Functional Analysis

Gerald Teschl

Gerald Teschl
Fakultät für Mathematik
Nordbergstraße 15
Universität Wien
1090 Wien, Austria

E-mail: Gerald.Teschl@univie.ac.at
URL: http://www.mat.univie.ac.at/~gerald/

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Abstract. This manuscript provides a brief introduction to Functional Analysis. It covers basic Hilbert and Banach space theory including Lebesgue spaces and their duals (no knowledge about Lebesgue integration is assumed).

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Preface

The present manuscript was written for my course *Functional Analysis* given at the University of Vienna in Winter 2004.

It is available from

`http://www.mat.univie.ac.at/~gerald/ftp/book-fa/`

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Gerald Teschl

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Chapter 0

Introduction

Functional analysis is an important tool in the investigation of all kind of problems in pure mathematics, physics, biology, economics, etc.. In fact, it is hard to find a branch in science where functional analysis is not used.

The main objects are (infinite dimensional) linear spaces with different concepts of convergence. The classical theory focuses on linear operators (i.e., functions) between these spaces but nonlinear operators are of course equally important. However, since one of the most important tools in investigating nonlinear mappings is linearization (differentiation), linear functional analysis will be our first topic in any case.

0.1. Linear partial differential equations

Rather than overwhelming you with a vast number of classical examples I want to focus on one: linear partial differential equations. We will use this example as a guide throughout this first chapter and will develop all necessary method for a successful treatment of our particular problem.

In his investigation of heat conduction Fourier was lead to the (one dimensional) **heat** or diffusion equation

$$\frac{\partial}{\partial t} u(t,x) = \frac{\partial^2}{\partial x^2} u(t,x), \qquad (0.1)$$

Here $u(t,x)$ is the temperature distribution at time t at the point x. It is usually assumed, that the temperature at $x = 0$ and $x = 1$ is fixed, say $u(t,0) = a$ and $u(t,1) = b$. By considering $u(t,x) \to u(t,x) - a - (b-a)x$ it is clearly no restriction to assume $a = b = 0$. Moreover, the initial temperature distribution $u(0,x) = u_0(x)$ is assumed to be know as well.

Since finding the solution seems at first sight not possible, we could try to find at least some solutions of (0.1) first. We could for example make an ansatz for $u(t,x)$ as a product of two functions, each of which depends on only one variable, that is,

$$u(t,x) = w(t)y(x). \tag{0.2}$$

This ansatz is called **separation of variables**. Plugging everything into the heat equation and bringing all t, x dependent terms to the left, right side, respectively, we obtain

$$\frac{\dot{w}(t)}{w(t)} = \frac{y''(x)}{y(x)}. \tag{0.3}$$

Here the dot refers to differentiation with respect to t and the prime to differentiation with respect to x.

Now if this equation should hold for all t and x, the quotients must be equal to a constant $-\lambda$. That is, we are lead to the equations

$$-\dot{w}(t) = \lambda w(t) \tag{0.4}$$

and

$$-y''(x) = \lambda y(x), \qquad y(0) = y(1) = 0 \tag{0.5}$$

which can easily be solved. The first one gives

$$w(t) = c_1 \mathrm{e}^{-\lambda t} \tag{0.6}$$

and the second one

$$y(x) = c_2 \cos(\sqrt{\lambda} x) + c_3 \sin(\sqrt{\lambda} x). \tag{0.7}$$

However, $y(x)$ must also satisfy the boundary conditions $y(0) = y(1) = 0$. The first one $y(0) = 0$ is satisfied if $c_2 = 0$ and the second one yields (c_3 can be absorbed by $w(t)$)

$$\sin(\sqrt{\lambda}) = 0, \tag{0.8}$$

which holds if $\lambda = (\pi n)^2$, $n \in \mathbb{N}$. In summary, we obtain the solutions

$$u_n(t,x) = c_n \mathrm{e}^{-(\pi n)^2 t} \sin(n\pi x), \qquad n \in \mathbb{N}. \tag{0.9}$$

So we have found a large number of solutions, but we still have not dealt with our initial condition $u(0,x) = u_0(x)$. This can be done using the superposition principle which holds since our equation is linear. In fact, choosing

$$u(t,x) = \sum_{n=1}^{\infty} c_n \mathrm{e}^{-(\pi n)^2 t} \sin(n\pi x), \tag{0.10}$$

where the coefficients c_n decay sufficiently fast, we obtain further solutions of our equation. Moreover, these solutions satisfy

$$u(0,x) = \sum_{n=1}^{\infty} c_n \sin(n\pi x) \tag{0.11}$$

and expanding the initial conditions into Fourier series

$$u_0(x) = \sum_{n=1}^{\infty} u_{0,n} \sin(n\pi x), \tag{0.12}$$

we see that the solution of our original problem is given by (0.10) if we choose $c_n = u_{0,n}$.

Of course for this last statement to hold we need to ensure that the series in (0.10) converges and that we can interchange summation and differentiation. You are asked to do so in Problem 0.1.

In fact many equations in physics can be solved in a similar way:

- **Reaction-Diffusion equation**:

$$\begin{aligned}
&\frac{\partial}{\partial t} u(t,x) - \frac{\partial^2}{\partial x^2} u(t,x) + q(x) u(t,x) = 0,\\
&u(0,x) = u_0(x),\\
&u(t,0) = u(t,1) = 0.
\end{aligned} \tag{0.13}$$

Here $u(t,x)$ could be the density of some gas in a pipe and $q(x) > 0$ describes that a certain amount per time is removed (e.g., by a chemical reaction).

- **Wave equation**:

$$\begin{aligned}
&\frac{\partial^2}{\partial t^2} u(t,x) - \frac{\partial^2}{\partial x^2} u(t,x) = 0,\\
&u(0,x) = u_0(x), \quad \frac{\partial u}{\partial t}(0,x) = v_0(x)\\
&u(t,0) = u(t,1) = 0.
\end{aligned} \tag{0.14}$$

Here $u(t,x)$ is the displacement of a vibrating string which is fixed at $x = 0$ and $x = 1$. Since the equation is of second order in time, both the initial displacement $u_0(x)$ and the initial velocity $v_0(x)$ of the string need to be known.

- **Schrödinger equation**:

$$\begin{aligned}
&\mathrm{i}\frac{\partial}{\partial t} u(t,x) = -\frac{\partial^2}{\partial x^2} u(t,x) + q(x) u(t,x),\\
&u(0,x) = u_0(x),\\
&u(t,0) = u(t,1) = 0.
\end{aligned} \tag{0.15}$$

Here $|u(t,x)|^2$ is the probability distribution of a particle trapped in a box $x \in [0,1]$ and $q(x)$ is a given external potential which describes the forces acting on the particle.

All these problems (and many others) leads to the investigation of the following problem

$$Ly(x) = \lambda y(x), \qquad L = -\frac{d^2}{dx^2} + q(x), \tag{0.16}$$

subject to the **boundary conditions**

$$y(a) = y(b) = 0. \tag{0.17}$$

Such a problem is called **Sturm–Liouville boundary value problem**. Our example shows that we should prove the following facts about our Sturm–Liouville problems:

(i) The Sturm–Liouville problem has a countable number of eigenvalues E_n with corresponding eigenfunctions $u_n(x)$, that is, $u_n(x)$ satisfies the boundary conditions and $Lu_n(x) = E_n u_n(x)$.

(ii) The eigenfunctions u_n are complete, that is, any *nice* function $u(x)$ can be expanded into a generalized Fourier series

$$u(x) = \sum_{n=1}^{\infty} c_n u_n(x).$$

This problem is very similar to the eigenvalue problem of a matrix and we are looking for a generalization of the well-known fact that every symmetric matrix has an orthonormal basis of eigenvectors. However, our linear operator L is now acting on some space of functions which is not finite dimensional and it is not at all what even orthogonal should mean for functions. Moreover, since we need to handle infinite series, we need convergence and hence define the distance of two functions as well.

Hence our program looks as follows:

- What is the distance of two functions? This automatically leads us to the problem of convergence and completeness.
- If we additionally require the concept of orthogonality, we are lead to Hilbert spaces which are the proper setting for our eigenvalue problem.
- Finally, the spectral theorem for compact symmetric operators will be the solution of our above problem

Problem 0.1. *Find conditions for the initial distribution $u_0(x)$ such that (0.10) is indeed a solution (i.e., such that interchanging the order of summation and differentiation is admissible).*

Chapter 1

A first look at Banach and Hilbert spaces

1.1. Warm up: Metric and topological spaces

Before we begin I want to recall some basic facts from metric and topological spaces. I presume that you are familiar with these topics from your calculus course.

A **metric space** is a space X together with a function $d : X \times X \to \mathbb{R}$ such that

(i) $d(x, y) \geq 0$

(ii) $d(x, y) = 0$ if and only if $x = y$

(iii) $d(x, y) = d(y, x)$

(iv) $d(x, z) \leq d(x, y) + d(y, z)$ (triangle inequality)

If (ii) does not hold, d is called a **semi-metric**.

Example. Euclidean space $\mathbb{R}^n$ together with $d(x, y) = (\sum_{k=1}^n (x_k - y_k)^2)^{1/2}$ is a metric space and so is $\mathbb{C}^n$ together with $d(x, y) = (\sum_{k=1}^n |x_k - y_k|^2)^{1/2}$. $\diamond$

The set

$$B_r(x) = \{y \in X | d(x, y) < r\} \tag{1.1}$$

is called an **open ball** around x with radius $r > 0$. A point x of some set U is called an **interior point** of U if U contains some ball around x. If x is an interior point of U, then U is also called a **neighborhood** of x. A point x is called a **limit point** of U if $B_r(x) \cap (U \backslash \{x\}) \neq \emptyset$ for every ball. Note that a limit point need not lie in U, but U contains points arbitrarily close to x. Moreover, x is not a limit point of U if and only if it is an interior point of the complement of U.

Example. Consider $\mathbb{R}$ with the usual metric and let $U = (-1, 1)$. Then every point $x \in U$ is an interior point of U. The points ± 1 are limit points of U. ⋄

A set consisting only of interior points is called **open**. The family of open sets $\mathcal{O}$ satisfies the following properties

(i) $\emptyset, X \in \mathcal{O}$

(ii) $O_1, O_2 \in \mathcal{O}$ implies $O_1 \cap O_2 \in \mathcal{O}$

(iii) $\{O_\alpha\} \subseteq \mathcal{O}$ implies $\bigcup_\alpha O_\alpha \in \mathcal{O}$

That is, $\mathcal{O}$ is closed under finite intersections and arbitrary unions.

In general, a space X together with a family of sets $\mathcal{O}$, the open sets, satisfying (i)–(iii) is called a **topological space**. The notions of interior point, limit point, and neighborhood carry over to topological spaces if we replace open ball by open set.

There are usually different choices for the topology. Two usually not very interesting examples are the **trivial topology** $\mathcal{O} = \{\emptyset, X\}$ and the **discrete topology** $\mathcal{O} = \mathfrak{P}(X)$ (the powerset of X). Given two topologies $\mathcal{O}_1$ and $\mathcal{O}_2$ on X, $\mathcal{O}_1$ is called **weaker** (or **coarser**) than $\mathcal{O}_2$ if and only if $\mathcal{O}_1 \subseteq \mathcal{O}_2$.

Example. Note that different metrics can give rise to the same topology. For example, we can equip $\mathbb{R}^n$ (or $\mathbb{C}^n$) with the Euclidean distance as before, or we could also use

$$\tilde{d}(x, y) = \sum_{k=1}^{n} |x_k - y_k| \tag{1.2}$$

Since

$$\frac{1}{\sqrt{n}} \sum_{k=1}^{n} |x_k| \le \sqrt{\sum_{k=1}^{n} |x_k|^2} \le \sum_{k=1}^{n} |x_k| \tag{1.3}$$

shows $B_{r/\sqrt{n}}((x, y)) \subseteq \tilde{B}_r((x, y)) \subseteq B_r((x, y))$, where B, $\tilde{B}$ are balls computed using d, $\tilde{d}$, respectively. Hence the topology is the same for both metrics. ⋄

Example. We can always replace a metric d by the bounded metric

$$\tilde{d}(x, y) = \frac{d(x, y)}{1 + d(x, y)} \tag{1.4}$$

without changing the topology. ⋄

Every subspace Y of a topological space X becomes a topological space of its own if we call $O \subseteq Y$ open if there is some open set $\tilde{O} \subseteq X$ such that $O = \tilde{O} \cap Y$ (**induced topology**).

Example. The set $(0,1] \subseteq \mathbb{R}$ is not open in the topology of $X = \mathbb{R}$, but it is open in the incuded topology when considered as a subset of $Y = [-1,1]$. $\diamond$

A family of open sets $\mathcal{B} \subseteq \mathcal{O}$ is called a **base** for the topology if for each x and each neighborhood $U(x)$, there is some set $O \in \mathcal{B}$ with $x \in O \subseteq U$. Since $O = \bigcap_{x\in O} U(x)$ we have

Lemma 1.1. *If $\mathcal{B} \subseteq \mathcal{O}$ is a base for the topology, then every open set can be written as a union of elements from $\mathcal{B}$.*

If there exists a countable base, then X is called **second countable**.

