$  such that  $x_n \rightarrow x$  and  $Sx_n \rightarrow \bar{S}x$ . Since  $\bar{S} \subset T^*$ , we have  $x_n \rightarrow x$  in  $[D(T^*)]$ , hence  $x \in \overline{D(S)}$ . Conversely, if  $x \in \overline{D(S)}$ , there exist  $x_n \in D(S)$  such that  $x_n \rightarrow x$  in  $[D(T^*)]$ , that is,  $x_n \rightarrow x$  and  $Sx_n (= T^*x_n) \rightarrow T^*x$ , hence  $x \in D(\bar{S})$ . This shows that  $D(\bar{S}) = \overline{D(S)}$ , and  $\bar{S}$  is the restriction of  $T^*$  to this domain.)

Clearly  $\bar{S}$  is the *minimal* closed symmetric extension of  $S$ , and  $S$  is closed iff  $S = \bar{S}$ .

Note that  $T$  and  $\bar{T}$  have equal adjoints, since

$$\Gamma(\bar{T}^*) = (Q\Gamma(\bar{T}))^\perp = (Q\overline{\Gamma(T)})^\perp = (\overline{Q\Gamma(T)})^\perp = (Q\Gamma(T))^\perp = \Gamma(T^*).$$

(The  $\perp$  signs and the closure signs in the third and fourth expressions refer to the Hilbert space  $X \times X$ .)

Therefore,  $T$  and  $\bar{T}$  have the *same family* of closed symmetric extensions (namely, the restrictions of  $T^*$  to closed symmetric subspaces of  $[D(T^*)]$ ).

We are interested in *the family of selfadjoint extensions of  $T$* , which is contained in the family of closed symmetric extensions of  $T$ . We may then assume without loss of generality that  $T$  is a *closed* symmetric operator.

By the orthogonal decomposition theorem for the Hilbert space  $[D(T^*)]$ ,

$$[D(T^*)] = D(T) \oplus D(T)^\perp. \quad (1)$$

**Definition 10.10.** Let  $T$  be a closed densely defined symmetric operator. The kernels

$$D^+ := \ker(I + iT^*); \quad D^- := \ker(I - iT^*)$$

are called the positive and negative deficiency spaces of  $T$  (respectively). Their (Hilbert) dimensions  $n^+$  and  $n^-$  (in the Hilbert space  $[D(T^*)]$ ) are called the deficiency indices of  $T$ .

Note that

$$D^+ = \{y \in D(T^*); T^*y = iy\}; \quad D^- = \{y \in D(T^*); T^*y = -iy\}. \quad (2)$$

In particular,  $(x, y)^* = 2(x, y)$  on  $D^+$  and on  $D^-$ , so that the Hilbert dimensions  $n^+$  and  $n^-$  may be taken with respect to  $X$  (the deficiency spaces are also closed in  $X$ , as can be seen from their definition and the fact that  $T^*$  is a closed operator).

We have  $D^+ \perp D^-$ , because if  $x \in D^+$  and  $y \in D^-$ , then

$$(x, y)^* = (x, y) + (T^*x, T^*y) = (x, y) + (ix, -iy) = 0.$$

If  $y \in D^+$ , then for all  $x \in D(T)$ ,

$$\begin{aligned} (x, y)^* &= (x, y) + (T^*x, T^*y) = (x, y) + (Tx, iy) = (x, y) + (x, iT^*y) \\ &= (x, y) - (x, y) = 0, \end{aligned}$$

and similarly for  $y \in D^-$ . Hence

$$D^+ \oplus D^- \subset D(T)^\perp. \quad (3)$$

On the other hand, if  $y \in D(T)^\perp$ , we have

$$0 = (x, y)^* = (x, y) + (Tx, T^*y) \quad (x \in D(T)),$$

hence  $(Tx, T^*y) = -(x, y)$  is a continuous function of  $x$  on  $D(T)$ , that is,  $T^*y \in D(T^*)$  and  $T^*(T^*y) = -y$ . It follows that

$$(I - iT^*)(I + iT^*)y = (I + iT^*)(I - iT^*)y = 0. \quad (4)$$

Therefore

$$y - iT^*y \in \ker(I + iT^*) := D^+; \quad y + iT^*y \in \ker(I - iT^*) := D^-.$$

Consequently

$$y = (1/2)(y - iT^*y) + (1/2)(y + iT^*y) \in D^+ \oplus D^-.$$

This shows that  $D(T)^\perp \subset D^+ \oplus D^-$ , and we conclude from (3) and (1) that

$$D(T)^\perp = D^+ \oplus D^-, \quad (5)$$

and

$$[D(T^*)] = D(T) \oplus D^+ \oplus D^-. \quad (6)$$

It follows trivially from (6) that  $T$  is selfadjoint if and only if  $n^+ = n^- = 0$ .

Let  $D$  be a closed symmetric subspace of  $[D(T^*)]$  containing  $D(T)$ . By the orthogonal decomposition theorem for the Hilbert space  $D$  with respect to its closed subspace  $D(T)$ ,  $D = D(T) \oplus W$ , where  $W = D \ominus D(T) := D \cap D(T)^\perp$  is a closed symmetric subspace of  $D(T)^\perp$ . Conversely, given such a subspace  $W$ , the subspace  $D := D(T) \oplus W$  is a closed symmetric subspace of  $D(T^*)$ . By (5), the problem of finding all the closed symmetric extensions  $S$  of  $T$  is now reduced to the problem of finding all the closed symmetric subspaces  $W$  of  $D^+ \oplus D^-$ . Let  $x_k, k = 1, 2$  be the components of  $x \in W$  in  $D^+$  and  $D^-$  ( $x = x_1 + x_2$  corresponds as usual to the element  $[x_1, x_2] \in D^+ \times D^-$ ). The symmetry of  $D$  means that  $(T^*x, x)$  is real on  $W$ . However

$$\begin{aligned} (T^*x, x) &= (T^*x_1 + T^*x_2, x_1 + x_2) = i(x_1 - x_2, x_1 + x_2) \\ &= i(\|x_1\|^2 - \|x_2\|^2) - 2\Im(x_1, x_2) \end{aligned}$$

is real iff  $\|x_1\| = \|x_2\|$ . Thus,  $(T^*x, x)$  is real on  $W$  iff the map  $U : x_1 \rightarrow x_2$  is a (linear) isometry of a (closed) subspace  $D(U)$  of  $D^+$  onto a (closed) subspace  $R(U)$  of  $D^-$ . Thus,  $W$  is a closed symmetric subspace of  $D(T)^\perp$  iff

$$W = \{[x_1, Ux_1]; x_1 \in D(U)\}$$

is the graph of a linear isometry  $U$  as above. (Note that since  $\|x\|^* = \sqrt{2}\|x\|$  on  $D^+$  and  $D^-$ ,  $U$  is an isometry in both Hilbert spaces  $X$  and  $[D(T^*)]$ .)

Suppose  $D(U)$  is a proper (closed) subspace of  $D^+$ . Let then  $0 \neq y \in D^+ \cap D(U)^\perp$ . Necessarily,  $y \in D(S)^\perp$ , so that for all  $x \in D(S)$

$$0 = (x, y)^* = (x, y) + (Sx, T^*y) = (x, y) - i(Sx, y).$$

Hence  $(Sx, y) = -i(x, y)$  is a continuous function of  $x$  on  $D(S)$ , that is,  $y \in D(S^*)$ . Since  $0 \neq y \in D(S)^\perp$ , this shows that  $S \neq S^*$ . The same conclusion is obtained if  $R(U)$  is a proper subspace of  $D^-$  (same argument!). In other words, a *necessary condition* for  $S$  to be selfadjoint is that  $U$  be an *isometry of  $D^+$  onto  $D^-$* . Thus, if  $T$  has a selfadjoint extension, there exists a (linear) isometry of  $D^+$  onto  $D^-$  (equivalently,  $n^+ = n^-$ ).

On the other hand, if there exists a (linear) isometry  $U$  of  $D^+$  onto  $D^-$ , define  $S$  as the restriction of  $T^*$  to  $D(S) := D(T) \oplus \Gamma(U)$ . Since this domain  $D(S)$  is a closed symmetric subspace of  $D(T^*)$  (containing  $D(T)$ ),  $S$  is a closed symmetric extension of  $T$ . In particular,  $S \subset S^*$ , and we have the decomposition (6) for  $S$

$$D(S^*) = D(S) \oplus D^+(S) \oplus D^-(S). \quad (7)$$

Since  $S^* \subset T^*$ , the graph inner products for  $S^*$  and  $T^*$  coincide on  $D(S^*)$ ,  $D^+(S) \subset D^+$ , and  $D^-(S) \subset D^-$ .

If  $S \neq S^*$ , it follows from (7) that there exists  $0 \neq x \in D^+(S)$  (or  $\in D^-(S)$ ). Hence  $x + Ux \in D(S)$  (or  $U^{-1}x + x \in D(S)$ , respectively). Therefore by (7), since  $Ux \in D^-$  and  $x \in D^+$  ( $U^{-1}x \in D^+$  and  $x \in D^-$ , respectively),

$$0 = (x + Ux, x)^* = (x, x)^* = 2\|x\|^2 > 0$$

( $0 = (U^{-1}x + x, x)^* = (x, x)^* = 2\|x\|^2 > 0$ , respectively), contradiction. Hence  $S = S^*$ .

We proved the following.

**Theorem 10.11.** *Let  $T$  be a closed densely defined symmetric operator on the Hilbert space  $X$ . Then the closed symmetric extensions of  $T$  are the restrictions of  $T^*$  to the closed subspaces of  $[D(T^*)]$  of the form  $D(T) \oplus \Gamma(U)$ , where  $U$  is a linear isometry of a closed subspace of  $D^+$  onto a closed subspace of  $D^-$ , and  $\Gamma(U)$  is its graph. Such a restriction is selfadjoint if and only if  $U$  is an isometry of  $D^+$  onto  $D^-$ . In particular,  $T$  has a selfadjoint extension iff  $n^+ = n^-$  and has no proper closed symmetric extensions iff at least one of its deficiency indices vanishes.*

## Exercises

### The generator of a semigroup

1. (Notation as in Exercise 14, Chapter 9) The *generator*  $A$  of the  $C_0$ -semigroup  $T(\cdot)$  is its strong right derivative at 0 with *maximal domain*  $D(A)$ : denoting the (right) differential ratio at 0 by  $A_h$ , that is,  $A_h := h^{-1}[T(h) - I]$  ( $h > 0$ ), we have

$$Ax = \lim_{h \rightarrow 0+} A_h x \quad x \in D(A) = \{x \in X; \lim_h A_h x \text{ exists}\}.$$

Prove:

- (a)  $\bigcup_{t>0} V(t)X \subset D(A)$ , and for each  $t > 0$  and  $x \in X$ ,  $AV(t)x = T(t)x - x$ . Hint: Exercise 14(e), Chapter 9.
- (b)  $D(A)$  is dense in  $X$ . Hint: by Part (a),  $V(t)x \in D(A)$  for any  $t > 0$  and  $x \in X$  and  $AV(t)x = T(t)x - x$ . Apply Exercises 14(d) and 13(c) in Chapter 9.
- (c) For  $x \in D(A)$  and  $t > 0$ ,  $T(t)x \in D(A)$  and

$$AT(t)x = T(t)Ax = (d/dt)T(t)x,$$

where the right-hand side denotes the strong derivative at  $t$  of  $u := T(\cdot)x$ . Therefore  $u : [0, \infty) \rightarrow D(A)$  is a solution of class  $C^1$  of the *abstract Cauchy problem* (ACP)

$$(ACP) \quad u' = Au \quad u(0) = x.$$

Also

$$\int_0^t T(s)Ax \, ds = T(t)x - x \quad (x \in D(A)). \quad (*)$$

Hint: for left derivation, use Exercise 14(c), Chapter 9.

- (d)  $A$  is a closed operator. Hint: use the identity

$$V(t)Ax = AV(t)x = T(t)x - x \quad (x \in D(A); t > 0)$$

(cf. Part a) and Exercise 13(c), Chapter 9.

- (e) If  $v : [0, \infty) \rightarrow D(A)$  is a solution of class  $C^1$  of ACP, then  $v = T(\cdot)x$ . (This is the *uniqueness* of the solution of ACP when  $A$  is the generator of a  $C_0$ -semigroup.) In particular, the generator  $A$  determines the semigroup  $T(\cdot)$  uniquely. Hint: apply Exercise 13(d), Chapter 9, to  $V := T(\cdot)v(s - \cdot)$  on the interval  $[0, s]$ .

## Semigroups continuous in the u.o.t.

2. (Notation as in Exercise 1.) Suppose  $T(h) \rightarrow I$  in the u.o.t. (i.e.  $\|T(h) - I\| \rightarrow 0$  as  $h \rightarrow 0+$ ). Prove:
- (a)  $V(h)$  is non-singular for  $h$  small enough (which we fix from now on). Define  $A := [T(h) - I]V(h)^{-1} (\in B(X))$ .
  - (b)  $T(t) - I = V(t)A$  for all  $t \geq 0$  (with  $A$  as above). Conclude that  $A$  is the generator of  $T(\cdot)$  (in particular, the generator is a bounded operator).
  - (c) Conversely, if the generator  $A$  of  $T(\cdot)$  is a bounded operator, then  $T(t) = e^{tA}$  (defined by the usual absolutely convergent series in  $B(X)$ ) and  $T(h) \rightarrow I$  in the u.o.t. Hint: the exponential is a continuous semigroup (in the u.o.t.) with generator  $A$ ; use the uniqueness statement in Exercise 1(e).

## The resolvent of a semigroup generator

3. Let  $T(\cdot)$  be a  $C_0$ -semigroup on the Banach space  $X$ . Let  $A$  be its generator, and  $\omega$  its type (cf. Exercise 14(f), Chapter 9). Fix  $a > \omega$ . Prove:

- (a) The *Laplace transform*

$$L(\lambda) := \int_0^\infty e^{-\lambda t} T(t) dt$$

converges absolutely (in  $B(X)$ ) and  $\|L(\lambda)\| = O(1/(\Re \lambda - a))$  for  $\Re \lambda > a$ . (Cf. Exercises 13(e) and 14(c), Chapter 9.)

- (b)  $L(\lambda)(\lambda I - A)x = x$  for all  $x \in D(A)$  and  $\Re \lambda > a$ .
- (c)  $L(\lambda)X \subset D(A)$ , and  $(\lambda I - A)L(\lambda) = I$  for  $\Re \lambda > a$ .
- (d) Conclude that  $\sigma(A) \subset \{\lambda \in \mathbb{C}; \Re \lambda \leq \omega\}$  and  $R(\lambda; A) = L(\lambda)$  for  $\Re \lambda > \omega$ .
- (e) For any  $\lambda_k > a$  ( $k = 1, \dots, m$ ),

$$\left\| \prod_k (\lambda_k - a) R(\lambda_k; A) \right\| \leq M, \quad (1)$$

where  $M$  is a positive constant depending only on  $a$  and  $T(\cdot)$ . In particular

$$\|R(\lambda)^m\| \leq \frac{M}{(\lambda - a)^m} \quad (\lambda > a; m \in \mathbb{N}). \quad (2)$$

Hint: apply Part (d), and the multiple integral version of Exercise 13(e), Chapter 9.



- (f) Let  $A$  be *any* closed densely defined operator on  $X$  whose resolvent set contains a ray  $(a, \infty)$  and whose resolvent  $R(\cdot)$  satisfies  $\|R(\lambda)\| \leq M/(\lambda - a)$  for  $\lambda > \lambda_0$  (for some  $\lambda_0 \geq a$ ). (Such an  $A$  is sometimes called an *abstract potential*.) Consider the function  $A(\cdot) : (a, \infty) \rightarrow B(X)$ :

$$A(\lambda) := \lambda AR(\lambda) = \lambda^2 R(\lambda) - \lambda I.$$

Then, as  $\lambda \rightarrow \infty$ ,

$$\begin{aligned} \lim A(\lambda)x &= Ax \quad (x \in D(A)); \\ \lim \lambda R(\lambda) &= I \quad \text{and} \quad \lim AR(\lambda) = 0 \quad \text{in the s.o.t.} \end{aligned}$$

Note that these conclusions are valid if  $A$  is the generator of a  $C_0$ -semigroup, with  $a > \omega$  fixed. Cf. Exercise 3, Parts (d) and (e).

4. Let  $A$  be a closed densely defined operator on the Banach space  $X$  such that  $(a, \infty) \subset \rho(A)$  and (2) in Exercise 3(e) is satisfied. Define  $A(\cdot)$  as in Exercise 3(f) and denote  $T_\lambda(t) := e^{tA(\lambda)}$  (the usual power series). Prove:

- (a)  $\|T_\lambda(t)\| \leq M \exp(t(a\lambda/(\lambda - a)))$  for all  $\lambda > a$ . Conclude that

$$\|T_\lambda(t)\| \leq M e^{2at} \quad (\lambda > 2a) \quad (3)$$

and

$$\limsup_{\lambda \rightarrow \infty} \|T_\lambda(t)\| \leq M e^{at}. \quad (4)$$

- (b) If  $x \in D(A)$ , then uniformly for  $t$  in bounded intervals,

$$\lim_{2a < \lambda, \mu \rightarrow \infty} \|T_\lambda(t)x - T_\mu(t)x\| = 0. \quad (5)$$

Hint: apply Exercise 13(d), Chapter 9, to the function  $V(s) := T_\lambda(t - s)T_\mu(s)$  on the interval  $[0, t]$ ; Exercise 1(c) to the semigroups  $T_\lambda(\cdot)$  and  $T_\mu(\cdot)$ ; Part (a), and Exercise 3(f).

- (c) For each  $x \in X$ ,  $\{T_\lambda(t)x; \lambda \rightarrow \infty\}$  is Cauchy (uniformly for  $t$  in bounded intervals). (Use Part (b), the density of  $D(A)$ , and (3) in Part (a)) Define then

$$T(t) = \lim_{\lambda \rightarrow \infty} T_\lambda(t)$$

in the s.o.t. Then  $T(\cdot)$  is a strongly continuous semigroup such that  $\|T(t)\| \leq M e^{at}$  and

$$T(t)x - x = \int_0^t T(s)Ax \, ds \quad (x \in D(A)). \quad (6)$$

Hint: use (\*) in Exercise 1(c) for the semigroup  $e^{tA(\lambda)}$ , and apply Exercise 3(f).

- (d) If  $A'$  is the generator of the semigroup  $T(\cdot)$  defined in Part (c), then  $A \subset A'$ . Since  $\lambda I - A$  and  $\lambda I - A'$  are both one-to-one and onto for  $\lambda > a$  and coincide on  $D(A)$ , conclude that  $A' = A$ .
- (e) An operator  $A$  with domain  $D(A) \subset X$  is the generator of a  $C_0$ -semigroup satisfying  $\|T(t)\| \leq M e^{at}$  for some real  $a$  iff it is closed, densely defined,  $(a, \infty) \subset \rho(A)$  and (2) is satisfied. (Collect information from above!) This is the *Hille-Yosida theorem*. In particular (case  $M = 1$  and  $a = 0$ ),  $A$  is the generator of a contraction semigroup iff it is closed, densely defined, and  $\lambda R(\lambda)$  exist and are contractions for all  $\lambda > 0$ . (Terminology: the bounded operators  $A(\lambda)$  are called the *Hille-Yosida approximations* of the generator  $A$ .)

## Core for the generator

5. Let  $T(\cdot)$  be a  $C_0$ -semigroup on the Banach space  $X$ , and let  $A$  be its generator. Prove:
- (a)  $T(\cdot)$  is a  $C_0$ -semigroup on the Banach space  $[D(A)]$ . (Recall that the norm on  $[D(A)]$  is the graph norm  $\|x\|_A := \|x\| + \|Ax\|$ .)
- (b) Let  $D$  be a  $T(\cdot)$ -invariant subspace of  $D(A)$ , dense in  $X$ . For each  $x \in D$ , consider  $V(t)x := \int_0^t T(s)x \, ds$  (defined in the Banach space  $\bar{D}$ , the closure of  $D$  in  $[D(A)]$ ). Given  $x \in D(A)$ , let  $x_n \in D$  be such that  $x_n \rightarrow x$  (in  $X$ , by density of  $D$  in  $X$ ). Then  $V(t)x_n \rightarrow V(t)x$  in the graph-norm. Conclude that  $V(t)x \in \bar{D}$  for each  $t > 0$ , and therefore  $x \in \bar{D}$ , that is,  $D$  is dense in  $[D(A)]$ . (A dense subspace of  $[D(A)]$  is called a *core* for  $A$ .) Thus, a  $T(\cdot)$ -invariant subspace of  $D(A)$  which is dense in  $X$  is a core for  $A$ . (On the other hand, a core  $D$  for  $A$  is trivially dense in  $X$ , since  $D(A)$  is dense in  $X$  and  $D$  is  $\|\cdot\|_A$ -dense in  $D(A)$ .)
- (c) A  $C^\infty$ -vector for  $A$  is a vector  $x \in X$  such that  $T(\cdot)x$  is of class  $C^\infty$  ('strongly') on  $[0, \infty)$ . Let  $D^\infty$  denote the space of all  $C^\infty$ -vectors for  $A$ . Then

$$D^\infty = \bigcap_{n=1}^{\infty} D(A^n). \quad (7)$$

- (d) Let  $\phi_n \in C_c^\infty(\mathbb{R})$  be non-negative, with support in  $(0, 1/n)$  and integral equal to 1. Given  $x \in X$ , let  $x_n = \int \phi_n(t)T(t)x \, dt$ . Then
- (i)  $x_n \rightarrow x$  in  $X$ ;
- (ii)  $x_n \in D(A)$  and  $Ax_n = -\int \phi'_n(t)T(t)x \, dt$ ;
- (iii)  $x_n \in D(A^k)$  and  $A^k x_n = (-1)^k \int \phi_n^{(k)}(t)T(t)x \, dt$  for all  $k \in \mathbb{N}$ . In particular,  $x_n \in D^\infty$ .

Conclude that  $D^\infty$  is dense in  $X$  and is a core for  $A$ . (Cf. Part (b).)

## The Hille–Yosida space of an arbitrary operator.

6. Let  $A$  be an unbounded operator on the Banach space  $X$  with  $(a, \infty) \subset \rho(A)$ , for some real  $a$ . Denote its resolvent by  $R(\cdot)$ . Let  $\mathcal{A}$  be the multiplicative semigroup generated by the set  $\{(\lambda - a)R(\lambda); \lambda > a\}$ . Let  $Z := Z(\mathcal{A})$  (cf. Theorem 9.11), and consider  $A_Z$ , the part of  $A$  in  $Z$ . The *Hille–Yosida space* for  $A$ , denoted  $W$ , is the closure of  $D(A_Z)$  in the Banach subspace  $Z$ . Prove:
- (a)  $W$  is  $R(\lambda)$ -invariant for each  $\lambda > a$  and  $R(\lambda; A_W) = R(\lambda)|_W$ . In particular,  $A_W$  is closed as an operator in the Banach space  $W$ .
  - (b)  $\|R(\lambda; A_W)^m\|_{B(W)} \leq 1/(\lambda - a)^m$  for all  $\lambda > a$  and  $m \in \mathbb{N}$ .
  - (c)  $\lim_{\lambda \rightarrow \infty} \lambda R(\lambda; A_W)w = w$  in the  $Z$ -norm. Conclude that  $D(A_W)$  is dense in  $W$ .
  - (d)  $A_W$  generates a  $C_0$ -semigroup  $T(\cdot)$  on the Banach space  $W$ , such that  $\|T(t)\|_{B(W)} \leq e^{at}$ .
  - (e) If  $Y$  is a Banach subspace of  $X$  such that  $A_Y$  generates a  $C_0$ -semigroup on  $Y$  with the growth condition  $\|T(t)\|_{B(Y)} \leq e^{at}$ , then  $Y$  is a Banach subspace of  $W$ . (This is the *maximality* of the Hille–Yosida space.)

## Convergence of semigroups

7. Let  $\{T_s(\cdot); 0 \leq s < c\}$  be a family of  $C_0$ -semigroups on the Banach space  $X$ , such that

$$\|T_s(t)\| \leq M e^{at} \quad (t \geq 0; 0 \leq s < c) \quad (8)$$

for some  $M \geq 1$  and  $a \geq 0$ . Let  $A_s$  be the generator of  $T_s(\cdot)$ , and denote  $T(\cdot) = T_0(\cdot)$  and  $A = A_0$ . Note that (8) implies that

$$\|R(\lambda; A_s)\| \leq M/(\lambda - a) \quad (\lambda > a; s \in [0, c)). \quad (9)$$

Fix a core  $D$  for  $A$ . We say that  $A_s$  *graph-converge on  $D$  to  $A$*  (as  $s \rightarrow 0$ ) if for each  $x \in D$ , there exists a vector function  $s \in (0, c) \rightarrow x_s \in X$  such that  $x_s \in D(A_s)$  for each  $s$  and  $[x_s, A_s x_s] \rightarrow [x, Ax]$  in  $X \times X$ . Prove:

- (a)  $A_s$  graph-converge to  $A$  on  $D$  iff, for each  $\lambda > a$  and  $y \in (\lambda I - A)D$ , there exists a vector function  $s \rightarrow y_s$  such that  $[y_s, R(\lambda; A_s)y] \rightarrow [y, R(\lambda; A)y]$  in  $X \times X$  (as  $s \rightarrow 0$ ). Hint:  $y_s = (\lambda I - A_s)x_s$  and (9).
- (b) If  $A_s$  graph-converge to  $A$ , then as  $s \rightarrow 0$ ,  $R(\lambda; A_s) \rightarrow R(\lambda; A)$  in the s.o.t. for all  $\lambda > a$  (the later property is called *resolvents strong convergence*). Hint: show that  $(\lambda I - A)D$  is dense in  $X$ , and use Part (a) and (9).
- (c) Conversely, resolvents strong convergence implies graph-convergence on  $D$ . (Given  $y \in (\lambda I - A)D$ , choose  $y_s = y$  constant!)

- (d) If  $T'(\cdot)$  is also a  $C_0$ -semigroup satisfying (8), and  $A'$  is its generator, then

$$R(\lambda; A')[T'(t) - T(t)]R(\lambda; A) = \int_0^t T'(t-u)[R(\lambda; A') - R(\lambda; A)]T(u) du \quad (10)$$

for  $\Re \lambda > a$  and  $t \geq 0$ . Hint: verify that the integrand in (10) is the derivative with respect to  $u$  of the function  $-T'(t-u)R(\lambda; A')T(u)R(\lambda; A)$ .

- (e) Resolvents strong convergence implies *semigroups strong convergence*, that is, for each  $0 < \tau < \infty$ ,

$$\sup_{t \leq \tau} \|T_s(t)x - T(t)x\| \rightarrow 0 \quad (11)$$

as  $s \rightarrow 0$ . Hint: by (8), it suffices to consider  $x \in D(A) = R(\lambda; A)X$ . Write  $[T_s(t) - T(t)]R(\lambda; A)y = R(\lambda; A_s)[T_s(t) - T(t)]y + T_s(t)[R(\lambda; A) - R(\lambda; A_s)]y + [R(\lambda; A_s) - R(\lambda; A)]T(t)y$ . Estimate the norm of the first summand for  $y \in D(A)$  (hence  $y = R(\lambda; A)x$ ) using (10), and use the density of  $D(A)$  and (8)–(9). The second summand  $\rightarrow 0$  strongly, uniformly for  $t \leq \tau$ , by (8)–(9). For the third summand, consider again  $y \in D(A)$ , for which one can use the relation  $T(t)y = y + \int_0^t T(u)Ay du$ . Cf. Exercise 13(b), Chapter 9, and the Dominated Convergence theorem.

- (f) Conversely, semigroups strong convergence implies resolvents strong convergence. Hint: Use the Laplace integral representation of the resolvents.

Collecting, we conclude that *generators graph-convergence, resolvents strong convergence, and semigroups strong convergence are equivalent* (when Condition (8) is satisfied).

## Exponential formulas

8. Let  $A$  be the generator of a  $C_0$ -semigroup  $T(\cdot)$  of contractions on the Banach space  $X$ .

Let  $F : [0, \infty) \rightarrow B(X)$  be contraction-valued, such that  $F(0) = I$  and the (strong) right derivative of  $F(\cdot)x$  at 0 coincides with  $Ax$ , for all  $x$  in a core  $D$  for  $A$ . Prove:

- (a) Fix  $t > 0$  and define  $A_n$  as in Exercise 21(f), Chapter 6. Then  $e^{tA_n} - F(t/n)^n \rightarrow 0$  in the s.o.t. as  $n \rightarrow \infty$ .
- (b)  $s \rightarrow e^{sA_n}$  is a (uniformly continuous) contraction semigroup, for each  $n \in \mathbb{N}$ . (Cf. Exercise 21(a), Chapter 6.)
- (c) Suppose  $T(\cdot)$  is a *contraction*  $C_0$ -semigroup. As  $n \rightarrow \infty$ , the semigroups  $e^{sA_n}$  converge strongly to the semigroup  $T(s)$ , uniformly on compact intervals. (Cf. conclusion of Exercise 7 above; note that  $A_n x \rightarrow Ax$  for all  $x \in D$ .) Conclude that  $F(t/n)^n \rightarrow T(t)$  in the s.o.t., for each  $t \geq 0$ .

- (d) Let  $T(\cdot)$  be a  $C_0$ -semigroup such that  $\|T(t)\| \leq e^{at}$ , and consider the contraction semigroup  $S(t) := e^{-at}T(t)$  (with generator  $A - aI$ ;  $a \geq 0$ ). Choose  $F$  as follows:  $F(0) = I$  and for  $0 < s < 1/a$ ,

$$F(s) := (s^{-1} - a)R(s^{-1}; A) = (s^{-1} - a)R(s^{-1} - a; A - aI).$$

Verify that  $F$  satisfies the hypothesis stated at the beginning of the exercise, and conclude that

$$T(t) = \lim_{n \rightarrow \infty} \left[ \frac{n}{t} R\left(\frac{n}{t}; A\right) \right]^n \quad (12)$$

in the s.o.t., for each  $t > 0$ .

- (e) Let  $T(\cdot)$  be *any*  $C_0$ -semigroup. By Exercise 14(c), Chapter 9,  $\|T(t)\| \leq M e^{at}$  for some  $M \geq 1$  and  $a \geq 0$ . Consider the equivalent norm

$$|x| := \sup_{t \geq 0} e^{-at} \|T(t)x\| \quad (x \in X).$$

Then  $|T(t)x| \leq e^{at}|x|$ , and therefore (12) is valid over  $(X, |\cdot|)$ , hence over  $X$  (since the two norms are equivalent). Relation (12) (true for any  $C_0$ -semigroup!) is called *the exponential formula* for semigroups.

- (f) Let  $A, B, C$  generate contraction  $C_0$ -semigroups  $S(\cdot), T(\cdot), U(\cdot)$ , respectively, and suppose  $C = A + B$  on a core  $D$  for  $C$ . Then

$$U(t) = \lim_{n \rightarrow \infty} [S(t/n)T(t/n)]^n \quad (t \geq 0) \quad (13)$$

in the s.o.t. Hint: Choose  $F(t) = S(t)T(t)$  in Part (c).

## Groups of operators

9. A group of operators on the Banach space  $X$  is a map  $T(\cdot) : \mathbb{R} \rightarrow B(X)$  such that

$$T(s+t) = T(s)T(t) \quad (s, t \in \mathbb{R}).$$

We assume that it is of class  $C_0$ , that is, *the semigroup*  $T(\cdot)|_{[0, \infty)}$  is of class  $C_0$ . Let  $A$  be the generator of this semigroup. Prove:

- (a) The semigroup  $S(t) := T(-t), t \geq 0$ , is of class  $C_0$ , and has the generator  $-A$ .  
 (b)  $\sigma(A)$  is contained in the strip

$$\Omega : -\omega' \leq \Re \lambda \leq \omega,$$

where  $\omega, \omega'$  are the types of the semigroups  $T(\cdot)$  and  $S(\cdot)$ , respectively.

Fix  $a > \omega$  and  $a' > \omega'$ , and let

$$\Omega' = \{\lambda \in \mathbb{C}; -a' \leq \Re \lambda \leq a\}.$$

For  $\lambda \notin \Omega'$ ,

$$\|R(\lambda; A)^n\| \leq \frac{M}{d(\lambda, \Omega')}. \quad (14)$$

If  $A$  generates a *bounded*  $C_0$ -group, then  $\sigma(A) \subset i\mathbb{R}$  and

$$\|R(\lambda; A)^n\| \leq \frac{M}{|\Re \lambda|^n}$$

where  $M$  is a bound for  $\|T(\cdot)\|$ .

- (c) An operator  $A$  generates a  $C_0$ -group of operators iff it is closed, densely defined, has spectrum in a strip  $\Omega$  as in Part (b), and (14) is satisfied for all *real*  $\lambda \notin [-a', a]$ . Hint: apply the Hille–Yosida theorem (cf. Exercise 4(e)) separately in the half-planes  $\Re \lambda > a$  and  $\Re \lambda > a'$ .
- (d) Let  $T(\cdot)$  be a  $C_0$ -group of *unitary* operators on a Hilbert space  $X$ . Let  $H = -iA$ , where  $A$  is the generator of  $T(\cdot)$ . Then  $H$  is a (closed, densely defined) *symmetric* operator with *real* spectrum. In particular,  $iI - H$  and  $-iI - H$  are both *onto*, so that the deficiency indices of  $H$  are both zero. Therefore  $H$  is selfadjoint (cf. (6) following Definition 10.10).
- (e) Define  $e^{itH}$  by means of the operational calculus for the selfadjoint operator  $H$ . This is a  $C_0$ -group with generator  $iH = A$ , and therefore  $T(t) = e^{itH}$  (cf. Exercise 1(e): the generator determines the semigroup uniquely). This representation of unitary groups is *Stone's theorem*.

# Application I

## Probability

### I.1 Heuristics

A fundamental concept in Probability Theory is that of an *event*. The ‘real world’ content of the ‘event’ plays no role in the mathematical analysis. What matters is only the event’s *occurrence* or non-occurrence.

Two ‘extreme’ events are the *empty event*  $\emptyset$  (which cannot occur), and the *sure event*  $\Omega$  (which occurs always).

To each event  $A$ , one associates the *complementary event*  $A^c$ , which occurs if and only if  $A$  does not occur.

If the occurrence of the event  $A$  forces the occurrence of the event  $B$ , one says that  $A$  *implies*  $B$ , and one writes  $A \subset B$ . One has trivially  $\emptyset \subset A \subset \Omega$  for any event  $A$ .

The events  $A, B$  are *equivalent* (notation:  $A = B$ ) if they imply each other. Such events are identified.

The *intersection*  $A \cap B$  of the events  $A$  and  $B$  occurs if and only if  $A$  and  $B$  both occur. If  $A \cap B = \emptyset$  (i.e. if  $A$  and  $B$  cannot occur together), one says that the events are *mutually disjoint*; for example, for any event  $A$ , the events  $A$  and  $A^c$  are mutually disjoint.

The *union*  $A \cup B$  of the events  $A, B$  is the event that occurs iff at least one of the events  $A, B$  occurs. The operations  $\cap$  and  $\cup$  are trivially commutative, and satisfy the following relations:

$$A \cup A^c = \Omega;$$

$$A \cap B \subset A \subset A \cup B.$$

One verifies that the algebra of events satisfies the usual associative and distributive laws for the family  $\mathbb{P}(\Omega)$  of all subsets of a set  $\Omega$ , with standard operations between subsets, as well as the DeMorgan (dual) laws:

$$\left( \bigcup_k A_k \right)^c = \bigcap_k A_k^c; \quad \left( \bigcap_k A_k \right)^c = \bigcup_k A_k^c,$$

for any sequence of events  $\{A_k\}$ . Mathematically, we may then view the sure event  $\Omega$  as a given set (called the *sample space*), and the set of all events (for a particular probability problem) as an algebra of subsets of  $\Omega$ .

Since limiting processes are central in Probability Theory, countable unions of events should also be events. Therefore, in the set-theoretical model, the algebra of events is required to be a  $\sigma$ -algebra  $\mathcal{A}$ .

The second fundamental concept of Probability Theory is that of a *probability*. Each event  $A \in \mathcal{A}$  is assigned a probability  $P(A)$  (also denoted  $PA$ ), such that

- (1)  $0 \leq P(A) \leq 1$  for all  $A \in \mathcal{A}$ ;
- (2)  $P(\bigcup A_k) = \sum P(A_k)$  for any sequence of mutually disjoint events  $A_k$ ; and
- (3)  $P(\Omega) = 1$ .

In other words,  $P$  is a ‘normalized’ finite positive measure on the *measurable space*  $(\Omega, \mathcal{A})$ . The measure space  $(\Omega, \mathcal{A}, P)$  is called a *probability space*. Note that  $P(A^c) = 1 - P(A)$ .

### Examples.

- (1) The *trivial* probability space  $(\Omega, \mathcal{A}, P)$  has  $\Omega$  arbitrary,  $\mathcal{A} = \{\emptyset, \Omega\}$ ,  $P(\emptyset) = 0$ , and  $P(\Omega) = 1$ .
- (2) *Discrete probability space*.  $\Omega$  is the union of finitely many mutually disjoint events  $A_1, \dots, A_n$ , with probabilities  $P(A_k) = p_k$ ,  $p_k \geq 0$ , and  $\sum p_k = 1$ . The family  $\mathcal{A}$  consists of  $\emptyset$  and all finite unions  $A = \bigcup_{k \in J} A_k$ , where  $J \subset \{1, \dots, n\}$ . One lets  $P(\emptyset) = 0$  and  $P(A) = \sum_{k \in J} p_k$ .

This probability space is the (finite) discrete probability space. When  $p_k = p$  for all  $k$  (so that  $p = 1/n$ ), one gets the (finite) uniform probability space. The formula for the probability reduces in this special case to

$$P(A) = \frac{|A|}{n},$$

where  $|A|$  denotes the number of points in the index set  $J$  (i.e. the number of ‘elementary events’  $A_k$  contained in  $A$ ).

- (3) *Random sampling*. A sample of size  $s$  from a population  $\mathcal{P}$  of  $N \geq s$  objects is a subset  $S \subset \mathcal{P}$  with  $s$  elements ( $|S| = s$ ). The sampling is random if all  $\binom{N}{s}$  samples of size  $s$  are assigned the same probability (i.e. the corresponding probability space is a uniform probability space, where  $\Omega$  is the set of all samples of given size  $s$ ; this is actually the origin of the name ‘sample space’ given to  $\Omega$ ). The elementary event of getting any particular sample of size  $s$  has probability  $1/\binom{N}{s}$ .

Suppose the population  $\mathcal{P}$  is the disjoint union of  $m$  sub-populations (‘layers’)  $\mathcal{P}_i$  of size  $N_i$  ( $\sum N_i = N$ ). The number of size  $s$  samples with  $s_i$  objects from  $\mathcal{P}_i$  ( $i = 1, \dots, m$ ;  $\sum s_i = s$ ) is the product of the binomial coefficients  $\binom{N_i}{s_i}$ . Therefore, if  $A_{s_1, \dots, s_m}$  denotes the event of getting  $s_i$  objects from  $\mathcal{P}_i$  ( $i = 1, \dots, m$ ) in



a random sampling of  $s$  objects from the multi-layered population  $\mathcal{P}$ , then

$$P(A_{s_1, \dots, s_m}) = \frac{\binom{N_1}{s_1} \dots \binom{N_m}{s_m}}{\binom{N}{s}}.$$

An *ordered sample* of size  $s$  is an ordered  $s$ -tuple  $(x_1, \dots, x_s) \subset \mathcal{P}$  (we may think of  $x_i$  as the object drawn at the  $i$ th drawing from the population). The number of such samples is clearly  $N(N-1) \dots (N-s+1)$  (since there are  $N$  possible outcomes of the first drawing,  $N-1$  for the second, etc.). Fixing one specific object, let  $A$  denote the event of getting that object in some specific drawing. Since the procedure is equivalent to (ordered) sampling of size  $s-1$  from a population of size  $N-1$ , we have  $|A| = (N-1) \dots [(N-1) - (s-1) + 1]$ , and therefore, for random sampling (the uniform model!),

$$P(A) = |A|/|\Omega| = \frac{(N-1) \dots (N-s+1)}{N(N-1) \dots (N-s+1)} = 1/N.$$

This probability is *independent* of the drawing considered! This fact is referred to as the ‘equivalence law of ordered sampling’.

## I.2 Probability space

Let  $(\Omega, \mathcal{A}, P)$  be a probability space, that is, a normalized finite positive measure space. Following the above terminology, the ‘measurable sets’  $A \in \mathcal{A}$  are called the *events*;  $\Omega$  is the *sure event*;  $\emptyset$  is the *empty event*; the measure  $P$  is called the *probability*. One says *almost surely* (a.s.) instead of ‘almost everywhere’ (or ‘with probability one’, since the complement of the exceptional set has probability one).

If  $f$  is a real valued function on  $\Omega$ , it is desirable that the sets  $[f > c]$  be events for any real  $c$ , that is, that  $f$  be measurable. Such functions will be called real *random variables* (r.v.). Similarly, a complex r.v. is a complex measurable function on  $\Omega$ .

The simplest r.v. is the indicator  $I_A$  of an event  $A \in \mathcal{A}$ . We clearly have

$$I_{A^c} = 1 - I_A; \quad (1)$$

$$I_{A \cap B} = I_A I_B; \quad (2)$$

$$I_{A \cup B} = I_A + I_B - I_{A \cap B} \quad (3)$$

for any events  $A, B$ , and

$$I_{\bigcup_k A_k} = \sum_k I_{A_k} \quad (4)$$

for any sequence  $\{A_k\}$  of mutually disjoint events.

A finite linear combination of indicators is a ‘simple random variable’;  $L^1(P)$  is the space of ‘integrable r.v.’s’ (real or complex, as needed); the integral over  $\Omega$  of an integrable r.v.  $X$  is called its *expectation*, and is denoted by  $E(X)$  or  $EX$ :

$$E(X) := \int_{\Omega} X dP, \quad X \in L^1(P).$$

The functional  $E$  on  $L^1(P)$  is linear, positive, bounded (with norm 1), and  $E1 = 1$ . For any  $A \in \mathcal{A}$ ,

$$E(I_A) = P(A).$$

For a simple r.v.  $X$ ,  $EX$  is then the weighted arithmetical average of its values, with weights equal to the probabilities that  $X$  assume these values.

The obvious relations

$$P(A^c) = 1 - P(A); \quad P(A \cup B) = PA + PB - P(A \cap B),$$

parallel (1) and (3); however, the probability analogue of (2), namely  $P(A \cap B) = P(A)P(B)$ , is *not* true in general. One says that the events  $A$  and  $B$  are (stochastically) independent if

$$P(A \cap B) = P(A)P(B).$$

More generally, a family  $\mathcal{F} \subset \mathcal{A}$  of events is (stochastically) independent if

$$P\left(\bigcap_{k \in J} A_k\right) = \prod_{k \in J} P(A_k)$$

for any finite subset  $\{A_k; k \in J\} \subset \mathcal{F}$ .

The random variables  $X_1, \dots, X_n$  are (stochastically) independent if for any choice of Borel sets  $B_1, \dots, B_n$  in  $\mathbb{R}$  (or  $\mathbb{C}$ ), the events  $X_1^{-1}(B_1), \dots, X_n^{-1}(B_n)$  are independent.

**Theorem I.2.1.** *If  $X_1, \dots, X_n$  are (real) independent r.v.'s, and  $f_1, \dots, f_n$  are (real or complex) Borel functions on  $\mathbb{R}$ , then  $f_1(X_1), \dots, f_n(X_n)$  are independent r.v.'s.*

**Proof.** For simplicity of notation, we take  $n = 2$  (the general case is analogous). Thus,  $X, Y$  are independent r.v.'s, and  $f, g$  are real (or complex) Borel functions on  $\mathbb{R}$ . Let  $A, B$  be Borel subsets of  $\mathbb{R}$  (or  $\mathbb{C}$ ). Then

$$\begin{aligned} P(f(X)^{-1}(A) \cap g(Y)^{-1}(B)) &= P(X^{-1}[f^{-1}(A)] \cap Y^{-1}[g^{-1}(B)]) \\ &= P(X^{-1}[f^{-1}(A)])P(Y^{-1}[g^{-1}(B)]) = P(f(X)^{-1}(A))P(g(Y)^{-1}(B)). \end{aligned}$$

□

In particular, when  $X, Y$  are independent r.v.'s, the random variables  $aX + b$  and  $cY + d$  are independent for any constants  $a, b, c, d$ . For example, if  $X, Y$  are independent integrable r.v.'s, then  $X - EX$  and  $Y - EY$  are independent *central* (integrable) r.v.'s, where 'central' means 'with expectation zero'.

**Theorem I.2.2 (Multiplicativity of  $E$  on independent r.v.'s).** *If  $X_1, \dots, X_n$  are independent integrable r.v.'s, then  $\prod X_k$  is integrable and*

$$E\left(\prod X_k\right) = \prod E(X_k).$$

**Proof.** The proof is by induction on  $n$ . It suffices therefore to prove the theorem for two independent integrable r.v.'s  $X, Y$ .

*Case 1. Simple r.v.'s:*

$$X = \sum x_j I_{A_j}, \quad Y = \sum y_k I_{B_k},$$

with all  $x_j$  distinct, and all  $y_k$  distinct. Thus  $A_j = X^{-1}(\{x_j\})$  and  $B_k = Y^{-1}(\{y_k\})$  are independent events. Hence

$$\begin{aligned} E(XY) &= E\left(\sum_{j,k} x_j y_k I_{A_j} I_{B_k}\right) = \sum x_j y_k E(I_{A_j \cap B_k}) \\ &= \sum x_j y_k P(A_j \cap B_k) = \sum x_j y_k P(A_j) P(B_k) \\ &= \sum_j x_j P(A_j) \sum_k y_k P(B_k) = E(X)E(Y). \end{aligned}$$

*Case 2. Non-negative (integrable) r.v.'s  $X, Y$ :*

For  $n = 1, 2, \dots$ , let

$$A_{n,j} := X^{-1}\left(\left[\frac{j-1}{2^n}, \frac{j}{2^n}\right)\right)$$

and

$$B_{n,k} := Y^{-1}\left(\left[\frac{k-1}{2^n}, \frac{k}{2^n}\right)\right),$$

with  $j, k = 1, \dots, n2^n$ . Consider the simple r.v.'s

$$\begin{aligned} X_n &= \sum_{j=1}^{n2^n} \frac{j-1}{2^n} I_{A_{n,j}}, \\ Y_n &= \sum_{k=1}^{n2^n} \frac{k-1}{2^n} I_{B_{n,k}}. \end{aligned}$$

For each  $n$ ,  $X_n, Y_n$  are independent, so that by Case 1,  $E(X_n Y_n) = E(X_n)E(Y_n)$ . Since the non-decreasing sequences  $\{X_n\}$ ,  $\{Y_n\}$ , and  $\{X_n Y_n\}$  converge to  $X, Y$ , and  $XY$ , respectively, it follows from the Lebesgue Monotone Convergence theorem that

$$E(XY) = \lim E(X_n Y_n) = \lim E(X_n)E(Y_n) = E(X)E(Y)$$

(and in particular,  $XY$  is integrable).

*Case 3.  $X, Y$  real independent integrable r.v.'s:*

In this case,  $|X|, |Y|$  are independent (by Theorem I.2.1), and by Case 2,

$$E(|XY|) = E(|X|)E(|Y|) < \infty,$$

so that  $XY$  is integrable. Also by Theorem I.2.1,  $X', Y'$  are independent r.v.'s, where the prime stands for either  $+$  or  $-$ . Therefore, by Case 2,

$$\begin{aligned} E(XY) &= E((X^+ - X^-)(Y^+ - Y^-)) \\ &= E(X^+)E(Y^+) - E(X^-)E(Y^+) - E(X^+)E(Y^-) + E(X^-)E(Y^-) \\ &= [E(X^+) - E(X^-)][E(Y^+) - E(Y^-)] = E(X)E(Y). \end{aligned}$$

The case of complex  $X, Y$  follows from Case 3 in a similar fashion.  $\square$

**Definition I.2.3.** If  $X$  is a real r.v., its *characteristic function* (ch.f.) is defined by

$$f_X(u) := E(e^{iuX}) \quad (u \in \mathbb{R}).$$

Clearly  $f_X$  is a well-defined complex valued function,  $|f_X| \leq 1$ ,  $f_X(0) = 1$ , and one verifies easily that it is uniformly continuous on  $\mathbb{R}$ .

**Corollary I.2.4.** *The ch.f. of the sum of independent real r.v.'s is the product of their ch.f.'s.*

**Proof.** If  $X_1, \dots, X_n$  are independent real r.v.'s, it follows from Theorem I.2.1 that  $e^{iuX_1}, \dots, e^{iuX_n}$  are independent (complex) integrable r.v.'s, and therefore, if  $X := \sum X_k$  and  $u \in \mathbb{R}$ ,

$$f_X(u) := E(e^{iuX}) = E\left(\prod_k e^{iuX_k}\right) = \prod_k E(e^{iuX_k}) = \prod_k f_{X_k}(u)$$

by Theorem I.2.2.  $\square$

## $L^2$ -random variables

**Terminology I.2.5.** If  $X \in L^2(P)$ , Schwarz's inequality shows that

$$E(|X|) = E(1 \cdot |X|) \leq \|1\|_2 \|X\|_2 = \|X\|_2,$$

that is,  $X$  is integrable, and

$$\sigma(X) := \|X - EX\|_2 < \infty$$

is called the *standard deviation* (s.d.) of  $X$  (this is the  $L^2$ -distance from  $X$  to its expectation). The square of the s.d. is the *variance* of  $X$ .

If  $X, Y$  are real  $L^2$ -r.v.'s, the product  $(X - EX)(Y - EY)$  is integrable (by Schwarz's inequality). One defines the *covariance* of  $X$  and  $Y$  by

$$\text{cov}(X, Y) := E((X - EX)(Y - EY)).$$

In particular,

$$\text{cov}(X, X) = \sigma^2(X).$$

By Schwarz's inequality,

$$|\text{cov}(X, Y)| \leq \sigma(X)\sigma(Y).$$

The linearity of  $E$  implies that

$$\text{cov}(X, Y) = E(XY) - E(X)E(Y), \quad (5)$$

and in particular (for  $Y = X$ ),

$$\sigma^2(X) = E(X^2) - (EX)^2. \quad (6)$$

By (5),  $\text{cov}(X, Y) = 0$  if  $X, Y$  are independent (cf. Theorem I.2.2). The converse is false in general, as can be seen by simple counter-examples.

The  $L^2$ -r.v.s  $X, Y$  are said to be *uncorrelated* if  $\text{cov}(X, Y) = 0$ .

If  $X = I_A$  and  $Y = I_B$  ( $A, B \in \mathcal{A}$ ), then by (5),

$$\text{cov}(I_A, I_B) = E(I_A I_B) - E(I_A)E(I_B) = P(A \cap B) - P(A)P(B). \quad (7)$$

Thus, indicators are uncorrelated iff they are independent! Taking  $B = A$  (with  $PA = p$ , so that  $P(A^c) = 1 - p := q$ ), we see from (6) that

$$\sigma^2(I_A) = E(I_A) - E(I_A)^2 = p - p^2 = pq. \quad (8)$$

**Lemma I.2.6.** *Let  $X_1, \dots, X_n$  be real  $L^2$ -r.v.s. Then*

$$\sigma^2\left(\sum_k X_k\right) = \sum_k \sigma^2(X_k) + 2 \sum_{1 \leq j < k \leq n} \text{cov}(X_j, X_k).$$

*In particular, if  $X_j$  are pairwise uncorrelated, then*

$$\sigma^2\left(\sum_k X_k\right) = \sum_k \sigma^2(X_k)$$

*(BienAyme's identity).*

**Proof.**

$$\begin{aligned} \sigma^2\left(\sum_k X_k\right) &= E\left(\sum_k X_k - \sum_k EX_k\right)^2 = E\left(\sum [X_k - EX_k]\right)^2 \\ &= E\left[\sum (X_k - EX_k)^2 + 2 \sum_{j < k} (X_j - EX_j)(X_k - EX_k)\right] \\ &= \sum \sigma^2(X_k) + 2 \sum_{j < k} \text{cov}(X_j, X_k). \end{aligned}$$

□

**Example I.2.7.** Let  $\{A_k\} \subset \mathcal{A}$  be a sequence of pairwise independent events. Let

$$S_n := \sum_{k=1}^n I_{A_k}, \quad n = 1, 2, \dots$$

Then

$$ES_n = \sum_{k=1}^n PA_k; \quad \sigma^2(S_n) = \sum_{k=1}^n PA_k(1 - PA_k).$$

In particular, when  $PA_k = p$  for all  $k$  (the so-called ‘Bernoulli case’), we have

$$ES_n = np; \quad \sigma^2(S_n) = npq.$$

Note that for each  $\omega \in \Omega$ ,  $S_n(\omega)$  is the number of events  $A_k$  with  $k \leq n$  for which  $\omega \in A_k$  (‘the number of successes in the first  $n$  trials’).

For  $0 \leq j \leq n$ ,  $S_n(\omega) = j$  iff there are precisely  $j$  events  $A_k$ ,  $1 \leq k \leq n$ , such that  $\omega \in A_k$  (and  $\omega \in A_k^c$  for the remaining  $n - j$  events). Since there are  $\binom{n}{j}$  possibilities to choose  $j$  indices  $k$  from the set  $\{1, \dots, n\}$  (for which  $\omega \in A_k$ ), and these choices define mutually disjoint events, we have in the Bernoulli case

$$P[S_n = j] = \binom{n}{j} p^j q^{n-j}. \quad (*)$$

One calls  $S_n$  the ‘Bernoulli random variable’, and  $(*)$  is its *distribution*.

**Example I.2.8.** Consider random sampling from a two-layered population (see Section I.1, Example 3). Let  $B_k$  be the event of getting an object from the layer  $\mathcal{P}_1$  in the  $k$ th drawing, and let  $D_s = \sum_{k=1}^s I_{B_k}$ . In our previous notations (with  $m = 2$ ),

$$P[D_s = s_1] = \frac{\binom{N_1}{s_1} \binom{N_2}{s_2}}{\binom{N}{s}},$$

where  $s_1 + s_2 = s$  and  $N_1 + N_2 = N$ .

By the equivalence principle of ordered sampling (cf. Section I.1),  $PB_k = N_1/N$  for all  $k$ , and therefore

$$ED_s = \sum PB_k = s \frac{N_1}{N}.$$

Note that the events  $B_k$  are *dependent* (drawing without return!). In the case of drawings with returns, the events  $B_k$  are independent, and  $D_s$  is the Bernoulli r.v., for which we saw that  $ED_s = sp = s(N_1/N)$  (since  $p := PB_k$ ). Note that the expectation is the same in both cases (of drawings with or without returns).

For  $1 \leq j < k \leq s$ , one has

$$P(B_k \cap B_j) = \frac{N_1}{N} \frac{N_1 - 1}{N - 1}$$

by the equivalence principle of ordered sampling. Therefore, by (7),

$$\text{cov}(I_{B_k}, I_{B_j}) = \frac{N_1}{N} \frac{N_1 - 1}{N - 1} - \left( \frac{N_1}{N} \right)^2,$$

independently of  $k, j$ . By Lemma I.2.6,

$$\begin{aligned}\sigma^2(D_s) &= s \frac{N_1}{N} \left(1 - \frac{N_1}{N}\right) + s(s-1) \left[ \frac{N_1}{N} \frac{N_1-1}{N-1} - \left(\frac{N_1}{N}\right)^2 \right] \\ &= \frac{N-s}{N-1} s \frac{N_1}{N} \left(1 - \frac{N_1}{N}\right).\end{aligned}$$

Thus, the difference between the variances for the methods of drawing with or without returns appears in the *correcting factor*  $(N-s)/(N-1)$ , which is close to 1 when the sample size  $s$  is small relative to the population size  $N$ .

One calls  $D_s$  the *hypergeometric random variable*.

**Example I.2.9.** Suppose we mark  $N \geq 1$  objects with numbers  $1, \dots, N$ . In drawings without returns from this population of objects, let  $A_k$  denote the event of drawing precisely the  $k$ th object in the  $k$ th drawing ('matching' in the  $k$ th drawing). In this case, the r.v.

$$S = \sum_{k=1}^N I_{A_k}$$

'is' the number of matchings in  $N$  drawings.

By the equivalence principle of ordered sampling,  $PA_k = 1/N$  and  $P(A_k \cap A_j) = (1/N)(1/(N-1))$ , independently of  $k$  and  $j < k$ . Hence

$$\begin{aligned}ES &= \sum_{k=1}^N PA_k = 1, \\ \text{cov}(I_{A_k}, I_{A_j}) &= \frac{1}{N(N-1)} - \frac{1}{N^2},\end{aligned}$$

and consequently, by Lemma I.2.6,

$$\sigma^2(S) = N \frac{1}{N} \left(1 - \frac{1}{N}\right) + 2 \binom{N}{2} \left[ \frac{1}{N(N-1)} - \frac{1}{N^2} \right] = 1.$$

**Lemma I.2.10.** *Let  $X$  be any r.v. and  $\epsilon > 0$ .*

(1) *If  $X \in L^2(P)$ , then*

$$P[|X - EX| \geq \epsilon] \leq \frac{\sigma^2(X)}{\epsilon^2}.$$

*(Tchebichev's inequality).*

(2) *If  $|X| \leq 1$ , then*

$$P[|X| \geq \epsilon] \geq E(|X|^2) - \epsilon^2.$$

*(Kolmogorov's inequality).*

**Proof.** Denote  $A = [|X - EX| \geq \epsilon]$ . Since  $|X - EX|^2 \geq |X - EX|^2 I_A \geq \epsilon^2 I_A$ , the monotonicity of  $E$  implies that

$$\sigma^2(X) := E(|X - EX|^2) \geq \epsilon^2 E(I_A) = \epsilon^2 P(A),$$

and Part (1) is verified.

In case  $|X| \leq 1$ , denote  $A = [|X| \geq \epsilon]$ . Then

$$|X|^2 = |X|^2 I_A + |X|^2 I_{A^c} \leq I_A + \epsilon^2,$$

and Part (2) follows by applying  $E$ . □

**Corollary I.2.11.** *Let  $X$  be an integrable r.v. with  $|X - EX| \leq 1$ . Then for any  $\epsilon > 0$ ,*

$$\sigma^2(X) - \epsilon^2 \leq P[|X - EX| \geq \epsilon] \leq \frac{\sigma^2(X)}{\epsilon^2}.$$

(Note that  $X$  is bounded, hence in  $L^2(P)$ , so that we may apply Part (1) to  $X$  and Part (2) to  $X - EX$ .)

**Corollary I.2.12.** *Let  $\{A_k\}$  be a sequence of events, and let  $X_n = (1/n) \sum_{k=1}^n I_{A_k}$  (the occurrence frequency of the first  $n$  events). Then  $X_n - EX_n$  converge to zero in probability (i.e.  $P[|X_n - EX_n| \geq \epsilon] \rightarrow 0$  as  $n \rightarrow \infty$ , for any  $\epsilon > 0$ ) if and only if  $\sigma^2(X_n) \rightarrow 0$ .*

**Proof.** Since  $0 \leq I_{A_k}, PA_k \leq 1$ , we clearly have  $|I_{A_k} - PA_k| \leq 1$ , and therefore

$$|X_n - EX_n| \leq \frac{1}{n} \sum_{k=1}^n |I_{A_k} - PA_k| \leq 1.$$

The result follows then by applying Corollary I.2.11. □

**Example I.2.13.** Suppose the events  $A_k$  are pairwise independent and  $PA_k = p$  for all  $k$ . By Example I.2.7,  $EX_n = p$  and  $\sigma^2(X_n) = pq/n \rightarrow 0$ , and consequently, by Corollary I.2.12,  $X_n \rightarrow p$  in probability. This is the *Bernoulli Law of Large Numbers* (the ‘success frequencies’ converge in probability to the probability of success when the number of trials tends to  $\infty$ ).

**Example I.2.14.** Let  $\{X_k\}$  be a sequence of pairwise uncorrelated  $L^2$ -random variables, with

$$EX_k = \mu, \quad \sigma(X_k) = \sigma \quad (k = 1, 2, \dots)$$

(e.g.,  $X_k$  is the outcome of the  $k$ th random drawing from an infinite population, or from a finite population with returns). Let

$$M_n := \frac{1}{n} \sum_{k=1}^n X_k$$

(the ‘sample mean’ for a sample of size  $n$ ). Then, by Lemma I.2.6,

$$E(M_n) = \mu; \quad \sigma^2(M_n) = \sigma^2/n.$$



By Lemma I.2.10,

$$P[|M_n - \mu| \geq \epsilon] \leq \frac{\sigma^2}{n\epsilon^2}$$

for any  $\epsilon > 0$ , and therefore  $M_n \rightarrow \mu$  in probability (when  $n \rightarrow \infty$ ). This is the so-called *Weak Law of Large Numbers* (the sample means converge to the expectation  $\mu$  when the sample size tends to infinity). The special case  $X_k = I_{A_k}$  for pairwise independent events  $A_k$  with  $PA_k = p$  is precisely the Bernoulli Law of Large Numbers of Example I.2.13.

The Bernoulli Law is generalized to *dependent* events in the next section.

**Theorem I.2.15 (Generalized Bernoulli law of large numbers).** *Let  $\{A_k\}$  be a sequence of events. Set*

$$p_1(n) := \frac{1}{n} \sum_{k=1}^n PA_k,$$

$$p_2(n) := \frac{1}{\binom{n}{2}} \sum_{1 \leq j < k \leq n} P(A_k \cap A_j),$$

and

$$d_n := p_2(n) - p_1^2(n).$$

Let  $X_n$  be as in Corollary I.2.12 [the occurrence frequency of the first  $n$  events].

Then  $X_n - EX_n \rightarrow 0$  in probability if and only if  $d_n \rightarrow 0$ .