Example. By construction the open balls $B_{1/n}(x)$ are a base for the topology in a metric space. In the case of $\mathbb{R}^n$ (or $\mathbb{C}^n$) it even suffices to take balls with rational center and hence $\mathbb{R}^n$ (and $\mathbb{C}^n$) are second countable. $\diamond$

A topological space is called **Hausdorff space** if for two different points there are always two disjoint neighborhoods.

Example. Any metric space is a Hausdorff space: Given two different points x and y the balls $B_{d/2}(x)$ and $B_{d/2}(y)$, where $d = d(x,y) > 0$, are disjoint neighborhoods (a semi-metric space will not be Hausdorff). $\diamond$

The complement of an open set is called a **closed set**. It follows from de Morgan's rules that the family of closed sets $\mathcal{C}$ satisfies

(i) $\emptyset, X \in \mathcal{C}$
(ii) $C_1, C_2 \in \mathcal{C}$ implies $C_1 \cup C_2 \in \mathcal{C}$
(iii) $\{C_\alpha\} \subseteq \mathcal{C}$ implies $\bigcap_\alpha C_\alpha \in \mathcal{C}$

That is, closed sets are closed under finite unions and arbitrary intersections.

The smallest closed set containing a given set U is called the **closure**

$$\overline{U} = \bigcap_{C\in\mathcal{C}, U\subseteq C} C, \tag{1.5}$$

and the largest open set contained in a given set U is called the **interior**

$$U^\circ = \bigcup_{O\in\mathcal{O}, O\subseteq U} O. \tag{1.6}$$

It is straightforward to check that

Lemma 1.2. *Let X be a topological space, then the interior of U is the set of all interior points of U and the closure of U is the set of all limit points of U.*

A sequence $(x_n)_{n=1}^\infty \subseteq X$ is said to **converge** to some point $x \in X$ if $d(x, x_n) \to 0$. We write $\lim_{n\to\infty} x_n = x$ as usual in this case. Clearly the limit is unique if it exists (this is not true for a semi-metric).

Every convergent sequence is a **Cauchy sequence**, that is, for every $\varepsilon > 0$ there is some $N \in \mathbb{N}$ such that

$$d(x_n, x_m) \le \varepsilon \qquad n, m \ge N. \tag{1.7}$$

If the converse is also true, that is, if every Cauchy sequence has a limit, then X is called **complete**.

Example. Both $\mathbb{R}^n$ and $\mathbb{C}^n$ are complete metric spaces. ⋄

A point x is clearly a limit point of U if and only if there is some sequence $x_n \in U$ converging to x. Hence

Lemma 1.3. *A closed subset of a complete metric space is again a complete metric space.*

Note that convergence can also be equivalently formulated in terms of topological terms: A sequence x_n converges to x if and only if for every neighborhood U of x there is some $N \in \mathbb{N}$ such that $x_n \in U$ for $n \ge N$. In a Hausdorff space the limit is unique.

A metric space is called **separable** if it contains a countable dense set. A set U is called **dense**, if its closure is all of X, that is if $\overline{U} = X$.

Lemma 1.4. *Let X be a separable metric space. Every subset of X is again separable.*

Proof. Let $A = \{x_n\}_{n\in\mathbb{N}}$ be a dense set in X. The only problem is that $A \cap Y$ might contain no elements at all. However, some elements of A must be at least arbitrarily close: Let $J \subseteq \mathbb{N}^2$ be the set of all pairs (n, m) for which $B_{1/m}(x_n) \cap Y \ne \emptyset$ and choose some $y_{n,m} \in B_{1/m}(x_n) \cap Y$ for all $(n, m) \in J$. Then $B = \{y_{n,m}\}_{(n,m)\in J} \subseteq Y$ is countable. To see that B is dense choose $y \in Y$. Then there is some sequence x_{n_k} with $d(x_{n_k}, y) < 1/4$. Hence $(n_k, k) \in J$ and $d(y_{n_k,k}, y) \le d(y_{n_k,k}, x_{n_k}) + d(x_{n_k}, y) \le 2/k \to 0$. □

A function between metric spaces X and Y is called continuous at a point $x \in X$ if for every $\varepsilon > 0$ we can find a $\delta > 0$ such that

$$d_Y(f(x), f(y)) \le \varepsilon \quad \text{if} \quad d_X(x, y) < \delta. \tag{1.8}$$

If f is continuous at every point it is called **continuous**.

Lemma 1.5. *Let X be a metric space. The following are equivalent*

(i) *f is continuous at x (i.e, (1.8) holds).*

(ii) *$f(x_n) \to f(x)$ whenever $x_n \to x$*

(iii) *For every neighborhood V of $f(x)$, $f^{-1}(V)$ is a neighborhood of x.*

Proof. (i) $\Rightarrow$ (ii) is obvious. (ii) $\Rightarrow$ (iii): If (iii) does not hold there is a neighborhood V of $f(x)$ such that $B_\delta(x) \not\subseteq f^{-1}(V)$ for every δ. Hence we can choose a sequence $x_n \in B_{1/n}(x)$ such that $f(x_n) \notin f^{-1}(V)$. Thus $x_n \to x$ but $f(x_n) \not\to f(x)$. (iii) $\Rightarrow$ (i): Choose $V = B_\varepsilon(f(x))$ and observe that by (iii) $B_\delta(x) \subseteq f^{-1}(V)$ for some δ. □

The last item implies that f is continuous if and only if the inverse image of every open (closed) set is again open (closed).

Note: In a topological space, (iii) is used as definition for continuity. However, in general (ii) and (iii) will no longer be equivalent unless one uses generalized sequences, so called nets, where the index set $\mathbb{N}$ is replaced by arbitrary directed sets.

If X and X are metric spaces then $X \times Y$ together with

$$d((x_1, y_1), (x_2, y_2)) = d_X(x_1, x_2) + d_Y(y_1, y_2) \tag{1.9}$$

is a metric space. A sequence (x_n, y_n) converges to (x, y) if and only if $x_n \to x$ and $y_n \to y$. In particular, the projections onto the first $(x, y) \mapsto x$ respectively onto the second $(x, y) \mapsto y$ coordinate are continuous.

In particular, by

$$|d(x_n, y_n) - d(x, y)| \le d(x_n, x) + d(y_n, y) \tag{1.10}$$

we see that $d : X \times X \to \mathbb{R}$ is continuous.

Example. If we consider $\mathbb{R} \times \mathbb{R}$ we do not get the Euclidean distance of $\mathbb{R}^2$ unless we modify (1.9) as follows:

$$\tilde{d}((x_1, y_1), (x_2, y_2)) = \sqrt{d_X(x_1, x_2)^2 + d_Y(y_1, y_2)^2}. \tag{1.11}$$

As noted in our previous example, the topology (and thus also convergence/continuity) is independent of this choice. ⋄

If X and Y are just topological spaces, the **product topology** is defined by calling $O \subseteq X \times Y$ open if for every point $(x, y) \in O$ there are open neighborhoods U of x and V of y such that $U \times V \subseteq O$. In the case of metric spaces this clearly agrees with the topology defined via the product metric (1.9).

A **cover** of a set $Y \subseteq X$ is a family of sets $\{U_\alpha\}$ such that $Y \subseteq \bigcup_\alpha U_\alpha$. A cover is call open if all U_α are open. A subset of $\{U_\alpha\}$ is called a **subcover**.

A subset $K \subset X$ is called **compact** if every open cover has a finite subcover.

Lemma 1.6. *A topological space is compact if and only if it has the* **finite intersection property***: The intersection of a family of closed sets is empty if and only if the intersection of some finite subfamily is empty.*

Proof. By taking complements, to every family of open sets there is a corresponding family of closed sets and vice versa. Moreover, the open sets are a cover if and only if the corresponding closed sets have empty intersection. □

A subset $K \subset X$ is called **sequentially compact** if every sequence has a convergent subsequence.

Lemma 1.7. *Let X be a topological space.*

(i) *The continuous image of a compact set is compact.*
(ii) *Every closed subset of a compact set is compact.*
(iii) *If X is Hausdorff, any compact set is closed.*
(iv) *The product of compact sets is compact.*
(v) *A compact set is also sequentially compact.*

Proof. (i) Just observe that if $\{O_\alpha\}$ is an open cover for $f(Y)$, then $\{f^{-1}(O_\alpha)\}$ is one for Y.

(ii) Let $\{O_\alpha\}$ be an open cover for the closed subset Y. Then $\{O_\alpha\} \cup \{X \backslash Y\}$ is an open cover for X.

(iii) Let $Y \subseteq X$ be compact. We show that $X \backslash Y$ is open. Fix $x \in X \backslash Y$ (if $Y = X$ there is nothing to do). By the definition of Hausdorff, for every $y \in Y$ there are disjoint neighborhoods $V(y)$ of y and $U_y(x)$ of x. By compactness of Y, there are $y_1, \dots y_n$ such that $V(y_j)$ cover Y. But then $U(x) = \bigcap_{j=1}^n U_{y_j}(x)$ is a neighborhood of x which does not intersect Y.

(iv) Let $\{O_\alpha\}$ be an open cover for $X \times Y$. For every $(x, y) \in X \times Y$ there is some $\alpha(x, y)$ such that $(x, y) \in O_{\alpha(x,y)}$. By definition of the product topology there is some open rectangle $U(x, y) \times V(x, y) \subseteq O_{\alpha(x,y)}$. Hence for fixed x, $\{V(x, y)\}_{y \in Y}$ is an open cover of Y. Hence there are finitely many points $y_k(x)$ such $V(x, y_k(x))$ cover Y. Set $U(x) = \bigcap_k U(x, y_k(x))$. Since finite intersections of open sets are open, $\{U(x)\}_{x \in X}$ is an open cover and there are finitely many points x_j such $U(x_j)$ cover X. By construction, $U(x_j) \times V(x_j, y_k(x_j)) \subseteq O_{\alpha(x_j, y_k(x_j))}$ cover $X \times Y$.

(v) Let x_n be a sequence which has no convergent subsequence. Then $K = \{x_n\}$ has no limit points and is hence compact by (ii). For every n there is a ball $B_{\varepsilon_n}(x_n)$ which contains only finitely many elements of K. However, finitely many suffice to cover K, a contradiction. □

In a metric space compact and sequentially compact are equivalent.

Lemma 1.8. *Let X be a metric space. Then a subset is compact if and only if it is sequentially compact.*

Proof. First of all note that every cover of open balls with fixed radius $\varepsilon > 0$ has a finite subcover. Since if this were false we could construct a sequence $x_n \in X \setminus \bigcup_{m=1}^{n-1} B_\varepsilon(x_m)$ such that $d(x_n, x_m) > \varepsilon$ for $m < n$.

In particular, we are done if we can show that for every open cover $\{O_\alpha\}$ there is some $\varepsilon > 0$ such that for every x we have $B_\varepsilon(x) \subseteq O_\alpha$ for some $\alpha = \alpha(x)$. Indeed, choosing $\{x_k\}_{k=1}^n$ such that $B_\varepsilon(x_k)$ is a cover, we have that $O_{\alpha(x_k)}$ is a cover as well.

So it remains to show that there is such an ε. If there were none, for every $\varepsilon > 0$ there must be an x such that $B_\varepsilon(x) \not\subseteq O_\alpha$ for every α. Choose $\varepsilon = \frac{1}{n}$ and pick a corresponding x_n. Since X is sequentially compact, it is no restriction to assume x_n converges (after maybe passing to a subsequence). Let $x = \lim x_n$, then x lies in some O_α and hence $B_\varepsilon(x) \subseteq O_\alpha$. But choosing n so large that $\frac{1}{n} < \frac{\varepsilon}{2}$ and $d(x_n, x) < \frac{\varepsilon}{2}$ we have $B_{1/n}(x_n) \subseteq B_\varepsilon(x) \subseteq O_\alpha$ contradicting our assumption. □

Please also recall the **Heine-Borel theorem**:

Theorem 1.9 (Heine-Borel)**.** *In $\mathbb{R}^n$ (or $\mathbb{C}^n$) a set is compact if and only if it is bounded and closed.*

Proof. By Lemma 1.7 (ii) and (iii) it suffices to show that a closed interval in $I \subseteq \mathbb{R}$ is compact. Moreover, by Lemma 1.8 it suffices to show that every sequence in $I = [a, b]$ has a convergent subsequence. Let x_n be our sequence and divide $I = [a, \frac{a+b}{2}] \cup [\frac{a+b}{2}]$. Then at least one of these two intervals, call it I_1, contains infinitely many elements of our sequence. Let $y_1 = x_{n_1}$ be the first one. Subdivide I_1 and pick $y_2 = x_{n_2}$, with $n_2 > n_1$ as before. Proceeding like this we obtain a Cauchy sequence y_n (note that by construction $I_{n+1} \subseteq I_n$ and hence $|y_n - y_m| \leq \frac{b-a}{n}$ for $m \geq n$). □

A topological space is called **locally compact** if every point has a compact neighborhood.

Example. $\mathbb{R}^n$ is locally compact. ⋄

The **distance** between a point $x \in X$ and a subset $Y \subseteq X$ is

$$\operatorname{dist}(x, Y) = \inf_{y \in Y} d(x, y). \tag{1.12}$$

Note that $x \in \overline{Y}$ if and only if $\operatorname{dist}(x, Y) = 0$.