**Proof.** By Lemma I.2.6 and relations (7) and (8) preceding it, we obtain by a straightforward calculation

$$\begin{aligned} \sigma^2(X_n) &= \frac{1}{n^2} \sum_{k=1}^n [PA_k - (PA_k)^2] + \frac{2}{n^2} \sum_{1 \leq j < k \leq n} [P(A_k \cap A_j) - PA_k \cdot PA_j] \\ &= d_n + (1/n)[p_1(n) - p_2(n)]. \end{aligned} \quad (9)$$

Therefore

$$|\sigma^2(X_n) - d_n| \leq (1/n)|p_1(n) - p_2(n)|.$$

However  $p_i(n)$  are arithmetical means of numbers in the interval  $[0, 1]$ , hence they belong to  $[0, 1]$ ; therefore  $|p_1 - p_2| \leq 1$  and so

$$|\sigma^2(X_n) - d_n| \leq \frac{1}{n}. \quad (10)$$

In particular,  $\sigma^2(X_n) \rightarrow 0$  iff  $d_n \rightarrow 0$ , and the theorem follows then from Corollary I.2.12.  $\square$

**Remark I.2.16.** Note that when the events  $A_k$  are pairwise independent,

$$\begin{aligned} |d_n| &= \left| \frac{2}{n(n-1)} \sum_{1 \leq j < k \leq n} P(A_k)P(A_j) - \frac{2}{n^2} \sum_{j < k} P(A_k)P(A_j) - \frac{1}{n^2} \sum_{k=1}^n P(A_k)^2 \right| \\ &= (1/n) \left| \frac{1}{\binom{n}{2}} \sum_{j < k} P(A_k)P(A_j) - \frac{1}{n} \sum_{k=1}^n P(A_k)^2 \right|. \end{aligned}$$

Both arithmetical means between the absolute value signs are means of numbers in  $[0,1]$ , and are therefore in  $[0,1]$ . The distance between them is thus  $\leq 1$ , hence  $|d_n| \leq 1/n$ , and the condition of the theorem is satisfied. Hence  $X_n - EX_n$  converge in probability to zero, even *without* the assumption  $PA_k = p$  (for all  $k$ ) of the Bernoulli case.

We consider next the stronger property of *almost sure convergence to zero* of  $X_n - EX_n$ .

**Theorem I.2.17 (Borel's strong law of large numbers).** *With notations as in Theorem I.2.15, suppose that  $d_n = O(1/n)$ . Then  $X_n - EX_n$  converge to zero almost surely.*

This happens in particular when the events  $A_k$  are pairwise independent, hence in the Bernoulli case.

**Proof.** We first prove the following:

**Lemma.** *Let  $\{X_n\}$  be any sequence of r.v.'s such that*

$$\sum_n P[|X_n| \geq 1/m] < \infty$$

*for all  $m = 1, 2, \dots$*

*Then  $X_n \rightarrow 0$  almost surely.*

**Proof of lemma.** Observe that by definition of convergence to 0,

$$[X_n \rightarrow 0] = \bigcap_m \bigcup_n \bigcap_k [|X_{n+k}| < 1/m],$$

where all indices run from 1 to  $\infty$ . By DeMorgan's laws, we then have

$$[X_n \rightarrow 0]^c = \bigcup_m \bigcap_n \bigcup_k [|X_{n+k}| \geq 1/m]. \quad (11)$$

Denote the 'innermost' union in (11) by  $B_{nm}$ , and let  $B_m := \bigcap_n B_{nm}$ . We have (by the  $\sigma$ -subadditivity of  $P$ ):

$$PB_{nm} \leq \sum_k P[|X_{n+k}| \geq 1/m] = \sum_{r=n+1}^{\infty} P[|X_r| \geq 1/m] \rightarrow_{n \rightarrow \infty} 0$$

by the Lemma's hypothesis, for all  $m$ . Therefore, since  $PB_m \leq PB_{nm}$  for all  $n$ , we have  $PB_m = 0$  (for all  $m$ ), and consequently

$$P([X_n \rightarrow 0]^c) = P\left(\bigcup_m B_m\right) = 0.$$

□

Back to the proof of the theorem, recall (10) from Section I.2.15:

$$|\sigma^2(X_n) - d_n| \leq 1/n.$$

Since  $|d_n| \leq c/n$  by hypothesis, we have  $\sigma^2(X_n) \leq (c+1)/n$ . By Tchebichev's inequality,

$$\begin{aligned} \sum_k P[|X_{k^2} - EX_{k^2}| \geq 1/m] &\leq m^2 \sum_k \sigma^2(X_{k^2}) \\ &\leq (c+1)m^2 \sum_k (1/k^2) < \infty. \end{aligned}$$

By the lemma, we then have almost surely

$$X_{k^2} - EX_{k^2} \rightarrow 0.$$

For each  $n \in \mathbb{N}$ , let  $k$  be the unique  $k \in \mathbb{N}$  such that

$$k^2 \leq n < (k+1)^2.$$

Necessarily  $n - k^2 \leq 2k$  and  $k \rightarrow \infty$  when  $n \rightarrow \infty$ . We have

$$\begin{aligned} |X_n - X_{k^2}| &= \left| \left( \frac{1}{n} - \frac{1}{k^2} \right) \sum_{j=1}^{k^2} I_{A_j} + \frac{1}{n} \sum_{j=k^2+1}^n I_{A_j} \right| \\ &\leq \frac{n - k^2}{nk^2} k^2 + \frac{n - k^2}{n} = 2 \frac{n - k^2}{n} \leq \frac{4k}{k^2} = 4/k. \end{aligned}$$

Hence also

$$|EX_{k^2} - EX_n| = |E(X_{k^2} - X_n)| \leq 4/k,$$

and therefore

$$\begin{aligned} |X_n - EX_n| &\leq |X_n - X_{k^2}| + |X_{k^2} - EX_{k^2}| + |EX_{k^2} - EX_n| \\ &\leq 8/k + |X_{k^2} - EX_{k^2}| \rightarrow 0 \end{aligned}$$

almost surely, when  $n \rightarrow \infty$ . □

**I.2.18.** Let  $\{A_n\}$  be a sequence of events. Recall the notation

$$\limsup A_n := \bigcap_{k=1}^{\infty} \bigcup_{j=k}^{\infty} A_j.$$

This event occurs iff for each  $k \in \mathbb{N}$ , there exists  $j \geq k$  such that  $A_j$  occurs, that is, iff *infinitely many  $A_n$  occur*.

**Lemma I.2.19 (The Borel–Cantelli lemma).** *If*

$$\sum P(A_n) < \infty, \tag{*}$$

*then*

$$P(\limsup A_n) = 0.$$

**Proof.**

$$P(\limsup A_n) \leq P\left(\bigcup_{j=k}^{\infty} A_j\right) \leq \sum_{j=k}^{\infty} P(A_j)$$

for all  $k$ , and the conclusion follows from  $(*)$  by letting  $k \rightarrow \infty$ .  $\square$

**Example I.2.20 (the mouse problem).** Consider a row of three connected chambers, denoted  $L$  (left),  $M$  (middle), and  $R$  (right). Chamber  $R$  has also a right exit to ‘freedom’ ( $F$ ). Chamber  $L$  has also a left exit to a ‘death’ trap  $D$ . A mouse, located originally in  $M$ , moves to a chamber to its right (left) with a fixed probability  $p$  ( $q := 1 - p$ ). The moves between chambers are independent. Thus, after  $2^1$  moves, we have

$$P(F_1) = p^2; \quad P(D_1) = q^2; \quad P(M_1) = 2pq,$$

where  $F_1, D_1, M_1$  denote, respectively, the events that the mouse reaches  $F, D$ , or  $M$  after precisely  $2^1$  moves. In general, let  $M_k$  denote the event that the mouse reaches back  $M$  (for the first time) after precisely  $2^k$  moves. Clearly

$$P(M_k) = (2pq)^k.$$

Since  $\sum P(M_k) < \infty$ , the Borel–Cantelli lemma implies that

$$P(\limsup M_n) = 0,$$

that is, with probability 1, there exists  $k \in \mathbb{N} \cup \{0\}$  such that the mouse moves either to  $F$  or to  $D$  at its  $2^{k+1}$ -th move. The probability of these events is, respectively  $(2pq)^k p^2$  and  $(2pq)^k q^2$ . Denoting also by  $F$  (or  $D$ ) the event that the mouse reaches freedom (or death) in *some* move, then  $F$  is the disjoint union of the  $F_k$  (and similarly for  $D$ ). Thus

$$PF = \sum_k (2pq)^k p^2 = \frac{p^2}{1 - 2pq}$$

and similarly

$$PD = \frac{q^2}{1 - 2pq}.$$

This is coherent with the preceding observation (that with probability 1, either  $F$  or  $D$  occurs), since the sum of these two probabilities is clearly 1.

The case of events with  $\sum P(A_n) = \infty$  is considered in Theorem I.2.21. Notation is as in Theorem I.2.15.

**Theorem I.2.21 (Erdos–Renyi).** *Let  $\{A_n\}$  be a sequence of events such that*

$$\sum P(A_n) = \infty \tag{12}$$

and

$$\liminf \frac{p_2(n)}{p_1^2(n)} = 1. \quad (13)$$

Then

$$P(\limsup A_n) = 1.$$

**Proof.** Let  $X_n = (1/n) \sum_{k=1}^n I_{A_k}$ . Then

$$EX_n = p_1(n)$$

and

$$\sigma^2(X_n) = p_2(n) - p_1^2(n) + (1/n)[p_1(n) - p_2(n)]$$

(cf. (9) in Theorem I.2.15). By Tchebichev's inequality,

$$\begin{aligned} P[|X_n - p_1(n)| \geq p_1(n)/2] &\leq \frac{\sigma^2(X_n)}{[p_1(n)/2]^2} \\ &= 4 \left[ \left(1 - \frac{1}{n}\right) \frac{p_2(n)}{p_1^2(n)} - 1 + \frac{1}{np_1(n)} \right]. \end{aligned}$$

By (12),

$$np_1(n) = \sum_{k=1}^n PA_k \rightarrow \infty. \quad (14)$$

Hence by (13), the  $\liminf$  of the right-hand side is 0. Thus

$$\liminf P[|X_n - p_1(n)| \geq p_1(n)/2] = 0. \quad (15)$$

Clearly

$$[X_n < p_1(n)/2] \subset [|X_n - p_1(n)| \geq p_1(n)/2],$$

and therefore

$$\liminf P[X_n < p_1(n)/2] = 0. \quad (16)$$

We may then choose a sequence of integers  $1 \leq n_1 < n_2 < \dots$  such that

$$\sum_k P[X_{n_k} < p_1(n_k)/2] < \infty.$$

By the Borel-Cantelli lemma,

$$P(\limsup [X_{n_k} < p_1(n_k)/2]) = 0,$$

that is, with probability 1,  $X_{n_k} < p_1(n_k)/2$  for only finitely many  $k$ s. Thus, with probability 1, there exists  $k_0$  such that  $X_{n_k} \geq p_1(n_k)/2$  for all  $k > k_0$ , that is,

$$\sum_{j=1}^{n_k} I_{A_j} \geq n_k p_1(n_k)/2$$

for all  $k > k_0$ , and since the right-hand side diverges to  $\infty$  by (14), we have

$$\sum_{j=1}^{\infty} I_{A_j} = \infty$$

with probability 1, that is, infinitely many  $A_j$ s occur with probability 1.  $\square$

**Corollary I.2.22.** *Let  $\{A_n\}$  be a sequence of pairwise independent events such that*

$$\sum PA_n = \infty.$$

*Then*

$$P(\limsup A_n) = 1.$$

**Proof.** We show that Condition (13) of the Erdos–Renyi theorem is satisfied. We have

$$\begin{aligned} \binom{n}{2} p_2(n) &= \sum_{1 \leq j < k \leq n} P(A_k \cap A_j) = (1/2) \sum_{j \neq k, 1 \leq j, k \leq n} P(A_j)P(A_k) \\ &= (1/2) \left[ \sum_{j,k=1}^n P(A_j)P(A_k) - \sum_{k=1}^n P(A_k)^2 \right] \\ &= (1/2) \left[ n^2 p_1^2(n) - \sum_{k=1}^n P(A_k)^2 \right]. \end{aligned}$$

Therefore

$$\frac{p_2(n)}{p_1^2(n)} = \frac{n}{n-1} - \frac{\sum_{k=1}^n P(A_k)^2}{n(n-1)p_1^2(n)}.$$

However, since  $PA_k \leq 1$ , the sum above is  $\leq \sum_{k=1}^n PA_k := np_1(n)$ . Therefore, the second (non-negative) term on the right-hand side is  $\leq 1/[(n-1)p_1(n)] \rightarrow 0$  by (14) (consequence of the divergence hypothesis). Hence,  $\lim p_2(n)/p_1^2(n) = 1$ .  $\square$

**Corollary I.2.23 (Zero-one law).** *Let  $\{A_n\}$  be a sequence of pairwise independent events. Then the event  $\limsup A_n$  (that infinitely many  $A_n$ 's occur) has probability 0 or 1, according to whether the series  $\sum PA_n$  converges or diverges (respectively).*

### I.3 Probability distributions

Let  $(\Omega, \mathcal{A}, P)$  be a probability space, let  $X$  be a real- (or complex-, or  $\mathbb{R}^n$ -) valued random variable on  $\Omega$ , and let  $\mathcal{B}$  denote the Borel  $\sigma$ -algebra of the range space ( $\mathbb{R}$ , etc.). We set

$$P_X(B) := P[X \in B] = P(X^{-1}(B)) \quad (B \in \mathcal{B}).$$

The set function  $P_X$  is called the *distribution* of  $X$ .

**Theorem I.3.1** (stated for the case of a real r.v.).

- (1)  $(\mathbb{R}, \mathcal{B}, P_X)$  is a probability space.  
 (2) For any finite Borel function  $g$  on  $\mathbb{R}$ , the distribution of the r.v.  $g(X)$  is given by

$$P_{g(X)}(B) = P_X(g^{-1}(B)) \quad (B \in \mathcal{B}).$$

- (3) If the Borel function  $g$  is integrable with respect to  $P_X$ , then

$$E(g(X)) = \int_{\mathbb{R}} g dP_X.$$

**Proof.**

- (1) Clearly,  $0 \leq P_X \leq 1$ , and  $P_X(\mathbb{R}) = P(X^{-1}(\mathbb{R})) = P(\Omega) = 1$ . If  $\{B_k\} \subset \mathcal{B}$  is a sequence of mutually disjoint sets, then the sets  $X^{-1}(B_k)$  are mutually disjoint sets in  $\mathcal{A}$ , and therefore

$$\begin{aligned} P_X\left(\bigcup_k B_k\right) &:= P\left(X^{-1}\left(\bigcup_k B_k\right)\right) = P\left(\bigcup_k X^{-1}(B_k)\right) \\ &= \sum_k P(X^{-1}(B_k)) = \sum_k P_X(B_k). \end{aligned}$$

- (2) We have for all  $B \in \mathcal{B}$ :

$$P_{g(X)}(B) := P(g(X)^{-1}(B)) = P(X^{-1}(g^{-1}(B))) := P_X(g^{-1}(B)).$$

- (3) If  $g = I_B$  for some  $B \in \mathcal{B}$ , then  $g(X) = I_{X^{-1}(B)}$ , and therefore

$$E(g(X)) = P(X^{-1}(B)) := P_X(B) = \int_{\mathbb{R}} g dP_X.$$

By linearity, Statement (3) is then valid for simple Borel functions  $g$ . If  $g$  is a non-negative Borel function, there exists an increasing sequence of non-negative simple Borel functions converging pointwise to  $g$ , and (3) follows from the Monotone Convergence theorem (applied to the measures  $P$  and  $P_X$ ). If  $g$  is any (real) Borel function in  $L^1(P_X)$ , then, by the preceding case,  $E(|g(X)|) = \int_{\mathbb{R}} |g| dP_X < \infty$ , that is,  $g(X) \in L^1(P)$ , and  $E(g(X)) = E(g^+(X) - g^-(X)) = \int_{\mathbb{R}} g^+ dP_X - \int_{\mathbb{R}} g^- dP_X = \int_{\mathbb{R}} g dP_X$ .

The routine extension to complex  $g$  is omitted.  $\square$

**Definition I.3.2.** The *distribution function* of the real r.v.  $X$  is the function

$$F_X(x) := P_X((-\infty, x)) = P[X < x] \quad (x \in \mathbb{R}).$$

The integral  $\int_{\mathbb{R}} g dP_X$  is also denoted  $\int_{\mathbb{R}} g dF_X$ .

**Proposition I.3.3.**  $F_X$  is a non-decreasing, left-continuous function with range in  $[0, 1]$ , such that

$$F_X(-\infty) = 0; \quad F_X(\infty) = 1. \quad (*)$$

**Proof.** Exercise.

**Definition I.3.4.** Any function  $F$  with the properties listed in Proposition I.3.3 is called a *distribution function*. If Property (\*) is omitted, the function  $F$  is called a *quasi-distribution function*.

Any (quasi-) distribution function induces a unique finite positive Lebesgue–Stieltjes measure (it is a probability measure on  $\mathbb{R}$  if (\*) is satisfied), and integration with respect to that measure is denoted by  $\int_{\mathbb{R}} g dF$ . In case  $g$  is a bounded continuous function on  $\mathbb{R}$ , this integral coincides with the (improper) Riemann–Stieltjes integral  $\int_{-\infty}^{\infty} g(x) dF(x)$ .

The *characteristic function* of the (quasi-) distribution function  $F$  is defined by

$$f(u) := \int_{\mathbb{R}} e^{iux} dF(x) \quad (x \in \mathbb{R}).$$

By Theorem I.3.1, if  $F = F_X$  for a real r.v.  $X$ , then  $f$  coincides with the ch.f.  $f_X$  of Definition I.2.3.

In general, the ch.f.  $f$  is a uniformly continuous function on  $\mathbb{R}$ ,  $|f| \leq 1$ , and  $f(0) = 1$  in case  $F$  satisfies (\*).

**Proposition I.3.5.** Let  $X$  be a real r.v.,  $b > 0$ , and  $a \in \mathbb{R}$ . Then

$$f_{a+bX}(u) = e^{iua} f_X(bu) \quad (u \in \mathbb{R}).$$

**Proof.** Write  $Y = a + bX$  and  $y = a + bx$  ( $x \in \mathbb{R}$ ). Since  $b > 0$ ,

$$F_X(x) = P[X < x] = P[Y < y] = F_Y(y),$$

and therefore

$$f_Y(u) = \int_{-\infty}^{\infty} e^{iuy} dF_Y(y) = \int_{-\infty}^{\infty} e^{iu(a+bx)} dF_X(x) = e^{iua} f_X(bu).$$

□

We shall consider r.v.s of class  $L^r(P)$  for  $r \geq 0$ . The  $L^r$ -‘norm’, denoted  $\|X\|_r$ , satisfies (by Theorem I.3.1)

$$\|X\|_r^r := E(|X|^r) = \int_{-\infty}^{\infty} |x|^r dF_X(x).$$

This expression is called the  $r$ th *absolute central moment* of  $X$  (or of  $F_X$ ). It always exists, but could be infinite (unless  $X \in L^r$ ).

The  $r$ th central moment of  $X$  (or  $F_X$ ) is

$$m_r := E(X^r) = \int_{-\infty}^{\infty} x^r dF_X(x).$$

These concepts are used with any quasi-distribution function  $F$ , whenever they make sense.



**Lemma I.3.6.**

- (1) The function  $\phi(r) := \log E(|X|^r)$  is convex on  $[0, \infty)$ , and  $\phi(0) = 0$ .  
 (2)  $\|X\|_r$  is a non-decreasing function of  $r$ . In particular, if  $X$  is of class  $L^r$  for some  $r > 0$ , then it is of class  $L^s$  for all  $0 \leq s \leq r$ .

**Proof.** For any r.v.s  $Y, Z$ , Schwarz's inequality gives

$$E(|YZ|) \leq \|Y\|_2 \|Z\|_2.$$

For  $r \geq s \geq 0$ , choose  $Y = |X|^{(r-s)/2}$  and  $Z = |X|^{(r+s)/2}$ . Then

$$E(|X|^r) \leq [E(|X|^{r-s})E(|X|^{r+s})]^{1/2},$$

so that

$$\phi(r) \leq (1/2)[\phi(r-s) + \phi(r+s)],$$

and (1) follows.

The slope of the chord joining the points  $(0,0)$  and  $(r, \phi(r))$  on the graph of  $\phi$  is  $\phi(r)/r$ , and it increases with  $r$ , by convexity of  $\phi$ . Therefore  $\|X\|_r = e^{\phi(r)/r}$  increases with  $r$ .  $\square$

**Theorem I.3.7.** Let  $X$  be an  $L^r$ -r.v. for some  $r \geq 1$ , and let  $f = f_X$ . Then for all integers  $1 \leq k \leq r$ , the derivative  $f^{(k)}$  exists, is uniformly continuous and bounded by  $E(|X|^k)(< \infty)$ , and is given by

$$f^{(k)}(u) = i^k \int_{-\infty}^{\infty} e^{iux} x^k dF(x), \quad (*)$$

(where  $F := F_X$ ). In particular, the moment  $m_k$  exists and is given by

$$m_k = f^{(k)}(0)/i^k \quad (k = 1, \dots, [r]).$$

**Proof.** By Lemma I.3.6,  $E(|X|^k) < \infty$  for  $k \leq r$ , and therefore the integral in (\*) converges absolutely and defines a continuous function  $g_k(u)$ . Also, by Fubini's theorem,

$$\begin{aligned} \int_0^t g_k(u) du &= \int_{-\infty}^{\infty} \int_0^t i^k e^{iux} du x^k dF(x) \\ &= g_{k-1}(t) - g_{k-1}(0). \end{aligned}$$

Assuming (\*) for  $k-1$ , the last expression is equal to  $f^{(k-1)}(t) - f^{(k-1)}(0)$ . Since the left-hand side is differentiable, with derivative  $g_k(t)$ , it follows that  $f^{(k)}$  exists and equals  $g_k$ . Since (\*) reduces to the definition of  $f$  for  $k = 0$ , Relation (\*) for general  $k \leq r$  follows by induction.  $\square$

**Corollary I.3.8.** If the r.v.  $X$  is in  $L^k$  for all  $k = 1, 2, \dots$ , and if  $f := f_X$  is analytic in some real interval  $|u| < R$ , then

$$f(u) = \sum_{k=0}^{\infty} i^k m_k u^k / k! \quad (|u| < R).$$

**Example I.3.9 (discrete r.v.).** Let  $X$  be a *discrete* real r.v., that is, its range is the set  $\{x_k\}$ , with  $x_k \in \mathbb{R}$  distinct, and let  $P[X = x_k] = p_k$  ( $\sum p_k = 1$ ). Then

$$f_X(u) = \sum_k e^{iux_k} p_k. \quad (1)$$

For the Bernoulli r.v. with parameters  $n, p$ , we have  $x_k = k$  ( $0 \leq k \leq n$ ) and  $p_k = \binom{n}{k} p^k q^{n-k}$  ( $q := 1 - p$ ). A short calculation starting from (1) gives

$$f_X(u) = (pe^{iu} + q)^n.$$

By Theorem I.3.7,

$$\begin{aligned} m_1 &= f'_X(0)/i = np, \\ m_2 &= f''_X(0)/i^2 = (np)^2 + npq, \end{aligned}$$

and therefore

$$\sigma^2(X) = m_2 - m_1^2 = npq$$

(cf. Example I.2.7).

The *Poisson r.v.*  $X$  with parameter  $\lambda > 0$  assumes exclusively the values  $k$  ( $k = 0, 1, 2, \dots$ ) with

$$P[X = k] = e^{-\lambda} \lambda^k / k!.$$

Then  $F_X(x) = 0$  for  $x \leq 0$ , and  $= e^{-\lambda} \sum_{k < x} \lambda^k / k!$  for  $x > 0$ . Clearly  $F_X(\infty) = 1$ , and by (1),

$$f_X(u) = e^{-\lambda} \sum_k e^{iuk} \lambda^k / k! = e^{\lambda(e^{iu} - 1)}.$$

Note that  $E(|X|^n) = e^{-\lambda} \sum_k k^n \lambda^k / k! < \infty$  for all  $n$ , and Corollary I.3.8 implies then that  $m_k = f^{(k)}(0)/i^k$  for all  $k$ . Thus, for example, one calculates that

$$m_1 = \lambda, \quad m_2 = \lambda(\lambda + 1),$$

and therefore

$$\sigma^2(X) = \lambda.$$

(This can be reached of course directly from the definitions.)

The Poisson distribution is the limit of Bernoulli distributions in the following sense:

**Proposition I.3.10.** Let  $\{A_{k,n}; k = 0, \dots, n; n = 1, 2, \dots\}$  be a 'triangular array' of events such that, for each  $n = 1, 2, \dots$ ,  $\{A_{k,n}; k = 0, \dots, n\}$  is a Bernoulli system with parameter  $p_n = \lambda/n$  ( $\lambda > 0$  fixed) (cf. Example I.2.7). Let  $X_n := \sum_{k=0}^n I_{A_{k,n}}$  be the Bernoulli r.v. corresponding to the  $n$ th system. Then the distribution of  $X_n$  converges pointwise to the Poisson distribution when  $n \rightarrow \infty$ :

$$P[X_n = k] \rightarrow_{n \rightarrow \infty} e^{-\lambda} \lambda^k / k! \quad (k = 0, 1, 2, \dots).$$

**Proof.**

$$\begin{aligned}
 P[X_n = k] &= \binom{n}{k} (\lambda/n)^k (1 - \lambda/n)^{n-k} \\
 &= \left(1 - \frac{1}{n}\right) \cdots \left(1 - \frac{k-1}{n}\right) \frac{\lambda^k}{k!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-k} \\
 &\rightarrow_{n \rightarrow \infty} \frac{\lambda^k}{k!} e^{-\lambda}.
 \end{aligned}$$

□

**Example I.3.11 (Distributions with density).** If there exists an  $L^1(dx)$ -function  $h$  such that the distribution function  $F$  has the form

$$F(x) = \int_{-\infty}^x h(t) dt \quad (x \in \mathbb{R}),$$

then  $h$  is uniquely determined a.e. (with respect to Lebesgue measure  $dx$  on  $\mathbb{R}$ ), and one has a.e.  $F'(x) = h(x)$ . In particular,  $h \geq 0$  a.e., and since it is only determined a.e., one assumes that  $h \geq 0$  everywhere, and one calls  $h$  (or  $F'$ ) the *density* of  $F$ . For any  $g \in L^1(F)$ ,

$$\int_B g dF = \int_B g F' dx \quad (B \in \mathcal{B}).$$

We consider a few common densities.

**I.3.12. The normal density.**

The ‘standard’ normal (or Gaussian) density is

$$F'(x) = (2\pi)^{-1/2} e^{-x^2/2}.$$

To verify that the corresponding  $F$  is a distribution function, we need only to show that  $F(\infty) = 1$ . We write

$$F(\infty)^2 = (1/2\pi) \iint_{\mathbb{R}^2} e^{-(t^2+s^2)/2} dt ds = (1/2\pi) \int_0^{2\pi} \int_0^\infty e^{-r^2/2} r dr d\theta = 1,$$

where we used polar coordinates. The ch.f. is

$$\begin{aligned}
 f(u) &= (2\pi)^{-1/2} \int_{\mathbb{R}} e^{iux} e^{-x^2/2} dx = (2\pi)^{-1/2} \int_{\mathbb{R}} e^{-[(x-iu)^2+u^2]/2} dx \\
 &= c(u) e^{-u^2/2},
 \end{aligned}$$

where  $c(u) = (2\pi)^{-1/2} \int_{\mathbb{R}} e^{-(x-iu)^2/2} dx$ . The Cauchy Integral theorem is applied to the entire function  $e^{-z^2/2}$ , with the rectangular path having vertices at  $-M$ ,  $N$ ,  $N - iu$ , and  $-M - iu$ ; letting then  $M, N \rightarrow \infty$ , one sees that  $c(u) = (2\pi)^{-1/2} \int_{-\infty}^\infty e^{-x^2/2} dx = F(\infty) = 1$ . Thus the ch.f. of the standard normal distribution is  $e^{-u^2/2}$ .

If the r.v.  $X$  has the standard normal distribution, and  $Y = a + bX$  with  $a \in \mathbb{R}$  and  $b > 0$ , then for  $y = a + bx$ , we have (through the change of variable  $s = a + bt$ ):

$$F_Y(y) = F_X(x) = (2\pi)^{-1/2} \int_{-\infty}^x e^{-t^2/2} dt = (2\pi b^2)^{-1/2} \int_{-\infty}^y e^{-(s-a)^2/2b^2} ds.$$

This distribution is called the normal (or Gaussian) distribution with parameters  $a, b^2$  (or briefly, the  $N(a, b^2)$  distribution). By Proposition I.3.5, its ch.f. is

$$f_Y(u) = e^{iua} e^{-(bu)^2/2}.$$

Since  $E(|Y|^n) < \infty$  for all  $n$ , Corollary I.3.8 applies. In particular, one calculates that

$$m_1 = f'_Y(0)/i = a; \quad m_2 = -f''_Y(0) = a^2 + b^2,$$

so that

$$\sigma^2(Y) = m_2 - m_1^2 = b^2.$$

Thus, the parameters of the  $N(a, b^2)$  distribution are its expectation and its variance.

For the  $N(0, 1)$  distribution, we write the power series for  $f(u) = e^{-u^2/2}$  and deduce from Corollary I.3.8 that  $m_{2j+1} = 0$  and  $m_{2j} = (2j)!/(2^j j!)$ ,  $j = 0, 1, 2, \dots$

**Proposition.** *The sum of independent normally distributed r.v.s is normally distributed.*

**Proof.** Let  $X_k$  be  $N(a_k, b_k^2)$  distributed independent r.v.s ( $k = 1, \dots, n$ ), and let  $X = X_1 + \dots + X_n$ . By Corollary I.2.4,

$$f_X(u) = \prod_k f_{X_k}(u) = \prod_k e^{iua_k} e^{-b_k^2 u^2/2} = e^{iua} e^{-b^2 u^2/2}$$

with  $a = \sum_k a_k$  and  $b^2 = \sum_k b_k^2$ . By the Uniqueness theorem for ch.f.s (see below),  $X$  is  $N(a, b^2)$  distributed.  $\square$

**I.3.13.** *The Laplace density.*

$$F'(x) = (1/2b)e^{-|x-a|/b},$$

where  $a \in \mathbb{R}$  and  $b > 0$  are its 'parameters'. One calculates that

$$f(u) = \frac{e^{iua}}{1 + b^2 u^2}.$$

In particular  $F(\infty) = f(0) = 1$ , so that  $F$  is indeed a distribution function. One verifies that

$$m_1 = f'(0)/i = a; \quad m_2 = f''(0)/i^2 = a^2 + b^2,$$

so that  $\sigma^2 = b^2$ .

**I.3.14.** *The Cauchy density.*

$$F'(x) = (1/\pi) \frac{b}{b^2 + (x-a)^2},$$

where  $a \in \mathbb{R}$  and  $b > 0$  are its ‘parameters’. To calculate its ch.f., one uses the Residues theorem with positively oriented rectangles in  $\Re z \geq 0$  and in  $\Re z \leq 0$  for  $u \geq 0$  and  $u \leq 0$ , respectively, for the function  $e^{iuz}/(b^2 + (z-a)^2)$ . One gets

$$f(u) = e^{iua} e^{-b|u|}. \quad (2)$$

In particular,  $f(0) = 1$ , so that  $F(\infty) = 1$  as needed.

Note that  $f$  is not differentiable at 0. Also  $m_k$  do not exist for  $k \geq 1$ .

As in the case of normal r.v.s, it follows from (2) that the sum of independent Cauchy-distributed r.v.s is Cauchy-distributed, with parameters equal to the sum of the corresponding parameters. This property is not true however for Laplace-distributed r.v.s.

**I.3.15.** *The Gamma density.*

$$F'(x) = \frac{b^p}{\Gamma(p)} x^{p-1} e^{-bx} \quad (x > 0),$$

and  $F(x) = 0$  for  $x \leq 0$ . The ‘parameters’  $b, p$  are positive. The special case  $p = 1$  gives the *exponential distribution density*.

The function  $F$  is trivially non-decreasing, continuous,  $F(-\infty) = 0$ , and

$$F(\infty) = \Gamma(p)^{-1} \int_0^\infty e^{-bx} (bx)^{p-1} d(bx) = 1,$$

so that  $F$  is indeed a distribution function. We have

$$\begin{aligned} f(u) &= \frac{b^p}{\Gamma(p)} \int_0^\infty x^{p-1} e^{-(b-iu)x} dx \\ &= (b-iu)^{-p} \frac{b^p}{\Gamma(p)} \int_0^\infty [(b-iu)x]^{p-1} e^{-(b-iu)x} d(b-iu)x. \end{aligned}$$

By Cauchy’s Integral theorem, integration along the ray  $\{(b-iu)x; x \geq 0\}$  can be replaced by integration along the ray  $[0, \infty)$ , and therefore the above integral equals  $\Gamma(p)$ , and

$$f(u) = (1-iu/b)^{-p}. \quad (3)$$

Thus

$$f^{(k)}(u) = p(p+1) \cdots (p+k-1)(i/b)^k (1-iu/b)^{-(p+k)}$$

and

$$m_k = f^{(k)}(0)/i^k = p(p+1) \cdots (p+k-1)b^{-k}.$$

In particular,

$$m_1 = p/b, \quad m_2 = p(p+1)/b^2, \quad \sigma^2 = m_2 - m_1^2 = p/b^2.$$

As in the case of the normal distribution, (3) implies the following

**Proposition 1.** *The sum of independent Gamma-distributed r.v.s with parameters  $b, p_k, k = 1, \dots, n$ , is Gamma-distributed with parameters  $b, \sum p_k$ .*

Note the special case of the exponential distribution ( $p = 1$ ):

$$f(u) = (1 - iu/b)^{-1}; \quad m_1 = 1/b; \quad \sigma^2 = 1/b^2.$$

Another important special case has  $p = b = 1/2$ . The Gamma distribution with these parameters is called the *standard  $\chi^2$  distribution with one degree of freedom*. Its density equals 0 for  $x \leq 0$ , and since  $\Gamma(1/2) = \pi^{1/2}$ ,

$$F'(x) = (2\pi)^{-1/2} x^{-1/2} e^{-x/2}$$

for  $x > 0$ .

By Proposition 1, the sum of  $n$  independent random variables with the standard  $\chi^2$ -distribution has the *standard  $\chi^2$ -distribution with  $n$  degrees of freedom*, that is, the Gamma distribution with  $p = n/2$  and  $b = 1/2$ . Its density for  $x > 0$  is

$$F'(x) = \frac{1}{2^{n/2} \Gamma(n/2)} x^{n/2-1} e^{-x/2}.$$

Note that  $m_1 = p/b = n$  and  $\sigma^2 = p/b^2 = 2n$ .

The  $\chi^2$  distribution arises naturally as follows. Suppose  $X$  is a real r.v. with continuous distribution  $F_X$ , and let  $Y = X^2$ . Then  $F_Y(x) := P[Y < x] = 0$  for  $x \leq 0$ , and for  $x > 0$ ,

$$F_Y(x) = P[-x^{1/2} < X < x^{1/2}] = F_X(x^{1/2}) - F_X(-x^{1/2}).$$

If  $F'_X$  exists and is continuous on  $\mathbb{R}$ , then

$$F'_Y(x) = \frac{1}{2x^{1/2}} [F'_X(x^{1/2}) + F'_X(-x^{1/2})]$$

for  $x > 0$  (and trivially 0 for  $x < 0$ ). In particular, if  $X$  is  $N(0, 1)$ -distributed, then  $Y = X^2$  has the density  $(2\pi)^{-1/2} x^{-1/2} e^{-x/2}$  for  $x > 0$  (and 0 for  $x < 0$ ), which is precisely the standard  $\chi^2$  density for one degree of freedom. Consequently, we have

**Proposition 2.** *Let  $X_1, \dots, X_n$  be  $N(0, 1)$ -distributed independent r.v.s and let*

$$\chi^2 := \sum_{k=1}^n X_k^2.$$

*Then  $F_{\chi^2}$  is the standard  $\chi^2$  distribution with  $n$  degrees of freedom (denoted  $F_{\chi^2, n}$ ).*

If we start with  $N(\mu, \sigma^2)$  independent r.v.s  $X_k$ , the *standardized r.v.s*

$$Z_k := \frac{X_k - \mu}{\sigma}$$

are independent  $N(0, 1)$  variables, and therefore the sum  $V := \sum_{k=1}^n Z_k^2$  has the  $F_{\chi^2, n}$  distribution. Hence, if we let

$$\chi_\sigma^2 := \sum_{k=1}^n (X_k - \mu)^2,$$

then, for  $x > 0$ ,

$$F_{\chi_\sigma^2}(x) = P[V < x/\sigma^2] = F_{\chi^2, n}(x/\sigma^2).$$

This is clearly the Gamma distribution with parameters  $p = n/2$  and  $b = 1/2\sigma^2$ . In particular, we have then  $m_1 = p/b = n\sigma^2$  (not surprisingly!), and  $\sigma^2(\chi_\sigma^2) = p/b^2 = 2n\sigma^4$ .

## I.4 Characteristic functions

Let  $F$  be a distribution function on  $\mathbb{R}$ . Its *normalization* is the distribution function

$$F^*(x) := (1/2)[F(x-0) + F(x+0)] = (1/2)[F(x) + F(x+0)]$$

(since  $F$  is left-continuous). Of course,  $F^*(x) = F(x)$  at all continuity points  $x$  of  $F$ .

**Theorem I.4.1 (The inversion theorem).** *Let  $f$  be the ch.f. of the distribution function  $F$ . Then*

$$F^*(b) - F^*(a) = \lim_{U \rightarrow \infty} \frac{1}{2\pi} \int_{-U}^U \frac{e^{-iua} - e^{-iub}}{iu} f(u) du$$

for  $-\infty < a < b < \infty$ .

**Proof.** Let  $J_U$  denote the integral on the right-hand side. By Fubini's theorem,

$$\begin{aligned} J_U &= (1/2\pi) \int_{-U}^U \frac{e^{-iua} - e^{-iub}}{iu} \int_{-\infty}^{\infty} e^{iux} dF(x) du \\ &= \int_{-\infty}^{\infty} K_U(x) dF(x), \end{aligned}$$

where

$$\begin{aligned} K_U(x) &:= (1/2\pi) \int_{-U}^U \frac{e^{iu(x-a)} - e^{iu(x-b)}}{iu} du \\ &= (1/\pi) \int_{U(x-b)}^{U(x-a)} \frac{\sin t}{t} dt. \end{aligned}$$

The convergence of the Dirichlet integral  $\int_{-\infty}^{\infty} (\sin t/t) dt$  (to the value  $\pi$ ) implies that  $|K_U(x)| \leq M < \infty$  for all  $x \in \mathbb{R}$  and  $U > 0$ , and as  $U \rightarrow \infty$ ,

$$K_U(x) \rightarrow \phi(x) := I_{(a,b)}(x) + (1/2)[I_{\{a\}} + I_{\{b\}}](x)$$

pointwise. Therefore, by dominated convergence,

$$\lim_{U \rightarrow \infty} J_U = \int_{\mathbb{R}} \phi dF = F^*(b) - F^*(a).$$

□

**Theorem I.4.2 (The uniqueness theorem).** *A distribution function is uniquely determined by its ch.f.*

**Proof.** Let  $F, G$  be distribution functions with ch.f.s  $f, g$ , and suppose that  $f = g$ . By the Inversion theorem,

$$F^*(b) - F^*(a) = G^*(b) - G^*(a)$$

for all real  $a < b$ . Letting  $a \rightarrow -\infty$ , we get  $F^* = G^*$ , and therefore  $F = G$  at all points where  $F, G$  are both continuous. Since these points are dense on  $\mathbb{R}$  and  $F, G$  are left-continuous, it follows that  $F = G$ . □

**Definition I.4.3.** Let  $C_F$  denote the set of all continuity points of the quasi-distribution function  $F$  (its complement in  $\mathbb{R}$  is finite or countable). A sequence  $\{F_n\}$  of quasi-distribution functions converges *weakly* to  $F$  if  $F_n(x) \rightarrow F(x)$  for all  $x \in C_F$ . One writes then  $F_n \rightarrow_w F$ . In case  $F_n, F$  are distributions of r.v.s  $X_n, X$  respectively, we also write  $X_n \rightarrow_w X$ .

**Lemma I.4.4 (Helly–Bray).** *Let  $F_n, F$  be quasi-distribution functions,  $F_n \rightarrow_w F$ , and suppose  $a < b$  are such that  $F_n(a) \rightarrow F(a)$  and  $F_n(b) \rightarrow F(b)$ . Then*

$$\int_a^b g dF_n \rightarrow \int_a^b g dF$$

for any continuous function  $g$  on  $[a, b]$ .

**Proof.** Consider partitions

$$a = x_{m,1} < \cdots < x_{m,k_m+1} = b, \quad x_{m,j} \in C_F,$$

such that

$$\delta_m := \sup_j (x_{m,j+1} - x_{m,j}) \rightarrow 0$$

as  $m \rightarrow \infty$ . Let

$$g_m = \sum_{j=1}^{k_m} g(x_{m,j}) I_{[x_{m,j}, x_{m,j+1})}.$$

Then

$$\sup_{[a,b]} |g - g_m| \rightarrow_{m \rightarrow \infty} 0. \quad (1)$$

The hypothesis implies that

$$F_n(x_{m,j+1}) - F_n(x_{m,j}) \rightarrow F(x_{m,j+1}) - F(x_{m,j})$$



when  $n \rightarrow \infty$ , for all  $j = 1, \dots, k_m$ ;  $m = 1, 2, \dots$ . Therefore

$$\begin{aligned} \int_a^b g_m dF_n &= \sum_{j=1}^{k_m} g(x_{m,j}) [F_n(x_{m,j+1}) - F_n(x_{m,j})] \\ &\rightarrow_{n \rightarrow \infty} \sum_{j=1}^{k_m} g(x_{m,j}) [F(x_{m,j+1}) - F(x_{m,j})] \\ &= \int_a^b g_m dF \quad (m = 1, 2, \dots). \end{aligned} \quad (2)$$

Write

$$\begin{aligned} &\left| \int_a^b g dF_n - \int_a^b g dF \right| \\ &\leq \int_a^b |g - g_m| dF_n + \left| \int_a^b g_m dF_n - \int_a^b g_m dF \right| + \int_a^b |g_m - g| dF \\ &\leq 2 \sup_{[a,b]} |g - g_m| + \left| \int_a^b g_m dF_n - \int_a^b g_m dF \right|. \end{aligned}$$

If  $\epsilon > 0$ , we may fix  $m$  such that  $\sup_{[a,b]} |g - g_m| < \epsilon/4$  (by (1)); for this  $m$ , it follows from (2) that there exists  $n_0$  such that the second summand above is  $< \epsilon/2$  for all  $n > n_0$ . Hence

$$\left| \int_a^b g dF_n - \int_a^b g dF \right| < \epsilon \quad (n > n_0).$$

□

We consider next integration over  $\mathbb{R}$ .

**Theorem I.4.5 (Helly–Bray).** *Let  $F_n, F$  be quasi-distribution functions such that  $F_n \rightarrow_w F$ . Then for every  $g \in C_0(\mathbb{R})$  (the continuous functions vanishing at  $\infty$ ),*

$$\int_{\mathbb{R}} g dF_n \rightarrow \int_{\mathbb{R}} g dF.$$

In case  $F_n, F$  are distribution functions, the conclusion is valid for all  $g \in C_b(\mathbb{R})$  (the bounded continuous functions on  $\mathbb{R}$ ).

**Proof.** Let  $\epsilon > 0$ . For  $a < b$  in  $C_F$  and  $g \in C_b(\mathbb{R})$ , write

$$\left| \int_{\mathbb{R}} g dF_n - \int_{\mathbb{R}} g dF \right| \leq \int_{[a,b]^c} |g| d(F_n + F) + \left| \int_a^b g dF_n - \int_a^b g dF \right|.$$

In case of quasi-distribution functions and  $g \in C_0$ , we may choose  $[a, b]$  such that  $|g| < \epsilon/4$  on  $[a, b]^c$ ; the first term on the right-hand side is then  $< \epsilon/2$  for all

$n$ . The second term is  $< \epsilon/2$  for all  $n > n_0$ , by Lemma I.4.4, and the conclusion follows.

In the case of distribution functions and  $g \in C_b$ , let  $M = \sup_{\mathbb{R}} |g|$ . Then the first term on the right-hand side is

$$\leq M[F_n(a) + 1 - F_n(b) + F(a) + 1 - F(b)].$$

Letting  $n \rightarrow \infty$ , we have by Lemma I.4.4

$$\limsup \left| \int_{\mathbb{R}} g dF_n - \int_{\mathbb{R}} g dF \right| \leq 2M[F(a) + 1 - F(b)],$$

for any  $a < b$  in  $C_F$ . The right-hand side is arbitrarily small, since  $F(-\infty) = 0$  and  $F(\infty) = 1$ .  $\square$

**Corollary I.4.6.** *Let  $F_n, F$  be distribution functions such that  $F_n \rightarrow_w F$ , and let  $f_n, f$  be their respective ch.f.s. Then  $f_n \rightarrow f$  pointwise on  $\mathbb{R}$ .*

In order to prove a converse to this corollary, we need the following:

**Lemma I.4.7 (Helly).** *Every sequence of quasi-distribution functions contains a subsequence converging weakly to a quasi-distribution function ('weak sequential compactness of the space of quasi-distribution functions').*

**Proof.** Let  $\{F_n\}$  be a sequence of quasi-distribution functions, and let  $\mathbb{Q} = \{x_n\}$  be the sequence of all rational points on  $\mathbb{R}$ .

Since  $\{F_n(x_1)\} \subset [0, 1]$ , the Bolzano–Weierstrass theorem asserts the existence of a convergent subsequence  $\{F_{n_1}(x_1)\}$ . Again,  $\{F_{n_1}(x_2)\} \subset [0, 1]$ , and has therefore a convergent subsequence  $\{F_{n_2}(x_2)\}$ , etc. Inductively, we obtain subsequences  $\{F_{n_k}\}$  such that the  $k$ th subsequence is a subsequence of the  $(k-1)$ th subsequence, and converges at the points  $x_1, \dots, x_k$ . The diagonal subsequence  $\{F_{nn}\}$  converges therefore at all the rational points. Let  $F_{\mathbb{Q}} := \lim_n F_{nn}$ , defined pointwise on  $\mathbb{Q}$ . For arbitrary  $x \in \mathbb{R}$ , define

$$F(x) := \sup_{r \in \mathbb{Q}: r \leq x} F_{\mathbb{Q}}(r).$$

Clearly,  $F$  is non-decreasing, has range in  $[0, 1]$ , and coincides with  $F_{\mathbb{Q}}$  on  $\mathbb{Q}$ . Its left-continuity is verified as follows: given  $\epsilon > 0$ , there exists a rational  $r < x$  (for  $x \in \mathbb{R}$  given) such that  $F_{\mathbb{Q}}(r) > F(x) - \epsilon$ . If  $t \in (r, x)$ ,

$$F(x) \geq F(t) \geq F(r) = F_{\mathbb{Q}}(r) > F(x) - \epsilon,$$

so that  $0 \leq F(x) - F(t) < \epsilon$ .

Thus  $F$  is a quasi-distribution function.

Given  $x \in C_F$ , if  $r, s \in \mathbb{Q}$  satisfy  $r < x < s$ , then

$$F_{nn}(r) \leq F_{nn}(x) \leq F_{nn}(s).$$

Therefore

$$F(r) = F_{\mathbb{Q}}(r) \leq \liminf F_{nn}(x) \leq \limsup F_{nn}(x) \leq F_{\mathbb{Q}}(s) = F(s).$$

Hence

$$F(x) := \sup_{r \in \mathbb{Q}; r \leq x} F(r) \leq \liminf F_{nn}(x) \leq \limsup F_{nn}(x) \leq F(s),$$

and since  $x \in C_F$ , letting  $s \rightarrow x+$ , we conclude that  $F_{nn}(x) \rightarrow F(x)$ .  $\square$

**Theorem I.4.8 (Paul Levy continuity theorem).** *Let  $F_n$  be distribution functions such that their ch.f.s  $f_n$  converge pointwise to a function  $g$  continuous at zero. Then there exists a distribution function  $F$  (with ch.f.  $f$ ) such that  $F_n \rightarrow_w F$  and  $f = g$ .*

**Proof.** Since  $|f_n| \leq 1$  (ch.f.s!) and  $f_n \rightarrow g$  pointwise, it follows by Dominated Convergence that

$$\int_0^u f_n(t) dt \rightarrow \int_0^u g(t) dt \quad (u \in \mathbb{R}). \quad (3)$$

By Helly's lemma (Lemma I.4.7), there exists a subsequence  $\{F_{n_k}\}$  converging weakly to a quasi-distribution function  $F$ . Let  $f$  be its ch.f. By Fubini's theorem and Theorem I.4.5,

$$\begin{aligned} \int_0^u f_{n_k}(t) dt &= \int_{\mathbb{R}} \frac{e^{iux} - 1}{ix} dF_{n_k}(x) \\ &\rightarrow_{k \rightarrow \infty} \int_{\mathbb{R}} \frac{e^{iux} - 1}{ix} dF(x) = \int_0^u f(t) dt. \end{aligned}$$

By (3), it follows that  $\int_0^u g(t) dt = \int_0^u f(t) dt$  for all real  $u$ , and since both  $g$  and  $f$  are continuous at zero, it follows that  $f(0) = g(0) := \lim f_n(0) = 1$ , that is,  $F(\infty) - F(-\infty) = 1$ , hence necessarily  $F(-\infty) = 0$  and  $F(\infty) = 1$ . Thus  $F$  is a distribution function. By Corollary I.4.6, any distribution function that is the weak limit of some subsequence of  $\{F_n\}$  has the ch.f.  $g$ , and therefore, by the Uniqueness Theorem (Theorem I.4.2), the full sequence  $\{F_n\}$  converges weakly to  $F$ .  $\square$

We proceed now to prove Lyapounov's Central Limit theorem.

**Lemma I.4.9.** *Let  $X$  be a real r.v. of class  $L^r$  ( $r \geq 2$ ), and let  $f := f_X$ . Then for any non-negative integer  $n \leq r - 1$ ,*

$$f(u) = \sum_{k=0}^n m_k (iu)^k / k! + R_n(u),$$

with

$$|R_n(u)| \leq E(|X|^{n+1}) |u|^{n+1} / (n+1)!,$$

for all  $u \in \mathbb{R}$ .

In particular, if  $X$  is a central  $L^3$ -r.v., then

$$f(u) = 1 - \sigma^2 u^2 / 2 + R_2(u),$$

where  $\sigma^2 := \sigma^2(X)$  and

$$|R_2(u)| \leq E(|X|^3)|u|^3/3! \quad (u \in \mathbb{R}).$$

**Proof.** Apply Theorem I.3.7 and Taylor's formula.  $\square$

Consider next a sequence of *independent central real r.v.s*  $X_k, k = 1, 2, \dots$  of class  $L^3$ . Denote  $\sigma_k := \sigma(X_k)$ . We assume that  $\sigma_k \neq 0$  (i.e.  $X_k$  is 'non-degenerate', which means that  $X_k$  is not a.s. zero) for all  $k$ . We fix the following notation:

$$f_k := f_{X_k}; \quad S_n := \sum_{k=1}^n X_k; \quad s_n := \sigma(S_n).$$

Of course,  $s_n^2 = \sum_{k=1}^n \sigma_k^2$ . In particular,  $s_n \neq 0$  for all  $n$ , and we may consider the 'standardized' sums  $S_n/s_n$ . We denote their ch.f.s by  $\phi_n$ :

$$\phi_n(u) := f_{S_n/s_n}(u) = \prod_{k=1}^n f_k(u/s_n). \quad (4)$$

Finally, we let

$$M_n = s_n^{-3} \sum_{k=1}^n E(|X_k|^3) \quad (n = 1, 2, \dots).$$

**Lemma I.4.10.** *Let  $\{X_k\}$  be as above, and suppose that  $M_n \rightarrow 0$  ('Lyapounov's Condition'). Then  $\phi_n(u) \rightarrow e^{-u^2/2}$  pointwise everywhere on  $\mathbb{R}$ .*

**Proof.** By Lemma I.3.6, for all  $k$ ,

$$\sigma_k \leq \|X_k\|_3. \quad (5)$$

Therefore

$$\frac{\sigma_k}{s_n} \leq \left[ \frac{E(|X_k|^3)}{s_n^3} \right]^{1/3} \leq M_n^{1/3}, \quad k = 1, \dots, n; \quad n = 1, 2, \dots \quad (6)$$

By (4) and Lemma I.4.9,

$$\log \phi_n(u) = \sum_{k=1}^n \log \left[ 1 - \frac{\sigma_k^2}{s_n^2} u^2/2 + R_2^{[k]}(u/s_n) \right]$$

and

$$|R_2^{[k]}(u/s_n)| \leq \frac{E(|X_k|^3)}{s_n^3} |u|^3/3!. \quad (7)$$

Write

$$\begin{aligned}
 |\log \phi_n(u) - (-u^2/2)| &= \left| \sum_{k=1}^n \log \left[ 1 - \frac{\sigma_k^2}{s_n^2} u^2/2 + R_2^{[k]}(u/s_n) \right] \right. \\
 &\quad \left. - \sum_{k=1}^n \left[ -\frac{\sigma_k^2}{s_n^2} u^2/2 + R_2^{[k]}(u/s_n) \right] + \sum_{k=1}^n R_2^{[k]}(u/s_n) \right| \\
 &\leq \sum_{k=1}^n \left| \log(1 + z_{k,n}) - z_{k,n} \right| + \sum_{k=1}^n \left| R_2^{[k]}(u/s_n) \right|, \quad (8)
 \end{aligned}$$

where

$$z_{k,n} := -\frac{\sigma_k^2}{s_n^2} u^2/2 + R_2^{[k]}(u/s_n).$$

By (6) and (7),

$$|z_{k,n}| \leq M_n^{2/3} u^2/2 + M_n |u|^3/3! \quad (k \leq n). \quad (9)$$

Since  $M_n \rightarrow 0$  by hypothesis, there exists  $n_0$  such that the right-hand side of (9) is  $< 1/2$  for all  $n > n_0$ . Thus

$$|z_{k,n}| < 1/2 \quad (k = 1, \dots, n; n > n_0). \quad (10)$$

By Taylor's formula, for  $|z| < 1$ ,

$$|\log(1+z) - z| \leq \frac{|z|^2/2}{(1-|z|)^2}.$$

Hence, by (7) and (10),

$$\begin{aligned}
 |\log(1 + z_{k,n}) - z_{k,n}| &\leq 2|z_{k,n}|^2 \\
 &\leq \left( \frac{\sigma_k}{s_n} \right)^4 u^4/2 + \left( \frac{\sigma_k}{s_n} \right)^2 \frac{E(|X_k|^3)}{s_n^3} |u|^5/3 + \left( \frac{E(|X_k|^3)}{s_n^3} \right)^2 u^6/18,
 \end{aligned}$$

for  $k \leq n$  and  $n > n_0$ . By (6), the first summand above is

$$\leq M_n^{1/3} \frac{E(|X_k|^3)}{s_n^3} u^4/2.$$

The second summand is

$$\leq M_n^{2/3} \frac{E(|X_k|^3)}{s_n^3} |u|^5/3.$$

The third summand is

$$\leq M_n \frac{E(|X_k|^3)}{s_n^3} u^6/18.$$

Therefore (by (8)),

$$\begin{aligned} & |\log \phi_n(u) - (-u^2/2)| \\ & \leq [M_n^{1/3}u^4/2 + M_n^{2/3}|u|^5/3 + M_n u^6/18 + |u|^3/6] \sum_{k=1}^n \frac{E(|X_k|^3)}{s_n^3} \\ & = M_n^{4/3}u^4/2 + M_n^{5/3}|u|^5/3 + M_n^2 u^6/18 + M_n |u|^3/6 \rightarrow 0 \end{aligned}$$

as  $n \rightarrow \infty$ . □

**Theorem I.4.11 (The Lyapounov central limit theorem).** *Let  $\{X_k\}$  be a sequence of non-degenerate, real, central, independent,  $L^3$ -r.v.s, such that*

$$\lim_{n \rightarrow \infty} s_n^{-3} \sum_{k=1}^n E(|X_k|^3) = 0.$$

*Then the distribution function of*

$$\frac{\sum_{k=1}^n X_k}{s_n} \left( := \frac{S_n}{\sigma(S_n)} \right)$$

*converges pointwise to the standard normal distribution as  $n \rightarrow \infty$ .*

**Proof.** By Lemma I.4.10, the ch.f. of  $S_n/s_n$  converges pointwise to the ch.f.  $e^{-u^2/2}$  of the standard normal distribution (cf. section I.3.12). By the Paul Levy Continuity theorem (Theorem I.4.8) and the Uniqueness theorem (Theorem I.4.2), the distribution function of  $S_n/s_n$  converges pointwise to the standard normal distribution. □

**Corollary I.4.12 (Central limit theorem for uniformly bounded r.v.s).** *Let  $\{X_k\}$  be a sequence of non-degenerate, real, central, independent r.v.s such that  $|X_k| \leq K$  for all  $k \in \mathbb{N}$  and  $s_n(= \sigma(S_n)) \rightarrow \infty$ . Then the distribution functions of  $S_n/s_n$  converge pointwise to the standard normal distribution.*

**Proof.** We have  $E(|X_k|^3) \leq K\sigma_k^2$ . Therefore

$$s_n^{-3} \sum_{k=1}^n E(|X_k|^3) \leq K/s_n \rightarrow 0,$$

and Theorem I.4.11 applies. □

**Corollary I.4.13 (Laplace central limit theorem).** *Let  $\{A_k\}$  be a sequence of independent events with  $PA_k = p, 0 < p < 1, k = 1, 2, \dots$ . Let  $B_n$  be the (Bernoulli) r.v., whose value is the number of occurrences of the first  $n$  events ('number of successes'). Then the distribution function of the 'standardized Bernoulli r.v.'*

$$B_n^* := \frac{B_n - np}{(npq)^{1/2}}$$

*converges pointwise to the standard normal distribution.*

**Proof.** Let  $X_k = I_{A_k} - p$ . Then  $X_k$  are non-degenerate (since  $\sigma^2(X_k) = pq > 0$  when  $0 < p < 1$ ), real, central, independent r.v.s, and  $|X_k| < 1$ . Also  $s_n = (npq)^{1/2} \rightarrow \infty$  (since  $pq > 0$ ). By Corollary I.4.12, the distribution function of  $S_n/s_n = B_n^*$  converges pointwise to the standard normal distribution.  $\square$

**Corollary I.4.14 (Central limit theorem for equidistributed r.v.s).** *Let  $\{X_k\}$  be a sequence of non-degenerate, real, central, independent, equidistributed,  $L^3$ -r.v.s. Then the distribution function of  $S_n/s_n$  converges pointwise to the standard normal distribution.*

**Proof.** Denote (independently of  $k$ , since the r.v.s are equidistributed):

$$E(|X_k|^3) = \alpha; \quad \sigma^2(X_k) = \sigma^2(>0).$$

Since  $X_k$  are independent,  $s_n^2 = n\sigma^2$  by BienAyme's identity, and therefore, as  $n \rightarrow \infty$ ,

$$M_n = \frac{n\alpha}{(n\sigma^2)^{3/2}} = (\alpha/\sigma^3)n^{-1/2} \rightarrow 0.$$

The result follows now from Theorem I.4.11.  $\square$

## I.5 Vector-valued random variables

Let  $X = (X_1, \dots, X_n)$  be an  $\mathbb{R}^n$ -valued r.v. on the probability space  $(\Omega, \mathcal{A}, P)$ . We say that  $X$  has a density if there exists a non-negative Borel function  $h$  on  $\mathbb{R}^n$  (called a *density* of  $X$ , or a *joint density* of  $X_1, \dots, X_n$ ), such that

$$P[X \in B] = \int_B h \, dx \quad (B \in \mathcal{B}(\mathbb{R}^n)),$$

where  $dx = dx_1 \dots dx_n$  is Lebesgue measure on  $\mathbb{R}^n$ .

When  $X$  has a density  $h$ , the latter is uniquely determined on  $\mathbb{R}^n$  almost everywhere with respect to Lebesgue measure  $dx$  (we may then refer to *the* density of  $X$ ).

Suppose the density of  $X$  is of the form

$$h(x) = u(x_1, \dots, x_k)v(x_{k+1}, \dots, x_n), \quad (*)$$

for some  $1 \leq k < n$ , where  $u, v$  are the densities of  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$ , respectively. Then for any  $A \in \mathcal{B}(\mathbb{R}^k)$  and  $B \in \mathcal{B}(\mathbb{R}^{n-k})$ , we have by Fubini's theorem

$$\begin{aligned} & P[(X_1, \dots, X_k) \in A] \cdot P[(X_{k+1}, \dots, X_n) \in B] \\ &= \int_A u \, dx_1 \dots dx_k \cdot \int_B v \, dx_{k+1} \dots dx_n \\ &= \int_{A \times B} h(x_1, \dots, x_n) dx_1 \dots dx_n = P[(X_1, \dots, X_n) \in A \times B] \\ &= P[(X_1, \dots, X_k) \in A] \cap [(X_{k+1}, \dots, X_n) \in B]. \end{aligned}$$

Thus  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$  are independent. Conversely, if  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$  are independent with respective densities  $u, v$ , then for all ‘measurable rectangles’  $A \times B$  with  $A, B$  Borel sets in  $\mathbb{R}^k$  and  $\mathbb{R}^{n-k}$ , respectively,

$$\begin{aligned} P[(X_1, \dots, X_n) \in A \times B] &= P[(X_1, \dots, X_k) \in A] \cap [(X_{k+1}, \dots, X_n) \in B] \\ &= P[(X_1, \dots, X_k) \in A] \cdot P[(X_{k+1}, \dots, X_n) \in B] \\ &= \int_A u \, dx_1 \dots dx_k \cdot \int_B v \, dx_{k+1} \dots dx_n \\ &= \int_{A \times B} uv \, dx_1 \dots dx_n, \end{aligned}$$

and therefore

$$P[X \in H] = \int_H uv \, dx$$

for all  $H \in \mathcal{B}(\mathbb{R}^n)$ , that is,  $X$  has a density of the form (\*).

We proved the following:

**Proposition I.5.1.** *Let  $X$  be an  $\mathbb{R}^n$ -valued r.v., and suppose that for some  $k \in \{1, \dots, n-1\}$ ,  $X$  has a density of the form  $h = uv$ , where  $u = u(x_1, \dots, x_k)$  and  $v = v(x_{k+1}, \dots, x_n)$  are densities for the r.v.s  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$ , respectively. Then  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$  are independent. Conversely, if, for some  $k$  as above,  $(X_1, \dots, X_k)$  and  $(X_{k+1}, \dots, X_n)$  are independent with densities  $u, v$ , respectively, then  $h := uv$  is a density for  $(X_1, \dots, X_n)$ .*

If the  $\mathbb{R}^n$ -r.v.  $X$  has the density  $h$  and  $x = x(y)$  is a  $C^1$ -transformation of  $\mathbb{R}^n$  with inverse  $y = y(x)$  and Jacobian  $J \neq 0$ , we have

$$P[X \in B] = \int_{x^{-1}(B)} h(x(y)) |J(y)| dy$$

for all  $B \in \mathcal{B}(\mathbb{R}^n)$ .

**Example I.5.2.** *The distribution of a sum.*

Suppose  $(X, Y)$  has the density  $h : \mathbb{R}^2 \rightarrow [0, \infty)$ . Consider the transformation

$$x = u - v; \quad y = v$$

with Jacobian identically 1 and inverse

$$u = x + y; \quad v = y.$$

The set

$$B = \{(x, y) \in \mathbb{R}^2; x + y < c, y \in \mathbb{R}\}$$

corresponds to the set

$$B' = \{(u, v) \in \mathbb{R}^2; u < c, v \in \mathbb{R}\}.$$



Therefore, by Tonelli's theorem,

$$\begin{aligned} F_{X+Y}(c) &:= P[X + Y < c] = P[(X, Y) \in B] = \int_{B'} h(u - v, v) \, du \, dv \\ &= \int_{-\infty}^c \left( \int_{-\infty}^{\infty} h(u - v, v) \, dv \right) du. \end{aligned}$$

This shows that the r.v.  $X + Y$  has the density

$$h_{X+Y}(u) = \int_{-\infty}^{\infty} h(u - v, v) \, dv.$$

In particular, when  $X, Y$  are independent with respective densities  $h_X$  and  $h_Y$ , then  $X + Y$  has the density

$$h_{X+Y}(u) = \int_{\mathbb{R}} h_X(u - v) h_Y(v) \, dv := (h_X * h_Y)(u),$$

(the *convolution* of  $h_X$  and  $h_Y$ ).

**Example I.5.3.** *The distribution of a ratio.*

Let  $Y$  be a positive r.v., and  $X$  any real r.v. We assume that  $(X, Y)$  has a density  $h$ . The transformation

$$x = uv; \quad y = v$$

has the inverse

$$u = x/y; \quad v = y,$$

and Jacobian  $J = v > 0$ . Therefore

$$F_{X/Y}(c) := P[X/Y < c] = P[(X, Y) \in B],$$

where

$$B = \{(x, y); -\infty < x < cy, y > 0\}$$

corresponds to

$$B' = \{(u, v); -\infty < u < c, v > 0\}.$$

Therefore, by Tonelli's theorem,

$$F_{X/Y}(c) = \int_{B'} h(uv, v) v \, du \, dv = \int_{-\infty}^c \left( \int_0^{\infty} h(uv, v) v \, dv \right) du$$

for all real  $c$ . This shows that  $X/Y$  has the density

$$h_{X/Y}(u) = \int_0^{\infty} h(uv, v) v \, dv \quad (u \in \mathbb{R}).$$

When  $X, Y$  are independent, this formula becomes

$$h_{X/Y}(u) = \int_0^\infty h_X(uv)h_Y(v)v \, dv \quad (u \in \mathbb{R}).$$

Let  $X$  be an  $\mathbb{R}^n$ -valued r.v., and let  $g$  be a real Borel function on  $\mathbb{R}^n$ . The r.v.  $g(X)$  is called a *statistic*. For example,

$$\bar{X} := (1/n) \sum_{k=1}^n X_k$$

and

$$S^2 := (1/n) \sum_{k=1}^n (X_k - \bar{X})^2$$

are the statistics corresponding to the Borel functions

$$\bar{x}(x_1, \dots, x_n) := (1/n) \sum_{k=1}^n x_k$$

and

$$s^2(x_1, \dots, x_n) := (1/n) \sum_{k=1}^n [x_k - \bar{x}(x_1, \dots, x_n)]^2,$$

respectively.

The statistics  $\bar{X}$  and  $S^2$  are called the *sample mean* and the *sample variance*, respectively.

**Theorem I.5.4 (Fisher).** *Let  $X_1, \dots, X_n$  be independent  $N(0, \sigma^2)$ -distributed r.v.s. Let*

$$Z_k := \frac{X_k - \bar{X}}{S}, \quad k = 1, \dots, n.$$

*Then*

- (1)  $(Z_1, \dots, Z_{n-2}), \bar{X}$ , and  $S$  are independent;
- (2)  $(Z_1, \dots, Z_{n-2})$  has density independent of  $\sigma$ ;
- (3)  $\bar{X}$  is  $N(0, \sigma^2/n)$ -distributed; and
- (4)  $nS^2$  is  $\chi_\sigma^2$ -distributed, with  $n - 1$  degrees of freedom.

**Proof.** The map sending  $X_1, \dots, X_n$  to the statistics  $Z_1, \dots, Z_{n-2}, \bar{X}$ , and  $S^2$  is given by the following equations:

$$\begin{aligned} z_k &= \frac{x_k - \bar{x}}{s}, \quad k = 1, \dots, n-2; \\ \bar{x} &= (1/n) \sum_{k=1}^n x_k; \\ s^2 &= (1/n) \sum_{k=1}^n (x_k - \bar{x})^2. \end{aligned} \tag{1}$$

Note the relations (for  $z_k$  defined as in (1) for all  $k = 1, \dots, n$ ).

$$\sum_{k=1}^n z_k = 0; \quad \sum_{k=1}^n z_k^2 = n; \quad (2)$$

$$\sum_{k=1}^n x_k^2 = n(\bar{x})^2 + ns^2. \quad (3)$$

By (2),

$$z_{n-1} + z_n = u \left( := - \sum_{k=1}^{n-2} z_k \right)$$

and

$$z_{n-1}^2 + z_n^2 = w \left( := n - \sum_{k=1}^{n-2} z_k^2 \right).$$

Thus

$$(u - z_n)^2 + z_n^2 = w,$$

that is,

$$z_n^2 - uz_n + (u^2 - w)/2 = 0.$$

Therefore

$$z_n = (u + v)/2; \quad z_{n-1} = (u - v)/2,$$

where  $v := \sqrt{2w - u^2}$ ; a second solution has  $v$  replaced by  $-v$ . Note that

$$2w - u^2 = 2z_{n-1}^2 + 2z_n^2 - (z_{n-1} + z_n)^2 = (z_{n-1} - z_n)^2 \geq 0,$$

so that  $v$  is real.

The inverse transformations are then

$$x_k = \bar{x} + sz_k, k = 1, \dots, n-2;$$

$$x_{n-1} = \bar{x} + s(u - v)/2;$$

$$x_n = \bar{x} + s(u + v)/2,$$

with  $v$  replaced by  $-v$  in the second inverse;  $u, v$  are themselves functions of  $z_1, \dots, z_{n-2}$ .

The corresponding Jacobian

$$J := \frac{\partial(x_1, \dots, x_n)}{\partial(z_1, \dots, z_{n-2}, \bar{x}, s)}$$

has the form

$$s^{n-2} g(z_1, \dots, z_{n-2}),$$

where  $g$  is a function of  $z_1, \dots, z_{n-2}$  only, and does not depend on the parameter  $\sigma^2$  of the given normal distribution of the  $X_k (k = 1, \dots, n)$ .

Replacing  $v$  by  $-v$  only interchanges the last two columns of the determinant, so that  $|J|$  remains unchanged. Therefore, using (3),

$$\begin{aligned} h_X(x) dx &= (2\pi\sigma^2)^{-n/2} e^{-\sum_{k=1}^n x_k^2/2\sigma^2} dx_1 \dots dx_n \\ &= 2(2\pi\sigma^2)^{-n/2} e^{-((n(\bar{x})^2 + ns^2)/2\sigma^2)} s^{n-2} g(z_1, \dots, z_{n-2}) \\ &\quad \times dz_1 \dots dz_{n-2} d\bar{x} ds, \end{aligned}$$

where the factor 2 comes from the fact that the inverse is bi-valued, with the same value of  $|J|$  for both possible choices, hence doubling the ‘mass element’.

The last expression can be written in the form

$$h_1(\bar{x}) d\bar{x} h_2(s) ds h_3(z_1, \dots, z_{n-2}) dz_1 \dots dz_{n-2},$$

where

$$h_1(\bar{x}) = \frac{1}{\sqrt{2\pi}\sigma/\sqrt{n}} e^{-((\bar{x})^2/2(\sigma/\sqrt{n})^2)}$$

is the  $N(0, \sigma^2/n)$ -density;

$$h_2(s) = \frac{n^{(n-1)/2} s^{n-2}}{2^{(n-3)/2} \Gamma((n-1)/2) \sigma^{n-1}} e^{-ns^2/2\sigma^2} \quad (s > 0)$$

(and  $h_2(s) = 0$  for  $s \leq 0$ ) is seen to be a density (i.e. has integral = 1); and

$$h_3(z_1, \dots, z_{n-2}) = \frac{\Gamma((n-1)/2)}{n^{n/2} \pi^{(n-1)/2}} g(z_1, \dots, z_{n-2})$$

is necessarily the density for  $(Z_1, \dots, Z_{n-2})$ , and clearly does not depend on  $\sigma^2$ .

The above decomposition implies by Proposition I.5.1 that Statements 1–3 of the theorem are correct. Moreover, for  $x > 0$ , we have

$$F_{nS^2}(x) = P[nS^2 < x] = P[S < \sqrt{x/n}] = \int_0^{\sqrt{x/n}} h_2(s) ds.$$

Therefore

$$\begin{aligned} h_{nS^2}(x) &:= \frac{d}{dx} F_{nS^2}(x) = h_2(\sqrt{x/n}) (1/2) (x/n)^{-1/2} (1/n) \\ &= \frac{x^{(n-1)/2-1} e^{-x/2\sigma^2}}{(2\sigma^2)^{(n-1)/2} \Gamma((n-1)/2)}. \end{aligned}$$

This is precisely the  $\chi_\sigma^2$  density with  $n-1$  degrees of freedom. □

Fisher’s theorem (Theorem I.5.4) will be applied in the sequel to obtain the distributions of some important statistics.

**Theorem I.5.5.** *Let  $U, V$  be independent r.v.s, with  $U$  normal  $N(0, 1)$  and  $V$   $\chi_1^2$ -distributed with  $\nu$  degrees of freedom. Then the statistic*

$$T := \frac{U}{\sqrt{V/\nu}}$$

has the density

$$h_\nu(t) = \nu^{-1/2} B(1/2, \nu/2)^{-1} (1 + t^2/\nu)^{-(\nu+1)/2} \quad (t \in \mathbb{R}),$$

called the ‘*t*-density’ or the ‘Student density with  $\nu$  degrees of freedom’.

In the above formula,  $B(\cdot, \cdot)$  denotes the Beta function:

$$B(s, t) = \frac{\Gamma(s)\Gamma(t)}{\Gamma(s+t)},$$

( $s, t > 0$ ).

**Proof.** We apply Example I.5.3 with the independent r.v.s  $X = U$  and  $Y = \sqrt{V/\nu}$ . The distribution  $F_Y$  (for  $y > 0$ ) is given by

$$\begin{aligned} F_Y(y) &:= P[\sqrt{V/\nu} < y] = P[V < \nu y^2] \\ &= 2^{-\nu/2} \Gamma(\nu/2)^{-1} \int_0^{\nu y^2} s^{(\nu/2)-1} e^{-s/2} ds \end{aligned}$$

(cf. Section I.3.15, Proposition 2). The corresponding density is (for  $y > 0$ )

$$h_Y(y) := \frac{d}{dy} F_Y(y) = \frac{\nu^{\nu/2}}{2^{\nu/2-1} \Gamma(\nu/2)} y^{\nu-1} e^{-\nu y^2/2},$$

and of course  $h_Y(y) = 0$  for  $y \leq 0$ .

By hypothesis,  $h_X(x) = e^{-x^2/2}/\sqrt{2\pi}$ . By Example I.5.3, the density of  $T$  is

$$\begin{aligned} h_T(t) &= \int_0^\infty h_X(vt) h_Y(v) v dv \\ &= \frac{\nu^{\nu/2}}{\sqrt{2\pi} 2^{\nu/2-1} \Gamma(\nu/2)} \int_0^\infty e^{-v^2(t^2+\nu)/2} v^\nu dv. \end{aligned}$$

Write  $s = v^2(t^2 + \nu)/2$ :

$$h_T(t) = \frac{\nu^{\nu/2}}{\sqrt{\pi} \Gamma(\nu/2) (t^2 + \nu)^{(\nu+1)/2}} \int_0^\infty e^{-s} s^{(\nu+1)/2-1} ds.$$

Since  $\Gamma(1/2) = \sqrt{\pi}$ , the last expression coincides with  $h_\nu(t)$ . □

**Corollary I.5.6.** *Let  $X_1, \dots, X_n$  be independent  $N(\mu, \sigma^2)$ -r.v.s. Then the statistic*

$$T := \frac{\bar{X} - \mu}{S} \sqrt{n-1} \tag{*}$$

*has the Student distribution with  $\nu = n - 1$  degrees of freedom.*

**Proof.** Take  $U = (\bar{X} - \mu)/(\sigma/\sqrt{n})$  and  $V = nS^2/\sigma^2$ . By Fisher’s theorem,  $U, V$  satisfy the hypothesis of Theorem I.5.5, and the conclusion follows (since the statistic  $T$  in Theorem I.5.5 coincides in the present case with (\*)). □

**Corollary I.5.7.** Let  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_m$  be independent  $N(\mu, \sigma^2)$ -r.v.s. Let  $\bar{X}$ ,  $\bar{Y}$ ,  $S_X^2$ , and  $S_Y^2$  be the ‘sample means’ and ‘sample variances’ (for the ‘samples’  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_m$ ). Then the statistic

$$W := \frac{\bar{X} - \bar{Y}}{\sqrt{nS_X^2 + mS_Y^2}} \sqrt{(n+m-2)nm/(n+m)}$$

has the Student distribution with  $\nu = n + m - 2$  degrees of freedom.

**Proof.** The independence hypothesis implies that  $\bar{X}$  and  $\bar{Y}$  are independent normal r.v.s with parameters  $(\mu, \sigma^2/n)$  and  $(\mu, \sigma^2/m)$ , respectively. Therefore  $\bar{X} - \bar{Y}$  is  $N(0, \sigma^2(n+m)/nm)$ -distributed, and

$$U := \frac{\bar{X} - \bar{Y}}{\sigma \sqrt{(n+m)/nm}}$$

is  $N(0, 1)$ -distributed.

By Fisher’s theorem, the r.v.s  $nS_X^2/\sigma^2$  and  $mS_Y^2/\sigma^2$  are  $\chi_1^2$ -distributed with  $n-1$  and  $m-1$  degrees of freedom, respectively, and are *independent* (as Borel functions of the independent r.v.s  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_m$ , resp.). Since the  $\chi^2$ -distribution with  $r$  degrees of freedom is the Gamma distribution with  $p = r/2$  and  $b = 1/2$ , it follows from Proposition 1 in Section I.3.15 that the r.v.

$$V := \frac{nS_X^2 + mS_Y^2}{\sigma^2}$$

is  $\chi_1^2$ -distributed with  $\nu = (n-1) + (m-1)$  degrees of freedom.

Also, by Fisher’s theorem,  $U, V$  are independent. We may then apply Theorem I.5.5 to the present choice of  $U, V$ . An easy calculation shows that for this choice  $T = W$ , and the conclusion follows.  $\square$

**Remark I.5.8.** The statistic  $T$  is used in ‘testing hypothesis’ about the value of the mean  $\mu$  of a normal ‘population’, using the ‘sample outcomes’  $X_1, \dots, X_n$ . The statistic  $W$  is used in testing the ‘zero-hypothesis’ that two normal populations have the same mean (using the outcomes of samples taken from the respective populations). Its efficiency is enhanced by the fact that it is independent of the unknown parameters  $(\mu, \sigma^2)$  of the normal population (cf. Section I.6).

Tables of the Student distribution are usually available for  $\nu < 30$ . For  $\nu \geq 30$ , the normal distribution is a good approximation. The theorem behind this fact is the following:

**Theorem I.5.9.** Let  $h_\nu$  be the Student distribution density with  $\nu$  degrees of freedom. Then as  $\nu \rightarrow \infty$ ,  $h_\nu$  converges pointwise to the  $N(0, 1)$ -density, and

$$\lim_{\nu \rightarrow \infty} P[a \leq T < b] = (1/\sqrt{2\pi}) \int_a^b e^{-t^2/2} dt \quad (*)$$

for all real  $a < b$ .

**Proof.** By Stirling's formula,  $\Gamma(n)$  is asymptotically equal to  $(n/e)^n$ . Therefore (since  $\Gamma(1/2) = \sqrt{\pi}$ )

$$\begin{aligned} & \lim_{\nu \rightarrow \infty} \frac{\nu^{1/2} B(1/2, \nu/2)}{\sqrt{2\pi}} \\ &= \lim_{\nu} \frac{\Gamma(\nu/2)}{(\nu/2e)^{\nu/2}} \cdot \frac{((\nu+1)/2e)^{(\nu+1)/2}}{\Gamma((\nu+1)/2)} \cdot \frac{e^{1/2}}{(1+1/\nu)^{\nu/2+1/2}} = 1. \end{aligned}$$

Hence, as  $\nu \rightarrow \infty$ ,

$$h_{\nu}(t) := [\nu^{1/2} B(1/2, \nu/2)]^{-1} (1 + t^2/\nu)^{-\nu/2-1/2} \rightarrow (1/\sqrt{2\pi}) e^{-t^2/2}.$$

For real  $a < b$ ,

$$P[a \leq T < b] = [\nu^{1/2} B(1/2, \nu/2)]^{-1} \int_a^b \frac{dt}{(1 + t^2/\nu)^{(\nu+1)/2}}.$$

The coefficient before the integral was seen to converge to  $1/\sqrt{2\pi}$ ; the integrand converges pointwise to  $e^{-t^2/2}$  and is bounded by 1; therefore (\*) follows by dominated convergence.  $\square$

**Theorem I.5.10.** Let  $U_i$  be independent  $\chi_1^2$ -distributed r.v.s with  $\nu_i$  degrees of freedom ( $i = 1, 2$ ). Assume  $U_2 > 0$ , and consider the statistic

$$F := \frac{U_1/\nu_1}{U_2/\nu_2}.$$

Then  $F$  has the distribution density

$$h(u; \nu_1, \nu_2) = \frac{\nu_1^{\nu_1/2} \nu_2^{\nu_2/2}}{B(\nu_1/2, \nu_2/2)} \frac{u^{\nu_1/2-1}}{(\nu_1 u + \nu_2)^{(\nu_1+\nu_2)/2}}$$

for  $u > 0$  (and  $= 0$  for  $u \leq 0$ ).

The above density is called the 'F-density' or 'Snedecor density' with  $(\nu_1, \nu_2)$  degrees of freedom.

**Proof.** We take in Example I.5.3 the independent r.v.s  $X = U_1/\nu_1$  and  $Y = U_2/\nu_2$ . We have for  $x > 0$ :

$$F_X(x) = P[U_1 < \nu_1 x] = [2^{\nu_1/2} \Gamma(\nu_1/2)]^{-1} \int_0^{\nu_1 x} t^{\nu_1/2-1} e^{-t/2} dt,$$

and  $F_X(x) = 0$  for  $x \leq 0$ . Therefore, for  $x > 0$ ,

$$h_X(x) = \frac{d}{dx} F_X(x) = \frac{(\nu_1/2)^{\nu_1/2}}{\Gamma(\nu_1/2)} x^{\nu_1/2-1} e^{-\nu_1 x/2},$$

and  $h_X(x) = 0$  for  $x \leq 0$ . A similar formula is valid for  $Y$ , with  $\nu_2$  replacing  $\nu_1$ . By Example I.5.3, the density of  $F := X/Y$  is 0 for  $u \leq 0$ , and for  $u > 0$ ,

$$\begin{aligned} h_F(u) &= \int_0^\infty h_X(uv)h_Y(v)v \, dv \\ &= \frac{(\nu_1/2)^{\nu_1/2}(\nu_2/2)^{\nu_2/2}}{\Gamma(\nu_1/2)\Gamma(\nu_2/2)} \int_0^\infty (uv)^{\nu_1/2-1} v^{\nu_2/2} e^{-(\nu_1 uv + \nu_2 v)/2} \, dv. \quad (*) \end{aligned}$$

The integral is

$$u^{\nu_1/2-1} \int_0^\infty e^{-v(\nu_1 u + \nu_2)/2} v^{(\nu_1 + \nu_2)/2-1} \, dv.$$

Making the substitution  $v(\nu_1 u + \nu_2)/2 = s$ , the integral takes the form

$$\frac{u^{\nu_1/2-1}}{[(\nu_1 u + \nu_2)/2]^{(\nu_1 + \nu_2)/2}} \int_0^\infty e^{-s} s^{(\nu_1 + \nu_2)/2-1} \, ds.$$

Since the last integral is  $\Gamma((\nu_1 + \nu_2)/2)$ , it follows from (\*) that  $h_F(u) = h(u; \nu_1, \nu_2)$ , for  $h$  as in the theorem's statement.  $\square$

**Corollary I.5.11.** *Let  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_m$  be independent  $N(0, \sigma^2)$ -distributed r.v.s. Then the statistic*

$$F := \frac{S_X^2}{S_Y^2} (1 - 1/m)/(1 - 1/n)$$

*has Snedecor's density with  $\nu_1 = n - 1$  and  $\nu_2 = m - 1$  degrees of freedom.*

**Proof.** Let  $U_1 := nS_X^2/\sigma^2$  and  $U_2 := mS_Y^2/\sigma^2$ . By Fisher's theorem,  $U_i$  are  $\chi_1^2$ -distributed with  $\nu_i$  degrees of freedom ( $i = 1, 2$ ). Since they are independent, the r.v.  $F = (U_1/\nu_1)/(U_2/\nu_2)$  has the Snedecor density with  $(\nu_1, \nu_2)$  degrees of freedom, by Theorem I.5.10.  $\square$

**I.5.12.** The statistic  $F$  of Corollary I.5.11 is used, for example, to test the 'zero hypothesis' that two normal 'populations' have the same variance (see Section I.6). Statistical tables give  $u_\alpha$ , defined by

$$P[F \geq u_\alpha] \left( = \int_{u_\alpha}^\infty h(u; \nu_1, \nu_2) \, du \right) = \alpha,$$

for various values of  $\alpha$  and of the degrees of freedom  $\nu_i$ .

## I.6 Estimation and decision

We consider random sampling of size  $n$  from a given population. The  $n$  outcomes are a value of a  $\mathbb{R}^n$ -valued r.v.  $X = (X_1, \dots, X_n)$ , where  $X_k$  have the same distribution function (the 'population distribution')  $F(\cdot; \theta)$ ; the 'parameter vector'  $\theta$



is usually unknown. For example, a normal population has the parameter vector  $(\mu, \sigma^2)$ , etc.

*Estimation* is concerned with the problem of ‘estimating’  $\theta$  by using the sample outcomes  $X_1, \dots, X_n$ , say, by means of some Borel function of  $X_1, \dots, X_n$ :

$$\theta^* := g(X_1, \dots, X_n).$$

This statistic is called an *estimator* of  $\theta$ .

Consider the case of a single real parameter  $\theta$ .

A measure of the estimator’s precision is its mean square deviation from  $\theta$ ,

$$E(\theta^* - \theta)^2,$$

called the *risk function* of the estimator.

We have

$$\begin{aligned} E(\theta^* - \theta)^2 &= E[(\theta^* - E\theta^*) + (E\theta^* - \theta)]^2 \\ &= E(\theta^* - E\theta^*)^2 + 2E(\theta^* - E\theta^*) \cdot (E\theta^* - \theta) + (E\theta^* - \theta)^2. \end{aligned}$$

The middle term vanishes, so that the risk function is equal to

$$\sigma^2(\theta^*) + (E\theta^* - \theta)^2. \quad (*)$$

The difference  $\theta - E\theta^*$  is called the *bias* of the estimator  $\theta^*$ ; the estimator is *unbiased* if the bias is zero, that is, if  $E\theta^* = \theta$ . In this case the risk function is equal to the variance  $\sigma^2(\theta^*)$ .

**Example I.6.1.** We wish to estimate the expectation  $\mu$  of the population distribution. Any weighted average

$$\mu^* = \sum_{k=1}^n a_k X_k \quad \left( a_k > 0, \sum a_k = 1 \right)$$

is a reasonable unbiased estimator of  $\mu$ :

$$E\mu^* = \sum a_k EX_k = \sum a_k \mu = \mu.$$

By BienAyme’s identity (since the  $X_k$  are independent in random sampling), the risk function is given by

$$\sigma^2(\mu^*) = \sum_{k=1}^n a_k^2 \sigma^2(X_k) = \sum a_k^2 \sigma^2,$$

where  $\sigma^2$  is the population variance (assuming that it exists). However, since  $\sum a_k = 1$ ,

$$\sum a_k^2 = \sum (a_k - 1/n)^2 + 1/n \geq 1/n,$$

and the minimum  $1/n$  is attained when  $a_k = 1/n$  for all  $k$ . Thus, among all estimators of  $\mu$  that are weighted averages of the sample outcomes, the estimator

with minimal risk function is the arithmetical mean  $\mu^* = \bar{X}$ ; its risk function is  $\sigma^2/n$ .

**Example I.6.2.** As an estimator of the parameter  $p$  of a binomial population we may choose the ‘successes frequency’  $p^* := S_n/n$  (cf. Example I.2.7). The r.v.  $p^*$  takes on the values  $k/n$  ( $k = 0, \dots, n$ ), and

$$P[p^* = k/n] = P[S_n = k] = \binom{n}{k} p^k q^{n-k}.$$

The estimator  $p^*$  is unbiased, since

$$Ep^* = ES_n/n = np/n = p.$$

Its risk function is

$$\sigma^2(p^*) = npq/n^2 = pq/n \leq 1/4n.$$

By Corollary I.4.13,  $(p^* - p)/\sqrt{pq/n}$  is approximately  $N(0, 1)$ -distributed for  $n$  ‘large’. Thus, for example,

$$P[|p^* - p| < 2\sqrt{pq/n}] > 0.95,$$

and since  $pq \leq 1/4$ , we surely have

$$P[|p^* - p| < 1/\sqrt{n}] > 0.95.$$

Thus, the estimated parameter  $p$  lies in the interval  $(p^* - 1/\sqrt{n}, p^* + 1/\sqrt{n})$  with ‘confidence’ 0.95 (at least). We shall return to this idea later.

In comparing two binomial populations (e.g. in quality control problems), we may wish to estimate the difference  $p_1 - p_2$  of their parameters, using samples of sizes  $n$  and  $m$  from the respective populations. A reasonable estimator is the difference  $V = S_n/n - S'_m/m$  of the success frequencies in the two samples. The estimator  $V$  is clearly unbiased, and its risk function is (by BienAyme’s identity)

$$\sigma^2(V) = \sigma^2(S_n/n) + \sigma^2(S'_m/m) = p_1q_1/n + p_2q_2/m \leq 1/4n + 1/4m.$$

For large samples, we may use the normal approximation (cf. Corollary I.4.13) to test the ‘zero hypothesis’ that  $p_1 - p_2 = 0$  (i.e. that the two binomial populations are equidistributed) by using the statistic  $V$ .

**Example I.6.3.** Consider the two-layered population of Example I.2.8. An estimator of the proportion  $N_1/N$  of the layer  $\mathcal{P}_1$  in the population, could be the sample frequency  $U := D_s/s$  of  $\mathcal{P}_1$ -objects in a random sample of size  $s$ .

We have  $EU = (1/s)(sN_1/N) = N_1/N$ , so that  $U$  is unbiased.

The risk function is (cf. Example I.2.8)

$$\sigma^2(U) = (1/s) \frac{N-s}{N-1} \frac{N_1}{N} \left(1 - \frac{N_1}{N}\right) < 1/4s.$$

**Example I.6.4.** The sample average  $\bar{X}$  is a natural unbiased estimator  $\lambda^*$  of the parameter  $\lambda$  of a Poissonian population. Its risk function is  $\sigma^2(\lambda^*) = \lambda/n$  (cf. Example I.3.9).

**Example I.6.5.** Let  $X_1, \dots, X_n$  be independent  $N(\mu, \sigma^2)$ -r.v.s. For any weights  $a_1, \dots, a_n$ , the statistic

$$V := \frac{n}{n-1} \left[ \sum_{k=1}^n a_k X_k^2 - \bar{X}^2 \right]$$

is an unbiased estimator of  $\sigma^2$ . Indeed,

$$E(\bar{X})^2 = \sigma^2(\bar{X}) + [E\bar{X}]^2 = \sigma^2/n + \mu^2,$$

and therefore

$$EV = \frac{n}{n-1} \left[ \sum_k a_k (\sigma^2 + \mu^2) - (\sigma^2/n + \mu^2) \right] = \frac{n}{n-1} (1 - 1/n) \sigma^2 = \sigma^2.$$

When  $a_k = 1/n$  for all  $k$ , the estimator  $V$  is the so-called ‘sample error’

$$V = nS^2/(n-1) = \frac{1}{n-1} \sum_{k=1}^n (X_k - \bar{X})^2.$$

**Example I.6.6.** Let  $X_1, X_2, \dots$  be independent r.v.s with the same distribution. Assume that the moment  $\mu_{2r}$  of that distribution exists for some  $r \geq 1$ . For each  $n$ , the arithmetical means

$$m_{r,n} := \frac{1}{n} \sum_{k=1}^n X_k^r$$

are unbiased estimators of the  $r$ th moment  $\mu_r$  of the distribution. By Example I.2.14 applied to  $Y_k = X_k^r$ ,  $m_{r,n} \rightarrow \mu_r$  in probability (as  $n \rightarrow \infty$ ). We say that the sequence of estimators  $\{m_{r,n}\}_n$  of the parameter  $\mu_r$  is *consistent* (the general definition of consistency is the same, *mutatis mutandis*).

If  $\{\theta_n^*\}$  is a consistent sequence of estimators for  $\theta$  and  $\{a_n\}$  is a real sequence converging to 1, then  $\{a_n \theta_n^*\}$  is clearly consistent as well. Biased estimators could be consistent (start with any consistent sequence of unbiased estimators  $\theta_n^*$ ; then  $((n-1)/n)\theta_n^*$  are still consistent estimators, but their bias is  $\theta/n \neq 0$  (unless  $\theta = 0$ )).

### I.6.7. Maximum likelihood estimators (MLEs).

The distribution density  $f(x_1, \dots, x_n; \theta)$  of  $(X_1, \dots, X_n)$ , considered as a function  $L(\theta)$ , is called the *likelihood function* (for  $X_1, \dots, X_n$ ). The MLE (for  $X_1, \dots, X_n$ ) is  $\theta^*(X_1, \dots, X_n)$ , where  $\theta^* = \theta^*(x_1, \dots, x_n)$  is the value of  $\theta$  for which  $L(\theta)$  is maximal (if such a value exists), that is,

$$f(x_1, \dots, x_n; \theta^*) \geq f(x_1, \dots, x_n; \theta)$$

for all  $(x_1, \dots, x_n) \in \mathbb{R}^n$  and  $\theta$  in the relevant range. Hence

$$P_{\theta^*}[X \in B] \geq P_{\theta}[X \in B]$$

for all  $B \in \mathcal{B}(\mathbb{R}^n)$  and  $\theta$ , where the subscript of  $P$  means that the distribution of  $X$  is taken with the parameter indicated.

We consider the case when  $X_1, \dots, X_n$  are independent  $N(\mu, \sigma^2)$ -r.v.s. Thus

$$L(\theta) = (2\pi\sigma^2)^{-n/2} e^{-\sum (x_k - \mu)^2 / 2\sigma^2} \quad \theta = (\mu, \sigma^2).$$

Maximizing  $L$  is equivalent to maximizing the function

$$\phi(\theta) := \log L(\theta) = -(n/2) \log(2\pi\sigma^2) - \sum (x_k - \mu)^2 / 2\sigma^2.$$

*Case 1. MLE for  $\mu$  when  $\sigma$  is given.* The necessary condition for  $\mu = \mu^*$ ,

$$\frac{\partial \phi}{\partial \mu} = \sum (x_k - \mu) / \sigma^2 = 0, \quad (1)$$

implies that  $\mu^* = \mu^*(x_1, \dots, x_n) = \bar{x}$ . This is indeed a maximum point, since

$$\frac{\partial^2 \phi}{\partial \mu^2} = -n / \sigma^2 < 0.$$

Thus the MLE for  $\mu$  (when  $\sigma$  is given) is

$$\mu^*(X_1, \dots, X_n) = \bar{X}.$$

*Case 2. MLE for  $\sigma^2$  when  $\mu$  is given.* The solution of the equation

$$\frac{\partial \phi}{\partial (\sigma^2)} = -n/2\sigma^2 + (1/2\sigma^4) \sum (x_k - \mu)^2 = 0 \quad (2)$$

is

$$(\sigma^2)^* = (1/n) \sum (x_k - \mu)^2.$$

Since the second derivative of  $\phi$  at  $(\sigma^2)^*$  is equal to  $-n/2[(\sigma^2)^*]^2 < 0$ , we got indeed a maximum point of  $\phi$ , and the corresponding MLE for  $\sigma^2$  is

$$(\sigma^2)^*(X_1, \dots, X_n) = (1/n) \sum (X_k - \mu)^2.$$

*Case 3. MLE for  $\theta := (\mu, \sigma^2)$  (as an unknown vector parameter).* We need to solve the equations (1) and (2) simultaneously. From (1), we get  $\mu^* = \bar{x}$ ; from (2) we get

$$(\sigma^2)^* = (1/n) \sum (x_k - \mu^*)^2 = (1/n) \sum (x_k - \bar{x})^2 := s^2.$$

The solution  $(\bar{x}, s^2)$  is indeed a maximum point for  $\phi$ , since the Hessian for  $\phi$  at this point equals  $n^2/2s^6 > 0$ , and the second partial derivative of  $\phi$  with respect to  $\mu$  (at this point) equals  $-n/s^2 < 0$ . Thus the MLE for  $\theta$  is

$$\theta^*(X_1, \dots, X_n) = (\bar{X}, S^2).$$

Note that the estimator  $S^2$  is *biased*, since

$$ES^2 = \frac{n-1}{n}\sigma^2,$$

but *consistent*: indeed,

$$S^2 = (1/n) \sum_{k=1}^n X_k^2 - \left[ (1/n) \sum_{k=1}^n X_k \right]^2;$$

by the weak law of large numbers, the first average on the right-hand side converges in probability to the second moment  $\mu_2$ , while the second average converges to the first moment  $\mu_1 = \mu$ ; hence  $S^2$  converges in probability to  $\mu_2 - \mu^2 = \sigma^2$  (when  $n \rightarrow \infty$ ).

## Confidence intervals

**I.6.8.** Together with the estimator  $\theta^*$  of a real parameter  $\theta$ , it is useful to have an interval around  $\theta^*$  that contains  $\theta$  with some high probability  $1 - \alpha$  (called the *confidence* of the interval). In fact,  $\theta$  is *not a random variable*, and the rigorous approach is to find an interval  $(a(\theta), b(\theta))$  such that

$$P[\theta^* \in (a(\theta), b(\theta))] = 1 - \alpha. \quad (3)$$

The corresponding interval for  $\theta$  is a  $(1 - \alpha)$ -confidence interval for  $\theta$ .

**Example 1.** Consider an  $N(\mu, \sigma^2)$ -population with known variance. Let

$$Z := \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}.$$

Then  $Z$  is  $N(0, 1)$ -distributed. In our case, (3) for the estimator  $\bar{X}$  of  $\mu$  takes the equivalent form

$$P \left[ \frac{a(\mu) - \mu}{\sigma/\sqrt{n}} < Z < \frac{b(\mu) - \mu}{\sigma/\sqrt{n}} \right] = 1 - \alpha. \quad (4)$$

For simplicity, take a symmetric  $Z$ -interval,

$$\frac{b(\mu) - \mu}{\sigma/\sqrt{n}} = c; \quad \frac{a(\mu) - \mu}{\sigma/\sqrt{n}} = -c,$$

that is,

$$a(\mu) = \mu - c\sigma/\sqrt{n}; \quad b(\mu) = \mu + c\sigma/\sqrt{n}.$$

Let  $\Phi$  denote the  $N(0, 1)$ -distribution function. By symmetry of the normal density,

$$\Phi(-c) = 1 - \Phi(c),$$

and therefore (4) takes the form

$$1 - \alpha = \Phi(c) - \Phi(-c) = 2\Phi(c) - 1,$$

that is,

$$\Phi(c) = 1 - \alpha/2,$$

and we get the unique solution for  $c$ :

$$c = \Phi^{-1}(1 - \alpha/2) := z_{1-\alpha/2}. \quad (5)$$

By symmetry of the normal density,  $-c = z_{\alpha/2}$ . The interval for the estimator  $\bar{X}$  is then

$$\mu + z_{\alpha/2}\sigma/\sqrt{n} := a(\mu) < \bar{X} < b(\mu) := \mu + z_{1-\alpha/2}\sigma/\sqrt{n},$$

and the corresponding  $(1 - \alpha)$ -confidence interval for  $\mu$  is

$$\bar{X} - z_{1-\alpha/2}\sigma/\sqrt{n} < \mu < \bar{X} - z_{\alpha/2}\sigma/\sqrt{n}.$$

**Example 2.** Consider an  $N(\mu, \sigma^2)$ -population with *both parameters unknown*. We still use the MLE  $\mu^* = \bar{X}$ . By Corollary I.5.6, the statistic

$$T := \frac{\bar{X} - \mu}{S} \sqrt{n-1}$$

has the Student distribution with  $n - 1$  degrees of freedom. By symmetry of the Student density, the argument in Example 1 applies in this case by replacing  $\Phi$  with  $F_{T, n-1}$ , the Student distribution function for  $n - 1$  degrees of freedom, and  $\sigma/\sqrt{n}$  by  $S/\sqrt{n-1}$ . Let

$$t_{\gamma, n-1} := F_{T, n-1}^{-1}(\gamma).$$

Then a  $(1 - \alpha)$ -confidence interval for  $\mu$  is

$$\bar{X} - t_{1-\alpha/2, n-1}S/\sqrt{n-1} < \mu < \bar{X} - t_{\alpha/2, n-1}S/\sqrt{n-1}.$$

**Example 3.** In the context of Example 2, we look for a  $(1 - \alpha)$ -confidence interval for  $\sigma^2$ . By Fisher's theorem, the statistic  $V := nS^2/\sigma^2$  has the  $\chi^2_1$  distribution with  $n - 1$  degrees of freedom (denoted for simplicity by  $F_{n-1}$ ). Denote

$$\chi^2_{\gamma, n-1} = F_{n-1}^{-1}(\gamma) \quad (\gamma \in \mathbb{R}).$$

Choosing

$$a = \chi^2_{\alpha/2, n-1}; \quad b = \chi^2_{1-\alpha/2, n-1},$$

we get

$$P[a < V < b] = F_{n-1}(b) - F_{n-1}(a) = (1 - \alpha/2) - \alpha/2 = 1 - \alpha,$$

which is equivalent to

$$P\left[\frac{nS^2}{\chi^2_{1-\alpha/2, n-1}} < \sigma^2 < \frac{nS^2}{\chi^2_{\alpha/2, n-1}}\right] = 1 - \alpha,$$

from which we read off the wanted  $(1 - \alpha)$ -confidence interval for  $\sigma^2$ .

## Testing of hypothesis and decision

**I.6.9.** Let  $X_1, \dots, X_n$  be independent r.v.s with common distribution  $F(\cdot; \theta)$ , with an unknown parameter  $\theta$ . A *simple hypothesis* is a hypothesis of the form

$$H_0 : \theta = \theta_0.$$

This is the so-called ‘zero hypothesis’.

We may consider an ‘alternative hypothesis’ that is also simple, that is,

$$H_1 : \theta = \theta_1.$$

Let  $P_{\theta_i}$  ( $i = 0, 1$ ) denote the probability of any event ‘involving’  $X_1, \dots, X_n$ , under the assumption that their common distribution is  $F(\cdot; \theta_i)$ .

The set  $C \in \mathcal{B}(\mathbb{R}^n)$  is called the *rejection region* of a statistical test if the zero hypothesis is rejected when  $X \in C$ , where  $X = (X_1, \dots, X_n)$ .

The *significance* of the test is the probability  $\alpha$  of rejecting  $H_0$  when  $H_0$  is true.

For the simple hypothesis  $H_0$  above,

$$\alpha = \alpha(C) := P_{\theta_0}[X \in C].$$

Similarly, the probability

$$\beta = \beta(C) := P_{\theta_1}[X \in C]$$

is called the *power* of the test. It is the probability of rejecting  $H_0$  when the alternative hypothesis  $H_1$  is true.

It is clearly desirable to choose  $C$  such that  $\alpha$  is minimal and  $\beta$  is maximal. The following result goes in this direction.

**Lemma I.6.10 (Neyman–Pearson).** *Suppose the population distribution has the density  $h(\cdot; \theta)$ . For  $k \in \mathbb{R}$ , let*

$$C_k := \left\{ (x_1, \dots, x_n) \in \mathbb{R}^n; \prod_{j=1}^n h(x_j; \theta_1) > k \prod_{j=1}^n h(x_j; \theta_0) \right\}.$$

*Then among all  $C \in \mathcal{B}(\mathbb{R}^n)$  with  $\alpha(C) \leq \alpha(C_k)$ , the set  $C_k$  has maximal power.*

In symbols,  $\beta(C) \leq \beta(C_k)$  for all  $C$  with  $\alpha(C) \leq \alpha(C_k)$ .

**Proof.** Let  $C \in \mathcal{B}(\mathbb{R}^n)$  be such that  $\alpha(C) \leq \alpha(C_k)$ . Denote  $D = C \cap C_k$ . Since

$$C - D = C \cap C_k^c \subset C_k^c,$$

we have

$$\begin{aligned}
 \beta(C - D) &= \beta(C \cap C_k^c) := P_{\theta_1}[X \in C \cap C_k^c] \\
 &= \int_{C \cap C_k^c} \prod_j h(x_j; \theta_1) dx_1 \cdots dx_n \leq k \int_{C \cap C_k^c} \prod_j h(x_j; \theta_0) dx_1 \cdots dx_n \\
 &= kP_{\theta_0}[X \in C - D] = kP_{\theta_0}[X \in C] - kP_{\theta_0}[X \in D] \\
 &\leq kP_{\theta_0}[X \in C_k] - kP_{\theta_0}[X \in D] = kP_{\theta_0}[X \in C_k - D] \\
 &= k \int_{C_k - D} \prod_j h(x_j; \theta_0) dx_1 \cdots dx_n \\
 &\leq \int_{C_k - D} \prod_j h(x_j; \theta_1) dx_1 \cdots dx_n = P_{\theta_1}[X \in C_k - D] = \beta(C_k - D).
 \end{aligned}$$

Hence

$$\beta(C) = \beta(C - D) + \beta(D) \leq \beta(C_k - D) + \beta(D) = \beta(C_k).$$

□

Note that the proof does not depend on the special form of the joint density of  $(X_1, \dots, X_n)$ . Thus, if  $C_k$  is defined using the *joint density* (with the values  $\theta_i$  of the parameter), the Neyman–Pearson lemma is valid *without the independence assumption* on  $X_1, \dots, X_n$ .

**Application I.6.11.** Suppose  $F$  is the  $N(\mu, \sigma^2)$  distribution with  $\sigma^2$  known, and consider simple hypothesis

$$H_i : \mu = \mu_i, \quad i = 0, 1.$$

For  $k > 0$ , we have (by taking logarithms):

$$\begin{aligned}
 C_k &= \left\{ (x_1, \dots, x_n) \in \mathbb{R}^n; (-1/2\sigma^2) \sum_{j=1}^n [(x_j - \mu_1)^2 - (x_j - \mu_0)^2] > \log k \right\} \\
 &= \left\{ (x_1, \dots, x_n); (\mu_1 - \mu_0) \left[ \sum x_j - n(\mu_1 + \mu_0)/2 \right] > \sigma^2 \log k \right\}.
 \end{aligned}$$

Denote

$$k^* := \frac{\mu_1 + \mu_0}{2} + (\sigma^2/n) \frac{\log k}{\mu_1 - \mu_0}.$$

Then if  $\mu_1 > \mu_0$ ,

$$C_k = \{(x_1, \dots, x_n); \bar{x} > k^*\},$$

and if  $\mu_1 < \mu_0$ ,

$$C_k = \{(x_1, \dots, x_n); \bar{x} < k^*\}.$$

We choose the rejection region  $C = C_k$  for maximal power (by the Neyman–Pearson lemma). Note that it is determined by the statistic  $\bar{X}$ :  $H_0$  is rejected



if  $\bar{X} > k^*$  (in case  $\mu_1 > \mu_0$ ). The *critical value*  $k^*$  is found by means of the requirement that *the significance be equal to some given  $\alpha$* :

$$\alpha(C_k) = \alpha,$$

that is, when  $\mu_1 > \mu_0$ ,

$$P_{\mu_0}[\bar{X} > k^*] = \alpha.$$

Since  $\bar{X}$  is  $N(\mu_0, \sigma^2/n)$ -distributed under the hypothesis  $H_0$ , the statistic

$$Z := \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$$

is  $N(0, 1)$ -distributed. Using the notation of Example 1 in Section I.6.8, we get

$$\alpha = P_{\mu_0} \left[ Z > \frac{k^* - \mu_0}{\sigma/\sqrt{n}} \right] = 1 - \Phi \left( \frac{k^* - \mu_0}{\sigma/\sqrt{n}} \right),$$

so that

$$\frac{k^* - \mu_0}{\sigma/\sqrt{n}} = z_{1-\alpha},$$

and

$$k^* = \mu_0 + z_{1-\alpha}\sigma/\sqrt{n}.$$

We thus arrive to the ‘optimal’ rejection region

$$C_k = \{(x_1, \dots, x_n) \in \mathbb{R}^n; \bar{x} > \mu_0 + z_{1-\alpha}\sigma/\sqrt{n}\}$$

for significance level  $\alpha$  (in case  $\mu_1 > \mu_0$ ).

An analogous calculation for the case  $\mu_1 < \mu_0$  gives the ‘optimal’ rejection region at significance level  $\alpha$

$$C_k = \{(x_1, \dots, x_n); \bar{x} < \mu_0 + z_{\alpha}\sigma/\sqrt{n}\}.$$

**Application I.6.12.** Suppose again that  $F$  is the  $N(\mu, \sigma^2)$  distribution, this time with  $\mu$  known. Consider the simple hypothesis

$$H_i : \sigma = \sigma_i, \quad i = 0, 1.$$

We deal with the case  $\sigma_1 > \sigma_0$  (the other case is analogous).

For  $k > 0$  given, the Neyman–Pearson rejection region is (after taking logarithms)

$$C_k = \left\{ x = (x_1, \dots, x_n) \in \mathbb{R}^n; \sum_{j=1}^n (x_j - \mu)^2 > k^* \right\},$$

where

$$k^* := \frac{2 \log k + n \log(\sigma_1/\sigma_0)}{\sigma_0^{-2} - \sigma_1^{-2}}.$$

We require the significance level  $\alpha$ , that is,

$$\alpha = \alpha(C_k) = P_{\sigma_0} \left[ \sum (X_j - \mu)^2 > k^* \right].$$

Since  $(X_j - \mu)/\sigma_0^2$  are independent  $N(0, 1)$ -distributed r.v.s (under the hypothesis  $H_0$ ), the statistic

$$\chi^2 := (1/\sigma_0^2) \sum_{j=1}^n (X_j - \mu)^2$$

has the standard  $\chi_1^2$  distribution with  $n$  degrees of freedom (cf. Section I.3.15, Proposition 2). Thus

$$\alpha = P_{\sigma_0}[\chi^2 > k^*/\sigma_0^2] = 1 - F_{\chi_1^2}(k^*/\sigma_0^2).$$

Denote

$$c_\gamma = F_{\chi_1^2}^{-1}(\gamma) \quad (\gamma > 0)$$

(for  $n$  degrees of freedom). Then  $k^* = \sigma_0^2 c_{1-\alpha}$ , and the Neyman–Pearson rejection region for  $H_0$  at significance level  $\alpha$  is

$$C_k = \left\{ x \in \mathbb{R}^n; \sum (x_j - \mu)^2 > \sigma_0^2 c_{1-\alpha} \right\}.$$

## Tests based on a statistic

**I.6.13.** Suppose we wish to test the hypothesis

$$H_0 : \theta = \theta_0$$

against the alternative hypothesis

$$H_1 : \theta \neq \theta_0$$

about the parameter  $\theta$  of the population distribution  $F = F(\cdot; \theta)$ .

Let  $X_1, \dots, X_n$  be independent  $F$ -distributed r.v.s (i.e. a random sample from the population), and suppose that the distribution  $F_{g(X)}$  of some statistic  $g(X)$  (where  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  is a Borel function) is known explicitly. Denote this distribution, under the hypothesis  $H_0$ , by  $F_0$ . It is reasonable to reject  $H_0$  when  $g(X) > c$  ('one-sided test') or when either  $g(X) < a$  or  $g(X) > b$  ('two-sided test'), where  $c$  and  $a < b$  are some 'critical values'. The corresponding rejection regions are

$$C = \{x \in \mathbb{R}^n; g(x) > c\} \tag{6}$$

and

$$C = \{x \in \mathbb{R}^n; g(x) < a \text{ or } g(x) > b\}. \tag{7}$$

For the one-sided test, the significance  $\alpha$  requirement is (assuming that  $F_0$  is continuous):

$$\alpha = \alpha(C) := P_{\theta_0}[g(X) > c] = 1 - F_0(c),$$

and the corresponding *critical value* of  $c$  is

$$c_\alpha = F_0^{-1}(1 - \alpha).$$

In case (7) (which is more adequate for the ‘decision problem’ with the alternative hypothesis  $H_1$ ), it is convenient to choose the values  $a, b$  by requiring

$$P_{\theta_0}[g(X) < a] = P_{\theta_0}[g(X) > b] = \alpha/2,$$

which is sufficient for having  $\alpha(C) = \alpha$ . The *critical values* of  $a, b$  are then

$$a_\alpha = F_0^{-1}(\alpha/2); \quad b_\alpha = F_0^{-1}(1 - \alpha/2). \quad (8)$$

**Example I.6.14.** *The  $z$ -test*

Suppose  $F$  is the normal distribution with known variance. We wish to test the hypothesis

$$H_0 : \mu = \mu_0$$

against

$$H_1 : \mu \neq \mu_0.$$

Using the  $N(0, 1)$ -distributed statistic  $Z$  as in Application I.6.11, the two-sided critical values at significance level  $\alpha$  are

$$a_\alpha = z_{\alpha/2}; \quad b_\alpha = z_{1-\alpha/2}.$$

By symmetry of the normal density,  $a_\alpha = -b_\alpha$ . The zero hypothesis is rejected (at significance level  $\alpha$ ) if either  $Z < z_{\alpha/2}$  or  $Z > z_{1-\alpha/2}$ , that is, if  $|Z| > z_{1-\alpha/2}$ .

**Example I.6.15.** *The  $t$ -test*

Suppose *both parameters* of the *normal* distribution  $F$  are unknown, and consider the hypothesis  $H_i$  of Example I.6.14. By Corollary I.5.6, the statistic

$$T := \frac{\bar{X} - \mu}{S/\sqrt{n-1}}$$

has the Student distribution with  $n - 1$  degrees of freedom. With notation as in Section I.6.8, Example 2, the critical values (at significance level  $\alpha$ , for the test based on the statistic  $T$ ) are

$$a_\alpha = t_{\alpha/2, n-1}; \quad b_\alpha = t_{1-\alpha/2, n-1}.$$

By symmetry of the Student density, the zero hypothesis is rejected (at significance level  $\alpha$ ) if  $|T| > t_{1-\alpha/2, n-1}$ .

**Example I.6.16.** *Comparing the means of two normal populations.*

The zero hypothesis is that the two populations have the same normal distribution.

Two random samples  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_m$  are taken from the respective populations. Under the zero hypothesis  $H_0$ , the statistic  $W$  of Corollary I.5.7

has the Student distribution with  $\nu = n + m - 2$  degrees of freedom. By symmetry of this distribution, the two-sided test at significance level  $\alpha$  rejects  $H_0$  if  $|W| > t_{1-\alpha/2, \nu}$ .

**Example I.6.17.** *Comparing the variances of two normal populations.*

With  $H_0$ ,  $X$ , and  $Y$  as in Example I.6.16, the statistic  $F$  of Corollary I.5.11 has the Snedecor distribution with  $(n - 1, m - 1)$  degrees of freedom. If  $F_0$  is this distribution, the critical values at significance level  $\alpha$  are given by (8), Section I.6.13.

## I.7 Conditional probability

### Heuristics

Let  $(\Omega, \mathcal{A}, P)$  be a probability space, and  $A_i, B_j \in \mathcal{A}$ . Consider a two-stage experiment, with possible outcomes  $A_1, \dots, A_m$  in Stage 1 and  $B_1, \dots, B_n$  in Stage 2.

On the basis of the ‘counting principle’, it is intuitively acceptable that

$$P(A_i \cap B_j) = P(A_i)P(B_j|A_i), \quad (1)$$

where  $P(B_j|A_i)$  denotes the so-called *conditional probability* that  $B_j$  will occur in Stage 2, *when it is given that  $A_i$  occurred in Stage 1*. We take (1) as the *definition* of  $P(B_j|A_i)$  (whenever  $P(A_i) \neq 0$ ).

**Definition I.7.1.** If  $A \in \mathcal{A}$  has  $PA \neq 0$ , the conditional probability of  $B \in \mathcal{A}$  given  $A$  is

$$P(B|A) := \frac{P(A \cap B)}{PA}.$$

It is clear that  $P(\cdot|A)$  is a probability measure on  $\mathcal{A}$ . For any  $L^1(P)$  real r.v.  $X$ , the expectation of  $X$  relative to  $P(\cdot|A)$  makes sense. It is called the *conditional expectation* of  $X$  *given  $A$* , and is denoted by  $E(X|A)$ :

$$E(X|A) := \int_{\Omega} X dP(\cdot|A) = (1/PA) \int_A X dP. \quad (2)$$

Equivalently,

$$E(X|A)PA = \int_A X dP \quad (A \in \mathcal{A}, PA \neq 0). \quad (3)$$

Since  $P(B|A) = E(I_B|A)$ , we may take the conditional expectation as the basic concept, and view the conditional probability as a derived concept.

Let  $\{A_i\} \subset \mathcal{A}$  be a partition of  $\Omega$  (with  $PA_i \neq 0$ ), and let  $\mathcal{A}_0$  be the  $\sigma$ -algebra generated by  $\{A_i\}$ . Denote

$$E(X|\mathcal{A}_0) := \sum_i E(X|A_i)I_{A_i}. \quad (4)$$

This is an  $\mathcal{A}_0$ -measurable function, which takes the constant value  $E(X|A_i)$  on  $A_i$  ( $i = 1, 2, \dots$ ). Any  $A \in \mathcal{A}_0$  has the form  $A = \bigcup_{i \in J} A_i$ , where  $J \subset \mathbb{N}$ . By (3),

$$\begin{aligned} \int_A E(X|\mathcal{A}_0) dP &= \sum_i E(X|A_i) P(A_i \cap A) = \sum_{i \in J} E(X|A_i) P A_i \\ &= \sum_{i \in J} \int_{A_i} X dP = \int_A X dP \quad (A \in \mathcal{A}_0). \end{aligned} \quad (5)$$

Relation (5) may be used to *define* the conditional expectation of  $X$ , *given the (arbitrary)  $\sigma$ -subalgebra  $\mathcal{A}_0$  of  $\mathcal{A}$* :

**Definition I.7.2.** Let  $\mathcal{A}_0$  be a  $\sigma$ -subalgebra of  $\mathcal{A}$ , and let  $X$  be an  $L^1(P)$ -real r.v. The conditional expectation of  $X$  given  $\mathcal{A}_0$  is the ( $P$ -a.s. determined)  $\mathcal{A}_0$ -measurable function  $E(X|\mathcal{A}_0)$  satisfying the identity

$$\int_A E(X|\mathcal{A}_0) dP = \int_A X dP \quad (A \in \mathcal{A}_0). \quad (6)$$

Note that the right-hand side of (6) defines a real-valued measure  $\nu$  on  $\mathcal{A}_0$ , absolutely continuous with respect to  $P$  (restricted to  $\mathcal{A}_0$ ). By the Radon–Nikodym theorem, there exists a  $P$ -a.s. determined  $\mathcal{A}_0$ -measurable function, integrable on  $(\Omega, \mathcal{A}_0, P)$ , such that (6) is valid. Actually,  $E(X|\mathcal{A}_0)$  is the Radon–Nikodym derivative of  $\nu$  with respect to (the restriction of)  $P$ .

The conditional probability of  $B \in \mathcal{A}$  given  $\mathcal{A}_0$  is then defined by

$$P(B|\mathcal{A}_0) := E(I_B|\mathcal{A}_0).$$

By (6), it is the  $P$ -a.s. determined  $\mathcal{A}_0$ -measurable function satisfying

$$\int_A P(B|\mathcal{A}_0) dP = P(A \cap B) \quad (A \in \mathcal{A}_0). \quad (7)$$

We show that  $E(X|\mathcal{A}_0)$  defined by (6) coincides with the function defined before for the special case of a  $\sigma$ -subalgebra generated by a sequence of mutually disjoint *atoms*. The idea is included in the following:

**Theorem I.7.3.** *The conditional expectation  $E(X|\mathcal{A}_0)$  has a.s. the constant value  $E(X|A)$  on each  $P$ -atom  $A \in \mathcal{A}_0$ . ( $A \in \mathcal{A}_0$  is a  $P$ -atom if  $PA > 0$ , and  $A$  is not the disjoint union of two  $\mathcal{A}_0$ -measurable sets with positive  $P$ -measure.)*

**Proof.** Suppose  $f : \Omega \rightarrow [-\infty, \infty]$  is  $\mathcal{A}_0$ -measurable, and let  $A \in \mathcal{A}_0$  be a  $P$ -atom. We show that  $f$  is a.s. constant on  $A$ .

Denote  $A_x := \{\omega \in A; f(\omega) < x\}$ , for  $x > -\infty$ .

By monotonicity of  $P$ , if  $-\infty \leq y \leq x \leq \infty$  and  $PA_x = 0$ , then also  $PA_y = 0$ .

Let

$$h = \sup\{x; PA_x = 0\}.$$

Then  $PA_x = 0$  for all  $x < h$ . Since

$$A_h = \bigcup_{r < h; r \in \mathbb{Q}} A_r,$$

we have

$$PA_h = 0. \quad (8)$$

By definition of  $h$ , we have  $PA_x > 0$  for  $x > h$ , and since  $A$  is a  $P$ -atom and  $A_x \in \mathcal{A}_0$  (because  $f$  is  $\mathcal{A}_0$ -measurable) is a subset of  $A$ , it follows that  $P\{\omega \in A; f(\omega) \geq x\} = 0$  for all  $x > h$ . Writing

$$\{\omega \in A; f(\omega) > h\} = \bigcup_{r>h; r \in \mathbb{Q}} \{\omega \in A; f(\omega) \geq r\},$$

we see that  $P\{\omega \in A; f(\omega) > h\} = 0$ . Together with (8), this proves that

$$P\{\omega \in A; f(\omega) \neq h\} = 0,$$

that is,  $f(\omega) = h$   $P$ -a.s. on  $A$ .

Applying the conclusion to  $f = E(X|\mathcal{A}_0)$ , we see from (6) that this constant value is necessarily  $E(X|A)$  (cf. (2)).  $\square$

We collect some elementary properties of the conditional expectation in the following

**Theorem I.7.4.**

- (1)  $E(E(X|\mathcal{A}_0)) = EX$ .
- (2) If  $X$  is  $\mathcal{A}_0$ -measurable, then  $E(X|\mathcal{A}_0) = X$  a.s. (this is true in particular for  $X$  constant, and for any r.v.  $X$  if  $\mathcal{A}_0 = \mathcal{A}$ ).
- (3) *Monotonicity:* for real r.v.s  $X, Y \in L^1(P)$  such that  $X \leq Y$  a.s.,  $E(X|\mathcal{A}_0) \leq E(Y|\mathcal{A}_0)$  a.s. (in particular, since  $-|X| \leq X \leq |X|$ ,  $|E(X|\mathcal{A}_0)| \leq E(|X||\mathcal{A}_0)$  a.s.).
- (4) *Linearity:* for  $X, Y \in L^1(P)$  and  $\alpha, \beta \in \mathbb{C}$ ,

$$E(\alpha X + \beta Y|\mathcal{A}_0) = \alpha E(X|\mathcal{A}_0) + \beta E(Y|\mathcal{A}_0) \quad \text{a.s.}$$

**Proof.**

- (1) Take  $A = \Omega$  in (6).
- (2)  $X$  is  $\mathcal{A}_0$ -measurable (hypothesis!) and satisfies trivially (6).
- (3) The right-hand side of (6) is monotonic; therefore

$$\int_A E(X|\mathcal{A}_0) dP \leq \int_A E(Y|\mathcal{A}_0) dP$$

for all  $A \in \mathcal{A}_0$ , and the conclusion follows for example from the ‘Averages lemma’.

- (4) The right-hand side of the equation in property (4) is  $\mathcal{A}_0$ -measurable and its integral over  $A$  equals  $\int_A (\alpha X + \beta Y) dP$  for all  $A \in \mathcal{A}_0$ . By (6), it coincides a.s. with  $E(\alpha X + \beta Y|\mathcal{A}_0)$ .

$\square$

We show next that conditional expectations behave like ‘projections’ in an appropriate sense.

**Theorem I.7.5.** *Let  $\mathcal{A}_0 \subset \mathcal{A}_1$  be  $\sigma$ -subalgebras of  $\mathcal{A}$ . Then for all  $X \in L^1(P)$ ,*

$$E(E(X|\mathcal{A}_0)|\mathcal{A}_1) = E(E(X|\mathcal{A}_1)|\mathcal{A}_0) = E(X|\mathcal{A}_0) \quad \text{a.s.} \quad (9)$$

**Proof.**  $E(X|\mathcal{A}_0)$  is  $\mathcal{A}_0$ -measurable, hence also  $\mathcal{A}_1$ -measurable (since  $\mathcal{A}_0 \subset \mathcal{A}_1$ ), and therefore, by Theorem I.7.4, Part 2, the far left and far right in (9) coincide a.s.