Lemma 1.10. *Let X be a metric space, then*

$$|\operatorname{dist}(x, Y) - \operatorname{dist}(z, Y)| \leq \operatorname{dist}(x, z). \tag{1.13}$$

In particular, $x \mapsto \operatorname{dist}(x, Y)$ *is continuous.*

Proof. Taking the infimum in the triangle inequality $d(x,y) \le d(x,z) + d(z,y)$ shows $\operatorname{dist}(x,Y) \le d(x,z) + \operatorname{dist}(z,Y)$. Hence $\operatorname{dist}(x,Y) - \operatorname{dist}(z,Y) \le \operatorname{dist}(x,z)$. Interchanging x and z shows $\operatorname{dist}(z,Y) - \operatorname{dist}(x,Y) \le \operatorname{dist}(x,z)$. □

Lemma 1.11 (Urysohn)**.** *Suppose* C_1 *and* C_2 *are disjoint closed subsets of a metric space* X*. Then there is a continuous function* $f : X \to [0,1]$ *such that* f *is zero on* C_1 *and one on* C_2*.*

If X *is locally compact and* C_1 *is compact, one can choose* f *with compact support.*

Proof. To prove the first claim set $f(x) = \frac{\operatorname{dist}(x,C_2)}{\operatorname{dist}(x,C_1)+\operatorname{dist}(x,C_2)}$. For the second claim, observe that there is an open set O such that $\overline{O}$ is compact and $C_1 \subset O \subset \overline{O} \subset X \backslash C_2$. In fact, for every x, there is a ball $B_\varepsilon(x)$ such that $\overline{B_\varepsilon(x)}$ is compact and $\overline{B_\varepsilon(x)} \subset X \backslash C_2$. Since C_1 is compact, finitely many of them cover C_1 and we can choose the union of those balls to be O. Now replace C_2 by $X \backslash \overline{O}$. □

Note that Urysohn's lemma implies that a metric space is **normal**, that is, for any two disjoint closed sets C_1 and C_2, there are disjoint open sets O_1 and O_2 such that $C_j \subseteq O_j$, $j = 1, 2$. In fact, choose f as in Urysohn's lemma and set $O_1 = f^{-1}([0,1/2))$ respectively $O_2 = f^{-1}((1/2,1])$.

Lemma 1.12. *Let* X *be a locally compact metric space. Suppose* K *is a compact set and* $\{O_j\}_{j=1}^n$ *an open cover. Then there is a continuous functions* $h_j : X \to [0,1]$ *such that* h_j *has compact support contained in* O_j *and*

$$\sum_{j=1}^{n} h_j(x) \le 1 \tag{1.14}$$

with equality for $x \in K$*.*

Proof. For every $x \in K$ there is some ε and some j such that $\overline{B_\varepsilon(x)} \subseteq O_j$. By compactness of K, finitely many of these balls cover K. Let K_j be the union of those balls which lie inside O_j. By Urysohn's lemma there are functions $g_j : X \to [0,1]$ such that $g_j = 1$ on K_j and $g_j = 0$ on $X \backslash O_j$. Now set

$$h_j = g_j \prod_{k=1}^{j-1} (1 - g_k) \tag{1.15}$$

Then $h_j : X \to [0,1]$ has compact support contained in O_j and

$$\sum_{j=1}^{n} h_j(x) = 1 - \prod_{j=1}^{n} (1 - g_j(x)) \tag{1.16}$$

shows that the sum is one for $x \in K$, since $x \in K_j$ for some j implies $g_j(x) = 1$ and causes the product to vanish. □

1.2. The Banach space of continuous functions

So let us start with the set of continuous functions $C(I)$ on a compact interval $I = [a, b] \subset \mathbb{R}$. Since we want to handle complex models (e.g., the Schrödinger equation) as well, we will always consider complex valued functions!

One way of declaring a distance, well-known from calculus, is the **maximum norm**:

$$\|f(x) - g(x)\|_\infty = \max_{x\in I} |f(x) - g(x)|. \tag{1.17}$$

It is not hard to see that with this definition $C(I)$ becomes a normed linear space:

A **normed linear space** X is a vector space X over $\mathbb{C}$ (or $\mathbb{R}$) with a real-valued function (the **norm**) $\|.\|$ such that

- $\|f\| \geq 0$ for all $f \in X$ and $\|f\| = 0$ if and only if $f = 0$,
- $\|\lambda f\| = |\lambda| \, \|f\|$ for all $\lambda \in \mathbb{C}$ and $f \in X$, and
- $\|f + g\| \leq \|f\| + \|g\|$ for all $f, g \in X$ (**triangle inequality).**

From the triangle inequality we also get the inverse triangle inequality (Problem 1.1)

$$|\|f\| - \|g\|| \leq \|f - g\|. \tag{1.18}$$

Once we have a norm, we have a **distance** $d(f, g) = \|f-g\|$ and hence we know when a sequence of vectors f_n **converges** to a vector f. We will write $f_n \to f$ or $\lim_{n\to\infty} f_n = f$, as usual, in this case. Moreover, a mapping $F : X \to Y$ between two normed spaces is called **continuous** if $f_n \to f$ implies $F(f_n) \to F(f)$. In fact, it is not hard to see that the norm, vector addition, and multiplication by scalars are continuous (Problem 1.2).

In addition to the concept of convergence we have also the concept of a **Cauchy sequence** and hence the concept of completeness: A normed space is called **complete** if every Cauchy sequence has a limit. A complete normed space is called a **Banach space**.

Example. The space $\ell^1(\mathbb{N})$ of all sequences $a = (a_j)_{j=1}^\infty$ for which the norm

$$\|a\|_1 = \sum_{j=1}^\infty |a_j| \tag{1.19}$$

is finite, is a Banach space.

To show this, we need to verify three things: (i) $\ell^1(\mathbb{N})$ is a Vector space, that is closed under addition and scalar multiplication (ii) $\|.\|_1$ satisfies the three requirements for a norm and (iii) $\ell^1(\mathbb{N})$ is complete.

First of all observe

$$\sum_{j=1}^k |a_j + b_j| \le \sum_{j=1}^k |a_j| + \sum_{j=1}^k |b_j| \le \|a\|_1 + \|b\|_1 \tag{1.20}$$

for any finite k. Letting $k \to \infty$ we conclude that $\ell^1(\mathbb{N})$ is closed under addition and that the triangle inequality holds. That $\ell^1(\mathbb{N})$ is closed under scalar multiplication and the two other properties of a norm are straightforward. It remains to show that $\ell^1(\mathbb{N})$ is complete. Let $a^n = (a_j^n)_{j=1}^\infty$ be a Cauchy sequence, that is, for given $\varepsilon > 0$ we can find an N_ε such that $\|a^m - a^n\|_1 \le \varepsilon$ for $m, n \ge N_\varepsilon$. This implies in particular $|a_j^m - a_j^n| \le \varepsilon$ for any fixed j. Thus a_j^n is a Cauchy sequence for fixed j and by completeness of $\mathbb{C}$ has a limit: $\lim_{n\to\infty} a_j^n = a_j$. Now consider

$$\sum_{j=1}^k |a_j^m - a_j^n| \le \varepsilon \tag{1.21}$$

and take $m \to \infty$:

$$\sum_{j=1}^k |a_j - a_j^n| \le \varepsilon. \tag{1.22}$$

Since this holds for any finite k we even have $\|a - a_n\|_1 \le \varepsilon$. Hence $(a - a_n) \in \ell^1(\mathbb{N})$ and since $a_n \in \ell^1(\mathbb{N})$ we finally conclude $a = a_n + (a - a_n) \in \ell^1(\mathbb{N})$. ⋄

Example. The space $\ell^\infty(\mathbb{N})$ of all bounded sequences $a = (a_j)_{j=1}^\infty$ together with the norm

$$\|a\|_\infty = \sup_{j\in\mathbb{N}} |a_j| \tag{1.23}$$

is a Banach space (Problem 1.3). ⋄

Now what about convergence in this space? A sequence of functions $f_n(x)$ converges to f if and only if

$$\lim_{n\to\infty} \|f - f_n\| = \lim_{n\to\infty} \sup_{x\in I} |f_n(x) - f(x)| = 0. \tag{1.24}$$

That is, in the language of real analysis, f_n converges uniformly to f. Now let us look at the case where f_n is only a Cauchy sequence. Then $f_n(x)$ is clearly a Cauchy sequence of real numbers for any fixed $x \in I$. In particular, by completeness of $\mathbb{C}$, there is a limit $f(x)$ for each x. Thus we get a limiting function $f(x)$. Moreover, letting $m \to \infty$ in

$$|f_m(x) - f_n(x)| \le \varepsilon \qquad \forall m, n > N_\varepsilon,\ x \in I \tag{1.25}$$

we see

$$|f(x) - f_n(x)| \le \varepsilon \qquad \forall n > N_\varepsilon,\ x \in I, \tag{1.26}$$

that is, $f_n(x)$ converges uniformly to $f(x)$. However, up to this point we don't know whether it is in our vector space $C(I)$ or not, that is, whether it is continuous or not. Fortunately, there is a well-known result from real analysis which tells us that the uniform limit of continuous functions is again continuous. Hence $f(x) \in C(I)$ and thus every Cauchy sequence in $C(I)$ converges. Or, in other words

Theorem 1.13. *$C(I)$ with the maximum norm is a Banach space.*

Next we want to know if there is a basis for $C(I)$. In order to have only countable sums, we would even prefer a countable basis. If such a basis exists, that is, if there is a set $\{u_n\} \subset X$ of linearly independent vectors such that every element $f \in X$ can be written as

$$f = \sum_n c_n u_n, \qquad c_n \in \mathbb{C}, \tag{1.27}$$

then the **span** $\operatorname{span}\{u_n\}$ (the set of all finite linear combinations) of $\{u_n\}$ is dense in X. A set whose span is dense is called **total** and if we have a total set, we also have a countable dense set (consider only linear combinations with rational coefficients – show this). A normed linear space containing a countable dense set is called **separable**.

Example. The Banach space $\ell^1(\mathbb{N})$ is separable. In fact, the set of vectors δ^n, with $\delta^n_n = 1$ and $\delta^n_m = 0$, $n \neq m$ is total: Let $a \in \ell^1(\mathbb{N})$ be given and set $a^n = \sum_{k=1}^n a_k \delta^k$, then

$$\|a - a^n\|_1 = \sum_{j=n+1}^{\infty} |a_j| \to 0 \tag{1.28}$$

since $a^n_j = a_j$ for $1 \le j \le n$ and $a^n_j = 0$ for $j > n$. ⋄

Luckily this is also the case for $C(I)$:

Theorem 1.14 (Weierstraß). *Let I be a compact interval. Then the set of polynomials is dense in $C(I)$.*

Proof. Let $f(x) \in C(I)$ be given. By considering $f(x) - f(a) + (f(b) - f(a))(x - b)$ it is no loss to assume that f vanishes at the boundary points. Moreover, without restriction we only consider $I = [\frac{-1}{2}, \frac{1}{2}]$ (why?).

Now the claim follows from the lemma below using

$$u_n(x) = \frac{1}{I_n}(1-x^2)^n, \tag{1.29}$$

where

$$\begin{aligned} I_n &= \int_{-1}^{1}(1-x^2)^n dx = \frac{n!}{\frac{1}{2}(\frac{1}{2}+1)\cdots(\frac{1}{2}+n)} \\ &= \sqrt{\pi}\frac{\Gamma(1+n)}{\Gamma(\frac{3}{2}+n)} = \sqrt{\frac{\pi}{n}}(1+O(\frac{1}{n})). \end{aligned} \tag{1.30}$$

(Remark: The integral is known as Beta function and the asymptotics follow from Stirling's formula.) □

Lemma 1.15 (Smoothing)**.** *Let $u_n(x)$ be a sequence of nonnegative continuous functions on $[-1,1]$ such that*

$$\int_{|x|\le 1} u_n(x)dx = 1 \quad \textit{and} \quad \int_{\delta\le|x|\le 1} u_n(x)dx \to 0, \quad \delta>0. \tag{1.31}$$

(In other words, u_n has mass one and concentrates near $x=0$ as $n\to\infty$.)

Then for every $f \in C[-\frac{1}{2},\frac{1}{2}]$ which vanishes at the endpoints, $f(-\frac{1}{2}) = f(\frac{1}{2}) = 0$, we have that

$$f_n(x) = \int_{-1/2}^{1/2} u_n(x-y)f(y)dy \tag{1.32}$$

converges uniformly to $f(x)$.

Proof. Since f is uniformly continuous, for given ε we can find a δ (independent of x) such that $|f(x)-f(y)|\le\varepsilon$ whenever $|x-y|\le\delta$. Moreover, we can choose n such that $\int_{\delta\le|y|\le1} u_n(y)dy \le \varepsilon$. Now abbreviate $M=\max\{1,|f|\}$ and note

$$|f(x)-\int_{-1/2}^{1/2} u_n(x-y)f(x)dy| = |f(x)|\,|1-\int_{-1/2}^{1/2} u_n(x-y)dy| \le M\varepsilon. \tag{1.33}$$

In fact, either the distance of x to one of the boundary points $\pm\frac{1}{2}$ is smaller than δ and hence $|f(x)|\le\varepsilon$ or otherwise the difference between one and the integral is smaller than ε.