Next, for all  $A \in \mathcal{A}_0(\subset \mathcal{A}_1)$ , we have by (6)

$$\int_A E(E(X|\mathcal{A}_1)|\mathcal{A}_0) dP = \int_A E(X|\mathcal{A}_1) dP = \int_A X dP = \int_A E(X|\mathcal{A}_0) dP,$$

so that the middle and far right expressions in (9) coincide a.s.  $\square$

‘Almost sure’ versions of the usual convergence theorems for integrals are valid for the conditional expectation.

**Theorem I.7.6 (Monotone convergence theorem for  $E(\cdot|\mathcal{A}_0)$ ).** *Let  $0 \leq X_1 \leq X_2 \leq \dots$  (a.s.) be r.v.s such that  $\lim X_n := X \in L^1(P)$ . Then*

$$E(X|\mathcal{A}_0) = \lim E(X_n|\mathcal{A}_0) \quad \text{a.s.}$$

**Proof.** By Part (3) in Theorem I.7.4,

$$0 \leq E(X_1|\mathcal{A}_0) \leq E(X_2|\mathcal{A}_0) \leq \dots \quad \text{a.s.}$$

Therefore  $h := \lim_n E(X_n|\mathcal{A}_0)$  exists a.s., and is  $\mathcal{A}_0$ -measurable (after being extended as 0 on some  $P$ -null set). By the usual Monotone Convergence theorem, we have for all  $A \in \mathcal{A}_0$

$$\int_A h dP = \lim_n \int_A E(X_n|\mathcal{A}_0) dP = \lim_n \int_A X_n dP = \int_A X dP,$$

hence  $h = E(X|\mathcal{A}_0)$  a.s.  $\square$

**Corollary I.7.7 (Beppo Levi theorem for conditional expectation).** *Let  $X_n \geq 0$  (a.s.) be r.v.s such that  $\sum_n X_n \in L^1(P)$ . Then*

$$E\left(\sum_n X_n|\mathcal{A}_0\right) = \sum_n E(X_n|\mathcal{A}_0) \quad \text{a.s.}$$

Taking in particular  $X_n = I_{B_n}$  with mutually disjoint  $B_n \in \mathcal{A}$ , we obtain the a.s.  $\sigma$ -additivity of  $P(\cdot|\mathcal{A}_0)$ :

$$P\left(\bigcup_{n=1}^{\infty} B_n|\mathcal{A}_0\right) = \sum_{n=1}^{\infty} P(B_n|\mathcal{A}_0) \quad \text{a.s.} \quad (10)$$

**Theorem I.7.8 (Dominated convergence theorem for conditional expectation).** *Let  $\{X_n\}$  be a sequence of r.v.s such that*

$$X_n \rightarrow X \quad \text{a.s.}$$

*and*

$$|X_n| \leq Y \in L^1(P).$$

*Then*

$$E(X|\mathcal{A}_0) = \lim_n E(X_n|\mathcal{A}_0) \quad \text{a.s.}$$

**Proof.** By Properties (1) and (3) in Theorem I.7.4,  $E(|E(X_n|\mathcal{A}_0)|) \leq E(Y) < \infty$ , and therefore  $E(X_n|\mathcal{A}_0)$  is finite a.s., and similarly  $E(X|\mathcal{A}_0)$ . Hence  $E(X_n|\mathcal{A}_0) - E(X|\mathcal{A}_0)$  is well-defined and finite a.s., and has absolute value equal a.s. to

$$|E(X_n - X|\mathcal{A}_0)| \leq E(|X_n - X||\mathcal{A}_0) \leq E(Z_n|\mathcal{A}_0), \quad (11)$$

where

$$Z_n := \sup_{k \geq n} |X_k - X| (\in L^1(P)).$$

Since  $Z_n$  is a non-increasing sequence (with limit 0 a.s.), Property (3) in Theorem I.7.4 implies that  $E(Z_n|\mathcal{A}_0)$  is a non-increasing sequence a.s. Let  $h$  be its (a.s.) limit. After proper extension on a  $P$ -null set,  $h$  is a non-negative  $\mathcal{A}_0$ -measurable function. Since  $0 \leq Z_n \leq 2Y \in L^1(P)$ , the usual Dominated Convergence theorem gives

$$0 \leq \int_{\Omega} h \, dP \leq \int_{\Omega} E(Z_n|\mathcal{A}_0) \, dP = \int_{\Omega} Z_n \, dP \rightarrow_n 0,$$

hence  $h = 0$  a.s. By (11), this gives the conclusion of the theorem.  $\square$

Property (2) in Theorem I.7.4 means that  $\mathcal{A}_0$ -measurable functions behave like constants relative to the operation  $E(\cdot|\mathcal{A}_0)$ . This ‘constant like’ behaviour is a special case of the following:

**Theorem I.7.9.** *Let  $X, Y$  be r.v.s such that  $X, Y, XY \in L^1(P)$ . If  $X$  is  $\mathcal{A}_0$ -measurable, then*

$$E(XY|\mathcal{A}_0) = X E(Y|\mathcal{A}_0) \quad \text{a.s.} \quad (12)$$

**Proof.** If  $B \in \mathcal{A}_0$  and  $X = I_B$ , then for all  $A \in \mathcal{A}_0$ ,

$$\int_A X E(Y|\mathcal{A}_0) \, dP = \int_{A \cap B} E(Y|\mathcal{A}_0) \, dP = \int_{A \cap B} Y \, dP = \int_A XY \, dP,$$

so that (12) is valid for  $\mathcal{A}_0$ -measurable indicators, and by linearity, for all  $\mathcal{A}_0$ -measurable simple functions. For an arbitrary  $\mathcal{A}_0$ -measurable r.v.  $X \in L^1(P)$ , there exists a sequence  $\{X_n\}$  of  $\mathcal{A}_0$ -measurable simple functions such that  $X_n \rightarrow X$  and  $|X_n| \leq |X|$ . We have

$$E(X_n Y|\mathcal{A}_0) = X_n E(Y|\mathcal{A}_0) \quad \text{a.s.}$$



Since  $E(Y|\mathcal{A}_0)$  is  $P$ -integrable, it is a.s. finite, and therefore the right-hand side converges a.s. to  $X E(Y|\mathcal{A}_0)$ .

Since  $X_n Y \rightarrow XY$  and  $|X_n Y| \leq |XY| \in L^1(P)$ , the left-hand side converges a.s. to  $E(XY|\mathcal{A}_0)$  by Theorem I.7.8, and the result follows.  $\square$

## Conditioning by a r.v.

**I.7.10.** Given a r.v.  $X$ , it induces a  $\sigma$ -subalgebra  $\mathcal{A}_X$  of  $\mathcal{A}$ , where

$$\mathcal{A}_X := \{X^{-1}(B); B \in \mathcal{B}\},$$

and  $\mathcal{B}$  is the Borel algebra of  $\mathbb{R}$  (or  $\mathbb{C}$ ). It is then ‘natural’ to define

$$E(Y|X) := E(Y|\mathcal{A}_X) \quad (13)$$

for any integrable r.v.  $Y$ .

Thus  $E(Y|X)$  is the a.s. uniquely determined ( $\mathcal{A}_X$ )-measurable function such that

$$\int_{X^{-1}(B)} E(Y|X) dP = \int_{X^{-1}(B)} Y dP \quad (14)$$

for all  $B \in \mathcal{B}$ .

As a function of  $B$ , the right-hand side of (14) is a real (or complex) measure on  $\mathcal{B}$ , absolutely continuous with respect to the probability measure  $P_X$  [ $P_X(B) = 0$  means that  $P(X^{-1}(B)) = 0$ , which implies that the right-hand side of (14) is zero]. By the Radon–Nikodym theorem, there exists a unique (up to  $P_X$ -equivalence) Borel  $L^1(P_X)$ -function  $h$  such that

$$\int_B h dP_X = \int_{X^{-1}(B)} Y dP \quad (B \in \mathcal{B}). \quad (15)$$

We shall denote (for  $X$  real valued)

$$h(x) := E(Y|X = x) \quad (x \in \mathbb{R}), \quad (16)$$

and call this function the ‘conditional expectation of  $Y$ , given  $X = x$ ’. Thus, by definition,

$$\int_B E(Y|X = x) dP_X(x) = \int_{X^{-1}(B)} Y dP \quad (B \in \mathcal{B}). \quad (17)$$

Taking  $B = \mathbb{R}$  in (17), we see that

$$E_{P_X}(E(Y|X = x)) = E(Y), \quad (18)$$

where  $E_{P_X}$  denotes the expectation operator on  $L^1(\mathbb{R}, \mathcal{B}, P_X)$ .

The proof of Theorem I.7.3 shows that  $E(Y|X = x)$  is  $P_X$ -a.s. constant on each  $P_X$ -atom  $B \in \mathcal{B}$ . By (17), we have

$$E(Y|X = x) = \frac{1}{P(X^{-1}(B))} \int_{X^{-1}(B)} Y dP \quad (19)$$

$P_X$ -a.s. on  $B$ , for each  $P_X$ -atom  $B \in \mathcal{B}$ .

As before, the ‘conditional probability of  $A \in \mathcal{A}$  given  $X = x$ ’ is defined by

$$P(A|X = x) := E(I_A|X = x) \quad (x \in \mathbb{R}),$$

or directly by (17) for the special case  $Y = I_A$ :

$$\int_B P(A|X = x) dP_X(x) = P(A \cap X^{-1}(B)) \quad (B \in \mathcal{B}). \quad (20)$$

If  $B \in \mathcal{B}$  is a  $P_X$ -atom, we have by (19)

$$P(A|X = x) = \frac{P(A \cap X^{-1}(B))}{P(X^{-1}(B))} = P(A|[X \in B]) \quad (21)$$

$P_X$ -almost surely on  $B$ , where the right-hand side of (21) is the ‘elementary’ conditional probability of the event  $A \in \mathcal{A}$ , given the event  $[X \in B]$ . In particular, if  $B = \{x\}$  is a  $P_X$ -atom (i.e., if  $P_X(\{x\}) > 0$ ; i.e. if  $P[X = x] > 0$ ), then

$$P(A|X = x) = P(A|[X = x]) \quad (A \in \mathcal{A}), \quad (22)$$

so that the notation is ‘consistent’.

The relation between the  $\mathcal{A}_X$ -measurable function  $E(Y|X)$  and the Borel function  $h(x) := E(Y|X = x)$  is stated in the next.

**Theorem I.7.11.** *Let  $(\Omega, \mathcal{A}, P)$  be a probability space, and let  $X, Y$  be (real) r.v.s, with  $Y$  integrable. Then,  $P$ -almost surely,*

$$E(Y|X) = h(X),$$

where  $h(x) := E(Y|X = x)$ .

**Proof.** By (14) and (15),

$$\int_{X^{-1}(B)} E(Y|X) dP = \int_B h dP_X \quad (B \in \mathcal{B}). \quad (23)$$

We claim that

$$\int_B h dP_X = \int_{X^{-1}(B)} h(X) dP \quad (B \in \mathcal{B}) \quad (24)$$

for any real Borel  $P_X$ -integrable function  $h$  on  $\mathbb{R}$ . If  $h = I_C$  for  $C \in \mathcal{B}$ , (24) is valid, since

$$\begin{aligned} \int_B h dP_X &= P_X(B \cap C) = P(X^{-1}(B \cap C)) \\ &= P(X^{-1}(B) \cap X^{-1}(C)) = \int_{X^{-1}(B)} I_{X^{-1}(C)} dP = \int_{X^{-1}(B)} h(X) dP. \end{aligned}$$

By linearity, (24) is then valid for simple Borel functions  $h$ . If  $h$  is a non-negative Borel function, let  $\{h_n\}$  be a sequence of simple Borel functions such that  $0 \leq h_1 \leq h_2 \leq \dots$ , and  $\lim h_n = h$ . Then  $\{h_n(X)\}$  is a sequence of  $\mathcal{A}_X$ -measurable functions such that

$$0 \leq h_1(X) \leq h_2(X) \leq \dots$$

and  $\lim h_n(X) = h(X)$ . By the Monotone Convergence theorem applied in the measure spaces  $(\mathbb{R}, \mathcal{B}, P_X)$  and  $(\Omega, \mathcal{A}, P)$ , we have for all  $B \in \mathcal{B}$ :

$$\int_B h dP_X = \lim_n \int_B h_n dP_X = \lim_n \int_{X^{-1}(B)} h_n(X) dP = \int_{X^{-1}(B)} h(X) dP.$$

For any real  $P_X$ -integrable Borel function  $h$ , write  $h = h^+ - h^-$ ; then

$$\begin{aligned} \int_B h dP_X &:= \int_B h^+ dP_X - \int_B h^- dP_X \\ &= \int_{X^{-1}(B)} h^+(X) dP - \int_{X^{-1}(B)} h^-(X) dP \\ &= \int_{X^{-1}(B)} h(X) dP \quad (B \in \mathcal{B}). \end{aligned}$$

Thus (24) is verified, and by (23), we have

$$\int_A E(Y|X) dP = \int_A h(X) dP$$

for all  $A \in \mathcal{A}_X := \{X^{-1}(B); B \in \mathcal{B}\}$ .

Since both integrands are in  $L^1(\Omega, \mathcal{A}_X, P)$ , it follows that they coincide  $P$ -almost surely.  $\square$

**Theorem I.7.12.** *Let  $X, Y$  be (real) r.v.s, with  $Y \in L^2(P)$ . Then  $Z = E(Y|X)$  is the (real)  $\mathcal{A}_X$ -measurable solution in  $L^2(P)$  of the extremal problem*

$$\|Y - Z\|_2 = \min.$$

(Geometrically,  $Z$  is the orthogonal projection of  $Y$  onto  $L^2(\Omega, \mathcal{A}_X, P)$ .)

**Proof.** Write (for  $Y, Z \in L^2(P)$ ):

$$\begin{aligned} (Y - Z)^2 &= [Y - E(Y|X)]^2 + [E(Y|X) - Z]^2 \\ &\quad + 2[E(Y|X) - Z][Y - E(Y|X)]. \end{aligned} \tag{25}$$

In particular, the third term is  $\leq (Y - Z)^2$ . Similarly, we see that the negative of the third term is also  $\leq (Y - Z)^2$ . Hence this term has absolute value  $\leq (Y - Z)^2 \in L^1(P)$ . Since  $E(Y|X) \in L^1(P)$ , the functions  $U := E(Y|X) - Z$ ,  $V := Y - E(Y|X)$ , and  $UV$  are all in  $L^1(P)$ , and  $U$  is  $\mathcal{A}_X$ -measurable whenever  $Z$  is. By Theorem I.7.9 with  $\mathcal{A}_0 = \mathcal{A}_X$ ,

$$\begin{aligned} E(UV|X) &:= E(UV|\mathcal{A}_X) = UE(V|\mathcal{A}_X) = UE(V|X) \\ &= U[E(Y|X) - E(Y|X)] = 0. \end{aligned}$$

Hence by (25)

$$E([Y - Z]^2|X) = E(U^2|X) + E(V^2|X).$$

Applying  $E$ , we obtain

$$E((Y - Z)^2) = E(U^2) + E(V^2) \geq E(V^2),$$

that is,  $\|Y - Z\|_2 \geq \|Y - E(Y|X)\|_2$ , with the minimum attained when  $U = 0$  ( $P$ -a.s.), that is, when  $Z = E(Y|X)$  a.s.  $\square$

Applying Theorem I.7.11, we obtain the following extremal property of  $h = E(Y|X = \cdot)$ :

**Corollary I.7.13.** *Let  $X, Y$  be (real) r.v.s, with  $Y \in L^2(P)$ . Then the extremal problem for (real) Borel functions  $g$  on  $\mathbb{R}$  with  $g(X) \in L^2(P)$*

$$\|Y - g(X)\|_2 = \min$$

*has the solution*

$$g = h := E(Y|X = \cdot) \quad \text{a.s.}$$

Thus  $h(X)$  gives the best ‘mean square approximation’ of  $Y$  by ‘functions of  $X$ ’. The graph of the equation

$$y = h(x)(:= E(Y|X = x))$$

is called the *regression curve of  $Y$  on  $X$* .

**I.7.14. Linear regression.** We consider the extremal problem of Corollary I.7.13 with the stronger restriction that  $g$  be *linear*. Thus we wish to find values of the real parameters  $a, b$  such that

$$\|Y - (aX + b)\|_2 = \min,$$

where  $X, Y$  are given non-degenerate  $L^2(P)$ -r.v.s. Necessarily,  $X, Y$  have finite expectations  $\mu_k$  and standard deviations  $\sigma_k > 0$ , and we may define the so-called *correlation coefficient of  $X$  and  $Y$*

$$\rho = \rho(X, Y) := \frac{\text{cov}(X, Y)}{\sigma_1 \sigma_2}.$$

By I.2.5,  $|\rho| \leq 1$ .

We have

$$\begin{aligned} \|Y - (aX + b)\|_2^2 &= E([Y - \mu_2] - a[X - \mu_1] + [\mu_2 - (a\mu_1 + b)])^2 \\ &= E(Y - \mu_2)^2 + a^2 E(X - \mu_1)^2 + [\mu_2 - (a\mu_1 + b)]^2 - 2aE((X - \mu_1)(Y - \mu_2)) \\ &= \sigma_2^2 + a^2 \sigma_1^2 + [\mu_2 - (a\mu_1 + b)]^2 - 2a\rho\sigma_1\sigma_2 \\ &= (a\sigma_1 - \rho\sigma_2)^2 + (1 - \rho^2)\sigma_2^2 + [\mu_2 - (a\mu_1 + b)]^2 \geq (1 - \rho^2)\sigma_2^2, \end{aligned}$$

with equality (giving the minimal  $L^2$ -distance  $\sigma_2\sqrt{1-\rho^2}$ ) attained when

$$a\sigma_1 - \rho\sigma_2 = 0, \quad \mu_2 - (a\mu_1 + b) = 0,$$

that is, when

$$a = a^* := \rho\sigma_2/\sigma_1; \quad b = b^* := \mu_2 - a^*\mu_1.$$

In conclusion, the *linear solution* of our extremum problem (the so-called *linear regression of  $Y$  on  $X$* ) has the equation

$$y = \mu_2 + \rho \frac{\sigma_2}{\sigma_1} (x - \mu_1).$$

Note that the minimal  $L^2$ -distance vanishes iff  $|\rho| = 1$ ; in that case  $Y = a^*X + b^*$  a.s.

**I.7.15.** *Conditional distribution; discrete case.*

Let  $X, Y$  be discrete real r.v.'s, with respective ranges  $\{x_j\}$  and  $\{y_k\}$ . The vector-valued r.v.  $(X, Y)$  assumes the value  $(x_j, y_k)$  with the positive probability  $p_{jk}$ , where

$$\sum_{j,k} p_{jk} = 1.$$

The *joint distribution function* of  $(X, Y)$  is defined (in general, for any real r.v.'s) by

$$F(x, y) := P[X < x, Y < y] \quad (x, y \in \mathbb{R}).$$

In the discrete case above,

$$F(x, y) = \sum_{x_j < x, y_k < y} p_{jk}.$$

The *marginal distributions* are defined in general by

$$F_X(x) := P[X < x] = F(x, \infty); \quad F_Y(y) := P[Y < y] = F(\infty, y).$$

In our case,

$$F_X(x) = \sum_{x_j < x} p_{j.}; \quad F_Y(y) = \sum_{y_k < y} p_{.k},$$

where

$$p_{j.} := \sum_k p_{jk} = P[X = x_j]; \quad p_{.k} := \sum_j p_{jk} = P[Y = y_k].$$

Each singleton  $\{x_j\}$  is a  $P_X$ -atom (because  $P[X = x_j] = p_{j.} > 0$ ). By (22) in Section I.7.10 (with  $A = [Y = y_k]$ ), we have

$$P(Y = y_k | X = x_j) = \frac{p_{jk}}{p_{j.}} \quad (j, k = 1, 2, \dots)$$

and similarly

$$P(X = x_j | Y = y_k) = \frac{p_{jk}}{p_{\cdot k}}.$$

Note that

$$\sum_k P(Y = y_k | X = x_j) = 1,$$

and therefore the function of  $y$  given by

$$F(y | X = x_j) := \sum_{y_k < y} P(Y = y_k | X = x_j) = (1/p_{\cdot j}) \sum_{y_k < y} p_{jk} \quad (j = 1, 2, \dots)$$

is a distribution function. It is called the *conditional distribution of  $Y$ , given  $X = x_j$* .

**I.7.16. Conditional distribution; Continuous case.** Consider now the case where the vector-valued r.v.  $(X, Y)$  has a (joint) density  $h$  (cf. Section I.5). Then the distribution function of  $(X, Y)$  is given by

$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y h(s, t) ds dt \quad (x, y \in \mathbb{R}).$$

By Tonelli's theorem, the order of integration is irrelevant. At all continuity points  $(x, y)$  of  $h$ , one has  $h(x, y) = \partial^2 F / \partial x \partial y$ . The *marginal density functions* are defined by

$$h_X(x) := \int_{\mathbb{R}} h(x, y) dy; \quad h_Y(y) := \int_{\mathbb{R}} h(x, y) dx \quad (x, y \in \mathbb{R}).$$

These are densities for the distribution function  $F_X$  and  $F_Y$ , respectively.

If  $S := \{(x, y) \in \mathbb{R}^2; h_X(x) = 0\} (= h_X^{-1}(\{0\}) \times \mathbb{R})$ , then

$$P[(X, Y) \in S] = P[X \in h_X^{-1}(\{0\})] = \int_{h_X^{-1}(\{0\})} h_X(x) dx = 0,$$

so that  $S$  may be disregarded. On  $\mathbb{R}^2 - S$ , define

$$h(y|x) := \frac{h(x, y)}{h_X(x)}. \quad (26)$$

This function is called the *conditional distribution density of  $Y$ , given  $X = x$* . The terminology is motivated by the following

**Proposition I.7.17.** *In the above setting, we have  $P_X$ -almost surely*

$$E(Y | X = x) = \int_{\mathbb{R}} y h(y|x) dy.$$

**Proof.** For all  $B \in \mathcal{B}(\mathbb{R})$ , we have by Fubini's theorem:

$$\begin{aligned} \int_{X^{-1}(B)} Y \, dP &= \iint_{B \times \mathbb{R}} y h(x, y) \, dx \, dy = \int_B h_X(x) \int_{\mathbb{R}} y h(y|x) \, dy \, dx \\ &= \int_B \left( \int_{\mathbb{R}} y h(y|x) \, dy \right) dP_X(x), \end{aligned}$$

and the conclusion follows from (17) in Section I.7.10.  $\square$

If  $h$  is *continuous* on  $\mathbb{R}^2$ , we also have

**Proposition I.7.18.** *Suppose the joint distribution density  $h$  of  $(X, Y)$  is continuous on  $\mathbb{R}^2$ . Then for all  $x \in \mathbb{R}$  for which  $h_X(x) \neq 0$  and for all  $B \in \mathcal{B}(\mathbb{R})$ , we have*

$$\int_B h(y|x) \, dy = \lim_{\delta \rightarrow 0+} P(Y \in B \mid x - \delta < X < x + \delta).$$

**Proof.** For  $\delta > 0$ ,

$$\begin{aligned} P(Y \in B \mid x - \delta < X < x + \delta) &= \frac{P([Y \in B] \cap [x - \delta < X < x + \delta])}{P[x - \delta < X < x + \delta]} \\ &= \frac{\int_{x-\delta}^{x+\delta} \int_B h(s, y) \, dy \, ds}{\int_{x-\delta}^{x+\delta} h_X(s) \, ds}. \end{aligned}$$

Divide numerator and denominator by  $2\delta$  and let  $\delta \rightarrow 0$ . The continuity assumption implies that, for all  $x$  for which  $h_X(x) \neq 0$ , the last expression has the limit

$$\frac{\int_B h(x, y) \, dy}{h_X(x)} = \int_B h(y|x) \, dy.$$

$\square$

It follows from (26) that  $\int_{\mathbb{R}} h(y|x) \, dy = 1$ , so that  $h(y|x)$  (defined for all  $x$  such that  $h_X(x) \neq 0$ ) is the density of a distribution function:

$$F(y|x) := \int_{-\infty}^y h(t|x) \, dt,$$

called the *conditional distribution of  $Y$  given  $X = x$* .

**Example I.7.19 (The binormal distribution).** We say that  $X, Y$  are *binormally distributed* if they have the joint density function (called the *binormal density*) given by

$$h(x, y) = (1/c) \exp \left( -Q \left( \frac{x - \mu_1}{\sigma_1}, \frac{y - \mu_2}{\sigma_2} \right) \right),$$

where  $Q$  is the positive definite quadratic form

$$Q(s, t) := \frac{s^2 - 2\rho st + t^2}{2(1 - \rho^2)},$$

and

$$c = 2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}$$

( $\mu_k \in \mathbb{R}$ ,  $\sigma_k > 0$ ,  $-1 < \rho < 1$  are the parameters of the distribution).

Note that

$$Q(s, t) = [(s - \rho t)^2 + (1 - \rho^2)t^2]/[2(1 - \rho^2)] \geq 0$$

for all real  $s, t$ , with equality holding iff  $s = t = 0$ . Therefore  $h$  attains its absolute maximum  $1/c$  at the unique point  $(x, y) = (\mu_1, \mu_2)$ .

The sections of the surface  $z = h(x, y)$  with the planes  $z = a$  are empty for  $a > 1/c$  (and  $a \leq 0$ ); a single point for  $a = 1/c$ ; and ellipses for  $0 < a < 1/c$  (the surface is ‘bell-shaped’).

In order to calculate the integral  $\int_{\mathbb{R}^2} h(x, y) dx dy$ , we make the transformation

$$x = \mu_1 + \sigma_1 s = \mu_1 + \sigma_1(u + \rho t); \quad y = \mu_2 + \sigma_2 t,$$

where  $u := s - \rho t$ . Then  $(u, t)$  ranges over  $\mathbb{R}^2$  when  $(x, y)$  does, and

$$\frac{\partial(x, y)}{\partial(u, t)} = \sigma_1\sigma_2 > 0.$$

Therefore the above integral is equal to

$$\begin{aligned} (1/c) \iint_{\mathbb{R}^2} e^{-((u^2 + (1-\rho^2)t^2)/(2(1-\rho^2)))} \sigma_1\sigma_2 du dt \\ = (1/\sqrt{2\pi(1-\rho^2)}) \int_{\mathbb{R}} e^{-u^2/2(1-\rho^2)} du = 1, \end{aligned}$$

since the last integral is that of the  $N(0, 1 - \rho^2)$ -density.

Thus  $h$  is indeed the density of a two-dimensional distribution function.

Since  $Q(s, t) = s^2/2 + (t - \rho s)^2/2(1 - \rho^2)$ , we get (for  $x \in \mathbb{R}$  fixed, with  $s = (x - \mu_1)/\sigma_1$  and  $t = (y - \mu_2)/\sigma_2$ , so that  $dy = \sigma_2 dt$ ):

$$h_X(x) = (1/c)e^{-s^2/2} \int_{\mathbb{R}} e^{-(t-\rho s)^2/2(1-\rho^2)} \sigma_2 dt = \frac{1}{\sqrt{2\pi}\sigma_1} e^{-(x-\mu_1)^2/2\sigma_1^2}.$$

Thus  $h_X$  is the  $N(\mu_1, \sigma_1^2)$ -density. By symmetry, the marginal density  $h_Y$  is the  $N(\mu_2, \sigma_2^2)$ -density. In particular, the meaning of the parameters  $\mu_k$  and  $\sigma_k^2$  has been clarified (as the expectations and variance of  $X$  and  $Y$ ).

We have (with  $s, t$  related to  $x, y$  as before and  $c' = \sqrt{2\pi\sigma_2^2(1-\rho^2)}$ ):

$$\begin{aligned} h(y|x) &:= \frac{h(x, y)}{h_X(x)} = (1/c') \exp \left\{ -\frac{\rho^2 s^2 - 2\rho st + t^2}{2(1-\rho^2)} \right\} \\ &= (1/c') \exp \left\{ -(t - \rho s)^2/2(1-\rho^2) \right\} \\ &= (1/c') \exp \left\{ -\frac{[y - (\mu_2 + \rho(\sigma_2/\sigma_1)(x - \mu_1))]^2}{2\sigma_2^2(1-\rho^2)} \right\}. \end{aligned}$$



Thus  $h(y|x)$  is the  $N(\mu_2 + \rho(\sigma_2/\sigma_1)(x - \mu_1), \sigma_2^2(1 - \rho^2))$  density.

By Proposition I.7.17, for all real  $x$ ,

$$E(Y|X = x) = \mu_2 + \rho(\sigma_2/\sigma_1)(x - \mu_1),$$

with an analogous formula for  $E(X|Y = y)$ .

Thus, for binormally distributed  $X, Y$ , the regression curves  $y = E(Y|X = x)$  and  $x = E(X|Y = y)$  coincide with the linear regression curves (cf. Section I.7.14). They intersect at  $(\mu_1, \mu_2)$ , and the coefficient  $\rho$  here coincides with the correlation coefficient  $\rho(X, Y)$  (cf. Section I.7.14). Indeed (with previous notations),

$$\begin{aligned} \rho(X, Y) &= E\left(\frac{X - \mu_1}{\sigma_1} \cdot \frac{Y - \mu_2}{\sigma_2}\right) \\ &= (1/c) \iint_{\mathbb{R}^2} (u + \rho t)t \exp\left\{-\frac{u^2 + (1 - \rho^2)t^2}{2(1 - \rho^2)}\right\} \sigma_1 \sigma_2 du dt. \end{aligned}$$

The integrand splits as the sum of two terms. The term with the factor  $ut$  is odd in each variable; by Fubini's theorem, its integral vanishes. The remaining integral is

$$\rho \frac{1}{\sqrt{2\pi(1 - \rho^2)}} \int_{\mathbb{R}} e^{-u^2/2(1 - \rho^2)} du \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} t^2 e^{-t^2/2} dt = \rho.$$

Note in particular that if the binormally distributed r.v.s  $X, Y$  are uncorrelated, then  $\rho = \rho(X, Y) = 0$ , and therefore  $h(x, y) = h_X(x)h_Y(y)$ . By Proposition I.5.1, it follows that  $X, Y$  are independent. Since the converse is generally true (cf. Section I.2.5), we have

**Proposition.** *If the r.v.s  $X, Y$  are binormally distributed, then they are independent if and only if they are uncorrelated.*

## I.8 Series of $L^2$ random variables

The problem considered in this section is the a.s. convergence of series of independent r.v.s.

We fix the following notation:  $\{X_k\}_{k=1}^\infty$  is a sequence of real independent central  $L^2(P)$  random variables; for  $n = 1, 2, \dots$ , we let

$$S_n = \sum_{k=1}^n X_k; \quad s_n^2 = \sigma^2(S_n) = \sum_{k=1}^n \sigma_k^2; \quad \sigma_k^2 = \sigma^2(X_k);$$

and

$$T_n = \max_{1 \leq m \leq n} |S_m|.$$

**Lemma I.8.1 (Kolmogorov).** *For each  $\epsilon > 0$  and  $n = 1, 2, \dots$ ,*

$$P[T_n \geq \epsilon] \leq s_n^2/\epsilon^2.$$

**Proof.** Write

$$\begin{aligned} [T_n \geq \epsilon] &= [|S_1| \geq \epsilon] \cup \{|S_2| \geq \epsilon\} \cap [|S_1| < \epsilon] \cup \dots \\ &\cup \{|S_n| \geq \epsilon\} \cap [|S_k| < \epsilon; k = 1, \dots, n-1], \end{aligned}$$

and denote the  $k$ th set in this union by  $A_k$ .

By independence and centrality of  $X_k$ ,

$$\begin{aligned} \int_{A_k} S_n^2 dP &= \sum_{j=1}^n \int_{A_k} X_j^2 dP \geq \sum_{j=1}^k \int_{A_k} X_j^2 dP \\ &= \int_{A_k} S_k^2 dP \geq \epsilon^2 P(A_k). \end{aligned}$$

Since the sets  $A_k$  are mutually disjoint, we get

$$\begin{aligned} \epsilon^2 P[T_n \geq \epsilon] &= \sum_{k=1}^n \epsilon^2 P(A_k) \leq \sum_{k=1}^n \int_{A_k} S_n^2 dP \\ &= \int_{[T_n \geq \epsilon]} S_n^2 dP \leq E(S_n^2) = s_n^2. \end{aligned}$$

□

**Theorem I.8.2.** (For  $X_k$  as above.) If  $\sum_k \sigma_k^2 < \infty$ , then  $\sum_k X_k$  converges a.s.

**Proof.** Fix  $\epsilon > 0$ . For  $n, m \in \mathbb{N}$ , denote

$$A_{nm} = \left[ \max_{1 \leq k \leq n} \left| \sum_{j=m+1}^{m+k} X_j \right| > \epsilon \right]$$

and

$$A_m = \left[ \sup_{1 \leq k < \infty} \left| \sum_{j=m+1}^{m+k} X_j \right| > \epsilon \right].$$

Then  $A_m$  is the union of the increasing sequence  $\{A_{nm}\}_n$ , so that

$$P(A_m) = \lim_n P(A_{nm}).$$

By Lemma I.8.1,

$$P(A_{nm}) \leq (1/\epsilon^2) \sum_{k=m+1}^{m+n} \sigma_k^2 \leq (1/\epsilon^2) \sum_{k=m+1}^{\infty} \sigma_k^2.$$

Hence for all  $m$

$$P(A_m) \leq (1/\epsilon^2) \sum_{k=m+1}^{\infty} \sigma_k^2,$$

and therefore

$$P[\inf_m \sup_k |\sum_{j=m+1}^{m+k} X_j| > \epsilon] \leq (1/\epsilon^2) \sum_{k=m+1}^{\infty} \sigma_k^2.$$

The right-hand side tends to zero when  $m \rightarrow \infty$  (by hypothesis), and therefore the left-hand side equals 0. Thus

$$\inf_m \sup_k |\sum_{j=m+1}^{m+k} X_j| \leq \epsilon \quad \text{a.s.},$$

hence there exists  $m$ , such that for all  $k$ , one has  $|\sum_{j=m+1}^{m+k} X_j| < 2\epsilon$  (a.s.).  $\square$

For an a.s. *bounded* sequence of r.v.s, a converse result is proved below.

**Theorem I.8.3.** *Let  $\{X_k\}$  be a sequence of independent central r.v.s such that*

(i)  $|X_k| \leq c$  a.s.; and

(ii)  $P[\sum_{k=1}^{\infty} X_k \text{ converges}] > 0$ .

Then  $\sum_k \sigma_k^2 < \infty$ .

**Proof.** By (i),  $|S_n| \leq nc$  a.s.

Let  $A$  be the set on which  $\{S_n\}$  converges. Since  $PA > 0$  by (ii), it follows from Egoroff's theorem (Theorem 1.57) that  $\{S_n\}$  converges uniformly on some measurable subset  $B \subset A$  with  $PB > 0$ . Hence  $|S_n| \leq d$  for all  $n$  on some measurable subset  $E \subset B$  with  $PE > 0$ . Let

$$E_n = [|S_k| \leq d; 1 \leq k \leq n] \quad (n \in \mathbb{N}).$$

The sequence  $\{E_n\}$  is decreasing, with intersection  $E$ . Let  $\alpha_0 = 0$  and

$$\alpha_n := \int_{E_n} S_n^2 dP \quad (n \in \mathbb{N}).$$

Write

$$F_n = E_{n-1} - E_n (\subset E_{n-1}); \quad E_n = E_{n-1} - F_n,$$

so that

$$\alpha_n - \alpha_{n-1} = \int_{E_{n-1}} S_n^2 dP - \int_{F_n} S_n^2 dP - \int_{E_{n-1}} S_{n-1}^2 dP.$$

Since  $X_n$  and  $S_{n-1}$  are central and independent, we have by BienAyme's identity

$$\int_{E_{n-1}} S_n^2 dP = \int_{E_{n-1}} X_n^2 dP + \int_{E_{n-1}} S_{n-1}^2 dP,$$

and therefore

$$\alpha_n - \alpha_{n-1} = \int_{E_{n-1}} X_n^2 dP - \int_{F_n} S_n^2 dP.$$

On  $F_n(\subset E_{n-1})$ , we have

$$|S_n| \leq |X_n| + |S_{n-1}| \leq c + d \quad \text{a.s.}$$

Therefore

$$\alpha_n - \alpha_{n-1} \geq \int_{E_{n-1}} X_n^2 dP - (c + d)^2 P(F_n). \quad (1)$$

Since  $I_{E_{n-1}}$  (which is defined exclusively by means of  $X_1, \dots, X_{n-1}$ ) and  $X_n^2$  are independent, it follows from Theorem I.2.2 that

$$\int_{E_{n-1}} X_n^2 dP = E(I_{E_{n-1}} X_n^2) = P(E_{n-1}) \sigma^2(X_n) \geq P(E) \sigma_n^2.$$

Hence, by (1), and summing all the inequalities for  $n = 1, \dots, k$ , we obtain

$$\alpha_k \geq P(E) \sum_{n=1}^k \sigma_n^2 - (c + d)^2 \sum_{n=1}^k P(F_n).$$

However, the sets  $F_n$  are disjoint, so that

$$\sum_{n=1}^k P(F_n) = P\left(\bigcup_{n=1}^k F_n\right) \leq 1,$$

hence

$$\alpha_k \geq P(E) \sum_{n=1}^k \sigma_n^2 - (c + d)^2.$$

Since  $P(E) > 0$ , we obtain for all  $k \in \mathbb{N}$

$$\sum_{n=1}^k \sigma_n^2 \leq \frac{\alpha_k + (c + d)^2}{P(E)} \leq \frac{d^2 + (c + d)^2}{P(E)},$$

so that  $\sum_n \sigma_n^2 < \infty$ . □

We consider next the non-central case.

**Theorem I.8.4.** *Let  $\{X_k\}$  be a sequence of independent r.v.s, such that  $|X_k| \leq c, k = 1, 2, \dots$  a.s. Then  $\sum_k X_k$  converges a.s. if and only if the two (numerical) series  $\sum_k E(X_k)$  and  $\sum_k \sigma_k^2$  converge.*

**Proof.** Suppose the two ‘numerical’ series converge (for this part of the proof, the hypothesis  $|X_k| \leq c$  a.s. is not needed, and  $X_k$  are only assumed in  $L^2$ , our standing hypothesis). Let  $Y_k = X_k - E(X_k)$ . Then  $Y_k$  are independent central  $L^2$  random variables, and  $\sum_k \sigma^2(Y_k) = \sum_k \sigma^2(X_k) < \infty$ . By Theorem I.8.2,  $\sum_k Y_k$  converges a.s., and therefore  $\sum_k X_k = \sum_k [Y_k + E(X_k)]$  converges a.s., since  $\sum E(X_k)$  converges by hypothesis.

Conversely, suppose that  $\sum X_k$  converges a.s.

Define on the product probability space

$$(\Omega, \mathcal{A}, P) \times (\Omega, \mathcal{A}, P)$$

the random variables

$$Z_n(\omega_1, \omega_2) := X_n(\omega_1) - X_n(\omega_2).$$

Then  $Z_n$  are independent. They are central, since

$$\begin{aligned} E(Z_n) &= \int_{\Omega \times \Omega} [X_n(\omega_1) - X_n(\omega_2)] d(P \times P) \\ &= \int_{\Omega} X_n(\omega_1) dP(\omega_1) - \int_{\Omega} X_n(\omega_2) dP(\omega_2) = E(X_n) - E(X_n) = 0. \end{aligned}$$

Also  $|Z_n| \leq 2c$ .

Furthermore,  $\sum Z_n$  converges almost surely, because

$$\begin{aligned} &\left\{ (\omega_1, \omega_2) \in \Omega \times \Omega; \sum Z_n \text{ diverges} \right\} \\ &\subset \left\{ (\omega_1, \omega_2); \sum X_n(\omega_1) \text{ diverges} \right\} \cup \left\{ (\omega_1, \omega_2); \sum_n X_n(\omega_2) \text{ diverges} \right\}, \end{aligned}$$

and both sets in the union have  $P \times P$ -measure zero (by our a.s. convergence hypothesis on  $\sum X_n$ ).

By Theorem I.8.3, it follows that  $\sum \sigma^2(Z_n) < \infty$ . However,

$$\sigma^2(Z_n) = \int_{\Omega \times \Omega} [X_n(\omega_1) - X_n(\omega_2)]^2 d(P \times P).$$

Expanding the square and integrating, we see that

$$\sigma^2(Z_n) = 2[E(X_n^2) - E(X_n)^2] = 2\sigma^2(X_n).$$

Therefore  $\sum \sigma^2(X_n) < \infty$ , and since  $Y_n$  are central, and  $\sum \sigma^2(Y_n) = \sum \sigma^2(X_n) < \infty$ , we conclude from Theorem I.8.2 that  $\sum Y_n$  converges a.s.; but then  $\sum E(X_n) = \sum (X_n - Y_n)$  converges as well.  $\square$

We consider finally the general case of a series of independent  $L^2$  random variables.

**Theorem I.8.5 (Kolmogorov's 'three series theorem').** *Let  $\{X_n\}$  be a sequence of real independent  $L^2(P)$  random variables. For any real  $k > 0$  and  $n \in \mathbb{N}$ , denote*

$$E_n := [|X_n| \leq k]; \quad X'_n = I_{E_n} X_n.$$

Then the series  $\sum X_n$  converges a.s. if and only if the following numerical series (a), (b), and (c) converge

- (a)  $\sum_n P(E_n^c)$ ;
- (b)  $\sum_n E(X'_n)$ ;
- (c)  $\sum_n \sigma^2(X'_n)$ .

**Proof.** Consider the 'truncated' r.v.s

$$Y_n^+ := I_{E_n} X_n + k I_{E_n^c},$$

and  $Y_n^-$  defined similarly with  $-k$  instead of  $k$ .

If  $\sum X_n(\omega)$  converges, then  $X_n(\omega) \rightarrow 0$ , so that  $\omega \in E_n$  for  $n > n_0$ , and therefore  $Y_n^+(\omega) = Y_n^-(\omega) = X_n(\omega)$  for all  $n > n_0$ ; hence both series  $\sum Y_n^+(\omega)$  and  $\sum Y_n^-(\omega)$  converge. Conversely, if one of these two series converge (say the first), then  $Y_n^+(\omega) \rightarrow 0$ , so that  $Y_n^+(\omega) \neq k$  for  $n > n_0$ , hence necessarily  $Y_n^+(\omega) = X_n(\omega)$  for  $n > n_0$ , and so  $\sum X_n(\omega)$  converges.

We showed that  $\sum X_n(\omega)$  converges if and only if the series of  $Y_n^+$  and  $Y_n^-$  both converge at  $\omega$ ; therefore  $\sum X_n$  converges a.s. if and only if both  $Y$ -series converge a.s. Since  $Y_n^+$  and  $Y_n^-$  satisfy the hypothesis of Theorem I.8.4, we conclude from that theorem that  $\sum X_n$  converges a.s. if and only if the numerical series  $\sum E(Y_n^+)$ ,  $\sum \sigma^2(Y_n^+)$ , and the corresponding series for  $Y_n^-$  converge. It then remains to show that the convergence of these four series is equivalent to the convergence of the three series (a)–(c).

Since

$$E(Y_n^+) = E(X'_n) + kP(E_n^c)$$

(and a similar formula for  $Y_n^-$ ), we see by addition and subtraction that the convergence of the series (a) and (b) is equivalent to the convergence of the two series

$$\sum E(Y_n^+), \quad \sum E(Y_n^-).$$

Next

$$\begin{aligned} \sigma^2(Y_n^+) &= E\{(Y_n^+)^2\} - \{E(Y_n^+)\}^2 \\ &= E\{(X'_n)^2\} + k^2 P(E_n^c) - [E(X'_n) + kP(E_n^c)]^2 \\ &= \sigma^2(X'_n) + k^2 P(E_n)P(E_n^c) - 2kE(X'_n)P(E_n^c), \end{aligned} \quad (2)$$

and similarly for  $Y_n^-$  (with  $-k$  replacing  $k$ ).

If the series (a)–(c) converge, then we already know that  $\sum E(Y_n^+)$  and  $\sum E(Y_n^-)$  converge. The convergence of (b) also implies that  $|E(X'_n)| \leq M$  for all  $n$ , and therefore

$$|E(X'_n)P(E_n^c)| \leq MP(E_n^c),$$

so that  $\sum E(X'_n)P(E_n^c)$  converges (by convergence of (a)).

Since  $0 \leq P(E_n)P(E_n^c) \leq P(E_n^c)$ , the series  $\sum P(E_n)P(E_n^c)$  converges (by convergence of (a)).

Relation (2) and the convergence of (c) imply therefore that  $\sum \sigma^2(Y_n^+)$  converges (and similarly for  $Y_n^-$ ), as wanted.

Conversely, if the ‘four series’ above converge, we saw already that the series (a) and (b) converge, and this in turn implies the convergence of the series

$$\sum E(X'_n)P(E_n^c) \quad \text{and} \quad \sum P(E_n)P(E_n^c).$$

By Relation (2), the convergence of  $\sum \sigma^2(Y_n^+)$  implies therefore the convergence of the series (c) as well.  $\square$

## I.9 Infinite divisibility

**Definition I.9.1.** A random variable  $X$  (or its distribution function  $F$ , or its characteristic function  $f$ ) is *infinitely divisible* (i.d.) if, for each  $n \in \mathbb{N}$ ,  $F$  is the distribution function of a sum of  $n$  independent r.v.s with the same distribution function  $F_n$ .

Equivalently,  $X$  is i.d. if there exists a *triangular array* of r.v.s

$$\{X_{nk}; 1 \leq k \leq n, n = 1, 2, \dots\} \quad (1)$$

such that, for each  $n$ , the r.v.s of the  $n$ th row are independent and equidistributed, and their sum  $T_n =_d X$  (if  $X, Y$  are r.v.s, we write  $X =_d Y$  when they have the same distribution).

By the Uniqueness theorem for ch.f.s,  $X$  is i.d. if and only if, for each  $n$ , there exists a ch.f.  $f_n$  such that

$$f = f_n^n. \quad (2)$$

In terms of distribution functions, infinite divisibility of  $F$  means the existence, for each  $n$ , of a distribution function  $F_n$ , such that

$$F = F_n^{(n)}, \quad (3)$$

where  $G^{(n)} := G * \dots * G$  ( $n$  times) for any distribution function  $G$ . The *convolution*  $F * G$  of two distribution functions is defined by

$$(F * G)(x) := \int_{\mathbb{R}} F(x - y) dG(y) \quad (x \in \mathbb{R}).$$

It is clearly a distribution function. An application of Fubini's theorem shows that its ch.f. is precisely  $fg$  (where  $f, g$  are the ch.f.s of  $F, G$ , respectively). It then follows from the uniqueness theorem for ch.f.s (or directly!) that convolution of distribution functions is commutative and associative. We may then omit parenthesis and write  $F_1 * F_2 * \dots * F_n$  for the convolution of finitely many distribution functions. In particular, the repeated convolutions  $G^{(n)}$  mentioned above make sense, and criterion (3) is clearly equivalent to (2).

**Example I.9.2.**

- (1) The Poisson distribution is i.d.: take  $F_n$  to be Poisson with parameter  $\lambda/n$  (where  $\lambda$  is the parameter of  $F$ ). Then indeed (cf. Section I.3.9):

$$f_n^n(u) = [e^{(\lambda/n)(e^{iu}-1)}]^n = f(u).$$

- (2) The normal distribution (parameters  $\mu, \sigma^2$ ) is i.d.: take  $F_n$  to be the normal distribution with parameters  $\mu/n, \sigma^2/n$ . Then (cf. Section I.3.12)

$$f_n^n(u) = [e^{iu\mu/n} e^{-(\sigma u)^2/2n}]^n = f(u).$$

- (3) The Gamma distribution (parameters  $p, b$ ) is i.d.: take  $F_n$  to be the *Gamma* distribution with parameters  $p/n, b$ . Then (cf. Section I.3.15):

$$f_n^n(u) = \left[ \left( 1 - \frac{iu}{b} \right)^{-p/n} \right]^n = f(u).$$

It is also clear that the Cauchy distribution is i.d. (cf. Section I.3.14), while the Laplace distribution is not.

We have the following criterion for infinite divisibility (its necessity is obvious, by (1); we omit the proof of its sufficiency):

**Theorem I.9.3.** *A random variable  $X$  is i.d. if and only if there exists a triangular array (1) such that (cf. Definition I.4.3)*

$$T_n \rightarrow_w X. \quad (4)$$

Some elementary properties of i.d. random variables (or ch.f.s) are stated in the next theorem.

**Theorem I.9.4.**

- (a) If  $X$  is i.d., so is  $Y := a + bX$ .
- (b) If  $f, g$  are i.d. characteristic functions, so is  $fg$ .
- (c) If  $f$  is an i.d. characteristic function, so are  $\bar{f}$  and  $|f|^2$ .
- (d) If  $f$  is an i.d. characteristic function, then  $f \neq 0$  everywhere.
- (e) If  $\{f_k\}$  is a sequence of i.d. characteristic functions converging pointwise to a function  $g$  continuous at 0, then  $g$  is an i.d. ch.f.
- (f) If  $f$  is an i.d. characteristic function, then its representation (2) is unique (for each  $n$ ).

**Proof.**

- (a) By Proposition I.3.5 and (2), for all  $n \in \mathbb{N}$ ,

$$f_Y(u) = e^{iua} f(bu) = [e^{iua/n} f_n(bu)]^n = g_n^n,$$

where  $g_n$  is clearly a ch.f.



- (b) Represent  $f, g$  as in (2), for each  $n$ . Then  $fg = [f_n g_n]^n$ . By Corollary I.2.4,  $fg$  and  $f_n g_n$  are ch.f.s, and they satisfy (2) as needed.
- (c) First,  $\bar{f}$  is a ch.f., since

$$\begin{aligned}\bar{f}(u) &= \overline{\int_{\mathbb{R}} e^{iux} dF(x)} = \int_{\mathbb{R}} e^{-iux} dF(x) \\ &= \int_{\mathbb{R}} e^{iux} d[1 - F(-x)] = \int_{\mathbb{R}} e^{iux} dF_{-X}(x) = f_{-X}(u).\end{aligned}$$

If  $f$  is i.d., then by (2),  $\bar{f} = (\overline{f_n})^n$ , where  $\overline{f_n}$  is a ch.f., as needed. The conclusion about  $|f|^2$  follows then from (b).

- (d) Since it suffices to prove that  $|f|^2 \neq 0$ , and  $|f|^2$  is a non-negative i.d. ch.f. (by (c)), we may assume without loss of generality that  $f \geq 0$ . Let then  $f_n := f^{1/n}$  be the unique non-negative  $n$ th root of  $f$ . Then  $g := \lim f_n$  (pointwise) exists:

$$g = I_E, \quad E = f^{-1}(\mathbb{R}^+).$$

Since  $f(0) = 1$ , the point 0 belongs to the open set  $E$ , and so  $g = 1$  in a neighbourhood of 0; in particular,  $g$  is continuous at 0. By the Paul Levy Continuity theorem (Theorem I.4.8),  $g$  is a ch.f., and is therefore continuous everywhere. In particular, its range is connected; since it is a subset of  $\{0, 1\}$  containing 1, it must be precisely  $\{1\}$ , that is,  $g = 1$  everywhere. This means that  $f > 0$  everywhere, as claimed.

- (e) By the Paul Levy Continuity theorem,  $g$  is a ch.f. By (d),  $f_k \neq 0$  everywhere (for each  $k$ ), and has therefore a continuous logarithm  $\log f_k$ , uniquely determined by the condition  $\log f_k(0) = 0$ . Since  $f_k$  is i.d.,  $f_k = f_{k,n}^n$ , with  $f_{k,n}$  ch.f.s (by (2)). We have

$$e^{(1/n) \log f_k} = e^{(1/n) \log f_{k,n}^n} = f_{k,n}.$$

The left-hand side converges pointwise (as  $k \rightarrow \infty$ ) to  $e^{(1/n) \log g} := g_n$ . Since  $g$  is a ch.f.,  $g(0) = 1$ , and therefore  $\log g$  is continuous at 0, and the same is true of  $g_n (= \lim_k f_{k,n})$ . By Paul Levy's theorem,  $g_n$  is a ch.f., and clearly  $g_n^n = g$ . Hence  $g$  is i.d.

- (f) Fix  $n$ , and suppose  $g, h$  are ch.f.s such that

$$g^n = h^n = f. \quad (*)$$

By (d),  $h \neq 0$  everywhere, and therefore  $g/h$  is continuous, and  $(g/h)^n = 1$  everywhere. The continuity implies that  $g/h$  has a connected range, which is a subset of the finite set of  $n$ th roots of unity. Since  $g(0) = h(0) = 1$  (these are ch.f.s!), the range contains 1, and coincides therefore with the singleton  $\{1\}$ , that is,  $g/h = 1$  identically.  $\square$

By Example I.9.2(1) and Theorem I.9.4 (Part (a)), if  $Y = a + bX$  with  $X$  Poisson-distributed, then  $Y$  is i.d. We call the distribution  $F_Y$  of such an r.v.  $Y$  a *Poisson-type distribution*. By Proposition I.3.5 and Section I.3.9,

$$f_Y(u) = e^{iua + \lambda(e^{iub} - 1)}. \quad (5)$$

The Poisson-type distributions ‘generate’ all the i.d. distributions in the following sense (compare with Theorem I.9.3):

**Theorem I.9.5.** *A random variable  $X$  is infinitely divisible if and only if there exists an array*

$$\{X_{nk}; 1 \leq k \leq r(n), n = 1, 2, \dots\}$$

*such that, for each  $n$ , the r.v.s in the  $n$ th row are independent Poisson-type, and*

$$T_n := \sum_{k=1}^{r(n)} X_{nk} \rightarrow_w X.$$

**Proof.** *Sufficiency.* As we just observed, each  $X_{nk}$  is i.d., hence  $T_n$  are i.d. by Theorem I.9.4, Part (b), and therefore  $X$  is i.d. by Part (e) of Theorem I.9.4 and Corollary I.4.6.

*Necessity.* Let  $X$  be i.d. By (2), there exist ch.f.s  $f_n$  such that  $f := f_X = f_n^n$ ,  $n = 1, 2, \dots$ . By Theorem I.9.4, Part (f) and the proof of Part (e), the  $f_n$  are uniquely determined, and can be written as

$$f_n = e^{(1/n) \log f},$$

where  $\log f$  is continuous and uniquely determined by the condition  $\log f(0) = 0$ .

Fix  $u \in \mathbb{R}$ . Then

$$\begin{aligned} n[f_n(u) - 1] &= n[e^{(1/n) \log f(u)} - 1] = n[(1/n) \log f(u) + o(1/n)] \\ &\rightarrow_{n \rightarrow \infty} \log f(u), \end{aligned}$$

that is, if  $F_n$  denotes the distribution function with ch.f.  $f_n$ , then

$$\log f(u) = \lim_n n \int_{\mathbb{R}} (e^{iux} - 1) dF_n(x). \quad (6)$$

For each  $n$ , let  $m = m(n)$  be such that

$$1 - F_n(m) + F_n(-m) < \frac{1}{2n^2}.$$

Then for all  $u \in \mathbb{R}$ ,

$$\left| \int_{\mathbb{R}} (e^{iux} - 1) dF_n(x) - \int_{-m(n)}^{m(n)} (e^{iux} - 1) dF_n(x) \right| < \frac{1}{n^2}.$$

Approximate the Stieltjes integral over  $[-m(n), m(n)]$  by Riemann–Stieltjes sums, such that

$$\left| \int_{-m(n)}^{m(n)} (\cdots) - \sum_{k=1}^{r(n)} (e^{iu x_k} - 1) [F_n(x_k) - F_n(x_{k-1})] \right| < \frac{1}{n^2},$$

where  $x_k = x_k(n) = -m(n) + 2m(n)k/r(n)$ ,  $k = 1, \dots, r(n)$ . By (6),  $f$  is the pointwise limit (as  $n \rightarrow \infty$ ) of the products

$$\prod_{k=1}^{r(n)} \exp\{\lambda_{nk}(e^{ia_{nk}u} - 1)\}, \quad (7)$$

where  $\lambda_{nk} := n[F_n(x_k) - F_n(x_{k-1})]$  and  $a_{nk} := x_k(= x_k(n))$ .

The products in (7) are the ch.f.s of sums of  $r(n)$  independent Poisson-type r.v.s.  $\square$

## I.10 More on sequences of random variables

Let  $\{X_n\}$  be a sequence of (complex) random variables on the probability space  $(\Omega, \mathcal{A}, P)$ . For  $c > 0$ , set

$$m(c) := \sup_n \int_{||X_n| \geq c} |X_n| dP.$$

**Definition I.10.1.**  $\{X_n\}$  is *uniformly integrable* (u.i.) if

$$\lim_{c \rightarrow \infty} m(c) = 0.$$

For example, if  $|X_n| \leq g \in L^1(P)$  for all  $n$ , then  $[|X_n| \geq c] \subset [g \geq c]$ , so that

$$m(c) \leq \int_{[g \geq c]} g dP \rightarrow 0$$

as  $c \rightarrow \infty$ .

A less trivial example is given in the following

**Proposition.** Let  $\mathcal{A}_n$  be sub- $\sigma$ -algebras of  $\mathcal{A}$ , let  $Y \in L^1(P)$ , and

$$X_n = E(Y|\mathcal{A}_n) \quad n = 1, 2, \dots$$

Then  $X_n$  are u.i.

**Proof.** Since  $|X_n| \leq E(|Y||\mathcal{A}_n)$ , and  $X_n$  are  $\mathcal{A}_n$ -measurable (so that  $A_n := [|X_n| \geq c] \in \mathcal{A}_n$ ),

$$\int_{A_n} |X_n| dP \leq \int_{A_n} E(|Y||\mathcal{A}_n) dP = \int_{A_n} |Y| dP.$$

We have

$$P(A_n) \leq E(|X_n|)/c \leq E(|Y|)/c.$$

Given  $\epsilon > 0$ , choose  $K > 0$  such that  $\int_{[|Y|>K]} |Y| dP < \epsilon$ . Then

$$\begin{aligned} \int_{A_n} |Y| dP &= \left( \int_{A_n \cap [|Y| \leq K]} + \int_{A_n \cap [|Y| > K]} \right) |Y| dP \\ &\leq KP(A_n) + \epsilon \leq K\|Y\|_1/c + \epsilon, \end{aligned}$$

hence

$$m(c) \leq K\|Y\|_1/c + \epsilon.$$

Since  $\epsilon$  was arbitrary, it follows that  $\lim_{c \rightarrow \infty} m(c) = 0$ . □

**Theorem I.10.2.**  $\{X_n\}$  is u.i. iff

- (1)  $\sup_n \|X_n\|_1 < \infty$  ('norm-boundedness') and
- (2)  $\sup_n \int_A |X_n| dP \rightarrow 0$  when  $PA \rightarrow 0$  ('uniform absolute continuity').

**Proof.** If the sequence is u.i., let

$$M := \sup_{c>0} m(c) \quad (< \infty).$$

Then for any  $c > 0$ ,

$$\|X_n\|_1 = \left( \int_{[|X_n| < c]} + \int_{[|X_n| \geq c]} \right) |X_n| dP \leq c + m(c) \leq c + M,$$

so that (1) is valid.

Also, for any  $A \in \mathcal{A}$ ,

$$\int_A |X_n| dP = \left( \int_{A \cap [|X_n| < c]} + \int_{A \cap [|X_n| \geq c]} \right) |X_n| dP \leq cPA + m(c).$$

Given  $\epsilon > 0$ , fix  $c > 0$  such that  $m(c) < \epsilon/2$ . For this  $c$ , if  $PA < \epsilon/2c$ , then  $\int_A |X_n| dP < \epsilon$  for all  $n$ , proving (2).

Conversely, if (1) and (2) hold, then by (1), for all  $n$

$$cP[|X_n| \geq c] \leq \int_{[|X_n| \geq c]} |X_n| dP \leq \sup_n \|X_n\|_1 := R < \infty,$$

so that

$$P[|X_n| \geq c] \leq R/c.$$

Given  $\epsilon > 0$ , there exists  $\delta > 0$  such that  $\int_A |X_n| dP < \epsilon$  for all  $n$  whenever  $A \in \mathcal{A}$  has  $PA < \delta$  (by (2)). Therefore, if  $c > R/\delta$ , surely  $P[|X_n| \geq c] < \delta$ , and consequently  $m(c) < \epsilon$ . □

Fatou's lemma extends as follows to u.i. real r.v.s:

**Theorem I.10.3.** *Let  $X_n$  be u.i. real r.v.s. Then*

$$E(\liminf X_n) \leq \liminf E(X_n) \leq \limsup E(X_n) \leq E(\limsup X_n).$$

**Proof.** Since  $[X_n < -c] \subset [|X_n| > c]$ , we have

$$\left| \int_{[X_n < -c]} X_n dP \right| \leq m(c)$$

for any  $c > 0$ . Given  $\epsilon > 0$ , we may fix  $c > 0$  such that  $m(c) < \epsilon$  (since  $X_n$  are u.i.).

Denote  $A_n = [X_n \geq -c]$ .

We apply Fatou's Lemma to the *non-negative* measurable functions  $c + X_n I_{A_n}$ :

$$E(c + \liminf X_n I_{A_n}) \leq \liminf E(c + X_n I_{A_n}),$$

hence

$$E(\liminf X_n I_{A_n}) \leq \liminf E(X_n I_{A_n}). \quad (*)$$

However

$$E(X_n I_{A_n}) = E(X_n) - \int_{A_n^c} X_n dP < E(X_n) + \epsilon,$$

and therefore the right-hand side of (\*) is

$$\leq \liminf E(X_n) + \epsilon.$$

Since  $X_n I_{A_n} \geq X_n$ , the left-hand side of (\*) is  $\geq E(\liminf X_n)$ , and the left inequality of the theorem follows. The right inequality is then obtained by replacing  $X_n$  by  $-X_n$ .  $\square$

**Corollary I.10.4.** *Let  $X_n$  be u.i. (complex) r.v.s, such that  $X_n \rightarrow X$  a.s. or in probability. Then  $X_n \rightarrow X$  in  $L^1$  (and in particular,  $E(X_n) \rightarrow E(X)$ ).*

**Proof.** Let  $K := \sup_n \|X_n\|_1 < \infty$ , by Theorem I.10.2(1)). By Fatou's Lemma, if  $X_n \rightarrow X$  a.s., then

$$\|X\|_1 \leq \liminf_m \|X_m\|_1 \leq K,$$

so  $X \in L^1$  and

$$\sup_n \|X_n - X\|_1 \leq 2K.$$

Also, for  $A \in \mathcal{A}$ , again by Fatou's Lemma,

$$\int_A |X_n - X| dP \leq \liminf_m \int_A |X_n - X_m| dP,$$

hence

$$\sup_n \int_A |X_n - X| dP \leq 2 \sup_n \int_A |X_n| dP \rightarrow 0$$

as  $PA \rightarrow 0$ , by Theorem I.10.2(2).