Using this we have

$$\begin{aligned} |f_n(x) - f(x)| &\le \int_{-1/2}^{1/2} u_n(x-y)|f(y)-f(x)|dy + M\varepsilon \\ &\le \int_{|y|\le 1/2, |x-y|\le\delta} u_n(x-y)|f(y)-f(x)|dy \\ &\quad + \int_{|y|\le 1/2, |x-y|\ge\delta} u_n(x-y)|f(y)-f(x)|dy + M\varepsilon \\ &= \varepsilon + 2M\varepsilon + M\varepsilon = (1+3M)\varepsilon, \end{aligned} \qquad (1.34)$$

which proves the claim. □

Note that f_n will be as smooth as u_n, hence the title smoothing lemma. The same idea is used to approximate noncontinuous functions by smooth ones (of course the convergence will no longer be uniform in this case).

Corollary 1.16. *C(I) is separable.*

The same is true for $\ell^1(\mathbb{N})$, but not for $\ell^\infty(\mathbb{N})$ (Problem 1.4)!

Problem 1.1. *Show that* $|\|f\| - \|g\|| \le \|f - g\|$.

Problem 1.2. *Show that the norm, vector addition, and multiplication by scalars are continuous. That is, if* $f_n \to f$, $g_n \to g$, *and* $\lambda_n \to \lambda$ *then* $\|f_n\| \to \|f\|$, $f_n + g_n \to f + g$, *and* $\lambda_n g_n \to \lambda g$.

Problem 1.3. *Show that* $\ell^\infty(\mathbb{N})$ *is a Banach space.*

Problem 1.4. *Show that* $\ell^\infty(\mathbb{N})$ *is not separable (Hint: Consider sequences which take only the value one and zero. How many are there? What is the distance between two such sequences?).*

1.3. The geometry of Hilbert spaces

So it looks like $C(I)$ has all the properties we want. However, there is still one thing missing: How should we define orthogonality in $C(I)$? In Euclidean space, two vectors are called **orthogonal** if their scalar product vanishes, so we would need a scalar product:

Suppose $\mathfrak{H}$ is a vector space. A map $\langle ., .. \rangle : \mathfrak{H} \times \mathfrak{H} \to \mathbb{C}$ is called skew linear form if it is conjugate linear in the first and linear in the second argument, that is,

$$\begin{array}{rcl} \langle \lambda_1 f_1 + \lambda_2 f_2, g \rangle &=& \lambda_1^* \langle f_1, g \rangle + \lambda_2^* \langle f_2, g \rangle \\ \langle f, \lambda_1 g_1 + \lambda_2 g_2 \rangle &=& \lambda_1 \langle f, g_1 \rangle + \lambda_2 \langle f, g_2 \rangle \end{array}, \quad \lambda_1, \lambda_2 \in \mathbb{C}, \qquad (1.35)$$

where '$*$' denotes complex conjugation. A skew linear form satisfying the requirements

(i) $\langle f, f\rangle > 0$ for $f \neq 0$ (positive definite)

(ii) $\langle f, g\rangle = \langle g, f\rangle^*$ (symmetry)

is called **inner product** or **scalar product**. Associated with every scalar product is a norm

$$\|f\| = \sqrt{\langle f, f\rangle}. \tag{1.36}$$

The pair $(\mathfrak{H}, \langle ., ..\rangle)$ is called **inner product space**. If $\mathfrak{H}$ is complete it is called a **Hilbert space**.

Example. Clearly $\mathbb{C}^n$ with the usual scalar product

$$\langle a, b\rangle = \sum_{j=1}^{n} a_j^* b_j \tag{1.37}$$

is a (finite dimensional) Hilbert space. ⋄

Example. A somewhat more interesting example is the Hilbert space $\ell^2(\mathbb{N})$, that is, the set of all sequences

$$\Big\{(a_j)_{j=1}^{\infty} \Big| \sum_{j=1}^{\infty} |a_j|^2 < \infty\Big\} \tag{1.38}$$

with scalar product

$$\langle a, b\rangle = \sum_{j=1}^{\infty} a_j^* b_j. \tag{1.39}$$

(Show that this is in fact a separable Hilbert space! Problem 1.5) ⋄

Of course I still owe you a proof for the claim that $\sqrt{\langle f, f\rangle}$ is indeed a norm. Only the triangle inequality is nontrivial which will follow from the Cauchy-Schwarz inequality below.

A vector $f \in \mathfrak{H}$ is called **normalized** or **unit vector** if $\|f\| = 1$. Two vectors $f, g \in \mathfrak{H}$ are called **orthogonal** or **perpendicular** ($f \perp g$) if $\langle f, g\rangle = 0$ and **parallel** if one is a multiple of the other.

For two orthogonal vectors we have the **Pythagorean theorem**:

$$\|f + g\|^2 = \|f\|^2 + \|g\|^2, \qquad f \perp g, \tag{1.40}$$

which is one line of computation.

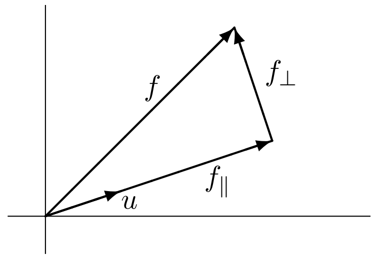
Suppose u is a unit vector, then the projection of f in the direction of u is given by

$$f_\parallel = \langle u, f\rangle u \tag{1.41}$$

and $f_\perp$ defined via

$$f_\perp = f - \langle u, f\rangle u \tag{1.42}$$

is perpendicular to u since $\langle u, f_\perp\rangle = \langle u, f - \langle u, f\rangle u\rangle = \langle u, f\rangle - \langle u, f\rangle\langle u, u\rangle = 0$.



Taking any other vector parallel to u it is easy to see

$$\|f - \lambda u\|^2 = \|f_\perp + (f_\| - \lambda u)\|^2 = \|f_\perp\|^2 + |\langle u, f\rangle - \lambda|^2 \tag{1.43}$$

and hence $f_\| = \langle u, f\rangle u$ is the unique vector parallel to u which is closest to f.

As a first consequence we obtain the **Cauchy-Schwarz-Bunjakowski** inequality:

Theorem 1.17 (Cauchy-Schwarz-Bunjakowski)**.** *Let $\mathfrak{H}_0$ be an inner product space, then for every $f, g \in \mathfrak{H}_0$ we have*

$$|\langle f, g\rangle| \le \|f\| \, \|g\| \tag{1.44}$$

with equality if and only if f and g are parallel.

Proof. It suffices to prove the case $\|g\| = 1$. But then the claim follows from $\|f\|^2 = |\langle g, f\rangle|^2 + \|f_\perp\|^2$. □

Note that the Cauchy-Schwarz inequality entails that the scalar product is continuous in both variables, that is, if $f_n \to f$ and $g_n \to g$ we have $\langle f_n, g_n\rangle \to \langle f, g\rangle$.

As another consequence we infer that the map $\|.\|$ is indeed a norm.

$$\|f + g\|^2 = \|f\|^2 + \langle f, g\rangle + \langle g, f\rangle + \|g\|^2 \le (\|f\| + \|g\|)^2. \tag{1.45}$$

But let us return to $C(I)$. Can we find a scalar product which has the maximum norm as associated norm? Unfortunately the answer is no! The reason is that the maximum norm does not satisfy the parallelogram law (Problem 1.7).

Theorem 1.18 (Jordan-von Neumann)**.** *A norm is associated with a scalar product if and only if the* **parallelogram law**

$$\|f + g\|^2 + \|f - g\|^2 = 2\|f\|^2 + 2\|g\|^2 \tag{1.46}$$

holds.

In this case the scalar product can be recovered from its norm by virtue of the **polarization identity**

$$\langle f, g\rangle = \frac{1}{4}\left(\|f + g\|^2 - \|f - g\|^2 + \mathrm{i}\|f - \mathrm{i}g\|^2 - \mathrm{i}\|f + \mathrm{i}g\|^2\right). \tag{1.47}$$

Proof. If an inner product space is given, verification of the parallelogram law and the polarization identity is straight forward (Problem 1.6).

To show the converse, we define

$$s(f,g)=\frac{1}{4}\left(\|f+g\|^2-\|f-g\|^2+\mathrm{i}\|f-\mathrm{i}g\|^2-\mathrm{i}\|f+\mathrm{i}g\|^2\right). \tag{1.48}$$

Then $s(f,f)=\|f\|^2$ and $s(f,g)=s(g,f)^*$ are straightforward to check. Moreover, another straightforward computation using the parallelogram law shows

$$s(f,g)+s(f,h)=2s(f,\frac{g+h}{2}). \tag{1.49}$$

Now choosing $h=0$ (and using $s(f,0)=0$) shows $s(f,g)=2s(f,\frac{g}{2})$ and thus $s(f,g)+s(f,h)=s(f,g+h)$. Furthermore, by induction we infer $\frac{m}{2^n}s(f,g)=s(f,\frac{m}{2^n}g)$, that is $\lambda s(f,g)=s(f,\lambda g)$ for every positive rational λ. By continuity (check this!) this holds for all $\lambda>0$ and $s(f,-g)=-s(f,g)$ respectively $s(f,\mathrm{i}g)=\mathrm{i}\,s(f,g)$ finishes the proof. □

Note that the parallelogram law and the polarization identity even hold for skew linear forms (Problem 1.6).

But how do we define a scalar product on $C(I)$? One possibility is

$$\langle f,g\rangle=\int_a^b f^*(x)g(x)dx. \tag{1.50}$$

The corresponding inner product space is denoted by $\mathcal{L}^2_{cont}(I)$. Note that we have

$$\|f\|\le\sqrt{|b-a|}\|f\|_\infty \tag{1.51}$$

and hence the maximum norm is stronger than the $\mathcal{L}^2_{cont}$ norm.

Suppose we have two norms $\|.\|_1$ and $\|.\|_2$ on a space X. Then $\|.\|_2$ is said to be **stronger** than $\|.\|_1$ if there is a constant $m>0$ such that

$$\|f\|_1\le m\|f\|_2. \tag{1.52}$$

It is straightforward to check that

Lemma 1.19. *If $\|.\|_2$ is stronger than $\|.\|_1$, then any $\|.\|_2$ Cauchy sequence is also a $\|.\|_1$ Cauchy sequence.*

Hence if a function $F: X\to Y$ is continuous in $(X,\|.\|_1)$ it is also continuous in $(X,\|.\|_2)$ and if a set is dense in $(X,\|.\|_2)$ it is also dense in $(X,\|.\|_1)$.

In particular, $\mathcal{L}^2_{cont}$ is separable. But is it also complete? Unfortunately the answer is no:

Example. Take $I = [0, 2]$ and define

$$f_n(x) = \begin{cases} 0, & 0 \le x \le 1 - \frac{1}{n} \\ 1 + n(x-1), & 1 - \frac{1}{n} \le x \le 1 \\ 1, & 1 \le x \le 2 \end{cases} \tag{1.53}$$

then $f_n(x)$ is a Cauchy sequence in $\mathcal{L}^2_{cont}$, but there is no limit in $\mathcal{L}^2_{cont}$! Clearly the limit should be the step function which is 0 for $0 \le x < 1$ and 1 for $1 \le x \le 2$, but this step function is discontinuous (Problem 1.8)! ⋄

This shows that in infinite dimensional spaces different norms will give raise to different convergent sequences! In fact, the key to solving problems in infinite dimensional spaces is often finding the right norm! This is something which cannot happen in the finite dimensional case.

Theorem 1.20. *If X is a finite dimensional case, then all norms are equivalent. That is, for given two norms $\|.\|_1$ and $\|.\|_2$ there are constants m_1 and m_2 such that*

$$\frac{1}{m_2}\|f\|_1 \le \|f\|_2 \le m_1 \|f\|_1. \tag{1.54}$$

Proof. Clearly we can choose a basis u_j, $1 \le j \le n$, and assume that $\|.\|_2$ is the usual Euclidean norm, $\|\sum_j \alpha_j u_j\|_2^2 = \sum_j |\alpha_j|^2$. Let $f = \sum_j \alpha_j u_j$, then by the triangle and Cauchy Schwartz inequalities

$$\|f\|_1 \le \sum_j |\alpha_j| \|u_j\|_1 \le \sqrt{\sum_j \|u_j\|_1^2} \|f\|_2 \tag{1.55}$$

and we can choose $m_2 = \sqrt{\sum_j \|u_j\|_1}$.

In particular, if f_n is convergent with respect to $\|.\|_2$ it is also convergent with respect to $\|.\|_1$. Thus $\|.\|_1$ is continuous with respect to $\|.\|_2$ and attains its minimum $m > 0$ on the unit sphere (which is compact by the Heine-Borel theorem). Now choose $m_1 = 1/m$. □

Problem 1.5. *Show that $\ell^2(\mathbb{N})$ is a separable Hilbert space.*

Problem 1.6. *Let $s(f, g)$ be a skew linear form and $p(f) = s(f, f)$ the associated quadratic form. Prove the parallelogram law*

$$p(f + g) + p(f - g) = 2p(f) + 2p(g) \tag{1.56}$$

and the polarization identity

$$s(f, g) = \frac{1}{4}\left(p(f + g) - p(f - g) + \mathrm{i}\, p(f - \mathrm{i}g) - \mathrm{i}\, p(f + \mathrm{i}g)\right). \tag{1.57}$$

Problem 1.7. *Show that the maximum norm (on $C[0, 1]$) does not satisfy the parallelogram law.*

Problem 1.8. *Prove the claims made about f_n, defined in (1.53), in the last example.*

1.4. Completeness

Since $\mathcal{L}^2_{cont}$ is not complete, how can we obtain a Hilbert space out of it? Well the answer is simple: take the **completion**.

If X is a (incomplete) normed space, consider the set of all Cauchy sequences $\tilde{X}$. Call two Cauchy sequences equivalent if their difference converges to zero and denote by $\bar{X}$ the set of all equivalence classes. It is easy to see that $\bar{X}$ (and $\tilde{X}$) inherit the vector space structure from X. Moreover,

Lemma 1.21. *If x_n is a Cauchy sequence, then $\|x_n\|$ converges.*

Consequently the norm of a Cauchy sequence $(x_n)_{n=1}^\infty$ can be defined by $\|(x_n)_{n=1}^\infty\| = \lim_{n\to\infty} \|x_n\|$ and is independent of the equivalence class (show this!). Thus $\bar{X}$ is a normed space ($\tilde{X}$ is not! why?).

Theorem 1.22. *$\bar{X}$ is a Banach space containing X as a dense subspace if we identify $x \in X$ with the equivalence class of all sequences converging to x.*

Proof. (Outline) It remains to show that $\bar{X}$ is complete. Let $\xi_n = [(x_{n,j})_{j=1}^\infty]$ be a Cauchy sequence in $\bar{X}$. Then it is not hard to see that $\xi = [(x_{j,j})_{j=1}^\infty]$ is its limit. □

Let me remark that the completion $\overline{X}$ is unique. More precisely any other complete space which contains X as a dense subset is isomorphic to $\overline{X}$. This can for example be seen by showing that the identity map on X has a unique extension to $\overline{X}$ (compare Theorem 1.25 below).

In particular it is no restriction to assume that a normed linear space or an inner product space is complete. However, in the important case of $\mathcal{L}^2_{cont}$ it is somewhat inconvenient to work with equivalence classes of Cauchy sequences and hence we will give a different characterization using the Lebesgue integral later.

1.5. Bounded operators

A linear map A between two normed spaces X and Y will be called a **(linear) operator**

$$A : \mathfrak{D}(A) \subseteq X \to Y. \tag{1.58}$$

The linear subspace $\mathfrak{D}(A)$ on which A is defined, is called the **domain** of A and is usually required to be dense. The **kernel**

$$\mathrm{Ker}(A) = \{f \in \mathfrak{D}(A) | Af = 0\} \tag{1.59}$$

and **range**

$$\mathrm{Ran}(A) = \{Af \mid f \in \mathfrak{D}(A)\} = A\mathfrak{D}(A) \tag{1.60}$$

are defined as usual. The operator A is called **bounded** if the following operator norm

$$\|A\| = \sup_{\|f\|_X = 1} \|Af\|_Y \tag{1.61}$$

is finite.

The set of all bounded linear operators from X to Y is denoted by $\mathfrak{L}(X,Y)$. If $X = Y$ we write $\mathfrak{L}(X,X) = \mathfrak{L}(X)$.

Theorem 1.23. *The space $\mathfrak{L}(X,Y)$ together with the operator norm (1.61) is a normed space. It is a Banach space if Y is.*

Proof. That (1.61) is indeed a norm is straightforward. If Y is complete and A_n is a Cauchy sequence of operators, then $A_n f$ converges to an element g for every f. Define a new operator A via $Af = g$. By continuity of the vector operations, A is linear and by continuity of the norm $\|Af\| = \lim_{n\to\infty} \|A_n f\| \le (\lim_{n\to\infty} \|A_n\|)\|f\|$ it is bounded. Furthermore, given $\varepsilon > 0$ there is some N such that $\|A_n - A_m\| \le \varepsilon$ for $n, m \ge N$ and thus $\|A_n f - A_m f\| \le \varepsilon \|f\|$. Taking the limit $m \to \infty$ we see $\|A_n f - Af\| \le \varepsilon\|f\|$, that is $A_n \to A$. □

By construction, a bounded operator is Lipschitz continuous

$$\|Af\|_Y \le \|A\| \|f\|_X \tag{1.62}$$

and hence continuous. The converse is also true

Theorem 1.24. *An operator A is bounded if and only if it is continuous.*

Proof. Suppose A is continuous but not bounded. Then there is a sequence of unit vectors u_n such that $\|Au_n\| \ge n$. Then $f_n = \frac{1}{n} u_n$ converges to 0 but $\|Af_n\| \ge 1$ does not converge to 0. □

Moreover, if A is bounded and densely defined, it is no restriction to assume that it is defined on all of X.

Theorem 1.25. *Let $A \in \mathfrak{L}(X,Y)$ and let Y be a Banach space. If $\mathfrak{D}(A)$ is dense, there is a unique (continuous) extension of A to X, which has the same norm.*

Proof. Since a bounded operator maps Cauchy sequences to Cauchy sequences, this extension can only be given by

$$Af = \lim_{n\to\infty} Af_n, \qquad f_n \in \mathfrak{D}(A), \quad f \in X. \tag{1.63}$$

To show that this definition is independent of the sequence $f_n \to f$, let $g_n \to f$ be a second sequence and observe

$$\|Af_n - Ag_n\| = \|A(f_n - g_n)\| \le \|A\|\|f_n - g_n\| \to 0. \tag{1.64}$$

From continuity of vector addition and scalar multiplication it follows that our extension is linear. Finally, from continuity of the norm we conclude that the norm does not increase. □

An operator in $\mathfrak{L}(X, \mathbb{C})$ is called a **bounded linear functional** and the space $X^* = \mathfrak{L}(X, \mathbb{C})$ is called the dual space of X.

Problem 1.9. *Show that the integral operator*

$$(Kf)(x) = \int_0^1 K(x, y) f(y) dy, \tag{1.65}$$

where $K(x, y) \in C([0, 1] \times [0, 1])$*, defined on* $\mathfrak{D}(K) = C[0, 1]$ *is a bounded operator both in* $X = C[0, 1]$ *(max norm) and* $X = \mathcal{L}^2_{cont}(0, 1)$*.*

Problem 1.10. *Show that the differential operator* $A = \frac{d}{dx}$ *defined on* $\mathfrak{D}(A) = C^1[0, 1] \subset C[0, 1]$ *is an unbounded operator.*

Chapter 2

Hilbert spaces

2.1. Orthonormal bases

In this section we will investigate orthonormal series and you will notice hardly any difference between the finite and infinite dimensional cases.

As our first task, let us generalize the projection into the direction of one vector:

A set of vectors $\{u_j\}$ is called **orthonormal set** if $\langle u_j, u_k\rangle = 0$ for $j \neq k$ and $\langle u_j, u_j\rangle = 1$.

Lemma 2.1. *Suppose* $\{u_j\}_{j=1}^n$ *is an orthonormal set. Then every* $f \in \mathfrak{H}$ *can be written as*

$$f = f_\parallel + f_\perp, \qquad f_\parallel = \sum_{j=1}^n \langle u_j, f\rangle u_j, \tag{2.1}$$

where $f_\parallel$ *and* $f_\perp$ *are orthogonal. Moreover,* $\langle u_j, f_\perp\rangle = 0$ *for all* $1 \leq j \leq n$. *In particular,*

$$\|f\|^2 = \sum_{j=1}^n |\langle u_j, f\rangle|^2 + \|f_\perp\|^2. \tag{2.2}$$

Moreover, every $\hat{f}$ *in the span of* $\{u_j\}_{j=1}^n$ *satisfies*

$$\|f - \hat{f}\| \geq \|f_\perp\| \tag{2.3}$$

with equality holding if and only if $\hat{f} = f_\parallel$. *In other words,* $f_\parallel$ *is uniquely characterized as the vector in the span of* $\{u_j\}_{j=1}^n$ *being closest to* f.

Proof. A straightforward calculation shows $\langle u_j, f - f_\parallel\rangle = 0$ and hence $f_\parallel$ and $f_\perp = f - f_\parallel$ are orthogonal. The formula for the norm follows by applying (1.40) iteratively.

Now, fix a vector

$$\hat{f} = \sum_{j=1}^{n} c_j u_j. \tag{2.4}$$

in the span of $\{u_j\}_{j=1}^n$. Then one computes

$$\begin{aligned} \|f - \hat{f}\|^2 &= \|f_\parallel + f_\perp - \hat{f}\|^2 = \|f_\perp\|^2 + \|f_\parallel - \hat{f}\|^2 \\ &= \|f_\perp\|^2 + \sum_{j=1}^{n} |c_j - \langle u_j, f\rangle|^2 \end{aligned} \tag{2.5}$$

from which the last claim follows. □

From (2.2) we obtain **Bessel's inequality**

$$\sum_{j=1}^{n} |\langle u_j, f\rangle|^2 \le \|f\|^2 \tag{2.6}$$

with equality holding if and only if f lies in the span of $\{u_j\}_{j=1}^n$.

Of course, since we cannot assume $\mathfrak{H}$ to be a finite dimensional vector space, we need to generalize Lemma 2.1 to arbitrary orthonormal sets $\{u_j\}_{j\in J}$. We start by assuming that J is countable. Then Bessel's inequality (2.6) shows that

$$\sum_{j\in J} |\langle u_j, f\rangle|^2 \tag{2.7}$$

converges absolutely. Moreover, for any finite subset $K \subset J$ we have

$$\|\sum_{j\in K} \langle u_j, f\rangle u_j\|^2 = \sum_{j\in K} |\langle u_j, f\rangle|^2 \tag{2.8}$$

by the Pythagorean theorem and thus $\sum_{j\in J}\langle u_j, f\rangle u_j$ is a Cauchy sequence if and only if $\sum_{j\in J} |\langle u_j, f\rangle|^2$ is. Now let J be arbitrary. Again, Bessel's inequality shows that for any given $\varepsilon > 0$ there are at most finitely many j for which $|\langle u_j, f\rangle| \ge \varepsilon$ (namely at most $\|f\|/\varepsilon$). Hence there are at most countably many j for which $|\langle u_j, f\rangle| > 0$. Thus it follows that

$$\sum_{j\in J} |\langle u_j, f\rangle|^2 \tag{2.9}$$

is well-defined and (by completeness) so is

$$\sum_{j\in J} \langle u_j, f\rangle u_j. \tag{2.10}$$

In particular, by continuity of the scalar product we see that Lemma 2.1 holds for arbitrary orthonormal sets without modifications.

Theorem 2.2. *Suppose $\{u_j\}_{j\in J}$ is an orthonormal set in an inner product space $\mathfrak{H}$. Then every $f \in \mathfrak{H}$ can be written as*

$$f = f_\parallel + f_\perp, \qquad f_\parallel = \sum_{j\in J} \langle u_j, f\rangle u_j, \tag{2.11}$$

where $f_\parallel$ and $f_\perp$ are orthogonal. Moreover, $\langle u_j, f_\perp\rangle = 0$ for all $j \in J$. In particular,

$$\|f\|^2 = \sum_{j\in J} |\langle u_j, f\rangle|^2 + \|f_\perp\|^2. \tag{2.12}$$

Moreover, every $\hat{f}$ in the span of $\{u_j\}_{j\in J}$ satisfies

$$\|f - \hat{f}\| \geq \|f_\perp\| \tag{2.13}$$

with equality holding if and only if $\hat{f} = f_\parallel$. In other words, $f_\parallel$ is uniquely characterized as the vector in the span of $\{u_j\}_{j\in J}$ being closest to f.

Note that from Bessel's inequality (which of course still holds) it follows that the map $f \to f_\parallel$ is continuous.

Of course we are particularly interested in the case where every $f \in \mathfrak{H}$ can be written as $\sum_{j\in J}\langle u_j, f\rangle u_j$. In this case we will call the orthonormal set $\{u_j\}_{j\in J}$ an **orthonormal basis**.

If $\mathfrak{H}$ is separable it is easy to construct an orthonormal basis. In fact, if $\mathfrak{H}$ is separable, then there exists a countable total set $\{f_j\}_{j=1}^N$. After throwing away some vectors we can assume that f_{n+1} cannot be expressed as a linear combinations of the vectors $f_1, \dots f_n$. Now we can construct an orthonormal set as follows: We begin by normalizing f_1

$$u_1 = \frac{f_1}{\|f_1\|}. \tag{2.14}$$

Next we take f_2 and remove the component parallel to u_1 and normalize again

$$u_2 = \frac{f_2 - \langle u_1, f_2\rangle u_1}{\|f_2 - \langle u_1, f_2\rangle u_1\|}. \tag{2.15}$$

Proceeding like this we define recursively

$$u_n = \frac{f_n - \sum_{j=1}^{n-1}\langle u_j, f_n\rangle u_j}{\|f_n - \sum_{j=1}^{n-1}\langle u_j, f_n\rangle u_j\|}. \tag{2.16}$$

This procedure is known as **Gram-Schmidt orthogonalization**. Hence we obtain an orthonormal set $\{u_j\}_{j=1}^N$ such that $\mathrm{span}\{u_j\}_{j=1}^n = \mathrm{span}\{f_j\}_{j=1}^n$ for any finite n and thus also for N. Since $\{f_j\}_{j=1}^N$ is total, we infer that $\{u_j\}_{j=1}^N$ is an orthonormal basis.