Consequently (by the same theorem)  $|X_n - X|$  are u.i., and since  $|X_n - X| \rightarrow 0$  a.s., Theorem I.10.3 applied to  $|X_n - X|$  shows that  $\|X_n - X\|_1 \rightarrow 0$ .

In case  $X_n \rightarrow X$  in probability, there exists a subsequence  $X_{n_k}$  converging a.s. to  $X$ . By the first part of the proof,  $X \in L^1$ , and  $\|X_{n_k} - X\|_1 \rightarrow 0$ . The previous argument with  $m = n_k$  shows that  $|X_n - X|$  are u.i. Therefore *any* subsequence of  $X_n$  has a subsequence converging to  $X$  in  $L^1$ . If we assume that  $\{X_n\}$  itself does not converge to  $X$  in  $L^1$ , then given  $\epsilon > 0$ , there exists a subsequence  $X_{n_k}$  such that  $\|X_{n_k} - X\|_1 \geq \epsilon$  for all  $k$ , a contradiction.  $\square$

**Definition I.10.5.** A *submartingale* is a sequence of ordered pairs  $(X_n, \mathcal{A}_n)$ , where

- (1)  $\{\mathcal{A}_n\}$  is an increasing sequence of sub- $\sigma$ -algebras of  $\mathcal{A}$ ;
- (2) the real r.v.  $X_n \in L^1$  is  $\mathcal{A}_n$ -measurable ( $n = 1, 2, \dots$ ); and
- (3)  $X_n \leq E(X_{n+1}|\mathcal{A}_n)$  a.s. ( $n = 1, 2, \dots$ ).

If equality holds in (3), the sequence is called a *martingale*. The sequence is a *supermartingale* if the inequality (3) is reversed.

By definition of conditional expectation, a sequence  $(X_n, \mathcal{A}_n)$  satisfying (1) and (2) is a submartingale iff

$$\int_A X_n dP \leq \int_A X_{n+1} dP \quad (A \in \mathcal{A}_n).$$

For example, if  $\mathcal{A}_n$  are as in (1) and  $Y$  is an  $L^1$ -r.v., then setting

$$X_n := E(Y|\mathcal{A}_n), \quad n = 1, 2, \dots,$$

the sequence  $(X_n, \mathcal{A}_n)$  is a martingale: indeed (2) is clear by definition, and equality in (3) follows from Theorem I.7.5.

An important example is given in the following:

**Proposition I.10.6.** Let  $\{Y_n\}$  be a sequence of  $L^1$ , central, independent r.v.s. Let  $\mathcal{A}_n$  be the smallest  $\sigma$ -algebra for which  $Y_1, \dots, Y_n$  are measurable, and let  $X_n = Y_1 + \dots + Y_n$ . Then  $(X_n, \mathcal{A}_n)$  is a martingale.

**Proof.** The requirements (1) and (2) are clear.

If  $Y_{n+1} = I_B$  with  $B \in \mathcal{A}$  independent of all  $A \in \mathcal{A}_n$ , then for all  $A \in \mathcal{A}_n$ ,

$$\int_A Y_{n+1} dP = P(A \cap B) = P(A)P(B) = \int_A E(Y_{n+1}) dP,$$

and this identity between the extreme terms remains true, by linearity, for all simple r.v.s  $Y_{n+1}$  independent of  $Y_1, \dots, Y_n$ . By monotone convergence, the identity is true for all  $Y_{n+1} \geq 0$ , and finally for all  $L^1$ -r.v.s  $Y_{n+1}$  independent of  $Y_1, \dots, Y_n$ . Therefore

$$E(Y_{n+1}|\mathcal{A}_n) = E(Y_{n+1}) \quad \text{a.s.,}$$

and in the central case,

$$E(Y_{n+1}|\mathcal{A}_n) = 0 \quad \text{a.s.}$$

Since  $X_n$  is  $\mathcal{A}_n$ -measurable,

$$E(X_n|\mathcal{A}_n) = X_n \quad \text{a.s.}$$

Adding these equations, we obtain  $E(X_{n+1}|\mathcal{A}_n) = X_n$  a.s.  $\square$

**Proposition I.10.7.** *If  $(X_n, \mathcal{A}_n)$  is a submartingale, and  $g: \mathbb{R} \rightarrow \mathbb{R}$  is a convex, increasing function such that  $g(X_n) \in L^1$  for all  $n$ , then  $(g(X_n), \mathcal{A}_n)$  is a submartingale. If  $(X_n, \mathcal{A}_n)$  is a martingale, the preceding conclusion is valid without assuming that  $g$  is increasing.*

**Proof.** Since  $g$  is increasing, and  $X_n \leq E(X_{n+1}|\mathcal{A}_n)$  a.s., we have a.s.

$$g(X_n) \leq g(E(X_{n+1}|\mathcal{A}_n)).$$

By Jensen's inequality for the convex function  $g$ , the right-hand side is  $\leq E(g(X_{n+1})|\mathcal{A}_n)$  a.s., proving the first statement. In the martingale case, since  $X_n = E(X_{n+1}|\mathcal{A}_n)$ , we get a.s.

$$g(X_n) = g(E(X_{n+1}|\mathcal{A}_n)) \leq E(g(X_{n+1})|\mathcal{A}_n). \quad \square$$

We omit the proof of the following important theorem:

**Theorem I.10.8 (Submartingale convergence theorem).** *If  $(X_n, \mathcal{A}_n)$  is a submartingale such that*

$$\sup_n E(X_n^+) < \infty,$$

*then there exists an  $L^1$ -r.v.  $X$  such that  $X_n \rightarrow X$  a.s.*

By the proposition following Definition I.10.1 and the comments following Definition I.10.5, if  $\mathcal{A}_n$  are increasing sub- $\sigma$ -algebras,  $Y$  is an  $L^1$ -r.v., and  $X_n := E(Y|\mathcal{A}_n)$ , then  $(X_n, \mathcal{A}_n)$  is a u.i. martingale. The converse is also true:

**Theorem I.10.9.**  *$(X_n, \mathcal{A}_n)$  is a u.i. martingale iff there exists an  $L^1$ -r.v.  $Y$  such that  $X_n = E(Y|\mathcal{A}_n)$  (a.s.) for all  $n$ . When this is the case,  $X_n \rightarrow Y = E(Y|\mathcal{A}_\infty)$  a.s. and in  $L^1$ , where  $\mathcal{A}_\infty$  is the  $\sigma$ -algebra generated by the algebra  $\bigcup_n \mathcal{A}_n$ .*

**Proof.** We just observed that if  $X_n = E(Y|\mathcal{A}_n)$ , then  $(X_n, \mathcal{A}_n)$  is a u.i. martingale. Conversely, let  $(X_n, \mathcal{A}_n)$  be a u.i. martingale. By Theorem I.10.2 (1),  $\sup_n \|X_n\|_1 < \infty$ , hence by Theorem I.10.8, there exists an  $L^1$ -r.v.  $Y$  such that  $X_n \rightarrow Y$  a.s. By Corollary I.10.4,  $X_n \rightarrow Y$  in  $L^1$  as well. Hence, for all  $A \in \mathcal{A}_n$  and  $m \geq n$ ,

$$\int_A X_n dP = \int_A X_m dP \rightarrow_m \int_A Y dP = \int_A E(Y|\mathcal{A}_n) dP$$

and it follows that  $X_n = E(Y|\mathcal{A}_n)$  a.s.

For any Borel set  $B \subset \mathbb{R}$ , we have  $X_n^{-1}(B) \in \mathcal{A}_n \subset \mathcal{A}_\infty$ . Hence  $X_n$  is  $\mathcal{A}_\infty$ -measurable for all  $n$ . If we give  $Y$  some arbitrary value on the null set where it is not determined,  $Y$  is  $\mathcal{A}_\infty$  measurable, and therefore  $E(Y|\mathcal{A}_\infty) = Y$ .  $\square$

# Application II

## Distributions

### II.1 Preliminaries

**II.1.1.** Let  $\Omega$  be an open subset of  $\mathbb{R}^n$ . For  $0 \leq k < \infty$ , let  $C^k(\Omega)$  denote the space of all complex functions on  $\Omega$  with continuous (mixed) partial derivatives of order  $\leq k$ . The intersection of all these spaces is denoted by  $C^\infty(\Omega)$ ; it is the space of all (complex) functions with continuous partial derivatives of *all* orders in  $\Omega$ . For  $0 \leq k \leq \infty$ ,  $C_c^k(\Omega)$  stands for the space of all  $f \in C^k(\Omega)$  with compact support (in  $\Omega$ ). We shall also use the notation  $C_c^k(\Delta)$  for the space of all functions in  $C^k(\mathbb{R}^n)$  with compact support in the *arbitrary* set  $\Delta \subset \mathbb{R}^n$ . (The latter notation is consistent with the preceding one when  $\Delta$  is *open*, since in that case any  $C^k$ -function in  $\Delta$  with compact support in  $\Delta$  extends trivially to a  $C^k$ -function on  $\mathbb{R}^n$ .)

Define  $f : \mathbb{R} \rightarrow [0, \infty)$  by  $f(t) = 0$  for  $t \geq 0$ , and  $f(t) = e^{1/t}$  for  $t < 0$ . Then  $f \in C^\infty(\mathbb{R})$ , and therefore, with a suitable choice of the constant  $\gamma$ , the function  $\phi(x) := \gamma f(|x|^2 - 1)$  on  $\mathbb{R}^n$  has the following properties:

- (1)  $\phi \in C_c^\infty(\mathbb{R}^n)$ ;
- (2)  $\text{supp } \phi = \{x \in \mathbb{R}^n; |x| \leq 1\}$ ;
- (3)  $\phi \geq 0$  and  $\int \phi \, dx = 1$ .

(For  $x \in \mathbb{R}^n$ ,  $|x|$  denotes the usual Euclidean norm;  $\int \cdot \, dx$  denotes integration over  $\mathbb{R}^n$  with respect to the  $n$ -dimensional Lebesgue measure  $dx$ .)

In the following,  $\phi$  will denote *any* fixed function with Properties (1)–(3).

If  $u : \mathbb{R}^n \rightarrow \mathbb{C}$  is locally integrable, then for any  $r > 0$ , we consider its *regularization*

$$u_r(x) := \int u(x - ry)\phi(y) \, dy = r^{-n} \int u(y)\phi\left(\frac{x-y}{r}\right) \, dy, \quad (1)$$

(i.e.  $u_r$  is the convolution of  $u$  with  $\phi_r := r^{-n}\phi(\cdot/r)$ ; note that the subscript  $r$  has different meanings when assigned to  $u$  and to  $\phi$ .)



**Theorem II.1.2.** *Let  $K$  be a compact subset of (the open set)  $\Omega \subset \mathbb{R}^n$ , and let  $u \in L^1(\mathbb{R}^n)$  vanish outside  $K$ . Then*

- (1)  $u_r \in C_c^\infty(\Omega)$  for all  $r < \delta := \text{dist}(K, \Omega^c)$ ;
- (2)  $\lim_{r \rightarrow 0} u_r = u$  in  $L^p$ -norm if  $u \in L^p$  ( $1 \leq p < \infty$ ), and uniformly if  $u$  is continuous.

**Proof.** Since  $u_r = u * \phi_r$ , one sees easily that (mixed) differentiation of any order can be performed on  $u_r$  by performing the operation on  $\phi_r$ ; the resulting convolution is clearly continuous. Hence  $u_r \in C^\infty(\Omega)$ .

Let

$$K_r := \{x \in \mathbb{R}^n; \text{dist}(x, K) \leq r\}. \quad (2)$$

It is a compact set, contained in  $\Omega$  for  $r < \delta$ . If  $y$  is in the closed unit ball  $S$  of  $\mathbb{R}^n$  and  $x - ry \in K$ , then  $\text{dist}(x, K) \leq |x - (x - ry)| = |ry| \leq r$ , that is,  $x \in K_r$ . Hence for  $x \notin K_r$ ,  $x - ry \notin K$  for all  $y \in S$ . Since  $u$  and  $\phi$  vanish outside  $K$  and  $S$  respectively, it follows from (1) that  $u_r = 0$  outside  $K_r$ . Therefore,  $\text{supp } u_r \subset K_r \subset \Omega$  (and so  $u_r \in C_c^\infty(\Omega)$ ) for  $r < \delta$ .

By Property (3) of  $\phi$ ,

$$u_r(x) - u(x) = \int_S [u(x - ry) - u(x)] \phi(y) dy.$$

If  $u$  is continuous (hence uniformly continuous, since its support is in  $K$ ), then for given  $\epsilon > 0$ , there exists  $\eta > 0$  such that  $|u(x - ry) - u(x)| < \epsilon$  for all  $x \in \mathbb{R}^n$  and all  $y \in S$  if  $r < \eta$ . Hence  $\|u_r - u\|_\infty \leq \epsilon$  for  $r < \eta$ .

In case  $u \in L^p$ , we have

$$\|u_r\|_p = \|u * \phi_r\|_p \leq \|u\|_p \|\phi_r\|_1 = \|u\|_p.$$

Fix  $v \in C_c(\Omega)$  such that  $\|u - v\|_p < \epsilon$  (by density of  $C_c(K)$  in  $L^p(K)$ ). Let  $M$  be a bound for the (Lebesgue) measure of  $(\text{supp } v)_r$  for all  $r < 1$ . Then for  $r < 1$ ,

$$\|u_r - u\|_p \leq \|(u - v)_r\|_p + \|v_r - v\|_p + \|v - u\|_p < 2\epsilon + \|v_r - v\|_\infty M^{1/p} < 3\epsilon$$

(by the preceding case), for  $r$  small enough.  $\square$

The inequality

$$\|f * g\|_p \leq \|f\|_p \|g\|_1 \quad (f \in L^p; g \in L^1) \quad (3)$$

used in the above proof, can be verified as follows:

$$\begin{aligned} \|f * g\|_p &= \sup \left\{ \left| \int (f * g) h dx \right|; h \in L^q, \|h\|_q \leq 1 \right\} \\ &\leq \sup_h \iint |f(x - y)| |g(y)| dy |h(x)| dx \\ &= \sup_h \iint |f(x - y)| |h(x)| dx |g(y)| dy \\ &\leq \sup_h \int \|f(\cdot - y)\|_p \|h\|_q |g(y)| dy = \sup_h \|f\|_p \|h\|_q \|g\|_1 = \|f\|_p \|g\|_1, \end{aligned}$$

where the suprema are taken over all  $h$  in the unit ball of  $L^q$ . (We used Theorems 4.6, 2.18, and 1.33, and the translation invariance of Lebesgue measure.)

### Corollary II.1.3.

- (1)  $C_c^\infty(\Omega)$  is dense in  $L^p(\Omega)$  ( $1 \leq p < \infty$ ).
- (2) A regular complex Borel measure  $\mu$  on  $\Omega$  is uniquely determined by the integrals  $\int_\Omega f d\mu$  with  $f \in C_c^\infty(\Omega)$ .

**Corollary II.1.4.** Let  $K$  be a compact subset of the open set  $\Omega \subset \mathbb{R}^n$ . Then there exists  $\psi \in C_c^\infty(\Omega)$  such that  $0 \leq \psi \leq 1$  and  $\psi = 1$  in a neighbourhood of  $K$ .

**Proof.** Let  $\delta$  be as in Theorem II.1.2,  $r < \delta/3$ , and  $\psi = u_r$ , where  $u$  is the indicator of  $K_{2r}$  (cf. (2)). Then  $\text{supp } \psi \subset (K_{2r})_r = K_{3r} \subset \Omega$ ,  $\psi \in C_c^\infty(\Omega)$ ,  $0 \leq \psi \leq 1$ , and  $\psi = 1$  on  $K_r$ .  $\square$

The last corollary is the special case  $k = 1$  of the following.

**Theorem II.1.5 (Partitions of unity in  $C_c^\infty(\Omega)$ ).** Let  $\Omega_1, \dots, \Omega_k$  be an open covering of the compact set  $K$  in  $\mathbb{R}^n$ . Then there exist  $\phi_j \in C_c^\infty(\Omega_j)$  ( $j = 1, \dots, k$ ) such that  $\phi_j \geq 0$ ,  $\sum_j \phi_j \leq 1$ , and  $\sum_j \phi_j = 1$  in a neighbourhood of  $K$ .

The set  $\{\phi_j\}$  is called a *partition of unity subordinate to the covering*  $\{\Omega_j\}$ .

**Proof.** There exist open sets with compact closures  $K_j \subset \Omega_j$  such that  $K \subset \bigcup_j K_j$ . Let  $\psi_j$  be associated with  $K_j$  and  $\Omega_j$  as in Corollary II.1.4. Define  $\phi_1 = \psi_1$  and  $\phi_j = \psi_j(1 - \psi_{j-1}) \dots (1 - \psi_1)$  for  $j = 2, \dots, k$  (as in the proof of Theorem 3.3). Then

$$\sum_j \phi_j = 1 - \prod_j \psi_j,$$

from which we read off the desired properties of  $\phi_j$ .  $\square$

## II.2 Distributions

### II.2.1. Topology on $C_c^\infty(\Omega)$ .

Let  $D_j := -i\partial/\partial x_j$  ( $j = 1, \dots, n$ ). For any ‘multi-index’  $\alpha = (\alpha_1, \dots, \alpha_n)$  ( $\alpha_j = 0, 1, 2, \dots$ ) and  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ , set

$$D^\alpha := D_1^{\alpha_1} \dots D_n^{\alpha_n},$$

$$\alpha! := \alpha_1! \dots \alpha_n!, |\alpha| := \sum \alpha_j, \text{ and } x^\alpha := x_1^{\alpha_1} \dots x_n^{\alpha_n}.$$

We denote also

$$f^{(\alpha)} = \frac{\partial^{|\alpha|} f}{\partial x_1^{\alpha_1} \dots \partial x_n^{\alpha_n}}$$

for any function  $f$  for which the above derivative makes sense.

Let  $K$  be a compact subset of the open set  $\Omega \subset \mathbb{R}^n$ . The sequence of semi-norms on  $C_c^\infty(K)$

$$\|\phi\|_k := \sum_{|\alpha| \leq k} \sup |D^\alpha \phi|, \quad k = 0, 1, \dots,$$

induces a (locally convex) topology on the vector space  $C_c^\infty(K)$ : it is the weakest topology on the space for which all these semi-norms are continuous. Basic neighbourhoods of 0 in this topology have the form

$$\{\phi \in C_c^\infty(K); \|\phi\|_k < \epsilon\}, \quad (1)$$

with  $\epsilon > 0$  and  $k = 0, 1, \dots$ .

A sequence  $\{\phi_j\}$  converges to  $\phi$  in  $C_c^\infty(K)$  iff  $D^\alpha \phi_j \rightarrow D^\alpha \phi$  uniformly, for all  $\alpha$ .

A linear functional  $u$  on  $C_c^\infty(K)$  is continuous iff there exist a constant  $C > 0$  and a non-negative integer  $k$  such that

$$|u(\phi)| \leq C \|\phi\|_k \quad (\phi \in C_c^\infty(K)) \quad (2)$$

(cf. Theorem 4.2).

Let  $\{\Omega_j\}$  be a sequence of open subsets of  $\Omega$ , with union  $\Omega$ , such that, for each  $j$ ,  $\Omega_j$  has compact closure  $K_j$  contained in  $\Omega_{j+1}$ .

Since  $C_c^\infty(\Omega) = \bigcup_j C_c^\infty(K_j)$ , we may topologize  $C_c^\infty(\Omega)$  in a natural way so that any linear functional  $u$  on  $C_c^\infty(\Omega)$  is continuous iff its restriction to  $C_c^\infty(K)$  is continuous for all compact  $K \subset \Omega$ . (This so-called ‘inductive limit topology’ will not be described here systematically.) We note the following facts:

- (i) a linear functional  $u$  on  $C_c^\infty(\Omega)$  is continuous iff for each compact  $K \subset \Omega$ , there exist  $C > 0$  and  $k \in \mathbb{N} \cup \{0\}$  such that (2) holds;
- (ii) a sequence  $\{\phi_j\} \subset C_c^\infty(\Omega)$  converges to 0 iff  $\{\phi_j\} \subset C_c^\infty(K)$  for some compact  $K \subset \Omega$ , and  $\phi_j \rightarrow 0$  in  $C_c^\infty(K)$ .
- (iii) a linear functional  $u$  on  $C_c^\infty(\Omega)$  is continuous iff  $u(\phi_j) \rightarrow 0$  for any sequence  $\{\phi_j\} \subset C_c^\infty(\Omega)$  converging to 0.

**Definition II.2.2.** The space  $C_c^\infty(\Omega)$  with the topology described above is denoted  $\mathcal{D}(\Omega)$  and is called *the space of test functions on  $\Omega$* . The elements of its dual  $\mathcal{D}'(\Omega)$  (= the space of all continuous linear functionals on  $\mathcal{D}(\Omega)$ ) are called *distributions* in  $\Omega$ .

The topology on  $\mathcal{D}'(\Omega)$  is the ‘weak\*’ topology: the net  $\{u_\nu\}$  of distributions converges to 0 iff  $u_\nu(\phi) \rightarrow 0$  for all  $\phi \in \mathcal{D}(\Omega)$ .

### II.2.3. Measures and functions.

If  $\mu$  is a regular complex Borel measure on  $\Omega$ , it may be identified with a continuous linear functional on  $C_c(\Omega)$  (through the Riesz representation theorem); since it is uniquely determined by its restriction to  $\mathcal{D}(\Omega)$  (cf. Corollary II.1.3), and this restriction is *continuous* (with respect to the stronger topology of  $\mathcal{D}(\Omega)$ ),

the measure  $\mu$  can (and will) be identified with the *distribution*  $u := \mu|_{\mathcal{D}(\Omega)}$ . We say in this case that the distribution  $u$  is a *measure*.

In the special case where  $d\mu = f dx$  with  $f \in L^1_{loc}(\Omega)$  (i.e.  $f$  ‘locally integrable’, that is, integrable on compact subsets), the function  $f$  is usually identified with the distribution it induces (as above) through the formula

$$f(\phi) := \int_{\Omega} \phi f dx \quad (\phi \in \mathcal{D}(\Omega)).$$

In such event, we say that the distribution is a *function*.

If  $\Omega'$  is an open subset of  $\Omega$ , the restriction of a distribution  $u$  in  $\Omega$  to  $\mathcal{D}(\Omega')$  is a distribution in  $\Omega'$ , denoted  $u|_{\Omega'}$  (and called *the restriction of  $u$  to  $\Omega'$* ). If the distributions  $u_1, u_2$  have equal restrictions to some open neighbourhood of  $x$ , one says that they are *equal in a neighbourhood of  $x$* .

**Proposition II.2.4.** *If two distributions in  $\Omega$  are equal in a neighbourhood of each point of  $\Omega$ , then they are equal.*

**Proof.** Fix  $\phi \in \mathcal{D}(\Omega)$ , and let  $K = \text{supp } \phi$ . Each  $x \in K$  has an open neighbourhood  $\Omega_x \subset \Omega$  in which the given distributions  $u_1, u_2$  are equal. By compactness of  $K$ , there exist open sets  $\Omega_j := \Omega_{x_j}$  ( $j = 1, \dots, m$ ) such that  $K \subset \bigcup_j \Omega_j$ . Let  $\{\phi_j\}$  be a partition of unity subordinate to the open covering  $\{\Omega_j\}$  of  $K$ . Then  $\phi = \sum_j \phi \phi_j$  and  $\phi \phi_j \in \mathcal{D}(\Omega_j)$ . Hence  $u_1(\phi \phi_j) = u_2(\phi \phi_j)$  by hypothesis, and therefore  $u_1(\phi) = u_2(\phi)$ .  $\square$

### II.2.5. The support.

For any distribution  $u$  in  $\Omega$ , the set

$$Z(u) := \{x \in \Omega; u = 0 \text{ in a neighbourhood of } x\}$$

is open, and  $u|_{Z(u)} = 0$  by Proposition II.2.4; furthermore,  $Z(u)$  is *the largest open subset  $\Omega'$  of  $\Omega$  such that  $u|_{\Omega'} = 0$*  (if  $x \in \Omega'$  for such a set  $\Omega'$ , then  $u = 0$  in the neighbourhood  $\Omega'$  of  $x$ , that is,  $x \in Z(u)$ ; hence  $\Omega' \subset Z(u)$ ). The *support* of  $u$ , denoted  $\text{supp } u$ , is the set  $\Omega - Z(u)$  (relatively *closed* in  $\Omega$ ). The previous statement may be rephrased as follows in terms of the support:  $\text{supp } u$  is the smallest relatively closed subset  $S$  of  $\Omega$  such that  $u(\phi) = 0$  for all  $\phi \in \mathcal{D}(\Omega)$  such that  $\text{supp } \phi \cap S = \emptyset$ .

If the distribution  $u$  is a measure or a function, its support as a distribution coincides with its support as a measure or a function, respectively (exercise).

### II.2.6. Differentiation.

Fix the open set  $\Omega \subset \mathbb{R}^n$ .

For  $j = 1, \dots, n$  and  $u \in \mathcal{D}' := \mathcal{D}'(\Omega)$ , we set

$$(D_j u)\phi = -u(D_j \phi) \quad (\phi \in \mathcal{D} := \mathcal{D}(\Omega)).$$

Then  $D_j u \in \mathcal{D}'$  and the map  $u \rightarrow D_j u$  is a continuous linear map of  $\mathcal{D}'$  into itself. Furthermore,  $D_k D_j = D_j D_k$  for all  $j, k \in \{1, \dots, n\}$ , and

$$(D^\alpha u)\phi = (-1)^{|\alpha|} u(D^\alpha \phi) \quad (\phi \in \mathcal{D}).$$

For example, if  $\delta$  denotes the Borel measure

$$\delta(E) = I_E(0) \quad (E \in \mathcal{B}(\mathbb{R}^n))$$

(the so called *delta measure at 0*), then for all  $\phi \in \mathcal{D}$ ,

$$(D^\alpha \delta)\phi = (-1)^{|\alpha|} \delta(D^\alpha \phi) = (-1)^{|\alpha|} (D^\alpha \phi)(0).$$

If  $u$  is a function such that  $\partial u / \partial x_j$  exists and is locally integrable in  $\Omega$ , then for all  $\phi \in \mathcal{D}$ , an integration by parts shows that

$$(D_j u)\phi := -u(D_j \phi) = i \int u(\partial \phi / \partial x_j) dx = -i \int (\partial u / \partial x_j) \phi dx := \left( \frac{1}{i} \frac{\partial u}{\partial x_j} \right) (\phi),$$

so that  $D_j u$  is the function  $(1/i)\partial u / \partial x_j$  in this case, as desired.

**Proposition II.2.7 (du Bois–Reymond).** *Let  $u, f \in C(\Omega)$ , and suppose  $D_j u = f$  (in the distribution sense). Then  $D_j u = f$  in the classical sense.*

**Proof.** Case  $u \in C_c(\Omega)$ . Let  $u_r, f_r$  be regularizations of  $u, f$  (using the same  $\phi$  as in II.1). Then

$$\begin{aligned} r^n D_j u_r(x) &= \int u(y) D_{x_j} \phi((x-y)/r) dy = - \int u(y) D_{y_j} \phi((x-y)/r) dy \\ &= \int (D_j u) \phi((x-y)/r) dy = r^n f_r. \end{aligned}$$

By Theorem II.1.2,  $u_r \rightarrow u$  and  $D_j u_r = f_r \rightarrow f$  uniformly as  $r \rightarrow 0$ . Therefore,  $D_j u = f$  in the classical sense.

$u \in C(\Omega)$  arbitrary. Let  $\psi \in \mathcal{D}(\Omega)$ . Then  $v := \psi u \in C_c(\Omega)$ , and  $D_j v = (D_j \psi)u + \psi f := g \in C(\Omega)$ , so by the first case above,  $D_j v = g$  in the classical sense. For any point  $x \in \Omega$ , we may choose  $\psi$  as above not vanishing at  $x$ . Then  $u = v/\psi$  is differentiable with respect to  $x_j$  at  $x$ , and  $D_j u = f$  at  $x$  (in the classical sense).  $\square$

Let  $\omega$  be an open set with compact closure contained in  $\Omega$  (this relation between  $\omega$  and  $\Omega$  is denoted  $\omega \subset\subset \Omega$ ), and let  $\rho = \text{diam}(\omega) := \sup\{|x-y|; x, y \in \omega\} (< \infty)$ . Denote the unit vectors in the  $x_j$ -direction by  $e_j$ . Given  $x \in \omega$ , let  $t_j$  be the smallest positive number  $t$  such that  $x + te_j \in \partial\omega$ . If  $\phi \in \mathcal{D}(\omega)$ , then  $\phi(x + t_j e_j) = 0$ , and by the mean value theorem, there exists  $0 < \tau_j < t_j$  such that

$$|\phi(x)| = |\phi(x) - \phi(x + t_j e_j)| = t_j |D_j \phi(x + \tau_j e_j)| \leq \rho \sup |D_j \phi|.$$

Hence

$$\sup |\phi| \leq \rho \sup |D_j \phi| \quad (\phi \in \mathcal{D}(\omega)).$$

Therefore, for any multi-index  $\alpha$  with  $|\alpha| \leq k$ ,

$$\sup |D^\alpha \phi| \leq \rho^{nk-|\alpha|} \sup |D_1^k \cdots D_n^k \phi|,$$

and consequently

$$\|\phi\|_k \leq C' \sup |D_1^k \cdots D_n^k \phi| \quad (\phi \in \mathcal{D}(\omega)),$$

where  $C' = C'(\rho, k, n)$  is a positive constant. Let  $u \in \mathcal{D}'(\Omega)$ . By (2), there exist a constant  $C > 0$  and a non-negative integer  $k$  such that

$$|u(\phi)| \leq CC' \sup |D_1^k \cdots D_n^k \phi| \quad (\phi \in \mathcal{D}(\omega)).$$

Write

$$(D_1^k \cdots D_n^k \phi)(x) = i^n \int_{[y < x]} D_1^{k+1} \cdots D_n^{k+1} \phi \, dy,$$

where  $[y < x] := \{y \in \mathbb{R}^n; y_j < x_j \ (j = 1, \dots, n)\}$  and  $\phi$  is extended to  $\mathbb{R}^n$  in the usual way ( $\phi = 0$  on  $\omega^c$ ). Writing  $s := k + 1$ , we have therefore

$$\sup |D_1^k \cdots D_n^k \phi| \leq \int_{\omega} |D_1^s \cdots D_n^s \phi| \, dy,$$

and consequently

$$|u(\phi)| \leq CC' \|D_1^s \cdots D_n^s \phi\|_{L^1(\omega)} \quad (\phi \in \mathcal{D}(\omega)).$$

This means that the linear functional

$$(-1)^{ns} D_1^s \cdots D_n^s \phi \rightarrow u(\phi)$$

is continuous with norm  $\leq CC'$  on the subspace  $D_1^s \cdots D_n^s \mathcal{D}(\omega)$  of  $L^1(\omega)$ . By the Hahn–Banach theorem, it has an extension as a continuous linear functional on  $L^1(\omega)$  (with the same norm). Therefore there exists  $f \in L^\infty(\omega)$  such that

$$u(\phi) = (-1)^{ns} \int_{\omega} f D_1^s \cdots D_n^s \phi \, dx \quad (\phi \in \mathcal{D}(\omega)).$$

This means that

$$u|_{\omega} = D_1^s \cdots D_n^s f.$$

We may also define

$$g(x) = i^n \int_{[y < x]} I_{\omega}(y) f(y) \, dy;$$

Then  $g$  is continuous, and one verifies easily that  $f = D_1 \cdots D_n g$  (in the distribution sense). Hence

$$u|_{\omega} = D_1^{s+1} \cdots D_n^{s+1} g$$

(with  $g$  continuous in  $\omega$ ). We proved

**Theorem II.2.8.** *Let  $u \in \mathcal{D}'(\Omega)$ . Then for any  $\omega \subset \subset \Omega$ , there exist a non-negative integer  $s$  and a function  $f \in L^\infty(\omega)$  such that  $u|_{\omega} = D_1^s \cdots D_n^s f$ . Moreover,  $f$  may be chosen to be continuous.*

**II.2.9. Leibnitz' formula.**

If  $\phi, \psi \in C^\infty(\Omega)$ , then for any multi-index  $\alpha$

$$D^\alpha(\phi\psi) = \sum_{\beta \leq \alpha} \frac{\alpha!}{\beta!(\alpha - \beta)!} D^\beta \phi D^{\alpha - \beta} \psi \quad (3)$$

(the sum goes over all multi-indices  $\beta$  with  $\beta_j \leq \alpha_j$  for all  $j$ ). This general Leibnitz formula follows by repeated application of the usual one-variable formula.

*Multiplication of a distribution by a function.*

Let  $u \in \mathcal{D}'(\Omega)$  and  $\psi \in C^\infty(\Omega)$ . Since  $\psi\phi \in \mathcal{D}(\Omega)$  for all  $\phi \in \mathcal{D}(\Omega)$ , the map  $\phi \rightarrow u(\psi\phi)$  is well defined on  $\mathcal{D}(\Omega)$ . It is trivially linear, and for any compact  $K \subset \Omega$ , if  $C$  and  $k$  are as in (2), then for all  $\phi \in C_c^\infty(K)$ , we have by (3)

$$|u(\psi\phi)| \leq C \|\psi\phi\|_k \leq C' \|\phi\|_k,$$

where  $C'$  is a constant depending on  $n, k, K$ , and  $\psi$ . This means that the map defined above is a distribution in  $\Omega$ ; it is denoted  $\psi u$  (and called the product of  $u$  by  $\psi$ ). Thus

$$(\psi u)(\phi) = u(\psi\phi) \quad (\phi \in \mathcal{D}(\Omega)) \quad (4)$$

for all  $\psi \in C^\infty(\Omega)$ .

This definition clearly coincides with the usual pointwise multiplication when  $u$  is a function or a measure.

We verify easily the inclusion  $Z(\psi) \cup Z(u) \subset Z(\psi u)$ , that is,

$$\text{supp}(\psi u) \subset \text{supp} \psi \cap \text{supp} u.$$

It follows from the definitions that

$$D_j(\psi u) = (D_j \psi)u + \psi(D_j u)$$

( $\psi \in C^\infty(\Omega)$ ,  $u \in \mathcal{D}'(\Omega)$ ,  $j = 1, \dots, n$ ), and therefore, by the same formal arguments as in the classical case, Leibnitz' formula (3) is valid in the present situation.

If  $P$  is a polynomial on  $\mathbb{R}^n$ , we denote by  $P(D)$  the differential operator obtained by substituting formally  $D$  for the variable  $x \in \mathbb{R}^n$ . Let  $P_\alpha(x) := x^\alpha$  for any multi-index  $\alpha$ . Since  $P_\alpha^{(\beta)}(x) = (\alpha! / (\alpha - \beta)!) x^{\alpha - \beta}$  for  $\beta \leq \alpha$  (and equals zero otherwise), we can rewrite (3) in the form

$$P(D)(\psi u) = \sum_{\beta} (1/\beta!) (D^\beta \psi) P^{(\beta)}(D) u \quad (5)$$

for the special polynomials  $P = P_\alpha$ , hence by linearity, for *all* polynomials  $P$ . This is referred to as the 'general Leibnitz formula'.

**II.2.10.** *The space  $\mathcal{E}(\Omega)$  and its dual.*

The space  $\mathcal{E}(\Omega)$  is the space  $C^\infty(\Omega)$  as a (locally convex) t.v.s. with the topology induced by the family of semi-norms

$$\phi \rightarrow \|\phi\|_{k,K} := \sum_{|\alpha| \leq k} \sup_K |D^\alpha \phi|, \quad (6)$$

with  $k = 0, 1, 2, \dots$ , and  $K$  varying over all compact subsets of  $\Omega$ .

A sequence  $\{\phi_k\}$  converges to 0 in  $\mathcal{E}(\Omega)$  iff  $D^\alpha \phi_k \rightarrow 0$  uniformly on every compact subset of  $\Omega$ , for all multi-indices  $\alpha$ .

A linear functional  $u$  on  $\mathcal{E}(\Omega)$  is continuous iff there exist constants  $k \in \mathbb{N} \cup \{0\}$  and  $C > 0$  and a compact set  $K$  such that

$$|u(\phi)| \leq C \|\phi\|_{k,K} \quad (\phi \in \mathcal{E}(\Omega)). \quad (7)$$

The dual space  $\mathcal{E}'(\Omega)$  of  $\mathcal{E}(\Omega)$  consists of all these continuous linear functionals, with the weak\*-topology: the net  $u_\nu$  converges to  $u$  in  $\mathcal{E}'(\Omega)$  if  $u_\nu(\phi) \rightarrow u(\phi)$  for all  $\phi \in \mathcal{E}(\Omega)$ .

If  $u \in \mathcal{E}'(\Omega)$ , then by (7),  $u(\phi) = 0$  whenever  $\phi \in \mathcal{E}(\Omega)$  vanishes in a neighbourhood of the compact set  $K$  (appearing in (7)). For all  $\phi \in \mathcal{D}(\Omega)$ ,  $|u(\phi)| \leq C \|\phi\|_{k,K} \leq C \|\phi\|_k$ , that is,  $\tilde{u} := u|_{\mathcal{D}(\Omega)} \in \mathcal{D}'(\Omega)$ . Also, if  $x \in \Omega - K$  and  $\omega$  is an open neighbourhood of  $x$  contained in  $\Omega - K$ , then for all  $\phi \in \mathcal{D}(\omega)$ ,  $\|\phi\|_{k,K} = 0$ , and (7) implies that  $u(\phi) = 0$ . This shows that  $\Omega - K \subset Z(\tilde{u})$ , that is,  $\text{supp } \tilde{u} \subset K$ , that is,  $\tilde{u}$  is a *distribution with compact support* in  $\Omega$ .

Conversely, let  $v$  be a distribution with compact support in  $\Omega$ , and let  $K$  be any compact subset of  $\Omega$  containing this support. Fix  $\psi \in \mathcal{D}(\Omega)$  such that  $\psi = 1$  in a neighbourhood of  $K$ . For any  $\phi \in \mathcal{E}(\Omega)$ , define  $u(\phi) = v(\phi\psi)$ . Then  $u$  is a well-defined linear functional on  $\mathcal{E}(\Omega)$ ,  $u(\phi) = 0$  whenever  $\phi = 0$  in a neighbourhood of  $K$ , and  $u(\phi) = v(\phi)$  for  $\phi \in \mathcal{D}(\Omega)$ . On the other hand, if  $w$  is a linear functional on  $\mathcal{E}(\Omega)$  with these properties, then for all  $\phi \in \mathcal{E}(\Omega)$ ,  $\phi\psi \in \mathcal{D}(\Omega)$  and  $\phi(1 - \psi) = 0$  in a neighbourhood of  $K$ , and consequently

$$w(\phi) = w(\phi\psi) + w(\phi(1 - \psi)) = v(\phi\psi) = u(\phi).$$

This shows that  $v$  has a *unique* extension as a linear functional on  $\mathcal{E}(\Omega)$  such that  $v(\phi) = 0$  whenever  $\phi$  vanishes in a neighbourhood of  $K$ .

Let  $Q = \text{supp } \psi$ . By (2) applied to the compact set  $Q$ , there exist  $C$  and  $k$  such that

$$|u(\phi)| = |v(\phi\psi)| \leq C \|\phi\psi\|_k$$

for all  $\phi \in \mathcal{E}(\Omega)$  (because  $\phi\psi \in \mathcal{D}(Q)$ ). Hence, by Leibnitz' formula,  $|u(\phi)| \leq C' \|\phi\|_{k,Q}$  for some constant  $C'$ , that is,  $u \in \mathcal{E}'(\Omega)$ .

We have established, therefore, that each distribution  $v$  with compact support has a unique extension as an element  $u \in \mathcal{E}'(\Omega)$ , and conversely, each  $u \in \mathcal{E}'(\Omega)$  restricted to  $\mathcal{D}(\Omega)$  is a distribution with compact support. This relationship allows us to *identify  $\mathcal{E}'(\Omega)$  with the space of all distributions with compact support in  $\Omega$ .*



**II.2.11. Convolution.**

Let  $u \in \mathcal{D}' := \mathcal{D}'(\mathbb{R}^n)$  and  $\phi \in \mathcal{D} := \mathcal{D}(\mathbb{R}^n)$ . For  $x \in \mathbb{R}^n$  fixed, let

$$(u * \phi)(x) := u(\phi(x - \cdot)).$$

The function  $u * \phi$  is called the *convolution of  $u$  and  $\phi$* .

For  $x$  fixed,  $h \neq 0$  real, and  $j = 1, \dots, n$ , the functions  $(ih)^{-1}[\phi(x + he_j - \cdot) - \phi(x - \cdot)]$  converge to  $(D_j \phi)(x - \cdot)$  (as  $h \rightarrow 0$ ) in  $\mathcal{D}$ . Therefore  $(ih)^{-1}[(u * \phi)(x + he_j) - (u * \phi)(x)] \rightarrow (u * (D_j \phi))(x)$ , that is,  $D_j(u * \phi) = u * (D_j \phi)$  in the classical sense. Also  $u * (D_j \phi) = (D_j u) * \phi$  by definition of the derivative of a distribution. Iterating, we obtain that  $u * \phi \in \mathcal{E} := \mathcal{E}(\mathbb{R}^n)$  and for any multi-index  $\alpha$ ,

$$D^\alpha(u * \phi) = u * (D^\alpha \phi) = (D^\alpha u) * \phi. \quad (8)$$

If  $\text{supp } u \cap \text{supp } \phi(x - \cdot) = \emptyset$ , then  $(u * \phi)(x) = 0$ . Equivalently, if  $(u * \phi)(x) \neq 0$  then  $\text{supp } u$  meets  $\text{supp } \phi(x - \cdot)$  at some point  $y$ , that is,  $x - y \in \text{supp } \phi$  and  $y \in \text{supp } u$ , that is,  $x \in \text{supp } u + \text{supp } \phi$ . This shows that

$$\text{supp } (u * \phi) \subset \text{supp } u + \text{supp } \phi. \quad (9)$$

Hence

$$\mathcal{E}' * \mathcal{D} \subset \mathcal{D} \quad (10)$$

and in particular

$$\mathcal{D} * \mathcal{D} \subset \mathcal{D}. \quad (11)$$

Let  $\phi_m \rightarrow 0$  in  $\mathcal{D}$ . There exists then a compact set  $K$  containing  $\text{supp } \phi_m$  for all  $m$ , and  $D^\alpha \phi_m \rightarrow 0$  uniformly for all  $\alpha$ . Let  $Q$  be any compact set in  $\mathbb{R}^n$ . It follows that  $\text{supp } \phi_m(x - \cdot) \subset Q - K := \{x - y; x \in Q, y \in K\}$  for all  $x \in Q$ . By (2) with the compact set  $Q - K$  and the distribution  $D^\alpha u$ , there exist  $C, k$  such that

$$|D^\alpha(u * \phi_m)(x)| = |(D^\alpha u)(\phi_m(x - \cdot))| \leq C \|\phi_m(x - \cdot)\|_k = C \|\phi_m\|_k \rightarrow 0$$

for all  $x \in Q$ . Hence  $D^\alpha(u * \phi_m) \rightarrow 0$  uniformly on  $Q$ , and we conclude that  $u * \phi_m \rightarrow 0$  in the topological space  $\mathcal{E}$ . In other words, *the (linear) map  $\phi \rightarrow u * \phi$  is sequentially continuous from  $\mathcal{D}$  to  $\mathcal{E}$* . If  $u \in \mathcal{E}'$ , the map is (sequentially) continuous from  $\mathcal{D}$  into itself (cf. (10)). In this case, *the definition of  $u * \phi$  makes sense for all  $\phi \in \mathcal{E}$ , and the (linear) map  $\phi \rightarrow u * \phi$  from  $\mathcal{E}$  into itself is continuous* (note that  $\mathcal{E}(\Omega)$  is metrizable, so there is no need to qualify the continuity; the metrizability follows from the fact that the topology of  $\mathcal{E}(\Omega)$  is induced by the countable family of semi-norms  $\{\|\cdot\|_{k, K_m}; k, m = 0, 1, 2, \dots\}$ , where  $\{K_m\}$  is a suitable sequence of compact sets with union equal to  $\Omega$ ).

If  $\phi, \psi \in \mathcal{D}$  and  $u \in \mathcal{D}'$ , it follows from (11) and the fact that  $u * \phi \in \mathcal{E}$  that both  $u * (\phi * \psi)$  and  $(u * \phi) * \psi$  make sense. In order to show that they coincide, we approximate  $(\phi * \psi)(x) = \int_Q \phi(x - y)\psi(y) dy$  (where  $Q$  is an  $n$ -dimensional cube containing the support of  $\psi$ ) by (finite) Riemann sums of the form

$$\chi_m(x) = m^{-n} \sum_{y \in \mathbb{Z}^n} \phi(x - y/m)\psi(y/m).$$

If  $\chi_m(x) \neq 0$  for some  $x$  and  $m$ , then there exists  $y \in \mathbb{Z}^n$  such that  $y/m \in \text{supp } \psi$  and  $x - y/m \in \text{supp } \phi$ , that is,  $x \in y/m + \text{supp } \phi \subset \text{supp } \psi + \text{supp } \phi$ . This shows that for all  $m$ ,  $\chi_m$  have support in the fixed compact set  $\text{supp } \psi + \text{supp } \phi$ . Also for all multi-indices  $\alpha$ ,

$$(D^\alpha \chi_m)(x) = m^{-n} \sum_y D^\alpha \phi(x - y/m) \psi(y/m) \rightarrow ((D^\alpha \phi) * \psi)(x) = (D^\alpha (\phi * \psi))(x)$$

uniformly (in  $x$ ). This means that  $\chi_m \rightarrow \phi * \psi$  in  $\mathcal{D}$ . By continuity of  $u$  on  $\mathcal{D}$ ,

$$\begin{aligned} [u * (\phi * \psi)](x) &:= u((\phi * \psi)(x - \cdot)) = \lim_m u(\chi_m(x - \cdot)) \\ &= \lim_m (u * \chi_m)(x) = \lim_m m^{-n} \sum_y (u * \phi)(x - y/m) \psi(y/m) \\ &= [(u * \phi) * \psi](x), \end{aligned}$$

for all  $x$ , that is,

$$u * (\phi * \psi) = (u * \phi) * \psi \quad (\phi, \psi \in \mathcal{D}; u \in \mathcal{D}'). \quad (12)$$

Fix  $\phi$  as in Section II.1, consider  $\phi_r$  as before, and define  $u_r := u * \phi_r$  for any  $u \in \mathcal{D}'$ .

**Proposition II.2.12 (Regularization of distributions).** *For any distribution  $u$  in  $\mathbb{R}^n$ ,*

- (i)  $u_r \in \mathcal{E}$  for all  $r > 0$ ;
- (ii)  $\text{supp } u_r \subset \text{supp } u + \{x; |x| \leq r\}$ ;
- (iii)  $u_r \rightarrow u$  in the space  $\mathcal{D}'$ .

**Proof.** Since  $\text{supp } \phi_r = \{x; |x| \leq r\}$ , (i) and (ii) are special cases of properties of the convolution discussed in II.2.11.

Denote  $J : \psi(x) \in \mathcal{D} \rightarrow \tilde{\psi}(x) := \psi(-x)$ . Then

$$u(\psi) = (u * \tilde{\psi})(0). \quad (13)$$

By Theorem II.1.2 applied to  $\tilde{\psi}$ ,  $\phi_r * \tilde{\psi} \rightarrow \tilde{\psi}$  in  $\mathcal{D}$ . Therefore, by (13),

$$\begin{aligned} u_r(\psi) &= [(u * \phi_r) * \tilde{\psi}](0) = [u * (\phi_r * \tilde{\psi})](0) \\ &= u(J(\phi_r * \tilde{\psi})) \rightarrow u(\psi) \end{aligned}$$

for all  $\psi \in \mathcal{D}$ , that is,  $u_r \rightarrow u$  in  $\mathcal{D}'$ . □

In particular,  $\mathcal{E}$  is sequentially dense in  $\mathcal{D}'$ .

Note also that if  $u * \mathcal{D} = \{0\}$ , then  $u_r = 0$  for all  $r > 0$ ; letting  $r \rightarrow 0$ , it follows that  $u = 0$ .

**II.2.13. Commutation with translations.**

Consider the *translation operators* (for  $h \in \mathbb{R}^n$ )

$$\tau_h : \phi(x) \rightarrow \phi(x - h)$$

from  $\mathcal{D}$  into itself. For any  $u \in \mathcal{D}'$ , it follows from the definitions that

$$\tau_h(u * \phi) = u * (\tau_h \phi),$$

that is, *convolution with  $u$  commutes with translations*. This commutation property, together with the previously observed fact that convolution with  $u$  is a (sequentially) continuous linear map from  $\mathcal{D}$  to  $\mathcal{E}$ , *characterizes convolution with distributions*. Indeed, let  $U : \mathcal{D} \rightarrow \mathcal{E}$  be linear, sequentially continuous, and commuting with translations. Define  $u(\phi) = (U(\tilde{\phi}))(0)$ , ( $\phi \in \mathcal{D}$ ). Then  $u$  is a linear functional on  $\mathcal{D}$ , and if  $\phi_k \rightarrow 0$  in  $\mathcal{D}$ , the sequential continuity of  $U$  on  $\mathcal{D}$  implies that  $u(\phi_k) \rightarrow 0$ . Hence  $u \in \mathcal{D}'$ . For any  $x \in \mathbb{R}^n$  and  $\phi \in \mathcal{D}$ ,

$$\begin{aligned} (U\phi)(x) &= [\tau_{-x}U\phi](0) = [U(\tau_{-x}\phi)](0) \\ &= u(J(\tau_{-x}\phi)) = [u * (\tau_{-x}\phi)](0) = [\tau_{-x}(u * \phi)](0) = (u * \phi)(x), \end{aligned}$$

that is,  $U\phi = u * \phi$ , as wanted.

**II.2.14. Convolution of distributions.**

Let  $u, v$  be distributions in  $\mathbb{R}^n$ , one of which has compact support. The map

$$W : \phi \in \mathcal{D} \rightarrow u * (v * \phi) \in \mathcal{E}$$

is linear, continuous, and commutes with translations. By II.2.13, there exists a distribution  $w$  such that  $W\phi = w * \phi$ . By the final observation in II.2.12,  $w$  is uniquely determined; we call it *the convolution of  $u$  and  $v$*  and denote it by  $u * v$ ; thus, *by definition*,

$$(u * v) * \phi = u * (v * \phi) \quad (\phi \in \mathcal{D}). \quad (14)$$

If  $v = \psi \in \mathcal{D}$ , the right-hand side of (14) equals  $(u * \psi) * \phi$  by (12) (where  $u * \psi$  is the ‘usual’ convolution of the distribution  $u$  with  $\psi$ ). Again by the final observation of II.2.12, it follows that the convolution of the two distributions  $u, v$  coincides with the previous definition when  $v$  is a function in  $\mathcal{D}$  (the same is true if  $u \in \mathcal{E}'$  and  $v = \psi \in \mathcal{E}$ ).

One verifies easily that

$$\text{supp}(u * v) \subset \text{supp } u + \text{supp } v. \quad (15)$$

(With  $\phi_r$  as in Section II.1, it follows from (9) and Proposition II.2.12 that

$$\begin{aligned} \text{supp}[(u * v) * \phi_r] &= \text{supp}[u * (v * \phi_r)] \subset \text{supp } u + \text{supp } v_r \\ &\subset \text{supp } u + \text{supp } v + \{x; |x| \leq r\}, \end{aligned}$$

and we obtain (15) by letting  $r \rightarrow 0$ .)

If  $u, v, w$  are distributions, two of which (at least) having compact support, then by (15) both convolutions  $(u * v) * w$  and  $u * (v * w)$  are well-defined distributions. Since their convolutions with any given  $\phi \in \mathcal{D}$  coincide, the ‘associative law’ for distributions follows (cf. end of Proposition II.2.12).

Convolution of *functions* in  $\mathcal{E}$  (one of which at least having compact support) is seen to be commutative by a change of variable in the defining integral. If  $u \in \mathcal{D}'$  and  $\psi \in \mathcal{D}$ , we have for all  $\phi \in \mathcal{D}$  (by definition and the associative law we just verified!):

$$\begin{aligned}(u * \psi) * \phi &:= u * (\psi * \phi) = u * (\phi * \psi) \\ &:= (u * \phi) * \psi = \psi * (u * \phi) = (\psi * u) * \phi,\end{aligned}$$

and therefore  $u * \psi = \psi * u$ . The same is valid (with a similar proof) when  $u \in \mathcal{E}'$  and  $\psi \in \mathcal{E}$ .

For any two distributions  $u, v$  (one of which at least having compact support), the *commutative law of convolution* follows now from the same formal calculation with  $\psi$  replaced by  $v$ .

Let  $\delta$  be the ‘delta measure at 0’ (cf. II.2.6). Then for any multi-index  $\alpha$  and  $u \in \mathcal{D}'$ ,

$$(D^\alpha \delta) * u = D^\alpha u. \quad (16)$$

Indeed, observe first that for all  $\phi \in \mathcal{D}$ ,  $(\delta * \phi)(x) = \int \phi(x - y) d\delta(y) = \phi(x)$ , that is  $\delta * \phi = \phi$ . Therefore, for any  $v \in \mathcal{D}'$ ,  $(v * \delta) * \phi = v * (\delta * \phi) = v * \phi$ , and consequently

$$v * \delta = v \quad (v \in \mathcal{D}'). \quad (17)$$

Now for all  $\phi \in \mathcal{D}$ ,

$$\begin{aligned}(u * D^\alpha \delta) * \phi &= u * (D^\alpha \delta * \phi) = u * (D^\alpha \phi * \delta) \\ &= u * D^\alpha \phi = (D^\alpha u) * \phi,\end{aligned}$$

and (16) follows (cf. end of II.2.12 and (8)).

Next, for any distributions  $u, v$  with one at least having compact support, we have by (16) and the associative and commutative laws for convolution:

$$\begin{aligned}D^\alpha(u * v) &= (D^\alpha \delta) * (u * v) = (D^\alpha \delta * u) * v = (D^\alpha u) * v \\ &= D^\alpha(v * u) = (D^\alpha v) * u = u * (D^\alpha v).\end{aligned}$$

This generalizes (8) to the case when both factors in the convolution are distributions.

## II.3 Temperate distributions

### II.3.1. The Schwartz space.

The Schwartz space  $\mathcal{S} = \mathcal{S}(\mathbb{R}^n)$  of *rapidly decreasing functions* consists of all  $\phi \in \mathcal{E} = \mathcal{E}(\mathbb{R}^n)$  such that

$$\|\phi\|_{\alpha, \beta} := \sup_x |x^\beta D^\alpha \phi(x)| < \infty.$$

The topology induced on  $\mathcal{S}$  by the family of semi-norms  $\|\cdot\|_{\alpha,\beta}$  (where  $\alpha, \beta$  range over all multi-indices) makes  $\mathcal{S}$  into a locally convex (metrizable) topological vector space. It follows from Leibnitz' formula that  $\mathcal{S}$  is a topological algebra for pointwise multiplication. It is also closed under multiplication by polynomials and application of any operator  $D^\alpha$ , with both operations continuous from  $\mathcal{S}$  into itself.

We have the topological inclusions

$$\mathcal{D} \subset \mathcal{S} \subset \mathcal{E}; \quad \mathcal{S} \subset L^1 := L^1(\mathbb{R}^n).$$

(Let  $p_n(x) = \prod_{j=1}^n (1 + x_j^2)$ ; since  $\|1/p_n\|_{L^1} = \pi^n$ , we have

$$\|\phi\|_{L^1} \leq \pi^n \sup_x |p_n(x)\phi(x)|. \quad (1)$$

If  $\phi_k \rightarrow 0$  in  $\mathcal{S}$ , it follows from (1) that  $\phi_k \rightarrow 0$  in  $L^1$ .)

Fix  $\phi$  as in II.1.1, and let  $\chi$  be the indicator of the ball  $\{x; |x| \leq 2\}$ . The function  $\psi := \chi * \phi$  belongs to  $\mathcal{D}$  (cf. Theorem II.1.2), and  $\psi = 1$  on the closed unit ball (because for  $|x| \leq 1$ ,  $\{y; |y| \leq 1 \text{ and } |x-y| \leq 2\} = \{y; |y| \leq 1\} = \text{supp } \phi$ , and therefore  $\psi(x) = \int_{\text{supp } \phi} \chi(x-y)\phi(y) dy = \int \phi(y) dy = 1$ ).

Now, for *any*  $\phi \in \mathcal{S}$  and the function  $\psi$  defined above, consider the functions  $\phi_r(x) := \phi(x)\psi(rx)$ , ( $r > 0$ ) (not to be confused with the functions defined in II.1.1). Then  $\phi_r \in \mathcal{D}$  and  $\phi - \phi_r = 0$  for  $|x| \leq 1/r$ . Therefore  $\|\phi - \phi_r\|_{\alpha,\beta} = \sup_{|x| > 1/r} |x^\beta D^\alpha(\phi - \phi_r)|$ . We may choose  $M$  such that  $\sup_x |x^\beta| x|^2 D^\alpha(\phi - \phi_r)| < M$  for all  $0 < r \leq 1$ . Then  $\|\phi - \phi_r\|_{\alpha,\beta} \leq \sup_{|x| > 1/r} \frac{M}{|x|^2} < Mr^2 \rightarrow 0$  when  $r \rightarrow 0$ . Thus  $\phi_r \rightarrow \phi$  in  $\mathcal{S}$ , and we conclude that  $\mathcal{D}$  is dense in  $\mathcal{S}$ . A similar argument shows that  $\phi_r \rightarrow \phi$  in  $\mathcal{E}$  for *any*  $\phi \in \mathcal{E}$ ; hence  $\mathcal{D}$  (and therefore  $\mathcal{S}$ ) is dense in  $\mathcal{E}$ .

### II.3.2. The Fourier transform on $\mathcal{S}$ .

Denote the inner product in  $\mathbb{R}^n$  by  $x \cdot y$  ( $x \cdot y := \sum_j x_j y_j$ ), and let  $F : f \rightarrow \hat{f}$  be the Fourier transform on  $L^1$ :

$$\hat{f}(y) = \int_{\mathbb{R}^n} e^{-ix \cdot y} f(x) dx \quad (f \in L^1). \quad (2)$$

(All integrals in the sequel are over  $\mathbb{R}^n$ , unless specified otherwise.)

If  $\phi \in \mathcal{S}$ ,  $x^\alpha \phi(x) \in \mathcal{S} \subset L^1$  for all multi-indices  $\alpha$ , and therefore, the integral  $\int e^{-ix \cdot y} (-x)^\alpha \phi(x) dx$  converges *uniformly* in  $y$ . Since this integral is the result of applying  $D^\alpha$  to the integrand of (2), we have  $\hat{\phi} \in \mathcal{E}$  and

$$D^\alpha \hat{\phi} = F[(-x)^\alpha \phi(x)]. \quad (3)$$

For any multi-index  $\beta$ , it follows from (3) (by integration by parts) that

$$y^\beta (D^\alpha \hat{\phi})(y) = F D^\beta [(-x)^\alpha \phi(x)]. \quad (4)$$

In particular ( $\alpha = 0$ )

$$y^\beta \hat{\phi}(y) = [F D^\beta \phi](y). \quad (5)$$

Since  $D^\beta[(-x)^\alpha \phi(x)] \in \mathcal{S} \subset L^1$ , it follows from (4) that  $y^\beta(D^\alpha \hat{\phi})(y)$  is a bounded function of  $y$ , that is,  $\hat{\phi} \in \mathcal{S}$ . Moreover, by (1) and (4),

$$\begin{aligned} \sup_y \left| y^\beta(D^\alpha \hat{\phi})(y) \right| &\leq \|D^\beta[(-x)^\alpha \phi(x)]\|_{L^1} \\ &\leq \pi^n \sup_x \left| p_n(x) D^\beta[(-x)^\alpha \phi(x)] \right|. \end{aligned}$$

This inequality shows that if  $\phi_k \rightarrow 0$  in  $\mathcal{S}$ , then  $\hat{\phi}_k \rightarrow 0$  in  $\mathcal{S}$ , that is, the map  $F : \phi \in \mathcal{S} \rightarrow \hat{\phi} \in \mathcal{S}$  is a *continuous* (linear) operator. Denote

$$M^\beta : \phi(x) \in \mathcal{S} \rightarrow x^\beta \phi(x) \in \mathcal{S}.$$

Then (4) can be written as the operator identity on  $\mathcal{S}$ :

$$M^\beta D^\alpha F = F D^\beta (-M)^\alpha. \quad (6)$$

A change of variables shows that

$$(F\psi(ry))(s) = r^{-n} \hat{\psi}(s/r) \quad (7)$$

for any  $\psi \in L^1$  and  $r > 0$ .

If  $\phi, \psi \in L^1$ , an application of Fubini's theorem gives

$$\int \hat{\phi}(y) \psi(y) e^{ix \cdot y} dy = \int \hat{\psi}(t - x) \phi(t) dt = \int \hat{\psi}(s) \phi(x + s) ds. \quad (8)$$

Replacing  $\psi(y)$  by  $\psi(ry)$  in (8), we obtain by (7) and the change of variable  $s = rt$

$$\int \hat{\phi}(y) \psi(ry) e^{ix \cdot y} dy = r^{-n} \int \hat{\psi}(s/r) \phi(x + s) ds = \int \hat{\psi}(t) \phi(x + rt) dt. \quad (9)$$

In case  $\phi, \psi \in \mathcal{S}(\subset L^1)$ , we have  $\hat{\phi}, \hat{\psi} \in \mathcal{S} \subset L^1$ , and  $\phi, \psi$  are bounded and continuous. By Lebesgue's Dominated Convergence theorem, letting  $r \rightarrow 0$  in (9) gives

$$\psi(0) \int \hat{\phi}(y) e^{ix \cdot y} dy = \phi(x) \int \hat{\psi}(t) dt \quad (10)$$

for all  $\phi, \psi \in \mathcal{S}$  and  $x \in \mathbb{R}^n$ .