Theorem 2.3. *Every separable inner product space has a countable orthonormal basis.*

Example. In $\mathcal{L}^2_{cont}(-1,1)$ we can orthogonalize the polynomial $f_n(x) = x^n$. The resulting polynomials are up to a normalization equal to the Legendre polynomials

$$P_0(x) = 1, \quad P_1(x) = x, \quad P_2(x) = \frac{3\,x^2 - 1}{2}, \quad \dots \tag{2.17}$$

(which are normalized such that $P_n(1) = 1$). ◇

Example. The set of functions

$$u_n(x) = \frac{1}{\sqrt{2\pi}} \mathrm{e}^{\mathrm{i}n\,x}, \qquad n \in \mathbb{Z}, \tag{2.18}$$

forms an orthonormal basis for $\mathfrak{H} = \mathcal{L}^2_{cont}(0, 2\pi)$. The corresponding orthogonal expansion is just the ordinary Fourier series. ◇

If fact, if there is one countable basis, then it follows that any other basis is countable as well.

Theorem 2.4. *If $\mathfrak{H}$ is separable, then every orthonormal basis is countable.*

Proof. We know that there is at least one countable orthonormal basis $\{u_j\}_{j\in J}$. Now let $\{u_k\}_{k\in K}$ be a second basis and consider the set $K_j = \{k \in K | \langle u_k, u_j\rangle \neq 0\}$. Since these are the expansion coefficients of u_j with respect to $\{u_k\}_{k\in K}$, this set is countable. Hence the set $\tilde{K} = \bigcup_{j\in J} K_j$ is countable as well. But $k \in K\backslash\tilde{K}$ implies $u_k = 0$ and hence $\tilde{K} = K$. □

It even turns out that, up to unitary equivalence, there is only one (separable) infinite dimensional Hilbert space:

A bijective operator $U \in \mathfrak{L}(\mathfrak{H}_1, \mathfrak{H}_2)$ is called **unitary** if U preserves scalar products:

$$\langle Ug, Uf\rangle_2 = \langle g, f\rangle_1, \qquad g, f \in \mathfrak{H}_1. \tag{2.19}$$

By the polarization identity this is the case if and only if U preserves norms: $\|Uf\|_2 = \|f\|_1$ for all $f \in \mathfrak{H}_1$. The two Hilbert space $\mathfrak{H}_1$ and $\mathfrak{H}_2$ are called **unitarily equivalent** in this case.

Let $\mathfrak{H}$ be an infinite dimensional Hilbert space and let $\{u_j\}_{j\in\mathbb{N}}$ be any orthogonal basis. Then the map $U : \mathfrak{H} \to \ell^2(\mathbb{N})$, $f \mapsto (\langle u_j, f\rangle)_{j\in\mathbb{N}}$ is unitary (by Theorem 2.6 (iii)). In particular,

Theorem 2.5. *Any separable infinite dimensional Hilbert space is unitarily equivalent to $\ell^2(\mathbb{N})$.*

To see that any Hilbert space has an orthonormal basis we need to resort to Zorn's lemma: The collection of all orthonormal sets in $\mathfrak{H}$ can be partially ordered by inclusion. Moreover, any linearly ordered chain has an upper bound (the union of all sets in the chain). Hence a fundamental result from axiomatic set theory, Zorn's lemma, implies the existence of a maximal element, that is, an orthonormal set which is not a proper subset of any other orthonormal set.

Theorem 2.6. *For an orthonormal set $\{u_j\}_{j\in J}$ in a Hilbert space $\mathfrak{H}$ the following conditions are equivalent:*
(i) $\{u_j\}_{j\in J}$ is a maximal orthogonal set.
(ii) For every vector $f \in \mathfrak{H}$ we have

$$f = \sum_{j\in J} \langle u_j, f\rangle u_j. \tag{2.20}$$

(iii) For every vector $f \in \mathfrak{H}$ we have

$$\|f\|^2 = \sum_{j\in J} |\langle u_j, f\rangle|^2. \tag{2.21}$$

(iv) $\langle u_j, f\rangle = 0$ for all $j \in J$ implies $f = 0$.

Proof. We will use the notation from Theorem 2.2.
(i) ⇒ *(ii)*: If $f_\perp \neq 0$ than we can normalize $f_\perp$ to obtain a unit vector $\tilde{f}_\perp$ which is orthogonal to all vectors u_j. But then $\{u_j\}_{j\in J} \cup \{\tilde{f}_\perp\}$ would be a larger orthonormal set, contradicting maximality of $\{u_j\}_{j\in J}$.
(ii) ⇒ *(iii)*: Follows since *(ii)* implies $f_\perp = 0$.
(iii) ⇒ *(iv)*: If $\langle f, u_j\rangle = 0$ for all $j \in J$ we conclude $\|f\|^2 = 0$ and hence $f = 0$.
(iv) ⇒ *(i)*: If $\{u_j\}_{j\in J}$ were not maximal, there would be a unit vector g such that $\{u_j\}_{j\in J} \cup \{g\}$ is larger orthonormal set. But $\langle u_j, g\rangle = 0$ for all $j \in J$ implies $g = 0$ by (iv), a contradiction. □

By continuity of the norm it suffices to check (iii), and hence also (ii), for f in a dense set.

2.2. The projection theorem and the Riesz lemma

Let $M \subseteq \mathfrak{H}$ be a subset, then $M^\perp = \{f | \langle g, f\rangle = 0, \forall g \in M\}$ is called the **orthogonal complement** of M. By continuity of the scalar product it follows that $M^\perp$ is a closed linear subspace and by linearity that $(\overline{\operatorname{span}(M)})^\perp = M^\perp$. For example we have $\mathfrak{H}^\perp = \{0\}$ since any vector in $\mathfrak{H}^\perp$ must be in particular orthogonal to all vectors in some orthonormal basis.

Theorem 2.7 (projection theorem)**.** *Let M be a closed linear subspace of a Hilbert space $\mathfrak{H}$, then every $f \in \mathfrak{H}$ can be uniquely written as $f = f_\| + f_\perp$ with $f_\| \in M$ and $f_\perp \in M^\perp$. One writes*

$$M \oplus M^\perp = \mathfrak{H} \tag{2.22}$$

in this situation.

Proof. Since M is closed, it is a Hilbert space and has an orthonormal basis $\{u_j\}_{j\in J}$. Hence the result follows from Theorem 2.2. □

In other words, to every $f \in \mathfrak{H}$ we can assign a unique vector $f_\|$ which is the vector in M closest to f. The rest $f - f_\|$ lies in $M^\perp$. The operator $P_M f = f_\|$ is called the **orthogonal projection** corresponding to M. Note that we have

$$P_M^2 = P_M \qquad \text{and} \qquad \langle P_M g, f\rangle = \langle g, P_M f\rangle \tag{2.23}$$

since $\langle P_M g, f\rangle = \langle g_\|, f_\|\rangle = \langle g, P_M f\rangle$. Clearly we have $P_{M^\perp} f = f - P_M f = f_\perp$.

Moreover, we see that the vectors in a closed subspace M are precisely those which are orthogonal to all vectors in $M^\perp$, that is, $M^{\perp\perp} = M$. If M is an arbitrary subset we have at least

$$M^{\perp\perp} = \overline{\operatorname{span}(M)}. \tag{2.24}$$

Note that by $\mathfrak{H}^\perp = \{\}$ we see that $M^\perp = \{0\}$ if and only if M is dense.

Finally we turn to **linear functionals**, that is, to operators $\ell : \mathfrak{H} \to \mathbb{C}$. By the Cauchy-Schwarz inequality we know that $\ell_g : f \mapsto \langle g, f\rangle$ is a bounded linear functional (with norm $\|g\|$). In turns out that in a Hilbert space every bounded linear functional can be written in this way.

Theorem 2.8 (Riesz lemma)**.** *Suppose ℓ is a bounded linear functional on a Hilbert space $\mathfrak{H}$. Then there is a unique vector $g \in \mathfrak{H}$ such that $\ell(f) = \langle g, f\rangle$ for all $f \in \mathfrak{H}$. In other words, a Hilbert space is equivalent to its own dual space $\mathfrak{H}^* = \mathfrak{H}$.*

Proof. If $\ell \equiv 0$ we can choose $g = 0$. Otherwise $\operatorname{Ker}(\ell) = \{f | \ell(f) = 0\}$ is a proper subspace and we can find a unit vector $\tilde{g} \in \operatorname{Ker}(\ell)^\perp$. For every $f \in \mathfrak{H}$ we have $\ell(f)\tilde{g} - \ell(\tilde{g})f \in \operatorname{Ker}(\ell)$ and hence

$$0 = \langle \tilde{g}, \ell(f)\tilde{g} - \ell(\tilde{g})f\rangle = \ell(f) - \ell(\tilde{g})\langle \tilde{g}, f\rangle. \tag{2.25}$$

In other words, we can choose $g = \ell(\tilde{g})^* \tilde{g}$. To see uniqueness, let g_1, g_2 be two such vectors. Then $\langle g_1 - g_2, f\rangle = \langle g_1, f\rangle - \langle g_2, f\rangle = \ell(f) - \ell(f) = 0$ for any $f \in \mathfrak{H}$, which shows $g_1 - g_2 \in \mathfrak{H}^\perp = \{0\}$. □

The following easy consequence is left as an exercise.

Corollary 2.9. *Suppose B is a bounded skew linear form, that is,*

$$|B(f,g)| \le C\|f\|\,\|g\|. \tag{2.26}$$

Then there is a unique bounded operator A such that

$$B(f,g) = \langle Af, g\rangle. \tag{2.27}$$

Problem 2.1. *Prove Corollary 2.9.*

2.3. Orthogonal sums and tensor products

Given two Hilbert spaces $\mathfrak{H}_1$ and $\mathfrak{H}_2$ we define their **orthogonal sum** $\mathfrak{H}_1 \oplus \mathfrak{H}_2$ to be the set of all pairs $(f_1, f_2) \in \mathfrak{H}_1 \times \mathfrak{H}_2$ together with the scalar product

$$\langle (g_1, g_2), (f_1, f_2)\rangle = \langle g_1, f_1\rangle_1 + \langle g_2, f_2\rangle_2. \tag{2.28}$$

It is left as an exercise to verify that $\mathfrak{H}_1 \oplus \mathfrak{H}_2$ is again a Hilbert space. Moreover, $\mathfrak{H}_1$ can be identified with $\{(f_1, 0) | f_1 \in \mathfrak{H}_1\}$ and we can regard $\mathfrak{H}_1$ as a subspace of $\mathfrak{H}_1 \oplus \mathfrak{H}_2$. Similarly for $\mathfrak{H}_2$. It is also custom to write $f_1 + f_2$ instead of (f_1, f_2).

More generally, let $\mathfrak{H}_j$ $j \in \mathbb{N}$, be a countable collection of Hilbert spaces and define

$$\bigoplus_{j=1}^{\infty} \mathfrak{H}_j = \{\sum_{j=1}^{\infty} f_j |\, f_j \in \mathfrak{H}_j,\ \sum_{j=1}^{\infty} \|f_j\|_j^2 < \infty\}, \tag{2.29}$$

which becomes a Hilbert space with the scalar product

$$\langle \sum_{j=1}^{\infty} g_j, \sum_{j=1}^{\infty} f_j\rangle = \sum_{j=1}^{\infty} \langle g_j, f_j\rangle_j. \tag{2.30}$$

Example. $\bigoplus_{j=1}^{\infty} \mathbb{C} = \ell^2(\mathbb{N})$. ⋄

Suppose $\mathfrak{H}$ and $\tilde{\mathfrak{H}}$ are two Hilbert spaces. Our goal is to construct their tensor product. The elements should be products $f \otimes \tilde{f}$ of elements $f \in \mathfrak{H}$ and $\tilde{f} \in \tilde{\mathfrak{H}}$. Hence we start with the set of all finite linear combinations of elements of $\mathfrak{H} \times \tilde{\mathfrak{H}}$

$$\mathcal{F}(\mathfrak{H}, \tilde{\mathfrak{H}}) = \{\sum_{j=1}^{n} \alpha_j (f_j, \tilde{f}_j) | (f_j, \tilde{f}_j) \in \mathfrak{H} \times \tilde{\mathfrak{H}},\ \alpha_j \in \mathbb{C}\}. \tag{2.31}$$

Since we want $(f_1 + f_2) \otimes \tilde{f} = f_1 \otimes \tilde{f} + f_2 \otimes \tilde{f}$, $f \otimes (\tilde{f}_1 + \tilde{f}_2) = f \otimes \tilde{f}_1 + f \otimes \tilde{f}_2$, and $(\alpha f) \otimes \tilde{f} = f \otimes (\alpha \tilde{f})$ we consider $\mathcal{F}(\mathfrak{H}, \tilde{\mathfrak{H}}) / \mathcal{N}(\mathfrak{H}, \tilde{\mathfrak{H}})$, where

$$\mathcal{N}(\mathfrak{H}, \tilde{\mathfrak{H}}) = \text{span}\{\sum_{j,k=1}^{n} \alpha_j \beta_k (f_j, \tilde{f}_k) - (\sum_{j=1}^{n} \alpha_j f_j, \sum_{k=1}^{n} \beta_k \tilde{f}_k)\} \tag{2.32}$$

and write $f \otimes \tilde{f}$ for the equivalence class of $(f, \tilde{f})$.