Choose for example  $\psi(x) = (2\pi)^{-n/2} e^{-|x|^2/2}$ . Then  $\hat{\psi}(t) = e^{-|t|^2/2}$  (cf. I.3.12) and  $\int \hat{\psi}(t) dt = (2\pi)^{n/2}$ . Substituting these values in (10), we obtain (for all  $\phi \in \mathcal{S}$ )

$$\phi(x) = (2\pi)^{-n} \int \hat{\phi}(y) e^{ix \cdot y} dy. \quad (11)$$

This is the *inversion formula for the Fourier transform*  $F$  on  $\mathcal{S}$ . It shows that  $F$  is an automorphism of  $\mathcal{S}$ , whose inverse is given by

$$F^{-1} = (2\pi)^{-n} JF, \quad (12)$$

where  $J : \phi \rightarrow \tilde{\phi}$ .

Note that  $JF = FJ$  and  $F^2 = (2\pi)^n J$ .

Also, by definition and the inversion formula,

$$\bar{\hat{\psi}}(y) = \int e^{ix \cdot y} \overline{\hat{\psi}(x)} dx = (2\pi)^n (F^{-1} \bar{\psi})(y),$$

that is,

$$F(\bar{\hat{\psi}}) = (2\pi)^n \bar{\psi} \quad (\psi \in \mathcal{S}). \quad (13)$$

(It is sometimes advantageous to define the Fourier transform by  $\mathcal{F} = (2\pi)^{-n/2} F$ ; the inversion formula for  $\mathcal{F} : \mathcal{S} \rightarrow \mathcal{S}$  is then  $\mathcal{F}^{-1} = J\mathcal{F}$ , and the last identities become  $\mathcal{F}^2 = J$  and  $\mathcal{F}C\mathcal{F} = C$ , where  $C : \psi \rightarrow \bar{\psi}$  is the conjugation operator.)

An application of Fubini's theorem shows that

$$F(\phi * \psi) = \hat{\phi} \hat{\psi} \quad (\phi, \psi \in \mathcal{S}). \quad (14)$$

(This is true actually for all  $\phi, \psi \in L^1$ .)

Replacing  $\phi, \psi$  by  $\hat{\phi}, \hat{\psi} (\in \mathcal{S})$  respectively, we get

$$\begin{aligned} F(\hat{\phi} * \hat{\psi}) &= (F^2 \phi)(F^2 \psi) = (2\pi)^{2n} (J\phi)(J\psi) \\ &= (2\pi)^{2n} J(\phi\psi) = (2\pi)^n F^2(\phi\psi). \end{aligned}$$

Hence

$$F(\phi\psi) = (2\pi)^{-n} \hat{\phi} * \hat{\psi} \quad (\phi, \psi \in \mathcal{S}). \quad (15)$$

For  $x = 0$ , the identity (8) becomes

$$\int \hat{\phi} \psi dx = \int \phi \hat{\psi} dx \quad (\phi, \psi \in L^1). \quad (16)$$

In case  $\psi \in \mathcal{S}$  (so that  $\hat{\psi} \in \mathcal{S} \subset L^1$ ), we replace  $\psi$  by  $\bar{\hat{\psi}}$  in (16); using (13), we get

$$\int \hat{\phi} \bar{\hat{\psi}} dx = (2\pi)^n \int \phi \bar{\psi} dx \quad (\phi, \psi \in \mathcal{S}). \quad (17)$$

This is *Parseval's formula* for the Fourier transform.

In terms of the operator  $\mathcal{F}$ , the formula takes the form

$$(\mathcal{F}\phi, \mathcal{F}\psi) = (\phi, \psi) \quad (\phi, \psi \in \mathcal{S}), \quad (18)$$

where  $(\cdot, \cdot)$  is the  $L^2$  inner product.

In particular,

$$\|\mathcal{F}\phi\|_2 = \|\phi\|_2, \quad (19)$$

where  $\|\cdot\|_2$  denotes here the  $L^2$ -norm.

Thus,  $\mathcal{F}$  is a (linear) isometry of  $\mathcal{S}$  onto itself, with respect to the  $L^2$ -norm on  $\mathcal{S}$ . Since  $\mathcal{S}$  is dense in  $L^2$  (recall that  $\mathcal{D} \subset \mathcal{S}$ , and  $\mathcal{D}$  is dense in  $L^2$ ), the operator  $\mathcal{F}$  extends uniquely as a linear isometry of  $L^2$  onto itself. This operator, also denoted  $\mathcal{F}$ , is called *the  $L^2$ -Fourier transform*.

**Example.** Consider the orthonormal sequence  $\{f_k; k \in \mathbb{Z}\}$  in  $L^2(\mathbb{R})$  defined in the second example of Section 8.11. Since  $\mathcal{F}$  is a Hilbert automorphism of  $L^2(\mathbb{R})$ , the sequence  $\{g_k := \mathcal{F}f_k; k \in \mathbb{Z}\}$  is orthonormal in  $L^2(\mathbb{R})$ . The fact that  $f_k$  are also in  $L^1(\mathbb{R})$  allows us to calculate as follows

$$\begin{aligned} g_k(y) &= (1/\sqrt{2\pi}) \int_{\mathbb{R}} e^{-ixy} f_k(x) dx = (1/2\pi) \int_{-\pi}^{\pi} e^{-ix(y-k)} dx \\ &= (-1)^k \frac{\sin \pi y}{\pi(y-k)}. \end{aligned}$$

Note in particular that

$$\int_{\mathbb{R}} \frac{\sin^2(\pi y)}{(\pi y)^2} dy = \|g_0\|_2^2 = 1,$$

that is,  $\int_{\mathbb{R}} \sin^2 t/t^2 dt = \pi$ . Integrating by parts, we have for all  $a < b$  real

$$\int_a^b \sin^2 t/t^2 dt = \int_a^b \sin^2 t d(-1/t) = \sin^2 a/a - \sin^2 b/b + \int_a^b \sin(2t)/t dt.$$

Letting  $a \rightarrow -\infty$  and  $b \rightarrow \infty$ , we see that the integral  $\int_{\mathbb{R}} \sin(2t)/t dt$  converges and has the value  $\pi$ , i.e.,

$$\int_{\mathbb{R}} \frac{\sin t}{t} dt = \pi.$$

This is the so-called *Dirichlet integral*.

If  $g = \mathcal{F}f$  is in the closure of the span of  $\{g_k\}$  in  $L^2(\mathbb{R})$ ,  $f$  is necessarily in the closure of the span of  $\{f_k\}$ , hence vanishing on  $(-\pi, \pi)^c$ ; in particular,  $f$  is also in  $L^1(\mathbb{R})$ , and therefore  $g$  is continuous. Also  $g$  has the unique  $L^2(\mathbb{R})$ -convergent generalized Fourier expansion  $g = \sum_{k \in \mathbb{Z}} a_k g_k$  (equality in  $L^2$ ). Since both  $\{a_k\}$  and  $\{\|g_k\|_{\infty}\}$  are in  $l^2(\mathbb{Z})$ , it follows (by Schwarz' inequality for  $l^2(\mathbb{Z})$ ) that the above series for  $g$  converges (absolutely and) uniformly on  $\mathbb{R}$ ; in particular,  $\sum a_k g_k$  is continuous. Since  $g$  is continuous as well,  $g = \sum a_k g_k$  everywhere, that is,

$$g(y) = \frac{\sin \pi y}{\pi} \sum_k (-1)^k a_k/(y-k).$$

Letting  $y \rightarrow n$  for any given  $n \in \mathbb{Z}$ , we get  $g(n) = a_n$ . Thus

$$g(y) = \frac{\sin \pi y}{\pi} \sum_k (-1)^k g(k)/(y-k).$$

### II.3.3. The dual space $\mathcal{S}'$ .

If  $u \in \mathcal{S}'$ , then  $u|_{\mathcal{D}} \in \mathcal{D}'$  (because the inclusion  $\mathcal{D} \subset \mathcal{S}$  is topological). Moreover, since  $\mathcal{D}$  is dense in  $\mathcal{S}$ ,  $u$  is uniquely determined by its restriction to  $\mathcal{D}$ . The one-to-one map  $u \in \mathcal{S}' \rightarrow u|_{\mathcal{D}} \in \mathcal{D}'$  allows us to identify  $\mathcal{S}'$  as a subspace of



$\mathcal{D}'$ ; its elements are called *temperate distributions*. We also have  $\mathcal{E}' \subset \mathcal{S}'$ :

$$\mathcal{E}' \subset \mathcal{S}' \subset \mathcal{D}',$$

topologically.

Note that  $L^p \subset \mathcal{S}'$  for any  $p \in [1, \infty]$ .

The *Fourier transform of the temperate distribution*  $u$  is defined by

$$\hat{u}(\phi) = u(\hat{\phi}) \quad (\phi \in \mathcal{S}). \quad (20)$$

If  $u$  is a function in  $L^1$ , it follows from (16) and the density of  $\mathcal{S}$  in  $L^1$  that its Fourier transform as a temperate distribution 'is the function  $\hat{u}$ ', the usual  $L^1$ -Fourier transform. Similarly, if  $u \in M(\mathbb{R}^n)$ , that is, if  $u$  is a (regular Borel) complex measure,  $\hat{u}$  'is' the usual *Fourier-Stieltjes transform* of  $u$ , defined by

$$\hat{u}(y) = \int_{\mathbb{R}^n} e^{-ix \cdot y} du(x) \quad (y \in \mathbb{R}^n).$$

In general, (20) defines  $\hat{u}$  as a temperate distribution, since the map  $F : \phi \rightarrow \hat{\phi}$  is a continuous linear map of  $\mathcal{S}$  into itself and  $\hat{u} := u \circ F$ . We shall write  $Fu$  for  $\hat{u}$  (using the same notation for the 'extended' operator);  $F$  is trivially a continuous linear operator on  $\mathcal{S}'$ .

Similarly, we define the (continuous linear) operators  $J$  and  $\mathcal{F}$  on  $\mathcal{S}'$  by

$$(Ju)(\phi) = u(J\phi) \quad (\phi \in \mathcal{S})$$

and  $\mathcal{F} = (2\pi)^{-n/2}F$ .

We have for all  $\phi \in \mathcal{S}$

$$(F^2u)(\phi) = u(F^2\phi) = (2\pi)^n u(J\phi) = (2\pi)^n (Ju)(\phi),$$

that is

$$F^2 = (2\pi)^n J$$

on  $\mathcal{S}'$ . It follows that  $F$  is a continuous automorphism of  $\mathcal{S}'$ ; its inverse is given by the Fourier inversion formula  $F^{-1} = (2\pi)^{-n}JF$  (equivalently,  $\mathcal{F}^{-1} = J\mathcal{F}$ ) on  $\mathcal{S}'$ .

It follows in particular that the restrictions of  $F$  to  $L^1$  and to  $M := M(\mathbb{R}^n)$  are one-to-one. This is the so-called *uniqueness property* of the  $L^1$ -Fourier transform and of the Fourier-Stieltjes transform, respectively.

If  $u \in L^2(\subset \mathcal{S}')$ , then for all  $\phi \in \mathcal{S}$

$$\begin{aligned} |(\mathcal{F}u)(\phi)| &= |u(\mathcal{F}\phi)| = \left| \int u \cdot \mathcal{F}\phi \, dx \right| \\ &\leq \|u\|_2 \|\mathcal{F}\phi\|_2 = \|u\|_2 \|\phi\|_2. \end{aligned}$$

Thus,  $\mathcal{F}u$  is a continuous linear functional on the dense subspace  $\mathcal{S}$  of  $L^2$  with norm  $\leq \|u\|_2$ . It extends uniquely as a continuous linear functional on  $L^2$  with

the same norm. By the ('little') Riesz representation theorem, there exists a unique  $g \in L^2$  such that  $\|g\|_2 \leq \|u\|_2$  and

$$(\mathcal{F}u)(\phi) = \int \phi g \, dx \quad (\phi \in \mathcal{S}).$$

This shows that  $\mathcal{F}u$  'is' the  $L^2$ -function  $g$ , that is,  $\mathcal{F}$  maps  $L^2$  into itself. The identity  $u = \mathcal{F}^2 Ju$  shows that  $\mathcal{F}L^2 = L^2$ . Also  $\|\mathcal{F}u\|_2 = \|g\|_2 \leq \|u\|_2$ , so that

$$\|u\|_2 = \|Ju\|_2 = \|\mathcal{F}^2 u\|_2 \leq \|\mathcal{F}u\|_2,$$

and the equality  $\|\mathcal{F}u\|_2 = \|u\|_2$  follows. This proves that  $\mathcal{F}|_{L^2}$  is a (linear) isometry of  $L^2$  onto itself. Its restriction to the dense subspace  $\mathcal{S}$  of  $L^2$  is the operator  $\mathcal{F}$  originally defined on  $\mathcal{S}$ ; therefore  $\mathcal{F}|_{L^2}$  coincides with the  $L^2$  Fourier transform defined at the end of II.3.2.

The formulae relating the operators  $F$ ,  $D^\alpha$ , and  $M^\beta$  on  $\mathcal{S}$  extend easily to  $\mathcal{S}'$ : if  $u \in \mathcal{S}'$ , then for all  $\phi \in \mathcal{S}$ ,

$$\begin{aligned} [FD^\beta u]\phi &= (D^\beta u)(F\phi) = u((-D)^\beta F\phi) \\ &= u(FM^\beta \phi) = (M^\beta Fu)(\phi), \end{aligned}$$

that is,  $FD^\beta = M^\beta F$  on  $\mathcal{S}'$ . By linearity of  $F$ , it follows that for any polynomial  $P$  on  $\mathbb{R}^n$ ,  $FP(D) = P(M)F$  on  $\mathcal{S}'$ .

**Theorem II.3.4.** *If  $u \in \mathcal{E}'$ ,  $\hat{u}$  'is the function'  $\psi(y) := u(e^{-ix \cdot y})$  ( $y$  is a parameter on the right of the equation), and extends to  $\mathbb{C}^n$  as the entire function  $\psi(z) := u(e^{-ix \cdot z})$ , ( $z \in \mathbb{C}^n$ ). In particular,  $\hat{u} \in \mathcal{E}$ .*

**Proof.**

- (i) *Case  $u \in \mathcal{D}(\subset \mathcal{E}')$ .* Since  $\psi(z) = \int_{\text{supp } u} e^{-ix \cdot z} u(x) \, dx$ , it is clear that  $\psi$  is entire, and coincides with  $\hat{u}$  on  $\mathbb{R}^n$ .
- (ii) *General case.* Let  $\phi$  be as in II.1.1, and consider the regularizations  $u_r := u * \phi_r \in \mathcal{D}$ . By Proposition II.2.12,  $u_r \rightarrow u$  in  $\mathcal{D}'$  and  $\text{supp } u_r \subset \text{supp } u + \{x; |x| \leq 1\} := K$  for all  $0 < r \leq 1$ . Given  $\phi \in \mathcal{E}$ , choose  $\phi' \in \mathcal{D}$  that coincides with  $\phi$  in a neighbourhood of  $K$ . Then for  $0 < r \leq 1$ ,  $u_r(\phi) = u_r(\phi') \rightarrow u(\phi') = u(\phi)$  as  $r \rightarrow 0$ , that is,  $u_r \rightarrow u$  in  $\mathcal{E}'$ , hence also in  $\mathcal{S}'$ . Therefore  $\hat{u}_r \rightarrow \hat{u}$  in  $\mathcal{S}'$  (by continuity of the Fourier transform on  $\mathcal{S}'$ ). By Case (i),  $\hat{u}_r(y) = u_r(e^{-ix \cdot y}) \rightarrow u(e^{-ix \cdot y})$ , since  $u_r \rightarrow u$  in  $\mathcal{E}'$ . More precisely, for any  $z \in \mathbb{C}^n$ ,

$$u_r(e^{-ix \cdot z}) = [(u * \phi_r) * e^{ix \cdot z}](0) = [u * (\phi_r * e^{ix \cdot z})](0).$$

However,

$$\begin{aligned} \phi_r * e^{ix \cdot z} &= r^{-n} \int e^{i(x-y) \cdot z} \phi(y/r) \, dy \\ &= e^{ix \cdot z} \int e^{-it \cdot rz} \phi(t) \, dt = e^{ix \cdot z} \hat{\phi}(rz). \end{aligned}$$

Therefore,

$$u_r(e^{-ix \cdot z}) = \hat{\phi}(rz)(u * e^{ix \cdot z})(0) = \hat{\phi}(rz)u(e^{-ix \cdot z}),$$

and so, for all  $0 < r \leq 1$ ,

$$\begin{aligned} |u_r(e^{-ix \cdot z}) - u(e^{-ix \cdot z})| &= |\hat{\phi}(rz) - 1| |u(e^{-ix \cdot z})| \\ &= \left| \int_{|x| \leq 1} (e^{-ix \cdot rz} - 1) \phi(x) dx \right| |u(e^{-ix \cdot z})| \\ &\leq C \|e^{-ix \cdot z}\|_{k,K} |z| e^{|\Im z| r}, \end{aligned}$$

with the constants  $C, k$  and the compact set  $K$  independent of  $r$ . Consequently  $u_r(e^{-ix \cdot z}) \rightarrow u(e^{-ix \cdot z})$  as  $r \rightarrow 0$ , uniformly with respect to  $z$  on compact subsets of  $\mathbb{C}^n$ . Since  $u_r(e^{-ix \cdot z})$  are entire (cf. Case (i)), it follows that  $u(e^{-ix \cdot z})$  is entire.

In order to verify that  $\hat{u}$  is the function  $u(e^{-ix \cdot y})$ , it suffices to show that

$$\hat{u}(\phi) = \int \phi(y) u(e^{-ix \cdot y}) dy$$

for arbitrary  $\phi \in \mathcal{D}$ , since  $\mathcal{D}$  is dense in  $\mathcal{S}$ . Let  $K$  be the compact support of  $\phi$ , and  $0 < r \leq 1$ . Since  $\phi(y) \hat{u}_r(y) \rightarrow \phi(y) u(e^{-ix \cdot y})$  uniformly on  $K$  as  $r \rightarrow 0$ , we get

$$\hat{u}(\phi) = \lim_{r \rightarrow 0} \hat{u}_r(\phi) = \lim_r \int_K \phi(y) \hat{u}_r(y) dy = \int_K \phi(y) u(e^{-ix \cdot y}) dy,$$

as desired.  $\square$

The entire function  $u(e^{-ix \cdot z})$  will be denoted  $\hat{u}(z)$ ; since the distribution  $\hat{u}$  'is' the above function restricted to  $\mathbb{R}^n$  (by Theorem II.3.4), the notation is justified. The function  $\hat{u}(z)$  is called the *Fourier-Laplace transform* of  $u \in \mathcal{E}'$ .

**Theorem II.3.5.** *If  $u \in \mathcal{E}'$  and  $v \in \mathcal{S}'$ , then  $u * v \in \mathcal{S}'$  and  $F(u * v) = \hat{u} \hat{v}$ .*

**Proof.** By Theorem II.3.4,  $\hat{u} \in \mathcal{E}$ , and therefore the product  $\hat{u} \hat{v}$  makes sense as a distribution (cf. (4), II.2.9). We prove next that  $u * v \in \mathcal{S}'$ ; then  $F(u * v)$  will make sense as well (and belong to  $\mathcal{S}'$ ), and it will remain to verify the identity.

For all  $\phi \in \mathcal{D}$ ,

$$\begin{aligned} (u * v)(\phi) &= [(u * v) * J\phi](0) = [u * (v * J\phi)](0) \\ &= u(J(v * J\phi)) = (Ju)(v * J\phi). \end{aligned} \tag{21}$$

Since  $Ju \in \mathcal{E}'$ , there exist  $K \subset \mathbb{R}^n$  compact and constants  $C > 0$  and  $k \in \mathbb{N} \cup \{0\}$  (all independent of  $\phi$ ) such that

$$|(u * v)(\phi)| \leq C \|v * J\phi\|_{k,K} \tag{22}$$

(cf. (7) in II.2.10).

For each multi-index  $\alpha$ ,  $D^\alpha v \in \mathcal{S}'$ . Therefore, there exist a constant  $C' > 0$  and multi-indices  $\beta, \gamma$  (all independent of  $\phi$  and  $x$ ), such that

$$|(D^\alpha v * J\phi)(x)| = |(D^\alpha v)(\phi(y - x))| \leq C' \sup_y |y^\beta (D^\gamma \phi)(y)|. \tag{23}$$

By (22) and (23),  $|(u * v)(\phi)|$  can be estimated by semi-norms of  $\phi$  in  $\mathcal{S}$  (for all  $\phi \in \mathcal{D}$ ). Since  $\mathcal{D}$  is dense in  $\mathcal{S}$ ,  $u * v$  extends uniquely as a continuous linear functional on  $\mathcal{S}$ . Thus,  $u * v \in \mathcal{S}'$ .

We shall verify now the identity of the theorem, first in the special case  $u \in \mathcal{D}$ . Then  $\hat{u} \in \mathcal{S}$ , and therefore  $\hat{u}\hat{v} \in \mathcal{S}'$  (by a simple application of Leibnitz' formula).

By (21), if  $\psi \in \mathcal{S}$  is such that  $\hat{\psi} \in \mathcal{D}$ ,

$$[F(u * v)](\psi) = (u * v)(\hat{\psi}) = (v * u)(\hat{\psi}) = v(J(u * J\hat{\psi})).$$

For  $f, g \in \mathcal{D}$ , it follows from the integral definition of the convolution that  $J(f * g) = (Jf) * (Jg)$ . Then, by (12) and (15) in II.3.2,

$$\begin{aligned} [F(u * v)](\psi) &= v((Ju) * \hat{\psi}) = v[(2\pi)^{-n}(F^2u) * F\psi]) \\ &= v(F(\hat{u}\psi)) = \hat{v}(\hat{u}\psi) = (\hat{u}\hat{v})(\psi), \end{aligned}$$

where the last equality follows from the definition of the product of the distribution  $\hat{v}$  by the function  $\hat{u} \in \mathcal{S} \subset \mathcal{E}$ . Hence  $F(u * v) = \hat{u}\hat{v}$  on the set  $\mathcal{S}_0 = \{\psi \in \mathcal{S}; \hat{\psi} \in \mathcal{D}\}$ . For  $\psi \in \mathcal{S}$  arbitrary, since  $\hat{\psi} \in \mathcal{S}$  and  $\mathcal{D}$  is dense in  $\mathcal{S}$ , there exists a sequence  $\phi_k \in \mathcal{D}$  such that  $\phi_k \rightarrow \hat{\psi}$  in  $\mathcal{S}$ . Let  $\psi_k = F^{-1}\phi_k$ . Then  $\psi_k \in \mathcal{S}$ ,  $F\psi_k = \phi_k \in \mathcal{D}$  (that is,  $\psi_k \in \mathcal{S}_0$ ), and  $\psi_k \rightarrow \psi$  by continuity of  $F^{-1}$  on  $\mathcal{S}$  (i.e.,  $\mathcal{S}_0$  is dense in  $\mathcal{S}$ ).

Since  $F(u * v)$  and  $\hat{u}\hat{v}$  are in  $\mathcal{S}'$  (as observed above) and coincide on  $\mathcal{S}_0$ , they are indeed equal.

Consider next the general case  $u \in \mathcal{E}'$ . If  $\phi \in \mathcal{D}$ ,  $\phi * u \in \mathcal{D}$  and  $v \in \mathcal{S}'$ , and also  $\phi \in \mathcal{D}$  and  $u * v \in \mathcal{S}'$ . Applying the special case to these two pairs, we get

$$\begin{aligned} (\hat{\phi}\hat{u})\hat{v} &= [F(\phi * u)]\hat{v} = F[(\phi * u) * v] \\ &= F[\phi * (u * v)] = \hat{\phi}F(u * v). \end{aligned}$$

For each point  $y$ , we can choose  $\phi \in \mathcal{D}$  such that  $\hat{\phi}(y) \neq 0$ ; it follows that  $\hat{u}\hat{v} = F(u * v)$ .  $\square$

The next theorem characterizes Fourier–Laplace transforms of distributions with compact support (cf. Theorem II.3.4).

**Theorem II.3.6 (Paley–Wiener–Schwartz).**

- (i) The entire function  $f$  on  $\mathbb{C}^n$  is the Fourier–Laplace transform of a distribution  $u$  with support in the ball  $S_A := \{x \in \mathbb{R}^n; |x| \leq A\}$  iff there exist a constant  $C > 0$  and a non-negative integer  $m$  such that

$$|f(z)| \leq C(1 + |z|)^m e^{A|\Im z|} \quad (z \in \mathbb{C}^n).$$

- (ii) The entire function  $f$  is the Fourier–Laplace transform of a function  $u \in \mathcal{D}(S_A)$  iff for each  $m$ , there exists a positive constant  $C = C(m)$  such that

$$|f(z)| \leq C(m)(1 + |z|)^{-m} e^{A|\Im z|} \quad (z \in \mathbb{C}^n).$$

**Proof.** Let  $r > 0$ , and suppose  $u \in \mathcal{E}'$  with  $\text{supp } u \subset S_A$ . By Theorem II.2.8 with  $\omega = \{x; |x| < A + r\}$ , there exists  $g \in L^\infty(\omega)$  such that  $u|_\omega = D_1^s \cdots D_n^s g$ . We may take  $g$  and  $s$  independent of  $r$  for all  $0 < r \leq 1$ . Extend  $g$  to  $\mathbb{R}^n$  by setting  $g = 0$  for  $|x| \geq A + 1$  (then of course  $\|g\|_{L^1} < \infty$ ). Since  $\text{supp } u \subset \omega$ , we have  $u = D_1^s \cdots D_n^s g$ , and therefore

$$\begin{aligned} |\hat{u}(z)| &= \left| \int_\omega e^{-ix \cdot z} D_1^s \cdots D_n^s g(x) dx \right| = \left| \int_\omega D_1^s \cdots D_n^s (e^{-ix \cdot z} g(x)) dx \right| \\ &= \left| \int_\omega (-1)^{sn} (z_1 \cdots z_n)^s e^{-ix \cdot z} g(x) dx \right| \leq (|z_1| \cdots |z_n|)^s \|g\|_{L^1} \sup_\omega e^{x \cdot \Im z} \\ &\leq C(1 + |z|)^m e^{(A+r)|\Im z|}, \end{aligned}$$

for any constant  $C > \|g\|_{L^1}$ ,  $m = sn$ , and  $r \leq 1$  (since  $|x \cdot \Im z| \leq |x| |\Im z| \leq (A+r)|\Im z|$  on  $\omega$ , by Schwarz's inequality). Letting  $r \rightarrow 0$ , we obtain the necessity of the estimate in (i).

If  $u \in \mathcal{D}$  with support in  $S_A$ , we have for any multi-index  $\beta$  and  $z \in \mathbb{C}^n$

$$|z^\beta \hat{u}(z)| = |(FD^\beta u)(z)| = \left| \int_{S_A} e^{-ix \cdot z} (D^\beta u)(x) dx \right| \leq \|D^\beta u\|_{L^1} e^{A|\Im z|},$$

and the necessity of the estimates in (ii) follows.

Suppose next that the entire function  $f$  satisfies the estimates in (ii). In particular, its restriction to  $\mathbb{R}^n$  is in  $L^1$ , and we may define  $u = (2\pi)^{-n/2} \mathcal{F}f|_{\mathbb{R}^n}$ . The estimates in (ii) show that  $y^\alpha f(y) \in L^1(\mathbb{R}^n)$  for all multi-indices  $\alpha$ , and therefore  $D^\alpha$  may be applied to the integral defining  $u$  under the integration sign; in particular,  $u \in \mathcal{E}$ .

The estimates in (ii) show also that the integral defining  $u$  can be shifted (by Cauchy's integral theorem) to  $\mathbb{R}^n + it$ , with  $t \in \mathbb{R}^n$  fixed (but arbitrary). Therefore

$$|u(x)| \leq C(m) \exp[A|t| - x \cdot t] \int (1 + |y|)^{-m} dy$$

for all  $x \in \mathbb{R}^n$  and  $m \in \mathbb{N}$ . Fix  $m$  so that the above integral converges, and choose  $t = \lambda x$  ( $\lambda > 0$ ). Then for a suitable constant  $C'$  and  $|x| > A$ ,  $|u(x)| \leq C' \exp[-\lambda|x|(|x| - A)] \rightarrow 0$  as  $\lambda \rightarrow \infty$ . This shows that  $\text{supp } u \subset S_A$ , and so  $u \in \mathcal{D}(S_A)$ . But then its Fourier-Laplace transform is entire, and coincides with the entire function  $f$  on  $\mathbb{R}^n$  (hence on  $\mathbb{C}^n$ ), by the Fourier inversion formula.

Finally, suppose the estimate in (i) is satisfied. Then  $f|_{\mathbb{R}^n} \in \mathcal{S}'$ , and therefore  $f|_{\mathbb{R}^n} = \hat{u}$  for a unique  $u \in \mathcal{S}'$ . It remains to show that  $\text{supp } u \subset S_A$ . Let  $\phi$  be as in II.1.1, and let  $u_r := u * \phi_r$  be the corresponding regularization of  $u$ . By Theorem II.3.5,  $\hat{u}_r = \hat{u} \hat{\phi}_r$ . Since  $\phi_r \in \mathcal{D}(S_r)$ , it follows from the necessity part of (ii) that  $|\hat{\phi}_r(z)| \leq C(m)(1 + |z|)^{-m} e^{r|\Im z|}$  for all  $m$ . Therefore, by the estimate in (i),

$$|f(z) \hat{\phi}_r(z)| \leq C'(k)(1 + |z|)^{-k} e^{(A+r)|\Im z|}$$

for all integers  $k$ . By the sufficiency part of (ii), the entire function  $f(z) \hat{\phi}_r(z)$  is the Fourier-Laplace transform of some  $\psi_r \in \mathcal{D}(S_{A+r})$ . Since  $F$  is injective on

$\mathcal{S}'$ , we conclude (by restricting to  $\mathbb{R}^n$ ) that the distribution  $u_r$  'is the function'  $\psi_r$ . In particular,  $\text{supp } u_r \subset S_{A+r}$  for all  $r > 0$ . If  $\chi \in \mathcal{D}$  has support in (the open set)  $S_A^c$ , there exists  $r_0 > 0$  such that  $\text{supp } \chi \subset S_{A+r_0}^c$ ; then for all  $0 < r \leq r_0$ ,  $u_r(\chi) = 0$  because the supports of  $u_r$  and  $\chi$  are contained in the disjoint sets  $S_{A+r}$  and  $S_{A+r_0}^c$  (respectively). Letting  $r \rightarrow 0$ , it follows that  $u(\chi) = 0$ , and we conclude that  $u$  has support in  $S_A$ .  $\square$

**Example.** Consider the orthonormal sequences  $\{f_k; k \in \mathbb{Z}\}$  and  $\{g_k; k \in \mathbb{Z}\}$  in  $L^2(\mathbb{R})$  defined in the example at the end of II.3.2. We saw that  $f \in L^2(\mathbb{R})$  belongs to the closure of the span of  $\{f_k\}$  in  $L^2(\mathbb{R})$  iff it vanishes in  $(-\pi, \pi)^c$ . Since  $\mathcal{F}$  is a Hilbert isomorphism of  $L^2(\mathbb{R})$  onto itself,  $g := \mathcal{F}f \in L^2(\mathbb{R})$  belongs to the closure of the span of  $\{g_k\}$  iff it extends to  $\mathbb{C}$  as an entire function of exponential type  $\leq \pi$  (by Theorem II.3.6). The expansion we found for  $g$  in the above example extends to  $\mathbb{C}$ :

$$g(z) = (1/\pi) \sin \pi z \sum_k (-1)^k g(k)/(z - k) \quad (z \in \mathbb{C}), \quad (24)$$

where the series converges uniformly in  $|\Re z| \leq r$ , for each  $r$ . We proved that any entire function  $g$  of exponential type  $\leq \pi$ , whose restriction to  $\mathbb{R}$  belongs to  $L^2(\mathbb{R})$ , admits the expansion (24).

Suppose  $h$  is an entire function of exponential type  $\leq \pi$ , whose restriction to  $\mathbb{R}$  is bounded. Let  $g(z) := [h(z) - h(0)]/z$  for  $z \neq 0$  and  $g(0) = h'(0)$ . Then  $g$  is entire of exponential type  $\leq \pi$ , and  $g|_{\mathbb{R}} \in L^2(\mathbb{R})$ . Applying (24) to  $g$ , we obtain

$$h(z) = h(0) + (1/\pi) \sin \pi z \left( h'(0) + \sum_{k \neq 0} (-1)^k [h(k) - h(0)][1/k + 1/(z - k)] \right). \quad (25)$$

The series  $s(z)$  in (25) can be differentiated term-by-term (because the series thus obtained converges uniformly in any strip  $|\Re z| \leq r$ ). We then obtain

$$h'(z) = \cos \pi z (h'(0) + s(z)) + (1/\pi) \sin \pi z \sum_{k \neq 0} (-1)^{k-1} [h(k) - h(0)]/(z - k)^2.$$

In particular,

$$\begin{aligned} h'(1/2) &= (1/\pi) \sum_{k \neq 0} (-1)^{k-1} [h(k) - h(0)]/(k - 1/2)^2 \\ &= (4/\pi) \sum_{k \in \mathbb{Z}} (-1)^{k-1} \frac{h(k) - h(0)}{(2k - 1)^2}. \end{aligned}$$

Since  $\sum_{k \in \mathbb{Z}} (-1)^{k-1}/(2k - 1)^2 = 0$ , we can rewrite the last formula in the form

$$h'(1/2) = (4/\pi) \sum_{k \in \mathbb{Z}} (-1)^{k-1} \frac{h(k)}{(2k - 1)^2}. \quad (26)$$

For  $t \in \mathbb{R}$  fixed, the function  $\tilde{h}(z) := h(z + t - 1/2)$  is entire of exponential type  $\leq \pi$ ,  $\sup_{\mathbb{R}} |\tilde{h}| = \sup_{\mathbb{R}} |h| := M < \infty$ , and  $\tilde{h}'(1/2) = h'(t)$ . Therefore by (26)

applied to  $\tilde{h}$ ,

$$h'(t) = (4/\pi) \sum_{k \in \mathbb{Z}} (-1)^{k-1} \frac{h(t+k-1/2)}{(2k-1)^2}. \quad (27)$$

Hence (cf. example at the end of Terminology 8.11)

$$|h'(t)| \leq (4/\pi)M \sum_{k \in \mathbb{Z}} \frac{1}{(2k-1)^2} = (4/\pi)M(\pi^2/4) = \pi M. \quad (28)$$

Thus, considering  $h$  restricted to  $\mathbb{R}$ ,

$$\|h'\|_\infty \leq \pi \|h\|_\infty. \quad (29)$$

Let  $1 \leq p < \infty$ , and let  $q$  be its conjugate exponent. For any simple measurable function  $\phi$  on  $\mathbb{R}$  with  $\|\phi\|_q = 1$ , we have by (27) and Holder's inequality

$$\begin{aligned} \left| \int_{\mathbb{R}} h' \phi \, dt \right| &= (4/\pi) \left| \sum_{k \in \mathbb{Z}} \frac{(-1)^{k-1}}{(2k-1)^2} \int h(t+k-1/2) \phi(t) \, dt \right| \\ &\leq (4/\pi) \sum_{k \in \mathbb{Z}} \frac{1}{(2k-1)^2} \|h\|_p \|\phi\|_q = \pi \|h\|_p. \end{aligned}$$

Taking the supremum over all such functions  $\phi$ , it follows that

$$\|h'\|_p \leq \pi \|h\|_p. \quad (30)$$

(For  $p = 1$ , (30) follows directly from (27): for any real numbers  $a < b$ ,

$$\int_a^b |h'(t)| \, dt \leq (4/\pi) \sum_{k \in \mathbb{Z}} \frac{1}{(2k-1)^2} \|h\|_1 = \pi \|h\|_1,$$

and (30) for  $p = 1$  is obtained by letting  $a \rightarrow -\infty$  and  $b \rightarrow \infty$ .)

If  $f$  is an entire function of exponential type  $\leq \nu > 0$  and is bounded on  $\mathbb{R}$ , the function  $h(z) := f(\pi z/\nu)$  is entire of exponential type  $\leq \pi$ ;  $\|h\|_\infty = \|f\|_\infty$  (norms in  $L^\infty(\mathbb{R})$ ); and  $h'(t) = (\pi/\nu)f'(\pi t/\nu)$ . A simple calculation starting from (30) for  $h$  shows that

$$\|f'\|_p \leq \nu \|f\|_p \quad (31)$$

for all  $p \in [1, \infty]$ . This is *Bernstein's inequality* (for  $f$  entire of exponential type  $\leq \nu$ , that is bounded on  $\mathbb{R}$ ).

## The spaces $\mathcal{W}_{p,k}$

### II.3.7. Temperate weights.

A (temperate) weight on  $\mathbb{R}^n$  is a *positive* function  $k$  on  $\mathbb{R}^n$  such that

$$\frac{k(x+y)}{k(y)} \leq (1+C|x|)^m \quad (x, y \in \mathbb{R}^n) \quad (32)$$

for some constants  $C, m > 0$ .

By (32),

$$(1 + C|x|)^{-m} \leq \frac{k(x+y)}{k(y)} \leq (1 + C|x|)^m \quad (x, y \in \mathbb{R}^n), \quad (33)$$

and it follows that  $k$  is *continuous*, satisfies the estimate

$$k(0)(1 + C|x|)^{-m} \leq k(x) \leq k(0)(1 + C|x|)^m \quad (x \in \mathbb{R}^n), \quad (34)$$

and  $1/k$  is also a weight.

For any real  $s$ ,  $k^s$  is a weight (trivial for  $s > 0$ , and since  $1/k$  is a weight, the conclusion follows for  $s < 0$  as well).

An elementary calculation shows that  $1 + |x|^2$  is a weight; therefore,  $k_s(x) := (1 + |x|^2)^{s/2}$  is a weight for any real  $s$ .

Sums and product of weights are weights. For any weight  $k$ , set

$$\underline{k}(x) := \sup_y \frac{k(x+y)}{k(y)}, \quad (35)$$

so that, by definition,

$$\underline{k}(x) \leq (1 + C|x|)^m \quad \text{and} \quad k(x+y) \leq \underline{k}(x)k(y). \quad (36)$$

Also

$$\underline{k}(x+y) \leq \underline{k}(x)\underline{k}(y). \quad (37)$$

By (36) and (37),  $\underline{k}$  is a weight with the additional ‘normal’ properties (37) and

$$1 = \underline{k}(0) \leq \underline{k}(x). \quad (38)$$

( $\underline{k}(0) = 1$  by (35); then by (37) and (36), for all  $r = 1, 2, \dots$ ,

$$1 = \underline{k}(0) = \underline{k}(rx - rx) \leq \underline{k}(x)^r \underline{k}(-rx) \leq \underline{k}(x)^r (1 + Cr|x|)^m;$$

taking the  $r$ th root and letting  $r \rightarrow \infty$ , we get  $1 \leq \underline{k}(x)$ .)

Given a weight  $k$  and  $t > 0$ , define

$$k^t(x) := \sup_y k(x-y) \exp(-t|y|) = \sup_y \exp(-t|x-y|)k(y).$$

( $x, y$  range in  $\mathbb{R}^n$ .)

We have

$$k(x) \leq k^t(x) \leq \sup_y (1 + C|y|)^m k(x) \exp(-t|y|) \leq C_t k(x),$$

that is,

$$1 \leq \frac{k^t(x)}{k(x)} \leq C_t \quad (x \in \mathbb{R}^n),$$

where  $C_t$  is a constant depending on  $t$ .



Also

$$\begin{aligned} k^t(x+x') &= \sup_y k(x+x'-y) \exp(-t|y|) \\ &\leq \sup_y (1+C|x'|)^m k(x-y) \exp(-t|y|) = (1+C|x'|)^m k^t(x), \end{aligned}$$

that is,  $k^t$  is a weight (with the constants  $C, m$  of  $k$ , whence independent of  $t$ ). By the last inequality,

$$(1 \leq) \underline{k}^t(x) \leq (1+C|x|)^m.$$

Since

$$k^t(x+x') = \sup_y \exp(-t|x+x'-y|) k(y) \leq e^{t|x'|} \sup_y \exp(-t|x-y|) k(y) = e^{t|x'|} k^t(x),$$

therefore

$$(1 \leq) \underline{k}^t(x) \leq e^{t|x|}.$$

In particular,  $\underline{k}^t \rightarrow 1$  as  $t \rightarrow 0+$ , uniformly on compact subsets of  $\mathbb{R}^n$ .

A weight associated with the differential operator  $P(D)$  (for any polynomial  $P$  on  $\mathbb{R}^n$ ) is defined by

$$k_P := \left[ \sum_{\alpha} |P^{(\alpha)}|^2 \right]^{1/2}, \quad (39)$$

where the (finite) sum extends over all multi-indices  $\alpha$ . The estimate (32) follows from Taylor's formula and Schwarz's inequality:

$$\begin{aligned} k_P^2(x+y) &= \sum_{\alpha} \left| \sum_{|\beta| \leq m} P^{(\alpha+\beta)}(y) x^{\beta} / \beta! \right|^2 \\ &\leq \sum_{\alpha} \sum_{\beta} |P^{(\alpha+\beta)}(y)|^2 \sum_{|\beta| \leq m} |x^{\beta} / \beta!|^2 \\ &\leq k_P^2(y) (1+C|x|)^{2m}, \end{aligned}$$

where  $m = \deg P$ .

Extending the sum in (39) over multi-indices  $\alpha \neq 0$  only, we get a weight  $k'_P$  (same verification!), that will also play a role in the sequel.

### II.3.8. Weighted $L^p$ -spaces.

For any (temperate) weight  $k$  and  $p \in [1, \infty]$ , consider the normed space

$$L_{p,k} := (1/k)L^p = \{f; kf \in L^p\}$$

(where  $L^p := L^p(\mathbb{R}^n)$ ), with the natural norm  $\|f\|_{L_{p,k}} = \|kf\|_p$  ( $\|\cdot\|_p$  denotes the  $L^p$ -norm). One verifies easily that  $L_{p,k}$  is a Banach space for all  $p \in [0, \infty]$ , and  $(L_{p,k})^*$  is isomorphic and isometric to  $L_{q,1/k}$  for  $1 \leq p < \infty$  ( $q$  denotes the

conjugate exponent of  $p$ ): if  $\Lambda \in (L_{p,k})^*$ , there exists a unique  $g \in L_{q,1/k}$  such that

$$\Lambda f = \int f g dy \quad (f \in L_{p,k})$$

and

$$\|\Lambda\| = \|g\|_{L_{q,1/k}}.$$

By (34),  $\mathcal{S} \subset L_{p,k}$  topologically.

Given  $f \in L_{p,k}$ , Holder's inequality shows that

$$\left| \int \phi f dx \right| \leq \|f\|_{L_{p,k}} \|\phi\|_{L_{q,1/k}} \quad (\phi \in \mathcal{S}). \quad (40)$$

Since  $\mathcal{S} \subset L_{q,1/k}$  topologically, it follows from (40) that the map  $\phi \rightarrow \int \phi f dx$  is continuous on  $\mathcal{S}$ , and belongs therefore to  $\mathcal{S}'$ . With the usual identification, this means that  $L_{p,k} \subset \mathcal{S}'$ , and it follows also from (40) that the inclusion is topological (if  $f_j \rightarrow 0$  in  $L_{p,k}$ , then  $\int \phi f_j dx \rightarrow 0$  by (40), for all  $\phi \in \mathcal{S}$ , that is,  $f_j \rightarrow 0$  in  $\mathcal{S}'$ ). We showed therefore that

$$\mathcal{S} \subset L_{p,k} \subset \mathcal{S}' \quad (41)$$

topologically.

### II.3.9. The spaces $\mathcal{W}_{p,k}$ .

Let

$$\mathcal{F} : u \in \mathcal{S}' \rightarrow \mathcal{F}u := (2\pi)^{-n/2} \hat{u} \in \mathcal{S}',$$

and consider the normed space (for  $p, k$  given as before)

$$\mathcal{W}_{p,k} := \mathcal{F}^{-1} L_{p,k} = \{u \in \mathcal{S}'; \mathcal{F}u \in L_{p,k}\} \quad (42)$$

with the norm

$$\|u\|_{p,k} := \|\mathcal{F}u\|_{L_{p,k}} = \|k\mathcal{F}u\|_p \quad (u \in \mathcal{W}_{p,k}). \quad (43)$$

Note that for any  $t > 0$ ,

$$\mathcal{W}_{p,k} = \mathcal{W}_{p,k^t}$$

(because of the inequality  $1 \leq k^t/k \leq C_t$ , cf. II.3.7).

By definition,  $\mathcal{F} : \mathcal{W}_{p,k} \rightarrow L_{p,k}$  is a (linear) surjective isometry, and therefore  $\mathcal{W}_{p,k}$  is a Banach space. Since  $\mathcal{F}^{-1}$  is a continuous automorphism of both  $\mathcal{S}$  and  $\mathcal{S}'$ , it follows from (41) in II.3.8 (and the said isometry) that

$$\mathcal{S} \subset \mathcal{W}_{p,k} \subset \mathcal{S}' \quad (44)$$

topologically.

Fix  $\phi$  as in II.1.1, and consider the regularizations  $u_r = u * \phi_r \in \mathcal{E}$  of  $u \in \mathcal{W}_{p,k}$ ,  $p < \infty$ . As  $r \rightarrow 0+$ ,  $k(x)\hat{u}_r(x) = k(x)\hat{u}(x)\hat{\phi}(rx) \rightarrow k(x)\hat{u}(x)$  pointwise, and  $|k\hat{u}_r| \leq |k\hat{u}| \in L^p$ ; therefore  $k\hat{u}_r \rightarrow k\hat{u}$  in  $L^p$ , that is,  $u_r \rightarrow u$  in  $\mathcal{W}_{p,k}$ . One verifies easily that  $u_r \in \mathcal{S}$ , and consequently  $\mathcal{S}$  is dense in  $\mathcal{W}_{p,k}$ . Since  $\mathcal{D}$  is dense in  $\mathcal{S}$ , and  $\mathcal{S}$  is topologically included in  $\mathcal{W}_{p,k}$ , we conclude that  $\mathcal{D}$  is dense in

$\mathcal{W}_{p,k}$  (and so  $\mathcal{W}_{p,k}$  is the completion of  $\mathcal{D}$  with respect to the norm  $\|\cdot\|_{p,k}$ ) for any  $1 \leq p < \infty$  and any weight  $k$ . The special space  $\mathcal{W}_{2,k_s}$  is called *Sobolev's space*, and is usually denoted  $\mathcal{H}^s$ .

Let  $L \in \mathcal{W}_{p,k}^*$  (for some  $1 \leq p < \infty$ ). Since  $\mathcal{F}$  is a linear isometry of  $\mathcal{W}_{p,k}$  onto  $L_{p,k}$ , the map  $\Lambda = L\mathcal{F}^{-1}$  is a continuous linear functional on  $L_{p,k}$  with norm  $\|L\|$ . By the preceding characterization of  $L_{p,k}^*$ , there exists a unique  $g \in L_{q,1/k}$  such that  $\|g\|_{L_{q,1/k}} = \|L\|$  and  $\Lambda f = \int f g dx$  for all  $f \in L_{p,k}$ . Define  $v = \mathcal{F}^{-1}g$ . For all  $u \in \mathcal{W}_{p,k}$ , denoting  $f = \mathcal{F}u$  ( $\in L_{p,k}$ ), we have  $\|v\|_{q,1/k} = \|L\|$  and

$$Lu = L\mathcal{F}^{-1}f = \Lambda f = \int (\mathcal{F}u)(\mathcal{F}v) dx. \quad (45)$$

The continuous functional  $L$  is uniquely determined by its restriction to the dense subspace  $\mathcal{S}$  of  $\mathcal{W}_{p,k}$ . For  $u \in \mathcal{S}$ , we may write (45) in the form

$$Lu = v(\mathcal{F}^2 u) = v(Ju) = (Jv)(u) \quad (u \in \mathcal{S}), \quad (46)$$

that is,  $L|_{\mathcal{S}} = Jv$ . Conversely, any  $v \in \mathcal{W}_{q,1/k}$  determines through (45) an element  $L \in \mathcal{W}_{p,k}^*$  such that  $\|L\| = \|v\|_{q,1/k}$  (and  $L|_{\mathcal{S}} = Jv$ ). We conclude that  $\mathcal{W}_{p,k}^*$  is isometrically isomorphic with  $\mathcal{W}_{q,1/k}$ . In particular,  $(\mathcal{H}^s)^*$  is isometrically isomorphic with  $\mathcal{H}^{-s}$ .

If the distribution  $u \in \mathcal{W}_{p,k}$  has compact support, and  $v \in \mathcal{W}_{\infty,k'}$ , then by Theorem II.3.5,  $u * v \in \mathcal{S}'$  and

$$\begin{aligned} \|u * v\|_{p,kk'} &= \|kk' \mathcal{F}(u * v)\|_p = (2\pi)^{n/2} \|(k\mathcal{F}u)(k'\mathcal{F}v)\|_p \\ &\leq (2\pi)^{n/2} \|k\mathcal{F}u\|_p \|k'\mathcal{F}v\|_{\infty} = (2\pi)^{n/2} \|u\|_{p,k} \|v\|_{\infty,k'}. \end{aligned}$$

In particular,

$$(\mathcal{W}_{p,k} \cap \mathcal{E}') * \mathcal{W}_{\infty,k'} \subset \mathcal{W}_{p,kk'}. \quad (47)$$

Let  $P$  be any polynomial on  $\mathbb{R}^n$ . We have  $P(D)\delta \in \mathcal{E}' \subset \mathcal{S}'$ , and

$$\mathcal{F}P(D)\delta = P(M)\mathcal{F}\delta = (2\pi)^{-n/2}P(M)1 = (2\pi)^{-n/2}P.$$

Thus

$$\|P(D)\delta\|_{\infty,k'} = (2\pi)^{-n/2} \|k'P\|_{\infty} < \infty,$$

for any weight  $k'$  such that  $k'P$  is bounded (we may take for example  $k' = 1/k_P$ , or  $k' = k_s$  with  $s \leq -m$ , where  $m = \deg P$ ).

Thus  $P(D)\delta \in \mathcal{W}_{\infty,k'}$  (for such  $k'$ ), and for any  $u \in \mathcal{W}_{p,k}$ ,  $P(D)u = (P(D)\delta) * u \in \mathcal{W}_{p,kk'}$ .

Formally stated, for any weight  $k'$  such that  $k'P$  is bounded,

$$P(D)\mathcal{W}_{p,k} \subset \mathcal{W}_{p,kk'}. \quad (48)$$

The above calculations show also that

$$\|P(D)u\|_{p,kk'} \leq \|k'P\|_{\infty} \|u\|_{p,k} \quad (u \in \mathcal{W}_{p,k}), \quad (49)$$

that is,  $P(D)$  is a continuous (linear) map of  $\mathcal{W}_{p,k}$  into  $\mathcal{W}_{p,kk'}$ .

If  $u \in \mathcal{W}_{p,k}$  and  $\phi \in \mathcal{D}$ , then  $\hat{u} \in L_{p,k}$  and  $\hat{\phi} \in \mathcal{S} \subset L_{q,1/k}$ , so that the convolution  $\hat{\phi} * \hat{u}$  makes sense as a usual integral. On the other hand,  $\phi u$  is well defined and belongs to  $\mathcal{S}'$ . Using Theorem II.3.5 and the Fourier inversion formula on  $\mathcal{S}'$ , we see that

$$(2\pi)^{n/2} \mathcal{F}(\phi u) = (\mathcal{F}\phi) * (\mathcal{F}u).$$

Hence (since  $k(x) \leq \underline{k}(x-y)k(y)$ )

$$\begin{aligned} (2\pi)^{n/2} \|\phi u\|_{p,k} &= \|k(\mathcal{F}\phi) * (\mathcal{F}u)\|_p \\ &\leq \|(\underline{k}|\mathcal{F}\phi|) * (k|\mathcal{F}u|)\|_p \leq \|\underline{k}\mathcal{F}\phi\|_1 \|k\mathcal{F}u\|_p, \end{aligned}$$

that is,

$$\|\phi u\|_{p,k} \leq (2\pi)^{-n/2} \|\phi\|_{1,\underline{k}} \|u\|_{p,k}. \quad (50)$$

Since  $\mathcal{D}$  is dense in  $\mathcal{S}$  and  $\mathcal{S} \subset \mathcal{W}_{1,\underline{k}}$ , it follows from (50) that the multiplication operator  $\phi \in \mathcal{D} \rightarrow \phi u \in \mathcal{W}_{p,k}$  extends uniquely as an operator from  $\mathcal{S}$  to  $\mathcal{W}_{p,k}$  (same notation!), that is,  $\mathcal{S}\mathcal{W}_{p,k} \subset \mathcal{W}_{p,k}$ , and (50) is valid for all  $\phi \in \mathcal{S}$  and  $u \in \mathcal{W}_{p,k}$ .

Apply (50) to the weights  $k^t$  associated with  $k$  (cf. II.3.7). We have (for any  $\phi \in \mathcal{S}$ )

$$\|\phi\|_{1,\underline{k}^t} = \int |\underline{k}^t \mathcal{F}\phi| dx.$$

As  $t \rightarrow 0+$ , the integrand converges pointwise to  $\mathcal{F}\phi$ , and are dominated by  $(1 + C|x|)^m |\mathcal{F}\phi| \in L^1$  (with  $C, m$  independent of  $t$ ). By Lebesgue's dominated convergence theorem, the integral tends to  $\|\mathcal{F}\phi\|_1 := \|\phi\|_{1,1}$ . There exists therefore  $t_0 > 0$  (depending on  $\phi$ ) such that  $\|\phi\|_{1,\underline{k}^t} \leq 2\|\phi\|_{1,1}$  for all  $t < t_0$ . Hence by (50)

$$\|\phi u\|_{p,k^t} \leq 2(2\pi)^{-n/2} \|\phi\|_{1,1} \|u\|_{p,k^t} \quad (51)$$

for all  $0 < t < t_0$ ,  $\phi \in \mathcal{S}$ , and  $u \in \mathcal{W}_{p,k} = \mathcal{W}_{p,k^t}$ , with  $t_0$  depending on  $\phi$ . This inequality will be used in the proof of Theorem II.7.2.

Let  $j$  be a non-negative integer. If  $|y|^j \in L_{q,1/k}$  for some weight  $k$  and some  $1 \leq q \leq \infty$ , then  $y^\alpha \in L_{q,1/k}$  for all multi-indices  $\alpha$  with  $|\alpha| \leq j$ . Consequently, for any  $u \in \mathcal{W}_{p,k}$  (with  $p$  conjugate to  $q$ ),  $y^\alpha \hat{u}(y) \in L^1$ . By the Fourier inversion formula,  $u(x) = (2\pi)^{-n} \int e^{ix \cdot y} \hat{u}(y) dy$ , and the integrals obtained by formal differentiations under the integral sign up to the order  $j$  converge absolutely and uniformly, and are equal therefore to the classical derivatives of  $u$ . In particular,  $u \in C^j$ . This shows that

$$\mathcal{W}_{p,k} \subset C^j \quad (52)$$

if  $|y|^j \in L_{q,1/k}$ . This is a *regularity property* of the distributions in  $\mathcal{W}_{p,k}$ .

## II.4 Fundamental solutions

**II.4.1.** Let  $P$  be a polynomial on  $\mathbb{R}^n$ . A *fundamental solution* for the (partial) differential operator  $P(D)$  is a distribution  $v$  on  $\mathbb{R}^n$  such that

$$P(D)v = \delta. \quad (1)$$

For any  $f \in \mathcal{E}'$ , the (well defined) distribution  $u := v * f$  is then a solution of the (partial) differential equation

$$P(D)u = f. \quad (2)$$

(Indeed,  $P(D)u = (P(D)v) * f = \delta * f = f$ .) The identity

$$P(D)(v * u) = v * (P(D)u) = u \quad (u \in \mathcal{E}') \quad (3)$$

means that the map  $V : u \in \mathcal{E}' \rightarrow v * u$  is the inverse of the map  $P(D) : \mathcal{E}' \rightarrow \mathcal{E}'$ .

**Theorem II.4.2 (Ehrenpreis–Malgrange–Hormander).** *Let  $P$  be a polynomial on  $\mathbb{R}^n$ , and  $\epsilon > 0$ . Then there exists a fundamental solution  $v$  for  $P(D)$  such that*

$$\text{sech}(\epsilon|x|)v \in \mathcal{W}_{\infty, k_P},$$

and  $\|\text{sech}(\epsilon|x|)v\|_{\infty, k_P}$  is bounded by a constant depending only on  $\epsilon, n$ , and  $m = \deg P$ .

Note that  $\text{sech}(\epsilon|x|) \in \mathcal{E}$  (since  $\cosh(\epsilon|x|) = \sum_k \epsilon^{2k} (x_1^2 + \cdots + x_n^2)^k / (2k)!$ ), and therefore its product with the distribution  $v$  is well defined.

For any  $\psi \in \mathcal{D}$  (and  $v$  as in the theorem), write

$$\psi v = [\psi \cosh(\epsilon|x|)][\text{sech}(\epsilon|x|)v].$$

The function in the first square brackets belongs to  $\mathcal{D}$ ; the distribution in the second square brackets belongs to  $\mathcal{W}_{\infty, k_P}$ . Hence  $\psi v \in \mathcal{W}_{\infty, k_P}$  by (50) in Section II.3.9. Denoting

$$\mathcal{W}_{p, k}^{\text{loc}} = \{u \in \mathcal{D}'; \psi u \in \mathcal{W}_{p, k} \text{ for all } \psi \in \mathcal{D}\},$$

the above observation means that *the operator  $P(D)$  has a fundamental solution in  $\mathcal{W}_{\infty, k_P}^{\text{loc}}$ .*

The basic estimate needed for the proof of the theorem is stated in the following.

**Lemma II.4.3 (Notation as in Theorem II.4.2).** *There exists a constant  $C > 0$  (depending only on  $\epsilon, n$ , and  $m$ ) such that, for all  $u \in \mathcal{D}$ ,*

$$|u(0)| \leq C \|\cosh(\epsilon|x|)P(D)u\|_{1, 1/k_P}.$$

**Proof of Theorem II.4.2.** Assuming the lemma, we proceed with the proof of the theorem. Consider the linear functional

$$w : P(D)u \rightarrow u(0) \quad (u \in \mathcal{D}). \quad (4)$$

By the lemma and the Hahn–Banach theorem,  $w$  extends as a continuous linear functional on  $\mathcal{D}$  such that

$$|w(\phi)| \leq C \|\cosh(\epsilon|x|)\phi\|_{1, 1/k_P} \quad (\phi \in \mathcal{D}). \quad (5)$$

Since  $\mathcal{D} \subset \mathcal{W}_{1,1/k_P}$  topologically, it follows that  $w$  is continuous on  $\mathcal{D}$ , that is,  $w \in \mathcal{D}'$ . By (5),

$$|[\operatorname{sech}(\epsilon|x|)w](\phi)| = |w(\operatorname{sech}(\epsilon|x|)\phi)| \leq C\|\phi\|_{1,1/k_P}$$

for all  $\phi \in \mathcal{D}$ . Since  $\mathcal{D}$  is dense in  $\mathcal{W}_{1,1/k_P}$ , the distribution  $\operatorname{sech}(\epsilon|x|)w$  extends uniquely to a continuous linear functional on  $\mathcal{W}_{1,1/k_P}$  with norm  $\leq C$ . Therefore

$$\operatorname{sech}(\epsilon|x|)w \in \mathcal{W}_{\infty,k_P} \quad (6)$$

and

$$\|\operatorname{sech}(\epsilon|x|)w\|_{\infty,k_P} \leq C. \quad (7)$$

Define  $v = Jw := \tilde{w}$ . Then the distribution  $\operatorname{sech}(\epsilon|x|)v \in \mathcal{W}_{\infty,k_P}$  has  $\|\cdot\|_{\infty,k_P}$ -norm  $\leq C$  (the constant in the lemma), and for all  $\phi \in \mathcal{D}$ , we have by (4)

$$\begin{aligned} (P(D)v)(\phi) &= [(P(D)v) * \tilde{\phi}](0) = (v * P(D)\tilde{\phi})(0) \\ &= [\tilde{w} * P(D)\tilde{\phi}](0) = \tilde{w}(J[P(D)\tilde{\phi}]) = w(P(D)\tilde{\phi}) \\ &= \tilde{\phi}(0) = \phi(0) = \delta(\phi), \end{aligned}$$

that is,  $P(D)v = \delta$ . □

**Proof of Lemma II.4.3.** (1) Let  $p$  be a monic polynomial of degree  $m$  in one complex variable, say  $p(z) = \sum_{j=0}^m a_j z^j$ ,  $a_m = 1$ . The polynomial  $q(z) = \sum_{j=0}^m \overline{a_j} z^{m-j}$  satisfies  $q(0) = 1$  and

$$|q(e^{it})| = \left| e^{imt} \overline{\sum_j a_j e^{ijt}} \right| = |p(e^{it})|.$$

If  $f$  is analytic on the closed unit disc, it follows from Cauchy's formula applied to the function  $fq$  that

$$|f(0)| = |f(0)q(0)| \leq \frac{1}{2\pi} \int_0^{2\pi} |f(e^{it})q(e^{it})| dt = \frac{1}{2\pi} \int_0^{2\pi} |f(e^{it})p(e^{it})| dt. \quad (8)$$

Writing  $p(z) = \prod_{j=1}^m (z + z_j)$ , we have for  $k \leq m$

$$p^{(k)}(z) = \sum_{n_1} \sum_{n_2 \notin \{n_1\}} \cdots \sum_{n_k \notin \{n_1, \dots, n_{k-1}\}} \prod_{j \notin \{n_1, \dots, n_k\}} (z + z_j), \quad (9)$$

where all indices range in  $\{1, \dots, m\}$ .

Using (8) with the analytic function

$$f(z) \prod_{j \notin \{n_1, \dots, n_k\}} (z + z_j)$$

and the polynomial

$$p(z) = \prod_{j \in \{n_1, \dots, n_k\}} (z + z_j),$$

we obtain

$$\left| f(0) \prod_{j \notin \{n_1, \dots, n_k\}} z_j \right| \leq \frac{1}{2\pi} \int_0^{2\pi} |f(e^{it})p(e^{it})| dt.$$

Since the number of summands in (9) is  $m(m-1) \cdots (m-k+1) = m!/(m-k)!$ , it follows that

$$|f(0)p^{(k)}(0)| \leq \frac{m!}{(m-k)!} \frac{1}{2\pi} \int_0^{2\pi} |f(e^{it})p(e^{it})| dt. \quad (10)$$

Since (10) remains valid when  $p$  is replaced by  $cp$  with  $c \neq 0$  complex, the inequality (10) is true for *any* polynomial  $p$  and any function  $f$  analytic on the closed unit disc.