Next we define

$$\langle f \otimes \tilde{f}, g \otimes \tilde{g} \rangle = \langle f, g \rangle \langle \tilde{f}, \tilde{g} \rangle \tag{2.33}$$

which extends to a skew linear form on $\mathcal{F}(\mathfrak{H}, \tilde{\mathfrak{H}})/\mathcal{N}(\mathfrak{H}, \tilde{\mathfrak{H}})$. To show that we obtain a scalar product, we need to ensure positivity. Let $f = \sum_i \alpha_i f_i \otimes \tilde{f}_i \neq 0$ and pick orthonormal bases u_j, $\tilde{u}_k$ for $\mathrm{span}\{f_i\}$, $\mathrm{span}\{\tilde{f}_i\}$, respectively. Then

$$f = \sum_{j,k} \alpha_{jk} u_j \otimes \tilde{u}_k, \quad \alpha_{jk} = \sum_i \alpha_i \langle u_j, f_i \rangle \langle \tilde{u}_k, \tilde{f}_i \rangle \tag{2.34}$$

and we compute

$$\langle f, f \rangle = \sum_{j,k} |\alpha_{jk}|^2 > 0. \tag{2.35}$$

The completion of $\mathcal{F}(\mathfrak{H}, \tilde{\mathfrak{H}})/\mathcal{N}(\mathfrak{H}, \tilde{\mathfrak{H}})$ with respect to the induced norm is called the **tensor product** $\mathfrak{H} \otimes \tilde{\mathfrak{H}}$ of $\mathfrak{H}$ and $\tilde{\mathfrak{H}}$.

Lemma 2.10. *If u_j, $\tilde{u}_k$ are orthonormal bases for $\mathfrak{H}$, $\tilde{\mathfrak{H}}$, respectively, then $u_j \otimes \tilde{u}_k$ is an orthonormal basis for $\mathfrak{H} \otimes \tilde{\mathfrak{H}}$.*

Proof. That $u_j \otimes \tilde{u}_k$ is an orthonormal set is immediate from (2.33). Moreover, since $\mathrm{span}\{u_j\}$, $\mathrm{span}\{\tilde{u}_k\}$ is dense in $\mathfrak{H}$, $\tilde{\mathfrak{H}}$, respectively, it is easy to see that $u_j \otimes \tilde{u}_k$ is dense in $\mathcal{F}(\mathfrak{H}, \tilde{\mathfrak{H}})/\mathcal{N}(\mathfrak{H}, \tilde{\mathfrak{H}})$. But the latter is dense in $\mathfrak{H} \otimes \tilde{\mathfrak{H}}$. □

Example. We have $\mathfrak{H} \otimes \mathbb{C}^n = \mathfrak{H}^n$. ⋄

It is straightforward to extend the tensor product to any finite number of Hilbert spaces. We even note

$$(\bigoplus_{j=1}^{\infty} \mathfrak{H}_j) \otimes \mathfrak{H} = \bigoplus_{j=1}^{\infty} (\mathfrak{H}_j \otimes \mathfrak{H}), \tag{2.36}$$

where equality has to be understood in the sense, that both spaces are unitarily equivalent by virtue of the identification

$$(\sum_{j=1}^{\infty} f_j) \otimes f = \sum_{j=1}^{\infty} f_j \otimes f. \tag{2.37}$$

2.4. Compact operators

A linear operator A defined on a normed space X is called **compact** if every sequence Af_n has a convergent subsequence whenever f_n is bounded. The set of all compact operators is denoted by $\mathfrak{C}(X)$. It is not hard to see that the set of compact operators is an ideal of the set of bounded operators (Problem 2.2):

Theorem 2.11. *Every compact linear operator is bounded. Linear combinations of compact operators are bounded and the product of a bounded and a compact operator is again compact.*

If X is a Banach space then this ideal is even closed:

Theorem 2.12. *Let X be a Banach space, and let A_n be a convergent sequence of compact operators. Then the limit A is again compact.*

Proof. Let $f_j^{(0)}$ be a bounded sequence. Choose a subsequence $f_j^{(1)}$ such that $A_1 f_j^{(1)}$ converges. From $f_j^{(1)}$ choose another subsequence $f_j^{(2)}$ such that $A_2 f_j^{(2)}$ converges and so on. Since $f_j^{(n)}$ might disappear as $n \to \infty$, we consider the diagonal sequence $f_j = f_j^{(j)}$. By construction, f_j is a subsequence of $f_j^{(n)}$ for $j \geq n$ and hence $A_n f_j$ is Cauchy for any fixed n. Now

$$\begin{aligned} \|A f_j - f_k\| &= \|(A - A_n)(f_j - f_k) + A_n(f_j - f_k)\| \\ &\leq \|A - A_n\| \|f_j - f_k\| + \|A_n f_j - A_n f_k\| \end{aligned} \tag{2.38}$$

shows that $A f_j$ is Cauchy since the first term can be made arbitrary small by choosing n large and the second by the Cauchy property of $A_n f_j$. □

Note that it suffices to verify compactness on a dense set.

Theorem 2.13. *Let X be a normed space and $A \in \mathfrak{C}(X)$. Let $\overline{X}$ be its completion, then $\overline{A} \in \mathfrak{C}(\overline{X})$, where $\overline{A}$ is the unique extension of A.*

Proof. Let $f_n \in \overline{X}$ be a given bounded sequence. We need to show that $\overline{A} f_n$ has a convergent subsequence. Pick $f_n^j \in X$ such that $\|f_n^j - f_n\| \leq \frac{1}{j}$ and by compactness of A we can assume that $A f_n^n \to g$. But then $\|\overline{A} f_n - g\| \leq \|A\| \|f_n - f_n^n\| + \|A f_n^n - g\|$ shows that $\overline{A} f_n \to g$. □

One of the most important examples of compact operators are integral operators:

Lemma 2.14. *The integral operator*

$$(Kf)(x) = \int_a^b K(x,y) f(y) dy, \tag{2.39}$$

where $K(x,y) \in C([a,b] \times [a,b])$, defined on $\mathcal{L}^2_{cont}(a,b)$ is compact.

Proof. First of all note that $K(.,..)$ is continuous on $[a,b] \times [a,b]$ and hence uniformly continuous. In particular, for every $\varepsilon > 0$ we can find a $\delta > 0$

such that $|K(y,t) - K(x,t)| \le \varepsilon$ whenever $|y - x| \le \delta$. Let $g(x) = Kf(x)$, then

$$\begin{aligned} |g(x) - g(y)| &\le \int_a^b |K(y,t) - K(x,t)|\,|f(t)|dt \\ &\le \varepsilon \int_a^b |f(t)|dt \le \varepsilon \|1\|\,\|f\|, \end{aligned} \tag{2.40}$$

whenever $|y - x| \le \delta$. Hence, if $f_n(x)$ is a bounded sequence in $\mathcal{L}^2_{cont}(a,b)$, then $g_n(x) = Kf_n(x)$ is equicontinuous and has a uniformly convergent subsequence by the Arzelà-Ascoli theorem (Theorem 2.15 below). But a uniformly convergent sequence is also convergent in the norm induced by the scalar product. Therefore K is compact. □

Note that (almost) the same proof shows that K is compact when defined on $C[a,b]$.

Theorem 2.15 (Arzelà-Ascoli)**.** *Suppose the sequence of functions $f_n(x)$, $n \in \mathbb{N}$, on a compact interval is (uniformly) equicontinuous, that is, for every $\varepsilon > 0$ there is a $\delta > 0$ (independent of n) such that*

$$|f_n(x) - f_n(y)| \le \varepsilon \quad \textit{if} \quad |x - y| < \delta. \tag{2.41}$$

If the sequence f_n is bounded, then there is a uniformly convergent subsequence.

Proof. Let $\{x_j\}_{j=1}^\infty$ be a dense subset of our interval (e.g., all rational in this set). Since $f_n(x_j)$ is bounded, we can choose a convergent subsequence $f_n^{(j)}(x_j)$ (Bolzano-Weierstraß). The diagonal subsequence $\tilde{f}_n = f_n^{(n)}(x)$ will hence converge for all $x = x_j$. We will show that it converges uniformly for all x:

Fix $\varepsilon > 0$ and chose δ such that $|f_n(x) - f_n(y)| \le \frac{\varepsilon}{3}$ for $|x - y| < \delta$. The balls $B_\delta(x_j)$ cover our interval and by compactness even finitely many, say $1 \le j \le p$ suffice. Furthermore, choose N_ε such that $|\tilde{f}_m(x_j) - \tilde{f}_n(x_j)| \le \varepsilon$ for $n, m \ge N_\varepsilon$ and $1 \le j \le p$.

Now pick x and note that $x \in B_\delta(x_j)$ for some j. Thus

$$|\tilde{f}_m(x) - \tilde{f}_n(x)| \le |\tilde{f}_m(x) - \tilde{f}_m(x_j)| + |\tilde{f}_m(x_j) - \tilde{f}_n(x_j)| + |\tilde{f}_n(x_j) - \tilde{f}_n(x)| \le \varepsilon \tag{2.42}$$

for $n, m \ge N_\varepsilon$, which shows that $\tilde{f}_n$ is Cauchy with respect to the maximum norm. □

Compact operators are very similar to (finite) matrices as we will see in the next section.

Problem 2.2. *Show that compact operators form an ideal.*

2.5. The spectral theorem for compact symmetric operators

Let $\mathfrak{H}$ be an inner product space. A linear operator A is called **symmetric** if its domain is dense and if

$$\langle g, Af\rangle = \langle Ag, f\rangle \qquad f, g \in \mathfrak{D}(A). \tag{2.43}$$

A number $z \in \mathbb{C}$ is called **eigenvalue** of A if there is a nonzero vector $u \in \mathfrak{D}(A)$ such that

$$Au = zu. \tag{2.44}$$

The vector u is called a corresponding **eigenvector** in this case. An eigenvalue is called **simple** if there is only one linearly independent eigenvector.

Theorem 2.16. *Let A be symmetric. Then all eigenvalues are real and eigenvectors corresponding to different eigenvalues are orthogonal.*

Proof. Suppose λ is an eigenvalue with corresponding normalized eigenvector u. Then $\lambda = \langle u, Au\rangle = \langle Au, u\rangle = \lambda^*$, which shows that λ is real. Furthermore, if $Au_j = \lambda_j u_j$, $j = 1, 2$, we have

$$(\lambda_1 - \lambda_2)\langle u_1, u_2\rangle = \langle Au_1, u_2\rangle - \langle u_1, Au_2\rangle = 0 \tag{2.45}$$

finishing the proof. □

Now we show that A has an eigenvalue at all (which is not clear in the infinite dimensional case)!

Theorem 2.17. *A symmetric compact operator has an eigenvalue α_1 which satisfies $|\alpha_1| = \|A\|$.*

Proof. We set $\alpha = \|A\|$ and assume $\alpha \neq 0$ (i.e, $A \neq 0$) without loss of generality. Since

$$\|A\|^2 = \sup_{f:\|f\|=1} \|Af\|^2 = \sup_{f:\|f\|=1} \langle Af, Af\rangle = \sup_{f:\|f\|=1} \langle f, A^2 f\rangle \tag{2.46}$$

there exists a normalized sequence u_n such that

$$\lim_{n\to\infty} \langle u_n, A^2 u_n\rangle = \alpha^2. \tag{2.47}$$

Since A is compact, it is no restriction to assume that $A^2 u_n$ converges, say $\lim_{n\to\infty} A^2 u_n = \alpha^2 u$. Now

$$\begin{aligned} \|(A^2 - \alpha^2)u_n\|^2 &= \|A^2 u_n\|^2 - 2\alpha^2\langle u_n, A^2 u_n\rangle + \alpha^4 \\ &\leq 2\alpha^2(\alpha^2 - \langle u_n, A^2 u_n\rangle) \end{aligned} \tag{2.48}$$

(where we have used $\|A^2 u_n\| \leq \|A\|\|Au_n\| \leq \|A\|^2\|u_n\| = \alpha^2$) implies $\lim_{n\to\infty}(A^2 u_n - \alpha^2 u_n) = 0$ and hence $\lim_{n\to\infty} u_n = u$. In addition, u is a normalized eigenvector of A^2 since $(A^2 - \alpha^2)u = 0$. Factorizing this last equation according to $(A-\alpha)u = v$ and $(A+\alpha)v = 0$ show that either $v \neq 0$ is an eigenvector corresponding to $-\alpha$ or $v = 0$ and hence $u \neq 0$ is an eigenvector corresponding to α. □

Note that for a bounded operator A, there cannot be an eigenvalue with absolute value larger than $\|A\|$, that is, the set of eigenvalues is bounded by $\|A\|$ (Problem 2.3).

Now consider a symmetric compact operator A with eigenvalue α_1 (as above) and corresponding normalized eigenvector u_1. Setting

$$\mathfrak{H}_1 = \{u_1\}^\perp = \{f \in \mathfrak{H} | \langle u_1, f\rangle = 0\} \tag{2.49}$$

we can restrict A to $\mathfrak{H}_1$ using

$$\mathfrak{D}(A_1) = \mathfrak{D}(A) \cap \mathfrak{H}_1 = \{f \in \mathfrak{D}(A) | \langle u_1, f\rangle = 0\} \tag{2.50}$$

since $f \in \mathfrak{D}(A_1)$ implies

$$\langle u_1, Af\rangle = \langle Au_1, f\rangle = \alpha_1 \langle u_1, f\rangle = 0 \tag{2.51}$$

and hence $Af \in \mathfrak{H}_1$. Denoting this restriction by A_1, it is not hard to see that A_1 is again a symmetric compact operator. Hence we can apply Theorem 2.17 iteratively to obtain a sequence of eigenvalues α_j with corresponding normalized eigenvectors u_j. Moreover, by construction, u_n is orthogonal to all u_j with $j < n$ and hence the eigenvectors $\{u_j\}$ form an orthonormal set. This procedure will not stop unless $\mathfrak{H}$ is finite dimensional. However, note that $\alpha_j = 0$ for $j \geq n$ might happen if $A_n = 0$.