(2) If  $f$  is entire, we apply (10) to the function  $f(rz)$  and the polynomial  $p(rz)$  for each  $r > 0$ . Then

$$|f(0)p^{(k)}(0)|2\pi r^k \leq \frac{m!}{(m-k)!} \int_0^{2\pi} |f(re^{it})p(re^{it})| dt. \quad (11)$$

Let  $g$  be a non-negative function with compact support, integrable with respect to Lebesgue measure on  $\mathbb{C}$ , and depending only on  $|z|$ . We multiply (11) by  $rg(re^{it})$  and integrate with respect to  $r$  over  $[0, \infty)$ . Thus

$$|f(0)p^{(k)}(0)| \int_{\mathbb{C}} |z|^k |g(z)| dz \leq \frac{m!}{(m-k)!} \int_{\mathbb{C}} |f(z)p(z)|g(z) dz, \quad (12)$$

where  $dz = r dr dt$  is the area measure in  $\mathbb{C}$ .

(3) The  $n$ -dimensional version of (12) is obtained by applying (12) ‘one variable at a time’: let  $f$  be an entire function on  $\mathbb{C}^n$ ,  $p$  a polynomial on  $\mathbb{C}^n$ , and  $g$  a non-negative function with compact support, integrable with respect to Lebesgue measure  $dz$  on  $\mathbb{C}^n$ , and depending only on  $|z_1|, \dots, |z_n|$ . Then

$$|f(0)p^{(\alpha)}(0)| \int_{\mathbb{C}^n} |z^\alpha| |g(z)| dz \leq \frac{m!}{(m-|\alpha|)!} \int_{\mathbb{C}^n} |f(z)p(z)|g(z) dz. \quad (13)$$

(4) Let  $u \in \mathcal{D}$ , fix  $y \in \mathbb{R}^n$ , and apply (13) to the entire function  $f(z) = \hat{u}(y+z)$ , the polynomial  $p(z) = P(y+z)$ , and the function  $g$  equal to the indicator of the ball  $B := \{z \in \mathbb{C}^n; |z| < \epsilon/2\}$ . Then

$$\left| \hat{u}(y)P^{(\alpha)}(y) \right| \int_B |z^\alpha| dz \leq \frac{m!}{(m-|\alpha|)!} \int_B |\hat{u}(y+z)P(y+z)| dz.$$

Therefore

$$\begin{aligned} k_P(y)|\hat{u}(y)| &\leq \sum_{|\alpha| \leq m} |P^{(\alpha)}(y)| |\hat{u}(y)| \\ &\leq C_1 \int_B |\hat{u}(y+z)P(y+z)| dz = C_1 (2\pi)^{n/2} \int_B |(\mathcal{F}P(D)u)(y+z)| dz \\ &= C_1 (2\pi)^{n/2} \int_B |\mathcal{F}[e^{-i x \cdot z} P(D)u](y)| dz, \end{aligned}$$

where  $C_1$  is a constant depending only on  $m$  and  $n$ . Denote  $k := 1/k_P$ . Then

$$\begin{aligned} (2\pi)^{n/2}|u(0)| &= (2\pi)^{-n/2} \left| \int_{\mathbb{R}^n} \hat{u}(y) dy \right| \\ &\leq C_1 \int_{\mathbb{R}^n} \int_B k(y) |\mathcal{F}[e^{-ix \cdot z} P(D)u](y)| dz dy \\ &= C_1 \int_B \|e^{-ix \cdot z} P(D)u\|_{1,k} dz \leq C_1 |B| \sup_{z \in B} \|e^{-ix \cdot z} P(D)u\|_{1,k}, \end{aligned}$$

where  $|B|$  denotes the  $\mathbb{C}^n$ -Lebesgue measure of  $B$  (depends only on  $n$  and  $\epsilon$ ).

For  $z \in B$ , all derivatives of the function  $\phi_z(x) := e^{-ix \cdot z} / \cosh(\epsilon|x|)$  are  $O(e^{-\epsilon|x|/2})$ , and therefore the family  $\phi_B := \{\phi_z; z \in B\}$  is *bounded* in  $\mathcal{S}$  (this means that given any zero neighborhood  $U$  in  $\mathcal{S}$ , there exists  $\eta > 0$  such that  $\lambda \phi_B \subset U$  for all scalars  $\lambda$  with modulus  $< \eta$ ). For a topological vector space with topology induced by a family of semi-norms, the above condition is equivalent to the boundedness of the semi-norms of the family on the set  $\phi_B$ ; thus, in the special case of  $\mathcal{S}$ , the boundedness of  $\phi_B$  means that  $\sup_{z \in B} \|\phi_z\|_{\alpha, \beta} < \infty$ . Since  $\mathcal{S} \subset \mathcal{W}_{p,k}$  topologically (for any  $p, k$ ), it follows that the set  $\phi_B$  is bounded in  $\mathcal{W}_{p,k}$  for any  $p, k$ . In particular,  $M_B := \sup_{z \in B} \|\phi_z\|_{1, \underline{k}} < \infty$ . ( $M_B$  depends only on  $n$  and  $\epsilon$ .) By (50) in II.3.9,

$$\begin{aligned} |u(0)| &\leq C_1 |B| \sup_{z \in B} (2\pi)^{-n/2} \|\phi_z[\cosh(\epsilon|x|)P(D)u]\|_{1,k} \\ &\leq (2\pi)^{-n} C_1 |B| \sup_{z \in B} \|\phi_z\|_{1, \underline{k}} \|\cosh(\epsilon|x|)P(D)u\|_{1,k} = C \|\cosh(\epsilon|x|)P(D)u\|_{1,k}, \end{aligned}$$

where  $C = (2\pi)^{-n} C_1 |B| M_B$  depends only on  $n, m$  and  $\epsilon$ . □

## II.5 Solution in $\mathcal{E}'$

**II.5.1.** Consider the operator  $P(D)$  *restricted to*  $\mathcal{E}'$ . We look for necessary and sufficient conditions on  $f \in \mathcal{E}'$  such that the equation  $P(D)u = f$  has a solution  $u \in \mathcal{E}'$ . A necessary condition is immediate. For any solution  $\phi \in \mathcal{E}$  of the so-called homogeneous ‘adjoint’ equation  $P(-D)\phi = 0$ , we have (for  $u$  as above)

$$f(\phi) = (P(D)u)(\phi) = u(P(-D)\phi) = u(0) = 0,$$

that is,  $f$  annihilates the null space of  $P(-D)|_{\mathcal{E}}$ . In particular,  $f$  annihilates elements of the null space of the special form  $\phi(x) = q(x)e^{ix \cdot z}$ , where  $z \in \mathbb{C}^n$  and  $q$  is a polynomial on  $\mathbb{R}^n$  (let us call such elements ‘exponential solutions’ of the homogeneous adjoint equation). The next theorem establishes that the later condition is also sufficient.

**Theorem II.5.2.** *The following statements are equivalent for  $f \in \mathcal{E}'$ :*

- (1) *The equation  $P(D)u = f$  has a solution  $u \in \mathcal{E}'$ .*



(2)  $f$  annihilates every exponential solution of the homogeneous adjoint equation.

(3) The function  $F(z) := \hat{f}(z)/P(z)$  is entire on  $\mathbb{C}^n$ .

**Proof.**  $1 \implies 2$ . See II.5.1.

$2 \implies 3$ . In order to make one-complex-variable arguments, we consider the function

$$F(t; z, w) := \frac{\hat{f}(tw + z)}{P(tw + z)} \quad (t \in \mathbb{C}; z, w \in \mathbb{C}^n). \quad (1)$$

Let  $P_m$  be the principal part of  $P$ , and fix  $w \in \mathbb{C}^n$  such that  $P_m(w) \neq 0$ . Since

$$\begin{aligned} P(tw + z) &= P_m(tw + z) + \text{terms of lower degree} \\ &= \sum_{\alpha} P_m^{(\alpha)}(tw) z^{\alpha} / \alpha! + \text{terms of lower degree} \\ &= P_m(tw) + \text{terms of lower degree} = t^m P_m(w) + \text{terms of lower degree}, \end{aligned}$$

and  $P_m(w) \neq 0$ ,  $P(tw + z)$  is a polynomial of degree  $m$  in  $t$  (for each given  $z$ ).

Fix  $z = z_0$ , and let  $t_0$  be a zero of order  $k$  of  $P(tw + z_0)$ . For  $j < k$ , set

$$\phi_j(x, t) := (x \cdot w)^j \exp(-ix \cdot (tw + z_0)). \quad (2)$$

Then

$$\begin{aligned} P(-D)\phi_j(x, t) &= P(-D) \left( i \frac{\partial}{\partial t} \right)^j \exp(-ix \cdot (tw + z_0)) \\ &= \left( i \frac{\partial}{\partial t} \right)^j P(-D) \exp(-ix \cdot (tw + z_0)) \\ &= \left( i \frac{\partial}{\partial t} \right)^j P(tw + z_0) \exp(-ix \cdot (tw + z_0)), \end{aligned}$$

and therefore  $P(-D)\phi_j(x, t_0) = 0$ , that is,  $\phi_j(\cdot, t_0)$  are exponential solutions of the homogeneous adjoint equation. By hypothesis, we then have  $f(\phi_j(\cdot, t_0)) = 0$  for all  $j < k$ . This means that, for all  $j < k$ ,

$$\frac{\partial^j}{\partial t^j} \Big|_{t=t_0} \hat{f}(tw + z_0) = 0,$$

that is,  $\hat{f}(tw + z_0)$  has a zero of order  $\geq k$  at  $t = t_0$ , and  $F(\cdot; z_0, w)$  is consequently entire (cf. (1)).

Choose  $r > 0$  such that  $P(tw + z_0) \neq 0$  on the circle  $\Gamma := \{t; |t| = r\}$  (this is possible since, as a polynomial in  $t$ ,  $P(tw + z_0)$  has finitely many zeros). By continuity,  $P(tw + z) \neq 0$  for all  $t$  on  $\Gamma$  and  $z$  in a neighborhood  $U$  of  $z_0$ . Therefore the function  $G(z) := 1/2\pi i \int_{\Gamma} F(t; z, w) dt/t$  is analytic in  $U$ , that is,  $G$  is entire (by the arbitrariness of  $z_0$ ). However, by Cauchy's integral theorem for the entire function  $F(\cdot; z, w)$ , we have  $G(z) = F(0; z, w) = \hat{f}(z)/P(z)$ , and Statement (3) is proved.

$3 \implies 1$ . By Theorem II.3.6 (the Paley–Wiener–Schwartz theorem), it suffices to show that  $F$  satisfies the estimate in Part (i) of Theorem II.3.6 (for then  $F$  is the Fourier–Laplace transform of some  $u \in \mathcal{E}'$ ; restricting to  $\mathbb{R}^n$ , we have therefore  $\mathcal{F}f = P\mathcal{F}u = \mathcal{F}P(D)u$ , hence  $P(D)u = f$ ).

Fix  $\zeta \in \mathbb{C}^n$  and apply (13) in the proof of Lemma II.4.3 to the entire function  $f(z) = F(\zeta + z)$ , the polynomial  $p(z) = P(\zeta + z)$ , and  $g$  the indicator of the unit ball  $B$  of  $\mathbb{C}^n$ . Then

$$|F(\zeta)P^{(\alpha)}(\zeta)| \leq C \int_B |\hat{f}(\zeta + z)| dz \leq C|B| \sup_{z \in B} |\hat{f}(\zeta + z)|,$$

where  $C$  is a constant depending only on  $n$  and  $m = \deg P$ . Choose  $\alpha$  such that  $P^{(\alpha)}$  is a non-zero constant. Then, by the necessity of the estimate in Part (i) of Theorem II.3.6 (applied to the distribution  $f \in \mathcal{E}'$ ), we have

$$|F(\zeta)| \leq C_1 \sup_{z \in B} (1 + |\zeta + z|)^k e^{A|\Im(\zeta + z)|} \leq C_2 (1 + |\zeta|)^k e^{A|\Im \zeta|},$$

as desired.  $\square$

## II.6 Regularity of solutions

**II.6.1.** Let  $\Omega$  be an open subset of  $\mathbb{R}^n$ . If  $\phi \in \mathcal{D}(\Omega)$  and  $u \in \mathcal{D}'(\Omega)$ , the product  $\phi u$  is a distribution with compact support in  $\Omega$ , and may then be considered as an element of  $\mathcal{E}' := \mathcal{E}'(\mathbb{R}^n) \subset \mathcal{S}' := \mathcal{S}'(\mathbb{R}^n)$ . Set (for any weight  $k$  and  $p \in [1, \infty]$ )

$$\mathcal{W}_{p,k}^{\text{loc}}(\Omega) := \{u \in \mathcal{D}'(\Omega); \phi u \in \mathcal{W}_{p,k} \text{ for all } \phi \in \mathcal{D}(\Omega)\}.$$

(cf. comments following Theorem II.4.2.) Note that if  $u \in \mathcal{E}(\Omega)$ , then  $\phi u \in \mathcal{D}(\Omega) \subset \mathcal{S} \subset \mathcal{W}_{p,k}$  for all  $\phi \in \mathcal{D}(\Omega)$ , that is,  $\mathcal{E}(\Omega) \subset \mathcal{W}_{p,k}^{\text{loc}}(\Omega)$  for all  $p, k$ . Conversely, if  $u \in \mathcal{W}_{p,k}^{\text{loc}}(\Omega)$  for all  $p, k$  (or even for some  $p$  and all weights  $k_s$ ), it follows from (52) in Section II.3.9 that  $u \in \mathcal{E}(\Omega)$ . This observation gives an approach for proving *regularity* of distribution solutions of the equation  $P(D)u = f$  in  $\Omega$  (for suitable  $f$ ): it would suffice to prove that the solutions  $u$  belong to all the spaces  $\mathcal{W}_{p,k}^{\text{loc}}(\Omega)$  (since then  $u \in \mathcal{E}(\Omega)$ ).

### II.6.2. Hypoellipticity.

The polynomial  $P$  (or the differential operator  $P(D)$ ) is *hypoelliptic* if there exist constants  $C, c > 0$  such that

$$\left| \frac{P(x)}{P^{(\alpha)}(x)} \right| \geq C|x|^{c|\alpha|} \quad (1)$$

as  $x \in \mathbb{R}^n \rightarrow \infty$ , for all multi-indices  $\alpha \neq 0$ .

Conditions equivalent to (1) are *any one* of the following conditions (2)–(4):

$$\lim_{|x| \rightarrow \infty} \frac{P^{(\alpha)}(x)}{P(x)} = 0 \quad (2)$$

for all  $\alpha \neq 0$ ;

$$\lim_{|x| \rightarrow \infty} \text{dist}(x, N(P)) = \infty, \quad (3)$$

where  $N(P) := \{z \in \mathbb{C}^n; P(z) = 0\}$ ;

$$\text{dist}(x, N(P)) \geq C|x|^c \quad (4)$$

as  $x \in \mathbb{R}^n \rightarrow \infty$ , for suitable positive constants  $C, c$  (these equivalent descriptions of hypoellipticity will not be used in the sequel). For example, if the principal part  $P_m$  of  $P$  does not vanish for  $0 \neq x \in \mathbb{R}^n$  (in this case,  $P$  and  $P(D)$  are said to be *elliptic*), Condition (2) is clearly satisfied; thus elliptic differential operators are hypoelliptic.

**Theorem II.6.3.** *Let  $P$  be a hypoelliptic polynomial, and let  $\Omega$  be an open subset of  $\mathbb{R}^n$ . If  $u \in \mathcal{D}'(\Omega)$  is a solution of the equation  $P(D)u = f$  with  $f \in \mathcal{W}_{p,k}^{\text{loc}}(\Omega)$ , then  $u \in \mathcal{W}_{p,kk_P}^{\text{loc}}(\Omega)$ . In particular, if  $f \in \mathcal{E}(\Omega)$ , then  $u \in \mathcal{E}(\Omega)$ .*

*[The following converse is also true (proof omitted): suppose that for some  $\Omega$ , some  $p \in [1, \infty]$ , and some weight  $k$ , every solution of the equation  $P(D)u = 0$  in  $\mathcal{W}_{p,k}^{\text{loc}}(\Omega)$  is in  $\mathcal{E}(\Omega)$ . Then  $P$  is hypoelliptic.]*

**Proof.** Fix  $\omega \subset \subset \Omega$ . For any  $u \in \mathcal{D}'(\Omega)$  and  $\phi \in \mathcal{D}(\omega)$ , we view  $\phi u$  as an element of  $\mathcal{E}'$  with support in  $\omega$  (cf. II.6.1). By the necessity of the estimate in Part (i) of the Paley–Wiener–Schwartz theorem (II.3.6),  $|\mathcal{F}(\phi u)(x)| \leq M(1 + |x|)^r$  for some constants  $M, r$  independent of  $\phi$ . Hence, for any given  $p$ , there exists  $s$  (independent of  $\phi$ ) such that  $k_{-s}\mathcal{F}(\phi u) \in L^p$ . Denote  $k' = k_{-s}$  for such an  $s$  (fixed from now on). Thus  $\phi u \in \mathcal{W}_{p,k'}^{\text{loc}}(\omega)$  for all  $\phi \in \mathcal{D}(\omega)$ , that is,  $u \in \mathcal{W}_{p,k'}^{\text{loc}}(\omega)$ .

The hypoellipticity condition (1) (Section II.6.2) implies the existence of a constant  $C' > 0$  such that  $|P^{(\alpha)}/P| \leq (1/C')|x|^{-c|\alpha|}$  for all  $\alpha \neq 0$ . Summing over all  $\alpha \neq 0$  with  $|\alpha| \leq m$ , we get that  $k'_P/|P| \leq (1/C'')(1 + |x|)^{-c}$  for some constant  $C'' > 0$  (cf. notation at the end of II.3.7). Hence

$$\frac{k_P}{k'_P} = 1 + \frac{|P|}{k'_P} \geq C''(1 + |x|)^c. \quad (5)$$

Given the weight  $k$ ,  $kk_P/k'$  is a weight, and therefore it is  $O((1 + |x|)^\nu)$  for some  $\nu$ . Consequently there exists a positive integer  $r$  (depending only on the ratio  $k/k'$ ) such that  $kk_P/k' \leq \text{const.}(1 + |x|)^{cr}$ . By (5), it then follows that

$$kk_P \leq Ck' \left( \frac{k_P}{k'_P} \right)^r \quad (6)$$

for some constant  $C$ .

**Claim.** *If  $k$  is any weight such that  $f \in \mathcal{W}_{p,k}^{\text{loc}}(\omega)$  and (6) with  $r = 1$  is valid (that is,  $k \leq C(k'/k'_P)$ ), then any solution  $u \in \mathcal{W}_{p,k'}^{\text{loc}}(\omega)$  of the equation  $P(D)u = f$  is necessarily in  $\mathcal{W}_{p,kk_P}^{\text{loc}}(\omega)$ .*

**Proof of claim.** We first observe that

$$P(D)\mathcal{W}_{p,k}^{\text{loc}}(\omega) \subset \mathcal{W}_{p,k/k_P}^{\text{loc}}(\omega) \quad (7)$$

(cf. II.3.9, Relation (48)). Therefore,  $P^{(\alpha)}(D)u \in \mathcal{W}_{p,k'/k'_P}^{\text{loc}}(\omega) \subset \mathcal{W}_{p,k}^{\text{loc}}(\omega)$  for all  $\alpha \neq 0$  (for  $u$  as in the claim, because  $k \leq C(k'/k'_P)$ ).

If  $\phi \in \mathcal{D}(\omega)$ , we have by Leibnitz' formula (II.2.9, (5))

$$P(D)(\phi u) = \phi f + \sum_{\alpha \neq 0} D^\alpha \phi P^{(\alpha)}(D)u / \alpha!.$$

The first term is in  $\mathcal{W}_{p,k}$  by hypothesis. The sum over  $\alpha \neq 0$  is in  $\mathcal{W}_{p,k}$  by the preceding observation. Hence  $P(D)(\phi u) \in \mathcal{W}_{p,k}$  (and has compact support).

Let  $v \in \mathcal{W}_{\infty,k_P}^{\text{loc}}(\mathbb{R}^n)$  be a fundamental solution for  $P(D)$  (by Theorem II.4.2 and the observation following its statement). Then

$$\phi u = v * [P(D)(\phi u)] \in \mathcal{W}_{p,kk_P}$$

since for any weights  $k, k_1$  (cf. II.3.9, (47))

$$\mathcal{W}_{\infty,k_1}^{\text{loc}}(\mathbb{R}^n) * [\mathcal{W}_{p,k} \cap \mathcal{E}'] \subset \mathcal{W}_{p,kk_1}.$$

This concludes the proof of the claim.

Suppose now that  $r > 1$ . Consider the weights

$$k_j = k \left( \frac{k'_P}{k_P} \right)^j \quad j = 0, \dots, r-1.$$

Since

$$k = k_0 \geq k_1 \geq \dots \geq k_{r-1},$$

we have  $f \in \mathcal{W}_{p,k_j}^{\text{loc}}(\omega)$  for all  $j = 0, \dots, r-1$ . Also by (6)

$$k_P k_{r-1} \left( = k_P k \left( \frac{k'_P}{k_P} \right)^{r-1} \right) \leq C k' \frac{k_P}{k'_P}.$$

We may then apply the claim with the weight  $k_{r-1}$  replacing  $k$ . Then  $u \in \mathcal{W}_{p,k_P k_{r-1}}^{\text{loc}}(\omega)$ ,  $f \in \mathcal{W}_{p,k_{r-2}}^{\text{loc}}(\omega)$ , and  $k_{r-2} = k_P k_{r-1} / k'_P$ . By the claim with the weights  $k, k'$  replaced by  $k_{r-2}, k_P k_{r-1}$  (respectively), it follows that  $u \in \mathcal{W}_{p,k_P k_{r-2}}^{\text{loc}}(\omega)$ . Repeating this argument, we obtain finally (since  $k_0 = k$ ) that  $u \in \mathcal{W}_{p,k_P k}^{\text{loc}}(\omega)$ . This being true for any  $\omega \subset\subset \Omega$ , we conclude that  $u \in \mathcal{W}_{p,k_P k}^{\text{loc}}(\Omega)$ .  $\square$

## II.7 Variable coefficients

**II.7.1.** The constant coefficients theory of II.4.1 and Theorem II.4.2 can be applied 'locally' to linear differential operators  $P(x, D)$  with (locally)  $C^\infty$ -coefficients. (This means that  $P(x, y)$  is a polynomial in  $y \in \mathbb{R}^n$ , with coefficients

that are  $C^\infty$ -functions of  $x$  in some neighborhood  $\Omega \subset \mathbb{R}^n$  of  $x^0$ .) Denote  $P_0 = P(x^0, \cdot)$ . We shall assume that there exist  $\epsilon > 0$  and  $0 < M < \infty$  such that the  $\epsilon$ -neighbourhood  $V$  of  $x^0$  is contained in  $\Omega$  and for all  $x \in V$

$$\frac{k_{P(x, \cdot)}}{k_{P_0}} \leq M. \quad (1)$$

The method described below regards the operator  $P(x, D)$  as a ‘perturbation’ of the operator  $P_0(D)$  for  $x$  in a ‘small’ neighbourhood of  $x^0$ .

Let  $r + 1$  be the (finite!) dimension of the space of polynomials  $Q$  such that  $k_Q/k_{P_0}$  is bounded, and choose a basis  $P_0, P_1, \dots, P_r$  for this space. By (1), we have a unique representation

$$P(x, \cdot) = P_0 + \sum_{j=1}^r c_j(x) P_j \quad (2)$$

for all  $x \in V$ . Necessarily  $c_j(x^0) = 0$  (take  $x = x^0$ ) and  $c_j \in C^\infty(V)$ .

By Theorem II.4.2, we may choose a fundamental solution  $v \in \mathcal{W}_{\infty, k_{P_0}}^{\text{loc}}$  for the operator  $P_0(D)$ . Fix  $\chi \in \mathcal{D}$  such that  $\chi = 1$  in a  $3\epsilon$ -neighbourhood of  $x^0$ . Then

$$w := \chi v \in \mathcal{W}_{\infty, k_{P_0}}, \quad (3)$$

and for all  $h \in \mathcal{E}'(V)$ ,

$$P_0(D)(w * h) = w * (P_0(D)h) = v * P_0(D)h = h. \quad (4)$$

(The second equality follows from the fact that  $\text{supp } P_0(D)h \subset V$  and  $w * g = v * g$  for all  $g \in \mathcal{E}'(V)$ .) By (2) and (4)

$$P(x, D)(w * h) = h + \sum_{j=1}^r c_j(x) P_j(D)(w * h) \quad (5)$$

for all  $h \in \mathcal{E}'(V)$ .

We localize to a suitable  $\delta$ -neighbourhood of  $x^0$  by fixing some function  $\phi \in \mathcal{D}$  such that  $\phi = 1$  for  $|x| \leq 1$  and  $\text{supp } \phi \subset \{x; |x| < 2\}$ , and letting  $\phi_\delta(x) = \phi((x - x^0)/\delta)$ . (Thus  $\phi_\delta = 1$  for  $|x - x^0| \leq \delta$  and  $\text{supp } \phi_\delta \subset \{x; |x - x^0| < 2\delta\}$ .)

By (5), whenever  $\delta < \epsilon$  and  $h \in \mathcal{E}'(V)$ ,

$$P(\cdot, D)(w * h) = h + \sum_{j=1}^r \phi_\delta c_j P_j(D)(w * h) \quad (6)$$

in  $|x - x^0| < \delta$ .

**Claim.** *There exists  $\delta_0 < \epsilon/2$  such that, for  $\delta < \delta_0$ , the equation*

$$h + \sum_{j=1}^r \phi_\delta c_j P_j(D)(w * h) = \phi_\delta f \quad (7)$$

(for any  $f \in \mathcal{E}'$ ) has a unique solution  $h \in \mathcal{E}'$ .

Assuming the claim, the solution  $h$  of (7) satisfies (by (6))

$$P(\cdot, D)(w * h) = \phi_\delta f = f \quad (8)$$

in  $V_\delta := \{x; |x - x^0| < \delta\}$ . (Since  $2\delta < \epsilon$ ,  $\text{supp } \phi_\delta \subset V$ , and therefore,  $\text{supp } h \subset V$  by (7), and (6) applies.)

In other words,  $u = w * h \in \mathcal{E}'$  solves the equation  $P(\cdot, D)u = f$  in  $V_\delta$ . Equivalently, the map

$$T : f \in \mathcal{E}' \rightarrow w * h \in \mathcal{E}' \quad (9)$$

(with  $h$  as in the ‘claim’) is ‘locally’ a right inverse of the operator  $P(\cdot, D)$ , that is,

$$P(x, D)Tf = f \quad (f \in \mathcal{E}'; x \in V_\delta). \quad (10)$$

The operator  $T$  is also a left inverse of  $P(\cdot, D)$  (in the above local sense). Indeed, given  $u \in \mathcal{E}'(V_\delta)$ , we take  $f := P(\cdot, D)u$  and  $h := P_0(D)u$ . By (4),  $w * h = w * P_0(D)u = u$  (since  $u \in \mathcal{E}'(V)$ ). Therefore, the left-hand side of (7) equals

$$P_0(D)u + \sum_j \phi_\delta c_j P_j(D)u = P(\cdot, D)u = f = \phi_\delta f$$

in  $V_\delta$  (since  $\phi_\delta = 1$  in  $V_\delta$ ). Thus ‘our’  $h$  is the (unique) solution of (7) (for ‘our’  $f$ ) in  $V_\delta$ . Consequently

$$TP(\cdot, D)u := w * h = u \quad (u \in \mathcal{E}'(V_\delta)) \quad (11)$$

in  $V_\delta$ . Since  $2\delta < \epsilon$ , (11) is true in  $V_\delta$  for all  $u \in \mathcal{E}'(\mathbb{R}^n)$ . Modulo the ‘claim’, we proved the first part of the following.

**Theorem II.7.2.** *Let  $P(\cdot, D)$  have  $C^\infty$ -coefficients and satisfy*

$$k_{P(x, \cdot)} \leq M k_{P_0} \quad (*)$$

in an  $\epsilon$ -neighbourhood of  $x^0$  (where  $M$  is a constant and  $P_0 := P(x^0, \cdot)$ ). Then there exists a  $\delta$ -neighbourhood  $V_\delta$  of  $x^0$  (with  $\delta < \epsilon$ ) and a linear map  $T : \mathcal{E}' \rightarrow \mathcal{E}'$  such that

$$P(\cdot, D)Tg = TP(\cdot, D)g = g \quad \text{in } V_\delta \quad (g \in \mathcal{E}').$$

Moreover, the restriction of  $T$  to the subspace  $\mathcal{W}_{p,k} \cap \mathcal{E}'$  of  $\mathcal{W}_{p,k}$  is a bounded operator into  $\mathcal{W}_{p,kk_{P_0}}$ , for any weight  $k$ .

**Proof.** We first prove the ‘claim’ (this will complete the proof of the first part of the theorem.)

For any  $\delta < \epsilon$ , consider the map

$$S_\delta : h \in \mathcal{S}' \rightarrow \sum_{j=1}^r \phi_\delta c_j P_j(D)(w * h).$$

Since  $w \in \mathcal{W}_{\infty, k_{P_0}}$  and  $k_{P_j}/k_{P_0}$  are bounded (by definition of  $P_j$ ), we have

$$|P_j \mathcal{F}w| \leq k_{P_j} |\mathcal{F}w| \leq \text{const} \cdot k_{P_0} |\mathcal{F}w| \leq C < \infty \quad (j = 0, \dots, r). \quad (12)$$

Let  $k$  be any given weight. By (51) in Section II.3.9, there exists  $t_0 > 0$  such that, for  $0 < t < t_0$ ,

$$\|S_\delta h\|_{p, k^t} \leq 2(2\pi)^{-n/2} \sum_{j=1}^r \|\phi_\delta c_j\|_{1,1} \|P_j(D)(w * h)\|_{p, k^t}. \quad (13)$$

By the inequality preceding (47) in II.3.9,

$$\|P_j(D)(w * h)\|_{p, k^t} = \|[P_j(D)w] * h\|_{p, k^t} \leq (2\pi)^{n/2} \|P_j(D)w\|_{\infty, 1} \|h\|_{p, k^t}$$

for all  $h \in \mathcal{W}_{p, k} = \mathcal{W}_{p, k^t}$ . Since by (12)

$$\|P_j(D)w\|_{\infty, 1} = \|\mathcal{F}[P_j(D)w]\|_\infty = \|P_j \mathcal{F}w\|_\infty \leq C$$

(for  $j = 1, \dots, r$ ), we obtain from (13)

$$\|S_\delta h\|_{p, k^t} \leq 2C \sum_{j=1}^r \|\phi_\delta c_j\|_{1,1} \|h\|_{p, k^t} \quad (14)$$

for all  $h \in \mathcal{W}_{p, k}$ .

Since  $c_j(x^0) = 0$ ,  $c_j = O(\delta)$  on  $\text{supp } \phi_\delta$  by the mean value inequality. Using the definition of  $\phi_\delta$ , it follows that  $c_j D^\alpha \phi_\delta = O(\delta^{1-|\alpha|})$ . Hence by Leibnitz' formula,  $D^\alpha(\phi_\delta c_j) = O(\delta^{1-|\alpha|})$ . Therefore, since the measure of  $\text{supp}(\phi_\delta c_j)$  is  $O(\delta^n)$ , we have

$$x^\alpha \mathcal{F}(\phi_\delta c_j) = \mathcal{F}[D^\alpha(\phi_\delta c_j)] = O(\delta^{n+1-|\alpha|}).$$

Hence

$$(1 + \delta|x|)^{n+1} \mathcal{F}(\phi_\delta c_j) = O(\delta^{n+1}).$$

Consequently,

$$\|\phi_\delta c_j\|_{1,1} := \int |\mathcal{F}(\phi_\delta c_j)| dx \leq \text{const} \cdot \delta^{n+1} \int \frac{dx}{(1 + \delta|x|)^{n+1}} = O(\delta).$$

We may then choose  $0 < \delta_0 < \epsilon/2$  such that

$$\sum_{j=1}^r \|\phi_\delta c_j\|_{1,1} < \frac{1}{4C} \quad (15)$$

for  $0 < \delta < \delta_0$ . By (14), we then have

$$\|S_\delta h\|_{p, k^t} \leq (1/2) \|h\|_{p, k^t}$$

for all  $h \in \mathcal{W}_{p, k} = \mathcal{W}_{p, k^t}$ . This means that for  $\delta < \delta_0$ , the operator  $S_\delta$  on the Banach space  $\mathcal{W}_{p, k^t}$  has norm  $\leq 1/2$ , and therefore  $I + S_\delta$  has a bounded

inverse ( $I$  is the identity operator). Thus, (7) (in II.7.1) has a unique solution  $h \in \mathcal{W}_{p,k}$  for each  $f \in \mathcal{W}_{p,k}$ . By the equation,  $h$  is necessarily in  $\mathcal{E}'$  (since  $\phi_\delta \in \mathcal{D}$ ). If  $f$  is an arbitrary distribution in  $\mathcal{E}'$ , the trivial part of the Paley–Wiener–Schwartz theorem (II.3.6) shows that  $\hat{f}(x) = O(1 + |x|)^N$  for some  $N$ , and therefore  $f \in \mathcal{W}_{p,k}$  for suitable weight  $k$  (e.g.  $k = k_{-s}$  with  $s$  large enough). Therefore (for  $\delta < \delta_0$ ) there exists a solution  $h$  of (7) in  $\mathcal{W}_{p,k} \cap \mathcal{E}'$ . The solution is unique (in  $\mathcal{E}'$ ), because if  $h, h' \in \mathcal{E}'$  are solutions, there exists a weight  $k$  such that  $f, h, h' \in \mathcal{W}_{p,k}$ , and therefore  $h = h'$  by the uniqueness of the solution in  $\mathcal{W}_{p,k}$ . This completes the proof of the claim.

Since  $\|S_\delta\| \leq 1/2$  (the norm is the  $B(\mathcal{W}_{p,k^t})$ -norm!), we have  $\|(I + S_\delta)^{-1}\| \leq 2$  (by the Neumann expansion of the resolvent!), and therefore  $\|h\|_{p,k^t} \leq 2\|\phi_\delta f\|_{p,k^t}$ . Consequently (with  $h$  related to  $f$  as in the ‘claim’, and  $t$  small enough), we have by the inequality preceding (47) and by (51) in II.3.9:

$$\begin{aligned} \|Tf\|_{p,k_{P_0}k^t} &= \|w * h\|_{p,k_{P_0}k^t} \leq (2\pi)^{n/2} \|w\|_{\infty,k_{P_0}} \|h\|_{p,k^t} \\ &\leq 2(2\pi)^{n/2} \|w\|_{\infty,k_{P_0}} \|\phi_\delta f\|_{p,k^t} \leq 4\|w\|_{\infty,k_{P_0}} \|\phi_\delta\|_{1,1} \|f\|_{p,k^t}. \end{aligned}$$

This proves the second part of the theorem, since the norms  $\|\cdot\|_{p,k}$  ( $\|\cdot\|_{p,k_{P_0}k}$ ) and  $\|\cdot\|_{p,k^t}$  ( $\|\cdot\|_{p,k_{P_0}k^t}$ , respectively) are equivalent.  $\square$

**Corollary II.7.3.** *For  $P(\cdot, D)$  and  $V_\delta$  as in Theorem II.7.2, the equation  $P(\cdot, D)u = f$  has a solution  $u \in C^\infty(V_\delta)$  for each  $f \in C^\infty(\mathbb{R}^n)$ .*

**Proof.** Fix  $\phi \in \mathcal{D}$  such that  $\phi = 1$  in a neighbourhood of  $\overline{V_\delta}$ . For  $f \in C^\infty$ ,  $\phi f \in \mathcal{W}_{p,k} \cap \mathcal{E}'$  for all  $k$ ; therefore  $u := T(\phi f)$  is a solution of  $P(\cdot, D)u = f$  in  $V_\delta$  (because  $\phi f = f$  in  $V_\delta$ ), which belongs to  $\mathcal{W}_{p,k_{P_0}k}$  for all weights  $k$ , hence  $u \in C^\infty(V_\delta)$ .  $\square$

## II.8 Convolution operators

Let  $h : \mathbb{R}^n \rightarrow \mathbb{C}$  be locally Lebesgue integrable, and consider the *convolution operator*

$$T : u \rightarrow h * u,$$

originally defined on the space  $L_c^1$  of integrable functions  $u$  on  $\mathbb{R}^n$  with compact support.

We set  $h^t(x) := t^n h(tx)$ , and make the following

### Hypothesis I.

$$\int_{|x| \geq 2} |h^t(x - y) - h^t(x)| dx \leq K < \infty \quad (|y| \leq 1; t > 0). \quad (1)$$

**Lemma II.8.1.** *If  $u \in L_c^1$  has support in the ball  $B(a, t)$  and  $\int u dx = 0$ , then*

$$\int_{B(a, 2t)^c} |Tu| dx \leq K \|u\|_1.$$



**Proof.** Denote  $u_a(x) = u(x + a)$ . Since  $(Tu)_a = Tu_a$ , we have

$$\begin{aligned} \int_{B(a, 2t)^c} |Tu| dx &= \int_{B(0, 2t)^c} |(Tu)_a| dx = \int_{B(0, 2t)^c} |h * u_a| dx \\ &= \int_{B(0, 2)^c} |h^t * (u_a)^t| dx. \end{aligned}$$

Since  $(u_a)^t(y) = t^n u(ty + a) = 0$  for  $|y| \geq 1$  and  $\int (u_a)^t(y) dy = \int u(x) dx = 0$ , the last integral is equal to

$$\begin{aligned} &\int_{B(0, 2)^c} \left| \int_{|y| < 1} [h^t(x - y) - h^t(x)] (u_a)^t(y) dy \right| dx \\ &\leq \int_{B(0, 2)^c} \int_{|y| < 1} |h^t(x - y) - h^t(x)| |(u_a)^t(y)| dy dx \\ &= \int_{|y| < 1} \left( \int_{B(0, 2)^c} |h^t(x - y) - h^t(x)| dx \right) |(u_a)^t(y)| dy \leq K \|(u_a)^t\|_1 = K \|u\|_1. \end{aligned}$$

□

We shall need the following version of the *Calderon–Zygmund decomposition lemma*.

**Lemma II.8.2.** Fix  $s > 0$ , and let  $u \in L^1(\mathbb{R}^n)$ . Then there exist disjoint open (hyper)cubes  $I_k$  and functions  $u_k, v \in L^1(\mathbb{R}^n)$  ( $k \in \mathbb{N}$ ), such that

- (1)  $\text{supp } u_k \subset I_k$  and  $\int u_k dx = 0$  for all  $k \in \mathbb{N}$ ;
- (2)  $|v| \leq 2^n s$  a.e.;
- (3)  $u = v + \sum_k u_k$ ;
- (4)  $\|v\|_1 + \sum_k \|u_k\|_1 \leq 3\|u\|_1$ ; and
- (5)  $\sum_k |I_k| \leq \|u\|_1/s$  (where  $|I_k|$  denotes the volume of  $I_k$ ).

**Proof.** We first partition  $\mathbb{R}^n$  into cubes of volume  $> \|u\|_1/s$ . For any such cube  $Q$ , the average on  $Q$  of  $|u|$ ,

$$A_Q(|u|) := |Q|^{-1} \int_Q |u| dx,$$

satisfies

$$A_Q(|u|) \leq |Q|^{-1} \|u\|_1 < s. \quad (2)$$

Subdivide  $Q$  into  $2^n$  congruent subcubes  $Q_i$  (by dividing each side of  $Q$  into two equal intervals). If  $A_{Q_i}(|u|) \geq s$  for all  $i$ , then

$$A_Q(|u|) = |Q|^{-1} \sum_i \int_{Q_i} |u| dx \geq |Q|^{-1} s \sum_i |Q_i| = s,$$

contradicting (2). Let  $Q_{1,j}$  be the *open* subcubes of  $Q$  on which the averages of  $|u|$  are  $\geq s$ , and let  $Q'_{1,l}$  be the remaining subcubes (there is at least one subcube of the latter kind). We have

$$s|Q_{1,j}| \leq \int_{Q_{1,j}} |u| dx \leq \int_Q |u| dx < s|Q| = 2^n s|Q_{1,j}|. \quad (3)$$

We define  $v$  on the the cubes  $Q_{1,j}$  as the *constant*  $A_{Q_{1,j}}(u)$  (for each  $j$ ), and we let  $u_{1,j}$  be equal to  $u - v$  on  $Q_{1,j}$  and to zero on  $Q_{1,j}^c$ .

For each cube  $Q'_{1,l}$  we repeat the construction we did with  $Q$  (since the average of  $|u|$  over such cubes is  $< s$ , as it was over  $Q$ ). We obtain the *open* subcubes  $Q_{2,j}$  (of the cubes  $Q'_{1,l}$ ) on which the average of  $|u|$  is  $\geq s$ , and the remaining subcubes  $Q'_{2,l}$  on which the average is  $< s$ . We then extend the definition of  $v$  to the subcubes  $Q_{2,j}$ , by assigning to  $v$  the constant value  $A_{Q_{2,j}}(u)$  on  $Q_{2,j}$  (for each  $j$ ). The functions  $u_{2,j}$  are then defined in the same manner as  $u_{1,j}$ , with  $Q_{2,j}$  replacing  $Q_{1,j}$ .

Continuing this process (and renaming), we obtain a sequence of mutually disjoint open cubes  $I_k$ , a sequence of measurable functions  $u_k$  defined on  $\mathbb{R}^n$ , and a measurable function  $v$  defined on  $\Omega := \bigcup I_k$  (which we extend to  $\mathbb{R}^n$  by setting  $v = u$  on  $\Omega^c$ ). By construction, Property 3 is satisfied.

Since the average of  $|u|$  on each  $I_k$  is  $\geq s$  (by definition), we have

$$s \sum |I_k| \leq \sum \int_{I_k} |u| dx = \int_{\Omega} |u| dx \leq \|u\|_1,$$

and Property 5 is satisfied.

If  $x \in \Omega$ , then  $x \in I_k$  for precisely one  $k$ , and therefore

$$|v(x)| = |A_{I_k}(u)| \leq A_{I_k}(|u|) \leq 2^n s$$

by (3) (which is true for all the cubes  $I_k$ , by construction). If  $x \notin \Omega$ , there is a sequence of open cubes  $J_k$  containing  $x$ , over which the average of  $|u|$  is  $< s$ , such that  $|J_k| \rightarrow 0$ . This implies that  $|u(x)| \leq s$  a.e. on  $\Omega^c$ , and since  $v = u$  on  $\Omega^c$ , we conclude that  $v$  has Property 2.

By construction,  $\text{supp } u_k \subset I_k$  and  $\int u_k dx = \int_{I_k} u dx - \int_{I_k} v dx = 0$  for all  $k \in \mathbb{N}$  (Property 1).

Since  $I_k$  are mutually disjoint and  $\text{supp } u_k \subset I_k$ , we have

$$\|v\|_1 + \sum \|u_k\|_1 = \int_{\Omega^c} |v| dx + \sum \int_{I_k} (|v| + |u_k|) dx.$$

However  $v = u$  on  $\Omega^c$  and  $u_k = u - v$  on  $I_k$ ; therefore, the right-hand side is

$$\leq \int_{\Omega^c} |u| dx + \sum_k \left( 2 \int_{I_k} |v| dx + \int_{I_k} |u| dx \right).$$

Since  $v$  has the constant value  $A_{I_k}(u)$  on  $I_k$ ,

$$\int_{I_k} |v| dx = |A_{I_k}(u)| |I_k| \leq \int_{I_k} |u| dx,$$

and Property 4 follows.  $\square$

Consider now  $u \in L^1(\mathbb{R}^n)$  with compact support and  $\|u\|_1 = 1$ . It follows from the construction in the last proof that  $v$  has compact support; by 1. in Lemma II.8.2,  $u_k$  have compact support as well, for all  $k$ . Therefore,  $Tv$  and  $Tu_k$  are well defined (for all  $k$ ), and

$$Tu = Tv + \sum_k Tu_k. \quad (4)$$

For any  $r > 0$ , we then have

$$[|Tu| > r] \subset [|Tv| > r/2] \cup \left[ \sum |Tu_k| > r/2 \right]. \quad (5)$$

Denote the sets in the above union by  $F_r$  and  $G_r$ .

Let  $B(a_k, t_k)$  be the smallest ball containing the cube  $I_k$  and let  $c_n$  be the ratio of their volumes (depends only on the dimension  $n$  of  $\mathbb{R}^n$ ). Since  $u_k$  has support in  $B(a_k, t_k)$  and  $\int u_k dx = 0$ , we have by Lemma II.8.1

$$\int_{B(a_k, 2t_k)^c} |Tu_k| dx \leq K \|u_k\|_1. \quad (6)$$

Let

$$E := \bigcup_{k=1}^{\infty} B(a_k, 2t_k).$$

Then (for  $s > 0$  given as in Lemma II.8.2)

$$\begin{aligned} |E| &\leq \sum_k |B(a_k, 2t_k)| = 2^n \sum_k |B(a_k, t_k)| \\ &= 2^n c_n \sum_k |I_k| \leq 2^n c_n / s. \end{aligned} \quad (7)$$

Therefore

$$\begin{aligned} |G_r| &= |G_r \cap E^c| + |G_r \cap E| \leq |G_r \cap E^c| + |E| \\ &\leq |G_r \cap E^c| + 2^n c_n / s. \end{aligned} \quad (8)$$

Since  $E^c \subset B(a_k, 2t_k)^c$  for all  $k$ , we have by (6)

$$\int_{E^c} |Tu_k| dx \leq K \|u_k\|_1 \quad (k = 1, 2, \dots),$$

and therefore

$$\begin{aligned} |G_r \cap E^c| &\leq (2/r) \int_{E^c} \sum |Tu_k| dx = (2/r) \sum \int_{E^c} |Tu_k| dx \\ &\leq (2/r) K \sum \|u_k\|_1 \leq (6/r) K \end{aligned}$$

by Property 4 of the functions  $u_k$  (cf. Lemma II.8.2). We then conclude from (8) that

$$|G_r| \leq 6K/r + 2^n c_n/s. \quad (9)$$

In order to get an estimate for  $|F_r|$ , we make the following.

**Hypothesis II.**

$$\|T\phi\|_2 \leq C\|\phi\|_2 \quad (\phi \in \mathcal{D}), \quad (10)$$

for some finite constant  $C > 0$ .

Since  $v$  is bounded a.e. (Property 2 in Lemma II.8.2) with compact support, it belongs to  $L^2$ , and it follows from (10) and the density of  $\mathcal{D}$  in  $L^2$  that

$$\|Tv\|_2^2 \leq C^2\|v\|_2^2 \leq C^2\|v\|_\infty\|v\|_1 \leq 3C^2 2^n s, \quad (11)$$

where Properties 2 and 4 in Lemma II.8.2 were used. Therefore

$$|F_r| \leq (4/r^2)\|Tv\|_2^2 \leq 12C^2 2^n s/r^2,$$

and we conclude from (5) and (9) that

$$|[Tu] > r| \leq 6K/r + 2^n c_n/s + 12C^2 2^n s/r^2.$$

The left-hand side being independent of  $s > 0$ , we may minimize the right-hand side with respect to  $s$ ; hence

$$|[Tu] > r| \leq C'/r, \quad (12)$$

where  $C' := 6K + 2^{n+2}\sqrt{3c_n}C$  depends linearly on the constants  $K$  and  $C$  of the hypothesis (1) and (10) (and on the dimension  $n$ ).

If  $u \in L^1$  (with compact support) is not necessarily normalized, we consider  $w = u/\|u\|_1$  (when  $\|u\|_1 > 0$ ). Then by (12) for  $w$ ,

$$|[Tu] > r| = |[Tw] > r/\|u\|_1| \leq C'\|u\|_1/r. \quad (13)$$

Since (13) is trivial when  $\|u\|_1 = 0$ , we proved the following.

**Lemma II.8.3.** *Let  $h : \mathbb{R}^n \rightarrow \mathbb{C}$  be locally Lebesgue integrable and satisfy Hypotheses I and II (where  $T$  denotes the convolution operator  $T : u \rightarrow h * u$ , originally defined on  $L_c^1$ ). Then there exists a positive constant  $C'$  depending linearly on the constants  $K$  and  $C$  of the hypothesis (and on the dimension  $n$ ) such that*

$$|[Tu] > r| \leq C'\|u\|_1/r \quad (r > 0)$$

for all  $u \in L_c^1$ .

Under the hypothesis of Lemma II.8.3, the linear operator  $T$  is of weak type  $(1, 1)$  with weak  $(1, 1)$ -norm  $\leq C'$ , and of strong (hence weak) type  $(2, 2)$  with strong (hence weak)  $(2, 2)$ -norm  $\leq C < C'$ . By the Marcinkiewicz Interpolation Theorem (Theorem 5.41), it follows that  $T$  is of strong type  $(p, p)$  with strong

$(p, p)$ -norm  $\leq A_p C'$ , for any  $p$  in the interval  $1 < p \leq 2$ , where  $A_p$  depends only on  $p$  (and is bounded when  $p$  is bounded away from 1).

Let  $\tilde{h}(x) := h(-x)$ , and let  $\tilde{T}$  be the corresponding convolution operator. Since  $\tilde{h}$  satisfies hypotheses (1) and (10) (when  $h$  does) with the same constants  $K$  and  $C$ , the operator  $\tilde{T}$  is of strong type  $(p, p)$  with strong  $(p, p)$ -norm  $\leq A_p C'$  for all  $p \in (1, 2]$ . Let  $q$  be the conjugate exponent of  $p$  (for a given  $p \in (1, 2]$ ). Then for all  $u \in \mathcal{D}$ ,

$$\begin{aligned} \|Tu\|_q &= \sup \left\{ \left| \int (Tu)v \, dx \right| ; v \in \mathcal{D}, \|v\|_p = 1 \right\} \\ &= \sup_v \left| \iint h(x-y)u(y) \, dy \, v(x) \, dx \right| \\ &= \sup_v \left| \iint \tilde{h}(y-x)v(x) \, dx \, u(y) \, dy \right| = \sup_v \left| \int (\tilde{T}v)u \, dy \right| \\ &\leq \sup_v \|\tilde{T}v\|_p \|u\|_q \leq A_p C' \|u\|_q. \end{aligned}$$

Thus,  $T$  is of strong type  $(q, q)$  with strong  $(q, q)$ -norm  $\leq A_p C'$ . Since  $q$  varies over the interval  $[2, \infty)$  when  $p$  varies in  $(1, 2]$ , we conclude that  $T$  is of strong type  $(p, p)$  with strong  $(p, p)$ -norm  $\leq A'_p C'$  for all  $p \in (1, \infty)$  ( $A'_p = A_p$  for  $p \in (1, 2]$  and  $A'_p = A_{p'}$  for  $p \in [2, \infty)$ , where  $p'$  is the conjugate exponent of  $p$ ). Observe that  $A'_p$  is a bounded function of  $p$  in any compact subset of  $(1, \infty)$ . We proved the following.

**Theorem II.8.4 (Hormander).** *Let  $h : \mathbb{R}^n \rightarrow \mathbb{C}$  be locally integrable, and let  $T : u \rightarrow h * u$  be the corresponding convolution operator (originally defined on  $L_c^1$ ). Assume Hypotheses I and II. Then for all  $p \in (1, \infty)$ ,  $T$  is of strong type  $(p, p)$  with strong  $(p, p)$ -norm  $\leq A_p C'$ , where the constant  $A_p$  depends only on  $p$  and is a bounded function of  $p$  in any compact subset of  $(1, \infty)$ , and  $C'$  depends linearly on the constants  $K$  and  $C$  of the hypothesis (and on the dimension  $n$ ).*

In order to apply Theorem II.8.4 to some special convolution operators, we need the following

**Lemma II.8.5.** *Let  $S := \{y \in \mathbb{R}^n ; 1/2 < |y| < 2\}$ . There exists  $\phi \in \mathcal{D}(S)$  with range in  $[0, 1]$  such that*

$$\sum_{k \in \mathbb{Z}} \phi(2^{-k}y) = 1 \quad (y \in \mathbb{R}^n - \{0\}).$$

(For each  $0 \neq y \in \mathbb{R}^n$ , at most two summands of the series are  $\neq 0$ .)

**Proof.** Fix  $\psi \in \mathcal{D}(S)$  such that  $\psi(x) = 1$  for  $3/4 \leq |x| \leq 3/2$ . Let  $y \in \mathbb{R}^n - \{0\}$  and  $k \in \mathbb{Z}$ . If  $\psi(2^{-k}y) \neq 0$ , then  $2^{-k}y \in S$ , that is,  $\log_2 |y| - 1 < k < \log_2 |y| + 1$ ; there are at most two values of the integer  $k$  in that range. Moreover,  $\psi(2^{-k}y) = 1$  if  $3/4 \leq 2^{-k}|y| \leq 3/2$ , that is, if  $\log_2(|y|/3) + 1 \leq k \leq \log_2(|y|/3) + 2$ ; there is at least one value of  $k$  in this range. It follows that

$$1 \leq \sum_{k \in \mathbb{Z}} \psi(2^{-k}y) < \infty$$

(there are at most two non-zero terms in the series, all terms are  $\geq 0$  and at least one term equals 1).

Define

$$\phi(y) = \frac{\psi(y)}{\sum_{j \in \mathbb{Z}} \psi(2^{-j}y)}.$$

Then  $\phi \in \mathcal{D}(S)$  has range in  $[0, 1]$  and for each  $y \neq 0$

$$\phi(2^{-k}y) = \frac{\psi(2^{-k}y)}{\sum_{j \in \mathbb{Z}} \psi(2^{-j-k}y)} = \frac{\psi(2^{-k}y)}{\sum_{j \in \mathbb{Z}} \psi(2^{-j}y)},$$

hence  $\sum_{k \in \mathbb{Z}} \phi(2^{-k}y) = 1$ . □

The following discussion will be restricted to the case  $n = 1$  for simplicity (a similar analysis can be done in the general case). Fix  $\phi$  as in Lemma II.8.5; let  $c := \max(1, \sup |\phi'|)$ , and denote  $\phi_k(y) := \phi(2^{-k}y)$  for  $k \in \mathbb{Z}$ .

Let  $f$  be a measurable complex function, locally square integrable on  $\mathbb{R}$ . Denote  $f_k := f\phi_k$ . Then

$$\begin{aligned} \text{supp } f_k &\subset 2^k \text{supp } \phi \subset 2^k S, \\ f &= \sum_{k \in \mathbb{Z}} f_k \text{ on } \mathbb{R} - \{0\}, \end{aligned}$$

and  $|f_k| \leq |f|$ .

Let  $I_k$  denote the indicator of the set  $2^k S$ . Then  $|f_k| = |f_k I_k| \leq |f I_k|$ , and therefore

$$\|f_k\|_2 \leq \|f I_k\|_2 < \infty. \quad (14)$$

Consider  $f$  and  $f_k$  as distributions, and suppose that  $Df$  (in distribution sense) is a locally square integrable (measurable) function. Since

$$|Df_k| \leq |Df|\phi_k + 2^{-k} \sup |\phi'| |f| \leq c(|Df| + 2^{-k}|f|),$$

it follows that

$$\begin{aligned} \|Df_k\|_2 &= \|(Df_k)I_k\|_2 \leq c(\|(Df)I_k\|_2 + 2^{-k}\|f I_k\|_2) \\ &= c2^{-k/2}[2^{-k/2}\|f I_k\|_2 + 2^{k/2}\|(Df)I_k\|_2]. \end{aligned} \quad (15)$$

**Notation.** We denote by  $\mathcal{H}$  the space of all measurable complex functions  $f$  on  $\mathbb{R}$ , locally square integrable on  $\mathbb{R}$ , for which

$$\|f\|_{\mathcal{H}} := \sup_{k \in \mathbb{Z}} [2^{-k/2}\|f I_k\|_2 + 2^{k/2}\|(Df)I_k\|_2] < \infty.$$

Assume  $f \in \mathcal{H}$ . It follows from (14) and (15) that

$$\|f_k\|_2 \leq 2^{k/2}\|f\|_{\mathcal{H}} \text{ and } \|Df_k\|_2 \leq c2^{-k/2}\|f\|_{\mathcal{H}} \quad (16)$$

for all  $k \in \mathbb{Z}$ .

Let  $g_k = \mathcal{F}^{-1}f_k$ . Since  $\mathcal{F}^{-1}$  is isometric on  $L^2$  and  $\mathcal{F}^{-1}D = -M\mathcal{F}^{-1}$  (where  $M$  denotes the operator of multiplication by the independent variable), we have by (16):

$$\begin{aligned} \int_{\mathbb{R}} (1 + 2^{2k}x^2)|g_k(x)|^2 dx &= \|g_k\|_2^2 + 2^{2k}\|(-M)g_k\|_2^2 \\ &= \|f_k\|_2^2 + 2^{2k}\|Df_k\|_2^2 \leq 2^k(1 + c^2)\|f\|_{\mathcal{H}}^2. \end{aligned} \quad (17)$$

Therefore, by Schwarz's inequality

$$\begin{aligned} \|g_k\|_1 &= \int_{\mathbb{R}} (1 + 2^{2k}x^2)^{-1/2}[(1 + 2^{2k}x^2)^{1/2}|g_k(x)|] dx \\ &\leq \left( \int_{\mathbb{R}} \frac{dx}{1 + 2^{2k}x^2} \right)^{1/2} \left( \int_{\mathbb{R}} (1 + 2^{2k}x^2)|g_k(x)|^2 dx \right)^{1/2} \\ &\leq \sqrt{1 + c^2}\|f\|_{\mathcal{H}} \left( \int_{\mathbb{R}} \frac{2^k dx}{1 + (2^k x)^2} \right)^{1/2} = c'\|f\|_{\mathcal{H}}, \end{aligned}$$

where  $c' = \sqrt{\pi(1 + c^2)}$ . Hence

$$|f_k| = |\mathcal{F}g_k| \leq \|g_k\|_1 \leq c'\|f\|_{\mathcal{H}} \quad (k \in \mathbb{Z}). \quad (18)$$

Consider now the 'partial sums'

$$s_m = \sum_{|k| \leq m} f_k \quad (m = 1, 2, \dots),$$

and let  $h_m := \mathcal{F}^{-1}s_m = \sum_{|k| \leq m} g_k$ .

Since at most two summands  $f_k$  are  $\neq 0$  at each point  $y \neq 0$  (and  $s_m(0) = 0$ ), we have by (18)

$$|s_m| \leq 2c'\|f\|_{\mathcal{H}} \quad (m = 1, 2, \dots). \quad (19)$$

Therefore, for all  $\psi \in \mathcal{S} := \mathcal{S}(\mathbb{R})$  and  $m \in \mathbb{N}$ ,

$$\begin{aligned} \|h_m * \psi\|_2 &= \|\mathcal{F}(h_m * \psi)\|_2 = \sqrt{2\pi}\|\mathcal{F}h_m\mathcal{F}\psi\|_2 \\ &= \sqrt{2\pi}\|s_m\mathcal{F}\psi\|_2 \leq c''\|f\|_{\mathcal{H}}\|\mathcal{F}\psi\|_2 = c''\|f\|_{\mathcal{H}}\|\psi\|_2, \end{aligned}$$

where  $c'' := 2\sqrt{2\pi}c'$ . Thus,  $h_m$  satisfy Hypothesis II of Theorem II.8.4, with  $C = c''\|f\|_{\mathcal{H}}$  independent of  $m$ .

**Claim.**  $h_m$  satisfies Hypothesis I of Theorem II.8.4 with  $K = K'\|f\|_{\mathcal{H}}$ , where  $K'$  is a constant independent of  $m$ .

Assuming the claim, it follows from Theorem II.8.4 that

$$\|h_m * \psi\|_p \leq C_p\|f\|_{\mathcal{H}}\|\psi\|_p \quad (m \in \mathbb{N}) \quad (20)$$

for all  $p \in (1, \infty)$ , where  $C_p$  is a constant depending only on  $p$ .

Let  $\psi, \chi \in \mathcal{S}$ . By (19)

$$|s_m \mathcal{F}\psi \overline{\mathcal{F}\chi}| \leq 2c' \|f\|_{\mathcal{H}} |\mathcal{F}\psi| |\mathcal{F}\chi| \in \mathcal{S} \subset L^1.$$

Since  $s_m \rightarrow f$  pointwise a.e., it follows from the Lebesgue dominated convergence theorem that

$$\lim_m \int_{\mathbb{R}} s_m \mathcal{F}\psi \overline{\mathcal{F}\chi} dx = \int_{\mathbb{R}} f \mathcal{F}\psi \overline{\mathcal{F}\chi} dx. \quad (21)$$

On the other hand, by Parseval's identity and (20) (with  $p' = p/(p-1)$ )

$$\begin{aligned} \sqrt{2\pi} \left| \int_{\mathbb{R}} s_m \mathcal{F}\psi \overline{\mathcal{F}\chi} dx \right| &= \left| \int_{\mathbb{R}} \mathcal{F}(h_m * \psi) \overline{\mathcal{F}\chi} dx \right| = \left| \int_{\mathbb{R}} (h_m * \psi) \bar{\chi} dx \right| \\ &\leq \|h_m * \psi\|_p \|\chi\|_{p'} \leq C_p \|f\|_{\mathcal{H}} \|\psi\|_p \|\chi\|_{p'} \end{aligned}$$

for all  $m \in \mathbb{N}$ . Hence

$$\left| \int_{\mathbb{R}} f \mathcal{F}\psi \overline{\mathcal{F}\chi} dx \right| \leq (2\pi)^{-1/2} C_p \|f\|_{\mathcal{H}} \|\psi\|_p \|\chi\|_{p'}. \quad (22)$$

Let  $u \in \mathcal{S}'$  be such that  $\mathcal{F}u \in \mathcal{H}$ . We may then apply (22) to  $f = \mathcal{F}u$ . Since  $f \mathcal{F}\psi = \mathcal{F}u \mathcal{F}\psi = (2\pi)^{-1/2} \mathcal{F}(u * \psi)$ , the integral on the left-hand side of (22) is equal to  $(2\pi)^{-1/2} \int_{\mathbb{R}} (u * \psi) \bar{\chi} dx$  (by Parseval's identity). Hence

$$\left| \int_{\mathbb{R}} (u * \psi) \bar{\chi} dx \right| \leq C_p \|f\|_{\mathcal{H}} \|\psi\|_p \|\chi\|_{p'} \quad (\psi, \chi \in \mathcal{S}).$$

Therefore

$$\|u * \psi\|_p \leq C_p \|f\|_{\mathcal{H}} \|\psi\|_p \quad (\psi \in \mathcal{S}). \quad (23)$$

This proves the following result (once the 'claim' above is verified).

**Theorem II.8.6.** *Let  $u \in \mathcal{S}'(\mathbb{R})$  be such that  $\mathcal{F}u \in \mathcal{H}$ . Then for each  $p \in (1, \infty)$ , the map  $T : \psi \in \mathcal{S} \rightarrow u * \psi$  is of strong type  $(p, p)$ , with strong  $(p, p)$ -norm  $\leq C_p \|\mathcal{F}u\|_{\mathcal{H}}$ , where the constant  $C_p$  depends only on  $p$ , and is a bounded function of  $p$  in any compact subset of  $(1, \infty)$ .*

**Proof of the 'claim'.** The change of variables  $tx \rightarrow x$  and  $ty \rightarrow y$  shows that we must prove the estimate

$$\int_{|x| \geq 2t} |h_m(x-y) - h_m(x)| dx \leq K \quad (24)$$

for all  $t > 0$  and  $|y| \leq t$ . Since  $h_m = \sum_{|k| \leq m} g_k$ , we consider the corresponding integrals for  $g_k$ .

$$\int_{|x| \geq 2t} |g_k(x-y) - g_k(x)| dx \leq \int_{|x| \geq 2t} |g_k(x-y)| dx + \int_{|x| \geq 2t} |g_k(x)| dx. \quad (25)$$

The change of variable  $x' = x - y$  in the first integral on the right-hand side transforms it to  $\int_{\{x'; |x'+y| \geq 2t\}} |g_k(x')| dx'$ . However, for  $|y| \leq t$ , we have



$\{x'; |x' + y| \geq 2t\} \subset \{x'; |x'| \geq t\}$ ; therefore the first integral (and trivially, the second as well) is  $\leq \int_{|x| \geq t} |g_k(x)| dx$ .

By the Cauchy–Schwarz’s inequality and (17),

$$\begin{aligned} \int_{|x| \geq t} |g_k(x)| dx &\leq \left( \int_{\mathbb{R}} (1 + 2^{2k} x^2) |g_k(x)|^2 dx \right)^{1/2} \left( \int_{|x| \geq t} \frac{dx}{2^{2k} x^2} \right)^{1/2} \\ &\leq (1 + c^2)^{1/2} \|f\|_{\mathcal{H}} 2^{-k/2} \left( \int_{|x| \geq t} x^{-2} dx \right)^{1/2} \\ &= \sqrt{2(1 + c^2)} \|f\|_{\mathcal{H}} (2^k t)^{-1/2}. \end{aligned}$$

Therefore, for all  $t > 0$  and  $|y| \leq t$ ,

$$\int_{|x| \geq 2t} |g_k(x - y) - g_k(x)| dx \leq 2\sqrt{2(1 + c^2)} \|f\|_{\mathcal{H}} (2^k t)^{-1/2}. \quad (26)$$

Another estimate of the integral on the left-hand side of (26) is obtained by writing it in the form

$$\int_{|x| \geq 2t} \left| \int_x^{x-y} g'_k(s) ds \right| dx. \quad (27)$$

The integrand of the outer integral is  $\leq \int_x^{x-y} |g'_k(s)| ds$  for  $y \leq 0$  ( $\leq \int_{x-y}^x |g'_k(s)| ds$  for  $y > 0$ ). Therefore, by Tonelli’s theorem, the expression in (27) is  $\leq \int_{\mathbb{R}} (\int_{s+y}^s dx) |g'_k(s)| ds$  for  $y \leq 0$  ( $\leq \int_{\mathbb{R}} (\int_s^{s+y} dx) |g'_k(s)| ds$  for  $y > 0$ , respectively)  $= |y| \|g'_k\|_1 \leq t \|g'_k\|_1$ .

Since  $\text{supp } f_k \subset 2^k S \subset B(0, 2^{k+1})$  and  $g_k = \mathcal{F}^{-1} f_k = \mathcal{F} \tilde{f}_k$ , the Paley–Wiener–Schwartz theorem (II.3.6) implies that  $g_k$  extends to  $\mathbb{C}$  as an entire function of exponential type  $\leq 2^{k+1}$ , and is bounded on  $\mathbb{R}$ . By Bernstein’s inequality (cf. Example in Section II.3.6, (31)) and (18)

$$\|g'_k\|_1 \leq 2^{k+1} \|g_k\|_1 \leq c' \|f\|_{\mathcal{H}} 2^{k+1},$$

and it follows that the integral on the left-hand side of (26) is  $\leq 2c' \|f\|_{\mathcal{H}} 2^k t$ . Hence (for all  $t > 0$  and  $|y| \leq t$ )

$$\int_{|x| \geq 2t} |g_k(x - y) - g_k(x)| dx \leq C \|f\|_{\mathcal{H}} \min(2^k t, (2^k t)^{-1/2}),$$

where  $C = 2\sqrt{\pi(1 + c^2)}$ . It follows that

$$\int_{|x| \geq 2t} |h_m(x - y) - h_m(x)| dx \leq C \|f\|_{\mathcal{H}} \sum_{k \in \mathbb{Z}} \min(2^k t, (2^k t)^{-1/2}). \quad (28)$$

We split the sum as  $\sum_{k \in J} + \sum_{k \in J^c}$ , where

$$J := \{k \in \mathbb{Z}; 2^k t \leq (2^k t)^{-1/2}\} = \{k; 2^k t \leq 1\} = \{k; k = -j, j \geq \log_2 t\},$$

and  $J^c := \mathbb{Z} - J$ . We have

$$\sum_{k \in J} = t \sum_{j \geq \log_2 t} \frac{1}{2^j} = \sum_{j - \log_2 t \geq 0} \frac{1}{2^{j - \log_2 t}} \leq \sum_{j - \log_2 t \geq 0} \frac{1}{2^{\lfloor j - \log_2 t \rfloor}} \leq 2.$$

Similarly

$$\begin{aligned} \sum_{k \in J^c} &= t^{-1/2} \sum_{k > -\log_2 t} (1/\sqrt{2})^k = \sum_{k + \log_2 t > 0} (1/\sqrt{2})^{k + \log_2 t} \\ &\leq \sum_{k + \log_2 t > 0} (1/\sqrt{2})^{\lfloor k + \log_2 t \rfloor} \leq \frac{1}{1 - (1/\sqrt{2})}. \end{aligned}$$

We then conclude from (28) that  $h_m$  satisfies Hypothesis I with  $K = K' \|f\|_{\mathcal{H}}$ , where  $K' = C(2 + (\sqrt{2}/\sqrt{2} - 1))$  is independent of  $m$ .  $\square$

**Notation.** Let

$$\mathcal{K} := \{f \in L^\infty; MDf \in L^\infty\},$$

where  $L^\infty := L^\infty(\mathbb{R})$ ,  $M$  denotes the multiplication by  $x$  operator, and  $D$  is understood in the distribution sense.

The norm on  $\mathcal{K}$  is

$$\|f\|_{\mathcal{K}} = \|f\|_\infty + \|MDf\|_\infty.$$

If  $f \in \mathcal{K}$ , we have for all  $k \in \mathbb{Z}$

$$2^{-k/2} \|f I_k\|_2 \leq 2^{-k/2} \|f\|_\infty |2^k S|^{1/2} = \sqrt{3} \|f\|_\infty,$$

and

$$\begin{aligned} 2^{k/2} \|(Df) I_k\|_2 &= 2^{k/2} \|(MDf)(1/x) I_k\|_2 \\ &\leq \|MDf\|_\infty \left( 2^k \int_{2^k S} x^{-2} dx \right)^{1/2} = \sqrt{3} \|MDf\|_\infty. \end{aligned}$$

Therefore

$$\|f\|_{\mathcal{H}} \leq \sqrt{3} \|f\|_{\mathcal{K}}$$

and  $\mathcal{K} \subset \mathcal{H}$ . We then have

**Corollary II.8.7.** *Let  $u \in \mathcal{S}'$  be such that  $\mathcal{F}u \in \mathcal{K}$ . Then for each  $p \in (1, \infty)$ , the map  $T : \psi \in \mathcal{S} \rightarrow u * \psi$  is of strong type  $(p, p)$ , with strong  $(p, p)$ -norm  $\leq C_p \|\mathcal{F}u\|_{\mathcal{K}}$ .*

(The constant  $C_p$  here may be taken as  $\sqrt{3}$  times the constant  $C_p$  in Theorem II.8.6.)

## II.9 Some holomorphic semigroups

**II.9.1.** We shall apply Corollary II.8.7 to the study of some holomorphic semigroups of operators and their boundary groups.

Let  $\mathbb{C}^+ = \{z \in \mathbb{C}; \Re z > 0\}$ . For any  $z \in \mathbb{C}^+$  and  $\epsilon > 0$ , consider the function

$$K_{z,\epsilon}(x) = \Gamma(z)^{-1} e^{-\epsilon x} x^{z-1} \quad (x > 0) \quad (1)$$

and  $K_{z,\epsilon}(x) = 0$  for  $x \leq 0$ .

Clearly,  $K_{z,\epsilon} \in L^1 \subset \mathcal{S}'$ , and a calculation using residues shows that

$$(\mathcal{F}K_{z,\epsilon})(y) = (1/\sqrt{2\pi})(\epsilon^2 + y^2)^{-z/2} e^{-iz \arctan(y/\epsilon)}. \quad (2)$$

We get easily from (2) that

$$(MD\mathcal{F}K_{z,\epsilon})(y) = -\frac{zy}{\epsilon + iy}(\mathcal{F}K_{z,\epsilon})(y). \quad (3)$$

Hence

$$|MD\mathcal{F}K_{z,\epsilon}| \leq |z| |\mathcal{F}K_{z,\epsilon}|. \quad (4)$$

Therefore, for all  $z = s + it$ ,  $s \in \mathbb{R}^+$ ,  $t \in \mathbb{R}$ ,

$$\|\mathcal{F}K_{z,\epsilon}\|_{\mathcal{K}} \leq (1 + |z|)\|\mathcal{F}K_{z,\epsilon}\|_{\infty} \leq (1/\sqrt{2\pi})\epsilon^{-s} e^{\pi|t|/2}(1 + |z|). \quad (5)$$

By Corollary II.8.7, it follows from (5) that the operator

$$T_{z,\epsilon} : f \rightarrow K_{z,\epsilon} * f$$

acting on  $L^p(\mathbb{R})$  ( $1 < p < \infty$ ) has  $B(L^p(\mathbb{R}))$ -norm  $\leq C \epsilon^{-s} e^{\pi|t|/2}(1 + |z|)$ , where  $C$  is a constant depending only on  $p$ . In the special case  $p = 2$ , the factor  $C(1 + |z|)$  can be omitted from the estimate, since

$$\begin{aligned} \|T_{z,\epsilon}f\|_2 &= \|K_{z,\epsilon} * f\|_2 = \|\mathcal{F}(K_{z,\epsilon} * f)\|_2 = \sqrt{2\pi}\|(\mathcal{F}K_{z,\epsilon})(\mathcal{F}f)\|_2 \\ &\leq \sqrt{2\pi}\|\mathcal{F}K_{z,\epsilon}\|_{\infty}\|f\|_2 \leq \epsilon^{-s} e^{\pi|t|/2}\|f\|_2, \end{aligned}$$

by (2) and the fact that  $\mathcal{F}$  is isometric on  $L^2(\mathbb{R})$ .

Consider  $L^p(\mathbb{R}^+)$  as the closed subspace of  $L^p(\mathbb{R})$  consisting of all (equivalence classes of) functions in  $L^p(\mathbb{R})$  vanishing a.e. on  $(-\infty, 0)$ . This is an *invariant subspace* for  $T_{z,\epsilon}$ , and

$$(T_{z,\epsilon}f)(x) = \Gamma(z)^{-1} \int_0^x (x-y)^{z-1} e^{-\epsilon(x-y)} f(y) dy \quad (f \in L^p(\mathbb{R}^+)). \quad (6)$$

We have for all  $f \in L^p(\mathbb{R}^+)$

$$\begin{aligned} \|T_{z,\epsilon}f\|_{L^p(\mathbb{R}^+)} &= \|T_{z,\epsilon}f\|_{L^p(\mathbb{R})} \leq \|T_{z,\epsilon}\|_{B(L^p(\mathbb{R}))}\|f\|_{L^p(\mathbb{R})} \\ &= \|T_{z,\epsilon}\|_{B(L^p(\mathbb{R}))}\|f\|_{L^p(\mathbb{R}^+)} \end{aligned}$$

hence

$$\|T_{z,\epsilon}\|_{B(L^p(\mathbb{R}^+))} \leq \|T_{z,\epsilon}\|_{B(L^p(\mathbb{R}))} \leq C\epsilon^{-s} e^{\pi|t|/2}(1+|z|) \quad (7)$$

for all  $z = s + it \in \mathbb{C}^+$ . (Again, the factor  $C(1+|z|)$  can be omitted in (7) in the special case  $p = 2$ .)

For  $z, w \in \mathbb{C}^+$  and  $y \in \mathbb{R}$ , we have by (14) in II.3.2 and (2)

$$\begin{aligned} [\mathcal{F}(K_{z,\epsilon} * K_{w,\epsilon})](y) &= \sqrt{2\pi}[(\mathcal{F}K_{z,\epsilon})(\mathcal{F}K_{w,\epsilon})](y) \\ &= (1/\sqrt{2\pi})(\epsilon^2 + y^2)^{-(z+w)/2} e^{-i(z+w)\arctan(y/\epsilon)} \\ &= [\mathcal{F}K_{z+w,\epsilon}](y). \end{aligned}$$

By the uniqueness property of the  $L^1$ -Fourier transform (cf. II.3.3), it follows that

$$K_{z,\epsilon} * K_{w,\epsilon} = K_{z+w,\epsilon}. \quad (8)$$

Therefore, for all  $f \in L^p$ ,

$$\begin{aligned} (T_{z,\epsilon}T_{w,\epsilon})f &= T_{z,\epsilon}(T_{w,\epsilon}f) = K_{z,\epsilon} * (K_{w,\epsilon} * f) \\ &= (K_{z,\epsilon} * K_{w,\epsilon}) * f = K_{z+w,\epsilon} * f = T_{z+w,\epsilon}f. \end{aligned}$$

Thus  $z \rightarrow T_{z,\epsilon}$  is a semigroup of operators on  $\mathbb{C}^+$ , i.e.,

$$T_{z,\epsilon}T_{w,\epsilon} = T_{z+w,\epsilon} \quad (z, w \in \mathbb{C}^+). \quad (9)$$

For any  $N > 0$ , consider the space  $L^p(0, N)$  (with Lebesgue measure) and the classical Riemann–Liouville fractional integration operators

$$(J^z f)(x) = \Gamma(z)^{-1} \int_0^x (x-y)^{z-1} f(y) dy \quad (f \in L^p(0, N)).$$

It is known that  $z \rightarrow J^z \in B(L^p(0, N))$  is strongly continuous in  $\mathbb{C}^+$ . Since

$$\|(T_{z,\epsilon} - T_{w,\epsilon})f\|_{L^p(0, N)} \leq \|(J^z - J^w)(e^{\epsilon x} f)\|_{L^p(0, N)} \quad (10)$$

the function  $z \rightarrow T_{z,\epsilon} \in B(L^p(0, N))$  is strongly continuous as well. For the space  $L^p(\mathbb{R}^+)$ , we have by (10)

$$\begin{aligned} \|T_{z,\epsilon}f - T_{w,\epsilon}f\|_{L^p(\mathbb{R}^+)}^p &\leq \|(J^z - J^w)(e^{\epsilon x} f)\|_{L^p(0, N)}^p \\ &\quad + \int_N^\infty e^{-\epsilon p x/2} |T_{z,\epsilon/2}(e^{\epsilon x/2} f)|^p dx \\ &\quad + \int_N^\infty e^{-\epsilon p x/2} |T_{w,\epsilon/2}(e^{\epsilon x/2} f)|^p dx. \end{aligned}$$

For  $f \in C_c(\mathbb{R}^+)$ ,  $e^{\epsilon x/2} f \in L^p(\mathbb{R}^+)$ , and therefore, for all  $z = s + it$  and  $w = u + iv$  in  $\mathbb{C}^+$ , it follows from (7) that the sum of the integrals over  $(N, \infty)$  can be estimated by

$$\begin{aligned} &e^{-\epsilon p N/2} C^p \left[ (\epsilon/2)^{-ps} e^{\pi p |t|/2} (1+|z|)^p \right. \\ &\quad \left. + (\epsilon/2)^{-pu} e^{\pi p |v|/2} (1+|w|)^p \right] \|e^{\epsilon x/2} f\|_{L^p(\mathbb{R}^+)}^p. \end{aligned}$$

Fix  $z \in \mathbb{C}^+$ , and let  $M$  be a bound for the expression in square brackets when  $w$  belongs to some closed disc  $\bar{B}(z, r) \subset \mathbb{C}^+$ . Given  $\delta > 0$ , we may choose  $N$  such that

$$e^{-\epsilon p N/2} C^p M \|e^{\epsilon x/2} f\|_{L^p(\mathbb{R}^+)}^p < \delta^p.$$

For this  $N$ , it then follows from the strong continuity of  $J^z$  on  $L^p(0, N)$  that

$$\limsup_{w \rightarrow z} \|(T_{z,\epsilon} - T_{w,\epsilon})f\|_{L^p(\mathbb{R}^+)} \leq \delta.$$

Thus,  $T_{w,\epsilon}f \rightarrow T_{z,\epsilon}f$  as  $w \rightarrow z$  for all  $f \in C_c(\mathbb{R}^+)$ . Since  $\|T_{w,\epsilon}\|_{B(L^p(\mathbb{R}^+))}$  is bounded for  $w$  in compact subsets of  $\mathbb{C}^+$  (by (7)) and  $C_c(\mathbb{R}^+)$  is dense in  $L^p(\mathbb{R}^+)$ , it follows that  $T_{w,\epsilon} \rightarrow T_{z,\epsilon}$  in the strong operator topology (as  $w \rightarrow z$ ). Thus the function  $z \rightarrow T_{z,\epsilon}$  is strongly continuous in  $\mathbb{C}^+$ . A similar argument (based on the corresponding known property of  $J^z$ ) shows that  $T_{z,\epsilon} \rightarrow I$  strongly, as  $z \in \mathbb{C}^+ \rightarrow 0$ . (Another way to prove this is to rely on the strong continuity of  $T_{z,\epsilon}$  at  $z = 1$ , which was proved above; by the semigroup property, it follows that  $T_{z,\epsilon}f \rightarrow f$  in  $L^p(\mathbb{R}^+)$ -norm for all  $f$  in the range of  $T_{1,\epsilon}$ , which is easily seen to be dense in  $L^p(\mathbb{R}^+)$ . Since the  $B(L^p(\mathbb{R}^+))$ -norms of  $T_{z,\epsilon}$  are uniformly bounded in the rectangle  $Q := \{z = s + it; 0 < s \leq 1, |t| \leq 1\}$ , the result follows.) An application of Morera's theorem shows now that  $T_{z,\epsilon}$  is an analytic function of  $z$  in  $\mathbb{C}^+$ . ( $T_{z,\epsilon}$  is said to be a *holomorphic semigroup on  $\mathbb{C}^+$* .)

Fix  $t \in \mathbb{R}$  and  $\delta > 0$ . For  $z \in B^+(it, \delta) := \{z \in \mathbb{C}^+; |z - it| < \delta\}$ , we have by (7)

$$\|T_{z,\epsilon}\| \leq C \max(\epsilon^{-\delta}, 1) e^{(\pi/2)(|t|+\delta)} (1 + |t| + \delta). \quad (11)$$

If  $z_n \in B^+(it, \delta)$  converge to  $it$  and  $f = T_{1,\epsilon}g$  for some  $g \in L^p(\mathbb{R}^+)$ , then

$$T_{z_n,\epsilon}f = T_{z_n+1,\epsilon}g \rightarrow T_{it+1,\epsilon}g$$

in  $L^p(\mathbb{R}^+)$ , by strong continuity of  $T_{z,\epsilon}$  at the point  $it + 1$ . Thus,  $\{T_{z_n,\epsilon}f\}$  is Cauchy in  $L^p(\mathbb{R}^+)$  for each  $f$  in the *dense* range of  $T_{1,\epsilon}$ , and therefore, by (11), it is Cauchy for all  $f \in L^p(\mathbb{R}^+)$ . If  $z'_n \in B^+(it, \delta)$  also converge to  $it$ , then  $T_{z_n,\epsilon}f - T_{z'_n,\epsilon}f \rightarrow 0$  in  $L^p(\mathbb{R}^+)$  for all  $f$  in the range of  $T_{1,\epsilon}$  (by strong continuity of  $T_{z,\epsilon}$  at  $z = 1 + it$ ), hence for all  $f \in L^p(\mathbb{R}^+)$ , by (11) and the density of the said range. Therefore the  $L^p$ -limit of the Cauchy sequence  $\{T_{z_n,\epsilon}f\}$  (for each  $f \in L^p(\mathbb{R}^+)$ ) exists and is independent of the particular sequence  $\{z_n\}$  in  $B^+(it, \delta)$ . This limit (denoted as usual  $\lim_{z \rightarrow it} T_{z,\epsilon}f$ ) defines a linear operator which will be denoted by  $T_{it,\epsilon}$ . By (11)

$$\|T_{it,\epsilon}f\|_p \leq C \max(\epsilon^{-\delta}, 1) e^{(\pi/2)(|t|+\delta)} (1 + |t| + \delta) \|f\|_p,$$

where the norms are  $L^p(\mathbb{R}^+)$ -norms. Since  $\delta > 0$  is arbitrary, we conclude that

$$\|T_{it,\epsilon}\| \leq C e^{\pi|t|/2} (1 + |t|) \quad (t \in \mathbb{R}) \quad (12)$$

where the norm is the  $B(L^p(\mathbb{R}^+))$ -norm and  $C$  depends only on  $p$ . (The factor  $C(1 + |t|)$  can be omitted in case  $p = 2$ .)

Since  $T_{w,\epsilon}$  is a bounded operator on  $L^p(\mathbb{R}^+)$  (cf. (7)), we have for each  $w \in \mathbb{C}^+$ ,  $t \in \mathbb{R}$ , and  $f \in L^p(\mathbb{R}^+)$ ,

$$T_{w,\epsilon}T_{it,\epsilon}f = \lim_{z \rightarrow it} T_{w,\epsilon}T_{z,\epsilon}f = \lim_{z \rightarrow it} T_{w+z,\epsilon}f = T_{w+it,\epsilon}f.$$

Also, by definition,

$$T_{it,\epsilon}T_{w,\epsilon}f = \lim_{z \rightarrow it} T_{z,\epsilon}T_{w,\epsilon}f = \lim_{z \rightarrow it} T_{z+w,\epsilon}f = T_{w+it,\epsilon}f.$$

Thus

$$T_{w,\epsilon}T_{it,\epsilon}f = T_{it,\epsilon}T_{w,\epsilon}f = T_{w+it,\epsilon}f \quad (13)$$

for all  $w \in \mathbb{C}^+$  and  $t \in \mathbb{R}$ . (In particular, for all  $s \in \mathbb{R}^+$  and  $t \in \mathbb{R}$ ,

$$T_{s+it,\epsilon} = T_{s,\epsilon}T_{it,\epsilon} = T_{it,\epsilon}T_{s,\epsilon})$$

Letting  $w \rightarrow is$  in (13) (for any  $s \in \mathbb{R}$ ), it follows from the definition of the operators  $T_{it,\epsilon}$  and their boundedness over  $L^p(\mathbb{R}^+)$  that

$$T_{is,\epsilon}T_{it,\epsilon}f = T_{i(s+t),\epsilon}f \quad (s, t \in \mathbb{R}; f \in L^p(\mathbb{R}^+)). \quad (14)$$

Thus  $\{T_{it,\epsilon}; t \in \mathbb{R}\}$  is a *group of operators*, called *the boundary group* of the holomorphic semigroup  $\{T_{z,\epsilon}; z \in \mathbb{C}^+\}$ . The boundary group is strongly continuous. (We use the preceding argument: for  $f = T_{1,\epsilon}g$  for some  $g \in L^p$ ,

$$T_{is,\epsilon}f = T_{1+is,\epsilon}g \rightarrow T_{1+it,\epsilon}g = T_{it,\epsilon}f$$

as  $s \rightarrow t$ , by strong continuity of  $T_{z,\epsilon}$  at  $z = 1 + it$ . By (12), the same is true for all  $f \in L^p$ , since  $T_{1,\epsilon}$  has dense range in  $L^p$ .) We formalize the above results as

**Theorem II.9.2.** *For each  $\epsilon > 0$  and  $p \in (1, \infty)$ , the family of operators  $\{T_{z,\epsilon}; z \in \mathbb{C}^+\}$  defined by (6) has the following properties:*

- (1) *It is a holomorphic semigroup of operators in  $L^p(\mathbb{R}^+)$ ;*
- (2)  *$\lim_{z \in \mathbb{C}^+ \rightarrow 0} T_{z,\epsilon} = I$  in the s.o.t.;*
- (3)  *$\|T_{z,\epsilon}\| \leq C(1 + |z|)\epsilon^{-s} e^{\pi|t|/2}$  for all  $z = s + it \in \mathbb{C}^+$ , where the norm is the operator norm on  $L^p(\mathbb{R}^+)$  and  $C$  is a constant depending only on  $p$  (in case  $p = 2$ , the factor  $C(1 + |z|)$  can be omitted).*
- (4) *The boundary group*

$$T_{it,\epsilon} := (\text{strong}) \lim_{z \in \mathbb{C}^+ \rightarrow it} T_{z,\epsilon} \quad (t \in \mathbb{R})$$

*exists, and is a strongly continuous group of operators in  $L^p(\mathbb{R}^+)$  with Properties (12) and (13).*

We may apply the theorem to the classical Riemann–Liouville semigroup  $J^z$  on  $L^p(0, N)$  ( $N > 0$ ). Elements of  $L^p(0, N)$  are regarded as elements of  $L^p(\mathbb{R}^+)$

vanishing outside the interval  $(0, N)$ . All  $p$ -norms below are  $L^p(0, N)$ -norms. The inequality (7) takes the form

$$\|T_{z,\epsilon}\|_{B(L^p(0,N))} \leq C(1+|z|)\epsilon^{-s} e^{\pi|t|/2} \quad (z = s + it \in \mathbb{C}^+). \quad (15)$$

For all  $f \in L^p(0, N)$ ,

$$\|J^z f\|_p \leq \|K_z\|_1 \|f\|_p,$$

where  $K_z(x) = \Gamma(z)^{-1} x^{z-1}$  for  $x \in (0, N)$ . Calculating the  $L^1$ -norm above we then have

$$\|J^z\| \leq \frac{N^s}{s|\Gamma(z)|} \quad (z = s + it \in \mathbb{C}^+). \quad (16)$$

Since  $0 \leq 1 - e^{-\epsilon(x-y)} \leq \epsilon(x-y)$  for  $0 \leq y \leq x \leq N$ , we get for  $z = s + it \in \mathbb{C}^+$

$$|J^z f - T_{z,\epsilon} f| \leq \epsilon \frac{\Gamma(s+1)}{|\Gamma(z)|} J^{s+1} |f|,$$

and therefore by (16)

$$\|J^z f - T_{z,\epsilon} f\|_p \leq \epsilon \frac{|z| N^{s+1}}{(s+1)|\Gamma(z+1)|} \|f\|_p. \quad (17)$$

By (15) and (17), we have for all  $z \in Q := \{z = s + it \in \mathbb{C}^+; s \leq 1, |t| \leq 1\}$

$$\|J^z f\|_p \leq \|J^z f - T_{z,1} f\|_p + \|T_{z,1} f\|_p \leq M \|f\|_p, \quad (18)$$

where

$$M = \sup_{0 < s \leq 1; |t| \leq 1} \frac{N^{s+1} \sqrt{2}}{|\Gamma(s+1+it)|} + C(1 + \sqrt{2})e^{\pi/2}.$$

Given  $t \in \mathbb{R}$ , let  $n = \lceil |t| \rceil + 1$ . Then  $z = s + it \in nQ$  for  $s \leq 1$ , and therefore, by the semigroup property (with  $w = z/n \in Q$ ),

$$\|J^z\| = \|J^{nw}\| = \|(J^w)^n\| \leq \|J^w\|^n \leq M^n = M^{\lceil |t| \rceil + 1}.$$

This inequality is surely valid in a neighbourhood  $B^+(it, \delta)$ , and the argument of the preceding section implies the existence of the boundary group  $J^{it}$ , defined as the strong limit of  $J^z$  as  $z \rightarrow it$ . By (17) and (15), we also have (for  $z = s + it \in \mathbb{C}^+$ )

$$\begin{aligned} \|J^z f\|_p &\leq \|J^z f - T_{z,\epsilon} f\|_p + \|T_{z,\epsilon} f\|_p \\ &\leq \epsilon \frac{|z| N^{s+1}}{(s+1)|\Gamma(z+1)|} \|f\|_p + C(1+|z|)\epsilon^{-s} e^{\pi|t|/2} \|f\|_p. \end{aligned}$$

Letting  $s \rightarrow 0$ , we get for all  $f \in L^p(0, N)$

$$\|J^{it} f\|_p \leq \epsilon \frac{|t| N}{|\Gamma(it+1)|} \|f\|_p + C(1+|t|)e^{\pi|t|/2} \|f\|_p.$$

Since  $\epsilon$  is arbitrary, we conclude that

$$\|J^{it}\| \leq C(1 + |t|)e^{\pi|t|/2} \quad (t \in \mathbb{R}). \quad (19)$$

As before, the factor  $C(1 + |t|)$  can be omitted in case  $p = 2$ . Formally

**Corollary II.9.3.** *For each  $p \in (1, \infty)$  and  $N > 0$ , the Riemann–Liouville semigroup  $J^z$  on  $L^p(0, N)$  has a boundary group  $J^{it}$  (defined as the strong limit of  $J^z$  as  $z \in \mathbb{C}^+ \rightarrow it$ ). The boundary group is strongly continuous, satisfies the identity  $J^{it}J^w = J^wJ^{it} = J^{w+it}$  (for all  $t \in \mathbb{R}$  and  $w \in \mathbb{C}^+$ ) and the growth relation (19).*



# Bibliography

- Akhiezer, N. I. and Glazman, I. M. (1962, 1963). *Theory of Linear Operators in Hilbert Space*, Vols I, II, Ungar, New York.
- Ash, R. B. (1972). *Real Analysis and Probability*, Academic Press, New York.
- Bachman, G. and Narici, L. (1966). *Functional Analysis*, Academic Press, New York.
- Banach, S. (1932). *Theorie des Operations Lineaires*, Hafner, New York.
- Berberian, S. K. (1961). *Introduction to Hilbert Space*, Oxford University Press, Fair Lawn, New Jersey.
- Berberian, S. K. (1965). *Measure and Integration*, Macmillan, New York.
- Berberian, S. K. (1974). *Lectures in Functional Analysis and Operator Theory*, Springer-Verlag, New York.
- Bonsall, F. F. and Duncan, J. (1973). *Complete Normed Algebras*, Springer-Verlag, Berlin.
- Bourbaki, N. (1953, 1955). *Espaces Vectoriels Topologiques*, livre V Hermann, Paris.
- Browder, A. (1968). *Introduction to Function Algebras*, Benjamin, New York.
- Brown, A. and Pearcy, C. (1977). *Introduction to Operator Theory*, Springer-Verlag, New York.
- Collojara, I. and Foias, C. (1968). *Theory of Generalized Spectral Operators*, Gordon and Breach, New York.
- Davies, E. B. (1980). *One-Parameter Semigroups*, Academic Press, London.
- Dieudonne, (1960). *Foundations of Modern Analysis*, Academic Press, New York.
- Dixmier, J. (1964). *Les  $C^*$ -algebres et leurs Representations*, Gauthier-Villar, Paris.
- Dixmier, J. (1969). *Les Algebres d'Operateurs dans l'Espace Hilbertien*, 2nd ed., Gauthier-Villar, Paris.
- Doob, J. L. (1953). *Stochastic Processes*, Wiley, New York.
- Douglas, R. G. (1972). *Banach Algebra Techniques in Operator Theory*, Academic Press, New York.
- Dowson, H. R. (1978). *Spectral Theory of Linear Operators*, Academic Press, London.

- Dunford, N. and Schwartz, J. T. (1958, 1963, 1972). *Linear Operators*, Parts I, II, III, Interscience, New York.
- Edwards, R. E. (1965). *Functional Analysis*, Holt, Rinehart, and Winston, Inc., New York.
- Friedman, A. (1963). *Generalized Functions and Partial Differential Equations*, Prentice-Hall, Englewood Cliffs, New Jersey.
- Friedman, A. (1970). *Foundations of Modern Analysis*, Holt, Rinehart, and Winston, New York.
- Gamelin, T. W. (1969). *Uniform Algebras*, Prentice-Hall, Englewood Cliffs, New Jersey.
- Goffman, C. and Pedrick, G. (1965). *First Course in Functional Analysis*, Prentice-Hall, Englewood Cliffs, New Jersey.
- Goldberg, S. (1966). *Unbounded Linear Operators*, McGraw-Hill, New York.
- Goldstein, J. A. (1985). *Semigroups of Operators and Applications*, Oxford, New York.
- Halmos, P. R. (1950). *Measure Theory*, Van Nostrand, New York.
- Halmos, P. R. (1951). *Introduction to Hilbert Space and the Theory of Spectral Multiplicity*, Chelsea, New York.
- Halmos, P. R. (1967). *A Hilbert Space Problem Book*, Van Nostrand-Reinhold, Princeton, New Jersey.
- Hewitt, E. and Ross, K. A. (1963, 1970). *Abstract Harmonic Analysis*, Vols I, II, Springer-Verlag, Berlin.
- Hewitt, E. and Stromberg, K. (1965). *Real and Abstract Analysis*, Springer-Verlag, New York.
- Hille, E. and Phillips, R. S. (1957). *Functional Analysis and Semigroups*, A.M.S. Colloq. Publ. 31, Providence, Rhode Island.
- Hirsch, F. and Lacombe, G. (1999). *Elements of Functional Analysis*, Springer, New York.
- Hoffman, K. (1962). *Banach Spaces of Analytic Functions*, Prentice-Hall, Englewood Cliffs, New Jersey.
- Hormander, L. (1960). Estimates for translation invariant operators, *Acta Math.*, **104**, 93–140.
- Hormander, L. (1963). *Linear Partial Differential Equations*, Springer-Verlag, Berlin.
- Kadison, R. V. and Ringrose, J. R. (1997). *Fundamentals of the Theory of Operator Algebras*, Vols I, II, III, A.M.S. Grad. Studies in Math., Providence, Rhode Island.
- Kantorovitz, S. (1983). *Spectral Theory of Banach Space Operators*, Lecture Notes in Math., Vol. 1012, Springer, Berlin.
- Kantorovitz, S. (1995). *Semigroups of Operators and Spectral Theory*, Pitman Research Notes in Math., Vol. 330, Longman, New York.
- Kato, T. (1966). *Perturbation Theory for Linear Operators*, Springer-Verlag, New York.

- Katznelson, Y. (1968). *An Introduction to Harmonic Analysis*, Wiley, New York.
- Kothe, G. (1969). *Topological Vector Spaces*, Springer-Verlag, New York.
- Lang, S. (1969). *Analysis II*, Addison-Wesley, Reading, Massachusetts.
- Larsen, R. (1973). *Banach Algebras*, Marcel Dekker, New York.
- Larsen, R. (1973). *Functional Analysis*, Marcel Dekker, New York.
- Loeve, M. (1963). *Probability Theory*, Van Nostrand-Reinhold, Princeton, New Jersey.
- Loomis, L. H. (1953). *An Introduction to Abstract Harmonic Analysis*, Van Nostrand, New York.
- Malliavin, P. (1995). *Integration and Probability*, Springer-Verlag, New York.
- Maurin, K. (1967). *Methods of Hilbert Spaces*, Polish Scientific Publishers, Warsaw.
- Meggison, R. E. (1998). *An Introduction to Banach Space Theory*, Springer, New York.
- Munroe, M. E. (1971). *Introduction to Measure and Integration*, 2nd ed., Addison-Wesley, Reading, Massachusetts.
- Naimark, M. A. (1959). *Normed Rings*, Noordhoff, Groningen.
- Pazy, A. (1983). *Semigroups of Linear Operators and Applications to Partial Differential Equations*, Springer, New York.
- Reed, M. and Simon, B. (1975). *Methods of Modern Mathematical Physics*, Vols I, II, Academic Press, New York.
- Rickart, C. E. (1960). *General Theory of Banach Algebras*, Van Nostrand-Reinhold, Princeton, New Jersey.
- Riesz, F. and Sz-Nagy, B. (1955). *Functional Analysis*, Ungar, New York.
- Royden, H. L. (1968). *Real Analysis*, 2nd ed., Macmillan, New York.
- Rudin, W. (1962). *Fourier Analysis on Groups*, Interscience-Wiley, New York.
- Rudin, W. (1973). *Functional Analysis*, McGraw-Hill, New York.
- Rudin, W. (1974). *Real and Complex Analysis*, 2nd ed., McGraw-Hill, New York.
- Schwartz, L. (1951). *Theorie des Distributions*, Vols I, II, Hermann, Paris.
- Stone, M. (1932). *Linear Transformations in Hilbert Space*, A.M.S. Colloq. Publ. 15, Providence, Rhode Island.
- Stout, E. L. (1971). *The Theory of Uniform Algebras*, Bogden and Quigley, New York.
- Taylor, A. E. (1958). *Introduction to Functional Analysis*, Wiley, New York.
- Tucker, H. G. (1967). *A Graduate Course in Probability*, Academic Press, New York.
- Weil, A. (1951). *L'integration dans les Groupes Topologiques et ses Applications*, 2nd ed., Act. Sci. et Ind. 869, 1145, Hermann et Cie, Paris.

- Wheeden, R. L. and Zygmund, A. (1977). *Measure and Integral*, Marcel Dekker, New York.
- Wilansky, A. (1964). *Functional Analysis*, Blaisdell, New York.
- Yosida, K. (1966). *Functional Analysis*, Springer-Verlag, Berlin.
- Zaanen, A. C. (1953). *Linear Analysis*, North-Holland, Amsterdam.
- Zygmund, A. (1959). *Trigonometric Series*, 2nd ed., Cambridge University Press, Cambridge.

# Index

- abelian groups 120
- abstract Cauchy problem (ACP) 275
- abstract measure theory 1
- abstract potential 277
- adjoints
  - of operator 165, 262
- Alaoglu's Theorem 137, 179
- Alexandroff one-point compactification 77
- algebra
  - commutative 170
  - generated by semi-algebra 58, 62
  - homomorphism 174, 191, 195, 202, 223, 239, 241, 270
- alternative hypothesis 331
- analytic operational calculus 243–7
- anti-automorphism 171
- anti-multiplicativity 181
- antisymmetric set 152
- approximate identity 121, 216, 218
- approximation almost everywhere by continuous functions 101
- Approximation Theorem 6, 13–14, 25–6, 56
- Arzela–Ascoli Theorem 249, 253
- Averages lemma 35–6, 46, 105–6, 338
- Axiom of Choice 67
  
- B\*-algebra 182–4, 188–90, 192, 195
  - commutative 182
  - general 185–9
- B\*-subalgebra 187
- Baire
  - Category Theorem 153–4
  - first category 154, 164, 219
  - second category 154, 156–8, 219–20
- Banach
  - adjoint 165
  - algebra 170, 174–7, 185, 197, 199, 202, 240, 243, 254
  - abstract 170
  - commutative 178–81, 201
  - elements, logarithms of 253–4
  - unital, complex, 170
  - space 27, 29, 31, 42, 121–2, 128, 136–9, 141, 150–1, 155–61, 164, 168, 170, 195, 206, 219–20, 226, 235, 238, 247–8, 256, 258, 261, 276–7, 280, 389–90, 403
  - reflexive 127, 240
  - uniformly convex 152
  - subalgebra 177, 196
  - subspace 235, 237, 239, 243
- Beppo Levi Theorem 13, 69–70
  - for conditional expectation 339
- Bernoulli
  - case 290, 294
  - distributions 302
  - Law of Large Numbers 292
  - Generalized 293
  - random variable 290 314
- Bernstein's inequality 387, 413
- Bessel
  - identity 209–10
  - inequality 205, 209
  - two-sided 222
- Beurling–Gelfand spectral radius formula 175

- BienAyme's identity 289, 315, 325–6, 351
- binormal density 347
- binormal distribution 347
- Bishop's Antisymmetry Theorem 142
- Bolzano–Weierstrass Theorem 310
- Borel
  - algebra 66, 84, 111, 224
  - $\sigma$ -algebra 75, 298
- Borel–Cantelli lemma 295–7
- Borel
  - functions 68, 75, 91, 94, 112, 142, 230, 233–4, 256, 286, 299, 315, 325, 334, 343–4
  - complex 225, 240, 265, 381
  - integrable 75
  - simple 68, 242
- map 2
- measurable function 113
- measure 88, 94, 96, 111–12, 224, 230, 233, 369
  - complex 11, 98, 109, 112, 122, 150, 152, 217–18, 227, 252–3, 366–7
  - positive 217
- set 2, 5, 94, 111–12, 265, 286, 316, 363
- subsets 92
- Strong Law of Large Numbers 294
- boundary group 418–420
- bounded operator 102, 153
- $C(\Gamma)$ -operational calculus 240, 268
- $C(K)$ -operational calculus 223–4, 226–7
- $C(\sigma(x))$ -operational calculus 184, 187
- calculus, fundamental theorem of 100
- Calderon–Zygmund decomposition
  - lemma 405
- canonical embedding 126–7, 153
- Cantor functions 99
- Caratheodory
  - condition 61
  - extension Theorem 62–4, 70
  - measurability condition 60
  - theory 81
- category 153
  - arguments 153
- Cauchy 28, 48, 91, 159, 236
  - condition 204
  - criterion 52
  - density 305
- Cauchy-distributed r.v.s. 305
- Cauchy distribution 356
- Cauchy integral formula 244, 246, 305, 385, 394, 397
- Cauchy Integral Theorem 244, 247, 303
- Cauchy sequences 27–8, 32, 42, 47, 104, 153, 158, 160–1, 194
- Cauchy–Schwarz inequality 30–1, 190, 194, 209, 413
- Cauchy transform 252–3
- Cayley transform 255–6, 267–9
- Central Limit Theorems *see also* Laplace
  - for equidistributed r.v.s. 315
  - for uniformly bounded r.v.s. 314
- Cesaro means 216
- characteristic functions 5
  - of r.v.s *see* random variables
- classical Volterra operator 199
- Closed Graph Theorem 159, 258, 260, 262
- closed operator 159, 259, 272, 274, 275
- commutation with translations 375
- commutants 214–15, 235
- compact closure 78–9, 90
- compact convex set 140
- compact Hausdorff space 170, 179, 181–2, 184, 201
- compact metric space 253
- compact neighbourhood 77
- compact operator 165, 248, 250
- complementary event 383
- complementary projection 206
- complete
  - measure space 21
  - normed space 27
- completeness hypothesis 157
- completion

- of measure space 21
- of normed space 27, 91, 391
- complex
  - integrable function 66
  - measure 39–46
- complex vector space 30, 103
- Composite Function Theorem 246
- conditional distribution 345–7
  - density 346
- conditional expectation 336–41, 362
- conditional probability 336–49
  - heuristics 336–7
- conditioning by a random variable 341
- confidence intervals 329–30
- conjugate automorphism 181
- conjugates
  - of  $C_c(X)$  109–11
  - of Lebesgue spaces *see* Lebesgue
- conjugate-homogeneous map 128
- conjugate-isomorphisms 128
- conjugate space 104, 123, 156 *see also*
  - normed dual 103
- continuous linear functional 102
- convergence 46–8
  - absolute 159–60
  - almost everywhere 15, 20, 28, 91
  - almost uniform 47, 48
  - in measure 48
  - modes of 46
  - on finite measure space 49–50
  - of semigroups 279
- convex hull 131, 140
- convolution 200, 373, 384
  - of distribution functions 355
  - of distributions 373, 375–6
  - and Fourier transform 75
  - of functions 376
  - on  $L^1$  75
  - on  $L^p$  121
  - operators 404–20
- core (for the generator of a
  - semigroup) 278
- covariance 288
- critical value 335
- cyclic vector 191, 234
- $\mathcal{D}, \mathcal{D}'$  373–6, 382, 385, 399
- decomposition, orthogonal 206, 208–9
- deficiency spaces 272
- delta measure 369
- DeMorgan Law 2, 294
- density 346
  - criterion 126
  - functions 346
- differentiability 93–7
- differentiation 368
- Dirichlet
  - formula 76
  - integral 219, 307, 380
- Distance theorem 32, 151
- distributions 364, 366–76
  - with compact support 372
  - with density 303
  - functions 50–3, 146
  - in  $\Omega$  367
  - see also* random variables
- divergence hypothesis 298
- Dominated Convergence theorem of
  - Lebesgue 19, 45, 280, 311, 323
  - for conditional expectation 340
- du Bois–Reymond proposition 369
- $\mathcal{E}, \mathcal{E}'$  372, 374–6, 382, 402
- Egoroff's Theorem 49, 56, 351
- Ehrenpreis–Malgrange–Hörmander
  - Theorem 393
- empty event 383
- equivalence
  - classes of integrable functions 17
  - principle of ordered sampling 290–1
- Erdős–Rényi Theorem 296
- estimation and decision 324–36
- estimator 325–7
- expectation 285
  - conditional 338
- exponential distribution density 305
- exponential formula (for semigroups
  - of operators) 280
- exponential type 386–7
- extremal points 139–43

- Fatou
  - lemma 12, 19, 27–8, 56, 194, 266, 360–1
  - Theorem 218–19
- F-density *see* Snedecor density
- finite inductive process 65
- Fisher's Theorem 318, 320–2, 324, 330
- Fourier
  - coefficients 216–17
  - expansion 210
  - inversion formula 385, 392
- Fourier–Laplace transform 383–5
- Fourier series 216, 219
- Fourier–Stieltjes transform 381
- Fourier transform 200
  - inversion formula for 378
  - on  $\mathcal{S}$  377
  - on  $\mathcal{S}'$  381
- $F_\sigma$  sets 88
- Fubini's Theorem 73, 76, 92, 118, 244, 252–3, 307, 311, 315, 347, 349, 355, 378–9
- Fubini–Tonelli Theorem 70
- Fuglede's Theorem 200
- functions, absolutely continuous 98
- fundamental solutions 392–6
  
- Gamelin, T. W. 238
- gamma density 305
- Gaussian density 303
- $G_\delta$  sets 88–9
- Gelfand–Mazur Theorem 174
- Gelfand–Naimark–Segal (GNS)
  - construction 190–5
  - Theorem 192
- Gelfand–Naimark Theorem 182–3, 185, 188
- Gelfand
  - representation 180, 182–3, 196
  - space 201
  - topology 179, 181, 183, 201
  - transform 179
- general Leibnitz formula 371
- generalized Fourier expansion 209
- generators, graph-convergence 280
- Goldstine's Theorem 137, 152
- graphs 159–60
  - norm 259
- groups of operators *see* operators
  
- $\mathcal{H}(K)$ -operational calculus 244–5
- Haar
  - functional 120
  - measure 113–20
- Hahn–Banach lemma 123, 132
- Hahn–Banach Theorem 123–7, 130, 143, 163, 166, 189, 193, 370, 393
- Hahn decomposition 46
- Hardy inequality 97–8
- Hartogs–Rosenthal Theorem 224, 229, 239, 253
- Hausdorff 88, 201
  - space 77, 111, 115, 121, 130, 141–2, 144–5, 220
  - locally compact 78–80, 89–91, 97, 107 *see also* Urysohn's lemma
  - topology 113, 133
- Heine–Borel Theorem 138
- Helly–Bray
  - lemma 308
  - Theorem 309–10
- Helly's lemma 310–11
- hermitian (positive) 188–9, 193
- hermitianity *see* sesquilinearity
- Hilbert
  - adjoint 167, 261–3
  - operation 185
  - automorphism 380
  - basis 210, 216, 220
  - dimension 211–12, 214
  - space 31–5, 102, 128–9, 185, 191–2, 194, 203–7, 209–12, 216, 220–2, 226, 229–30, 232, 234, 237, 253–4, 259, 264, 266, 272, 283
  - canonical model 213–14
  - deficiency indices 272
  - geometric properties of 32
  - isomorphism 212–13, 222, 233
  - symmetric operators in 271–4



- Hille–Yosida
  - approximations 278
  - space
    - of an arbitrary operator 279
    - maximality of 279
  - Theorem 278, 282
- Holder’s inequality 22–5, 29–30, 105, 147, 219, 387, 390
- holomorphic semigroups
  - boundary group of 418
  - of operators 415–20
- homeomorphism 113, 133, 138, 155, 171–2, 268
- homomorphism 120, 175, 179, 187, 195
- \*-homomorphism *see* isomorphism
- Hormander Theorem 409
- (hyper)cubes 405
- hypoellipticity 398–9
- hypothesis, testing of 331–6
- identity operator 98, 170, 226
- improper Riemann integrals 68
- improper Riemann–Stieltjes integral 51
- independence assumption 332
- indicator 5
- ‘inductive limit’ topology 367
- infinite divisibility 355–9
- inner derivation 198
- inner measure 82
- inner product 29–31, 191
  - space 56, 151, 203–4, 207, 220–1
- integrability 147
- integrable functions 15–22
- integration (with respect to vector measures) 224–6
- inversion formula 378, 381
- Inversion Theorem 307
- involution 181–3, 186, 190
- isometric automorphism 259, 262
- isomorphism 181, 185
  - isometric 121, 126, 130, 222
- \*-isomorphism, isometric 183–5, 195
- Jensen’s inequality 363
- joint density 315, 332
- joint distribution
  - density 347
  - function 345
- Jordan decomposition 43, 46, 95, 229
- Jordan Decomposition Theorem 98
- Kolmogorov
  - inequality 291
  - lemma 349
  - ‘Three Series Theorem’ 353–4
- Krein–Milman Theorem 139–40
- Kronecker’s delta 203
- $L^1$ 
  - convolution on 75
  - elements of 38
- $L^1$ -Fourier transform 381, 416
- $L^1$ 
  - functions, Fourier coefficients of 220
- $L^1$ -r.v. 363
- sequence of 362
- $L^2$ 
  - Fourier transform 379, 382
  - random variables 288
    - independent 353
    - series of 349–55
  - trigonometric Hilbert basis of 215
- Laplace
  - density 304
  - distribution 356
  - integral representation 280
  - transform 276
- Laplace Theorem 326 *see also* Central Limit Theorem
- Laurent series expansion 176
- Laurent Theorem 176, 246
- Lavrentiev’s Theorem 238
- Lax–Milgram Theorem 167
- Lebesgue
  - decomposition 36–8, 44, 94, 96
  - Theorem 34, 36
  - Dominated Convergence Theorem 20, 71, 231, 266, 378, 392, 412
  - integral 67–8

- Lebesgue (*cont.*)
  - measure 74–5, 92, 97, 121, 164, 200, 203, 215, 218, 224, 303, 315, 364, 366, 395, 416
  - see also* Monotone Convergence Theorem
- Lebesgue–Radon–Nikodym Theorem 34–9, 54
- Lebesgue
  - $\sigma$ -algebra 147
  - spaces 75, 97
    - conjugates of 104–9
    - operators between 145
- Lebesgue–Stieltjes measure 64, 98
- space 66
- Leibnitz
  - formula 371, 377, 384, 400, 403
- lim inf
  - of sequences 3–4, 12–13, 19, 27
  - of set 9, 155
- lim sup
  - of sequences 3–4, 19
  - of set 9, 55–6, 198
- linear functionals
  - continuous 102
  - positive 79–87, 120
- linear maps 102–4, 127–8
- Liouville’s Theorem 174
- Lipschitz constant 103
- ‘Little’ Riesz Representation Theorem 34, 129, 185
- locally compact Hausdorff space *see* Hausdorff
- locally convex t.v.s. 135
- $L^p$  381
  - elements of 419–20
- $L^p$ -spaces 22–9, 389
- Lusin’s Theorem 89, 233
- Lyapounov Central Limit Theorem 311, 314
- Lyapounov Condition 312
- Marcinkiewicz Interpolation Theorem 145, 149–50, 408
- marginal distributions 345
- martingale 362–3
- maximal element 123
- maximal ideal 178
- maximal ideal space 179
- maximum likelihood estimators (MLEs) 327–8, 330
- measurable function 1–7, 13–14, 22, 37, 46, 54–5, 66, 91, 105
  - bounded 106
  - non-negative 9
- measurable map 26, 66
- measurable sets 1–7, 73
  - structure of 63–4
- measurable simple functions 73
- measurable
  - spaces 1, 4–7, 26, 35, 54, 284
  - concept of 1
- measures
  - absolutely continuous 36, 38–9, 95
  - on algebras, extension of 62–3
  - complex 39–46, 54–5, 226
  - construction of 57
  - and functions 367–8
  - positive 7, 25, 39, 43
  - product of 70
  - space 19, 22, 55, 71, 81, 85, 97, 108, 122, 147
    - complete 21, 60, 66, 70, 73
    - extension of 21
    - positive 8–9, 12, 34, 36, 54, 104–5, 129, 164
    - properties 87
    - and topology 77
- Minkowski
  - functional 131–2
  - inequality 23
- modular function 120
- Monotone Convergence Theorem 12, 26, 51, 72–3, 107, 109, 287, 299, 339 *see also* Lebesgue
- Morera’s Theorem 417
- mouse problem 296
- multiplicativity of  $E$  on independent r.v.’s 286–8

- natural homomorphism 178
- natural embedding *see* canonical embedding
- Neumann
  - expansion 173, 404
  - series expansion 245
- Neyman–Pearson
  - lemma 331–2
  - rejection region 333–4
- Noether’s ‘First Homomorphism Theorem’ 178
- norm
  - definiteness of 6
  - topology 161–2
- normal distribution 303, 314–15, 335
- normal element 183–5
- normal operator 167, 221, 229–30, 235
- normed algebra 6, 170
- normed space 6, 17, 19, 56, 75, 91, 102–4, 109, 121, 124–6, 141, 150–2, 155–8, 160, 162, 164, 390
  - completion of 27, 159
- Open Mapping Theorem 153, 156–9
- operational calculus 228, 237
  - for a normal element 186
  - for bounded Borel functions 266–7
  - for unbounded selfadjoint operators 265–7
- operators
  - bounded 153, 276
  - closed 159, 259, 272, 274–5
  - compact 165, 248–51
  - complementary projection 168
  - conjugation 129
  - groups of 281, 418
  - linear 146, 258
  - norm 103, 153
  - normal 167
    - bounded 263
  - topologies 161–4
  - unbounded 258, 260
- Orthogonal Decomposition Theorem 33, 232, 273
- orthogonal projections 221, 226, 343
- orthonormal bases 208–11, 221
- orthonormal set 203–5, 211
  - complete 209
- outer measures 59–63, 81
- Paley–Wiener–Schwartz Theorem 384, 398, 404, 413
- parallelogram identity 30, 32
- Parseval
  - identity 209, 213, 412
  - formula for Fourier transform 379
- partition of unity 77–80, 366
- Paul Levy Continuity Theorem 311, 314, 357
- Poisson
  - distribution 302, 356
  - integrals 217–18
  - kernel 218
  - probability measure 168
  - r.v. 302
- Poisson-type distribution 358
- polarization identity 31
- probability
  - distributions 298–307
  - heuristics 283–5
  - measure 56
  - space 55, 284–98, 342
- product measure 69–73, 92
- projections 206–8
- Pythagores’ Theorem 203
- quasi-distribution function 300, 309–10
- quotient norm 160
- quotient space 160
- $\mathbb{R}^2$ -Lebesgue measure 253
- Radon–Nikodym derivative 39, 94–5
- Radon–Nikodym Theorem 34, 36, 106, 337, 341
- random variables (r.v.s) 285–6, 298, 328–9, 349, 356
  - characteristic functions (ch.f.s) of 288, 307–15, 355–6
  - distribution 290, 316–17
  - gamma-distributed 306
  - hypergeometric 291

- random variables (r.v.s) (*cont.*)
  - independent 286–8, 306, 327–8, 331, 334, 351
  - infinitely divisible 355, 358
  - Laplace-distributed 305
  - Poisson-type 359
  - real 287, 299, 344
  - sequences of 359–63
- reflexive (Banach space) 127–30, 152
- reflexivity 127–30
- regression curve 344–5
- regular
  - element 171
  - measure 88, 111–2
- regularity of solution 398–400
- regularization 364
  - of distributions 374
- rejection region 331
- renorming method 235–6
- Renorming Theorem 235
- resolution of the identity 229, 265–6
  - on  $\mathbb{Z}$  239–43
  - selfadjoint 226
- resolvent 173
  - identity 174, 260–1
  - set 172, 259
  - strong convergence 279–80
- Riemann integrals 68–9, 92, 256
- Riemann–Lebesgue lemma 76
- Riemann–Liouville
  - fractional integration operators 416
  - semigroup 418
- Riemann–Stieltjes sums 359
- Riemann versus Lebesgue 67–9
- Riesz–Dunford integral 244
- Riesz–Markov Representation Theorem 87, 100–1, 109
- Riesz
  - projection 247, 250
  - representation 226, 228
- Riesz Representation Theorem 34–5, 111–13, 143, 214, 227, 240–2, 367, 382
- Riesz–Schauder Theorem 250
- risk function 325–6
- Runge Theorem 252
- $\mathcal{S}, \mathcal{S}'$  380–2, 385, 392, 412
- sampling
  - ordered 290
  - random 284
- Schauder Theorem 249–50 *see also* Riesz
- Schwartz space 376
- Schwarz inequality 33, 35, 129, 169, 227, 288–9, 301, 380, 389, 411
- selfadjoint 181–3, 207
  - element 190
  - operator 145, 237, 255, 262, 282
  - spectral measure 226, 228–9
  - subalgebra 143, 145, 215
- semi-algebras 57–9, 63–4
- semi-continuity 99–100
- semigroups 277–8
  - generator of 275–6, 278
  - exponential formula for 281
  - of operators 256
  - strong convergence 280
- semi-inner product 30, 190
- semi-measure 57–8, 62, 64–6
- semi-norms 373
- semi-simplicity space 237–9
  - for unbounded operators in Banach space 267–71 *see also* Banach
- separation 130–2
- Separation Theorem 132
  - for topological vector spaces 134
  - strict 135, 139
- sesquilinearity 30
- simple Borel functions *see* Borel
- simple functions 11, 13, 27
- simple hypothesis 333
- $\sigma$ -additivity 7, 11, 62, 67, 83, 230, 339
- $\sigma$ -algebra 1, 8, 21, 54, 61, 66, 84, 336
  - generated 2
- $\sigma$ -compactness 88
- $\sigma$ -finite 122, 164
  - measure 54, 56, 63, 107
  - positive measure space 35, 39, 74
- $\sigma$ -subadditivity 16, 21, 233, 337
- singular function 99

- singular element 171
- Snedecor density 323–4
- Sobolev's space 391
- space of test functions 367
- spectral integral 225
- Spectral Mapping Theorems 175, 184–5, 245–6
- spectral measure 224, 226
  - on a Banach subspace 223–4
- spectral norm 180
- spectral operator 229
- spectral radius 173, 184, 257
- spectral representation 233–5
- spectral set 248
- Spectral Theorem
  - for normal operators 229–31
  - for unbounded selfadjoint operators 264–5
- spectrum 177, 186, 190, 199, 239–40, 248, 268
  - continuous 173, 231, 260
  - parts of 231–3
  - point 231, 260
  - residual 260
- standard deviation (of r.v.) 288
- star-algebra (\*-algebra) 181
- statistic 318
- Stieltjes integral 359
- Stirling's formula 323
- Stone's Theorem 282
- Stone–Weierstrass Theorem 143–5, 181, 183–4, 210, 233, 238–9
- strong operator topology (s.o.t.) 162–3, 222, 256–7
- student density 321–2, 331
- sub- $\sigma$ -algebras 362–3
- subadditivity 8, 110
- subalgebra 141, 142, 144
- submartingale 362–3
- Submartingale Convergence Theorem 363
- supermartingale 362
- support
  - compact 77
  - of a function 77
  - of a measure 92–3
  - of distribution 368
- symmetric subspace 271
  - operator 262
- $t$ -density 321
- $t$ -test 335
- Taylor formula 312–13, 389
- Taylor Theorem, non-commutative 254
- Tchebichev's inequality 291, 295, 297
- temperate distributions 376–92
- temperate weights 387–9
- Tonelli's Theorem 73, 147, 253, 317, 346, 413
- topological
  - algebra 243
  - group 113
  - space 2, 4
  - vector space 133–5
- total variation
  - of measure 42
  - of function 98
- total variation measure 42, 45
- translation invariance 66–7, 119
  - of Lebesgue measure 121
- translation invariant 114, 116–17, 120
- translation operators 375
- triangular array 356
  - of r.v.s 355
- 'truncated' r.v.s 354
- truncation of functions 52–3
- Tychonoff's Theorem 115, 137
- Uniform Boundedness Theorem 153, 155, 196, 256
  - general version 154–5
- uniform convexity 151
- uniform norm 5, 78, 91, 103, 109, 143
- uniform operator topology 161
- uniformly integrable 359
- unimodular locally compact group 120
- uniqueness of extension 64
- Uniqueness Theorem for characteristic functions 355
- unitary equivalence 212
- Urysohn's lemma 77–8, 86, 88, 90, 92, 201 *see also* Hausdorff

- variance (of r.v.) 288
- variable coefficients 400–4
- vector-valued random variables  
315–24
- Vitali–Caratheodory Theorem 100
- Von Neumann’s Double Commutant  
Theorem 214
  
- weak and strong types 145–9
- Weak Law of Large Numbers 293
- weak operator topology (w.o.t.)  
162–3, 214, 220, 222
  
- weak topology 135–9
- weak\*-topology 151, 165, 179
- weighted  $L^p$ -spaces 389
- Weierstrass Approximation Theorem,  
classical 143
- Wiener’s Theorem 255
- $\mathcal{W}_{p,k}$  spaces 387–92
  
- $z$ -test 335
- zero hypothesis 324, 326, 331, 335
- Zero-one law 298
- Zorn’s lemma 123, 140, 234