Theorem 2.18. *Suppose $\mathfrak{H}$ is a Hilbert space and $A : \mathfrak{H} \to \mathfrak{H}$ is a compact symmetric operator. Then there exists a sequence of real eigenvalues α_j converging to 0. The corresponding normalized eigenvectors u_j form an orthonormal set and every $f \in \mathfrak{H}$ can be written as*

$$f = \sum_{j=1}^{\infty} \langle u_j, f\rangle u_j + h, \tag{2.52}$$

where h is in the kernel of A, that is, $Ah = 0$.

In particular, if 0 is not an eigenvalue, then the eigenvectors form an orthonormal basis (in addition, $\mathfrak{H}$ needs not to be complete in this case).

Proof. Existence of the eigenvalues α_j and the corresponding eigenvectors has already been established. If the eigenvalues should not converge to zero, there is a subsequence such that $|\alpha_{j_k}| \geq \varepsilon$. Hence $v_k = \alpha_{j_k}^{-1} u_{j_k}$ is a bounded sequence ($\|v_k\| \leq \frac{1}{\varepsilon}$) for which Av_k has no convergent subsequence since $\|Av_k - Av_l\|^2 = \|u_{j_k} - u_{j_l}\|^2 = 2$.

Next, setting

$$f_n = \sum_{j=1}^{n} \langle u_j, f\rangle u_j, \tag{2.53}$$

we have

$$\|A(f - f_n)\| \le |\alpha_n| \|f - f_n\| \le |\alpha_n| \|f\| \tag{2.54}$$

since $f - f_n \in \mathfrak{H}_n$. Letting $n \to \infty$ shows $A(f_\infty - f) = 0$ proving (2.52). □

Remark: There are two cases where our procedure might fail to construct an orthonormal basis of eigenvectors. One case is where there is an infinite number of nonzero eigenvalues. In this case α_n never reaches 0 and all eigenvectors corresponding to 0 are missed. In the other case, 0 is reached, but there might not be a countable basis and hence again some of the eigenvectors corresponding to 0 are missed. In any case one can show that by adding vectors from the kernel (which are automatically eigenvectors), one can always extend the eigenvectors u_j to an orthonormal basis of eigenvectors.

This is all we need and it remains to apply these results to Sturm-Liouville operators.

Problem 2.3. *Show that if A is bounded, then every eigenvalue α satisfies $|\alpha| \le \|A\|$.*

2.6. Applications to Sturm-Liouville operators

Now, after all this hard work, we can show that our Sturm-Liouville operator

$$L = -\frac{d^2}{dx^2} + q(x), \tag{2.55}$$

where q is continuous and real, defined on

$$\mathfrak{D}(L) = \{f \in C^2[0,1] | f(0) = f(1) = 0\} \subset \mathcal{L}^2_{cont}(0,1) \tag{2.56}$$

has an orthonormal basis of eigenfunctions.

We first verify that L is symmetric:

$$\begin{aligned} \langle f, Lg\rangle &= \int_0^1 f(x)^*(-g''(x) + q(x)g(x))dx \\ &= \int_0^1 f'(x)^* g'(x)dx + \int_0^1 f(x)^* q(x)g(x)dx \\ &= \int_0^1 -f''(x)^* g(x)dx + \int_0^1 f(x)^* q(x)g(x)dx \\ &= \langle Lf, g\rangle. \end{aligned} \tag{2.57}$$

Here we have used integration by part twice (the boundary terms vanish due to our boundary conditions $f(0) = f(1) = 0$ and $g(0) = g(1) = 0$).

Of course we want to apply Theorem 2.18 and for this we would need to show that L is compact. But this task is bound to fail, since L is not even bounded (Problem 1.10)!

So here comes the trick: If L is unbounded its inverse L^{-1} might still be bounded. Moreover, L^{-1} might even be compact and this is the case here! Since L might not be injective (0 might be an eigenvalue), we consider $R_L(z) = (L - z)^{-1}$, $z \in \mathbb{C}$, which is also known as the **resolvent** of L.

Define

$$\begin{aligned} f(x) &= \frac{u_+(z,x)}{W(z)} \Big(\int_0^x u_-(z,t) g(t) dt \Big) \\ &\quad + \frac{u_-(z,x)}{W(z)} \Big(\int_x^1 u_+(z,t) g(t) dt \Big), \end{aligned} \tag{2.58}$$

where $u_\pm(z,x)$ are the solutions of the homogenous differential equation $-u_\pm''(z,x) + (q(x) - z) u_\pm(z,x) = 0$ satisfying the initial conditions $u_-(z,0) = 0$, $u_-'(z,0) = 1$ respectively $u_+(z,1) = 0$, $u_+'(z,1) = 1$ and

$$W(z) = W(u_+(z), u_-(z)) = u_-'(z,x) u_+(z,x) - u_-(z,x) u_+'(z,x) \tag{2.59}$$

is the Wronski determinant, which is independent of x (check this!).

Then clearly $f(0) = 0$ since $u_-(z,0) = 0$ and similarly $f(1) = 0$ since $u_+(z,1) = 0$. Furthermore, f is differentiable and a straightforward computation verifies

$$\begin{aligned} f'(x) &= \frac{u_+(z,x)'}{W(z)} \Big(\int_0^x u_-(z,t) g(t) dt \Big) \\ &\quad + \frac{u_-(z,x)'}{W(z)} \Big(\int_x^1 u_+(z,t) g(t) dt \Big). \end{aligned} \tag{2.60}$$

Thus we can differentiate once more giving

$$\begin{aligned} f''(x) &= \frac{u_+(z,x)''}{W(z)} \Big(\int_0^x u_-(z,t) g(t) dt \Big) \\ &\quad + \frac{u_-(z,x)''}{W(z)} \Big(\int_x^1 u_+(z,t) g(t) dt \Big) - g(x) \\ &= (q(x) - z) f(x) - g(x). \end{aligned} \tag{2.61}$$

In summary, f is in the domain of L and satisfies $(L - z) f = g$.

Note that z is an eigenvalue if and only if $W(z) = 0$. In fact, in this case $u_+(z,x)$ and $u_-(z,x)$ are linearly dependent and hence $u_-(z,1) = c\, u_+(z,1) = 0$ which shows that $u_-(z,x)$ satisfies both boundary conditions and is thus an eigenfunction.

Introducing the **Green function**

$$G(z,x,t) = \frac{1}{W(u_+(z),u_-(z))} \begin{cases} u_+(z,x)u_-(z,t), & x \geq t \\ u_+(z,t)u_-(z,x), & x \leq t \end{cases} \tag{2.62}$$

we see that $(L-z)^{-1}$ is given by

$$(L-z)^{-1}g(x) = \int_0^1 G(z,x,t)g(t)dt. \tag{2.63}$$

Moreover, from $G(z,x,t) = G(z,t,x)$ it follows that $(L-z)^{-1}$ is symmetric for $z \in \mathbb{R}$ (Problem 2.4) and from Lemma 2.14 it follows that it is compact. Hence Theorem 2.18 applies to $(L-z)^{-1}$ and we obtain:

Theorem 2.19. *The Sturm-Liouville operator L has a countable number of eigenvalues E_n. All eigenvalues are discrete and simple. The corresponding normalized eigenfunctions u_n form an orthonormal basis for $\mathcal{L}^2_{cont}(0,1)$.*

Proof. Pick a value $\lambda \in \mathbb{R}$ such that $R_L(\lambda)$ exists. By Theorem 2.18 there are eigenvalues α_n of $R_L(\lambda)$ with corresponding eigenfunctions u_n. Moreover, $R_L(\lambda)u_n = \alpha_n u_n$ is equivalent to $Lu_n = (\lambda + \frac{1}{\alpha_n})u_n$, which shows that $E_n = \lambda + \frac{1}{\alpha_n}$ are eigenvalues of L with corresponding eigenfunctions u_n. Now everything follows from Theorem 2.18 except that the eigenvalues are simple. To show this, observe that if u_n and v_n are two different eigenfunctions corresponding to E_n, then $u_n(0) = v_n(0) = 0$ implies $W(u_n, v_n) = 0$ and hence u_n and v_n are linearly dependent. □

Problem 2.4. *Show that the integral operator*

$$(Kf)(x) = \int_0^1 K(x,y)f(y)dy, \tag{2.64}$$

where $K(x,y) \in C([0,1]\times[0,1])$ is symmetric if $K(x,y)^* = K(y,x)$.

Problem 2.5. *Show that the resolvent $R_A(z) = (A-z)^{-1}$ (provided it exists and is densely defined) of a symmetric operator A is again symmetric for $z \in \mathbb{R}$. (Hint: $g \in \mathfrak{D}(R_A(z))$ if and only if $g = (A-z)f$ for some $f \in \mathfrak{D}(A)$.*

Chapter 3

Almost everything about Lebesgue integration

3.1. Borel measures in a nut shell

The first step in defining the Lebesgue integral is extending the notion of size from intervals to arbitrary sets. Unfortunately, this turns out to be too much, since a classical paradox by Banach and Tarski shows that one can break the unit ball in $\mathbb{R}^3$ into a finite number of (wild – choosing the pieces uses the Axiom of Choice and cannot be done with a jigsaw;-) pieces, rotate and translate them, and reassemble them to get two copies of the unit ball (compare Problem 3.1). Hence any reasonable notion of size (i.e., one which is translation and rotation invariant) cannot be defined for all sets!

A collection of subsets $\mathcal{A}$ of a given set X such that

- $X \in \mathcal{A}$,
- $\mathcal{A}$ is closed under finite unions,
- $\mathcal{A}$ is closed under complements.

is called an **algebra**. Note that $\emptyset \in \mathcal{A}$ and that, by de Morgan, $\mathcal{A}$ is also closed under finite intersections. If an algebra is closed under countable unions (and hence also countable intersections), it is called a **σ-algebra**.

Moreover, the intersection of any family of (σ-)algebras $\{\mathcal{A}_\alpha\}$ is again a (σ-)algebra and for any collection S of subsets there is a unique smallest

(σ-)algebra $\Sigma(S)$ containing S (namely the intersection of all (σ-)algebra containing S). It is called the (σ-)algebra generated by S.

If X is a topological space, the **Borel σ-algebra** of X is defined to be the σ-algebra generated by all open (respectively all closed) sets. Sets in the Borel σ-algebra are called **Borel sets**.

Example. In the case $X = \mathbb{R}^n$ the Borel σ-algebra will be denoted by $\mathfrak{B}^n$ and we will abbreviate $\mathfrak{B} = \mathfrak{B}^1$. $\diamond$

When checking wether a given algebra is in fact a σ-algebra it turns out that it suffices to check increasing (or decreasing) sequences of sets: We will write $A_n \nearrow A$ if $A_n \subseteq A_{n+1}$ (note $A = \bigcup_n A_n$) and $A_n \searrow A$ if $A_{n+1} \subseteq A_n$ (note $A = \bigcap_n A_n$).

A collections of sets $\mathcal{M}$ is called a **monotone class** if $A_n \nearrow A$ implies $A \in \mathcal{M}$ whenever $A_n \in \mathcal{M}$ and $A_n \searrow A$ implies $A \in \mathcal{M}$ whenever $A_n \in \mathcal{M}$. Every σ-algebra is a monotone class and the intersection of monotone classes is a monotone class. Hence every collection of sets S generates a smallest monotone class $\mathcal{M}(S)$.

Theorem 3.1. *Let $\mathcal{A}$ be an algebra. Then $\mathcal{M}(\mathcal{A}) = \Sigma(\mathcal{A})$.*

Proof. We first show that $\mathcal{M} = \mathcal{M}(\mathcal{A})$ is an algebra.

Put $M(A) = \{B \in \mathcal{M} | A \cup B \in \mathcal{M}\}$. If B_n is an increasing sequence of sets in $M(A)$ then $A \cup B_n$ is an increasing sequence in $\mathcal{M}$ and hence $\bigcup_n (A \cup B_n) \in \mathcal{M}$. Now

$$A \cup \Big(\bigcup_n B_n\Big) = \bigcup_n (A \cup B_n) \tag{3.1}$$

shows that $M(A)$ is closed under increasing sequences. Similarly, $M(A)$ is closed under decreasing sequences and hence it is a monotone class. But does it contain any elements? Well if $A \in \mathcal{A}$ we have $\mathcal{A} \subseteq M(A)$ implying $M(A) = \mathcal{M}$. Hence $A \cup B \in \mathcal{M}$ if at least one of the sets is in $\mathcal{A}$. But this shows $\mathcal{A} \subseteq M(A)$ and hence $M(A) = \mathcal{M}$ for any $A \in \mathcal{M}$. So $\mathcal{M}$ is closed under finite unions.

To show that we are closed under complements consider $M = \{A \in \mathcal{M} | X \backslash A \in \mathcal{M}\}$. If A_n is an increasing sequence then $X \backslash A_n$ is a decreasing sequence and $X \backslash \bigcup_n A_n = \bigcap_n X \backslash A_n \in \mathcal{M}$ if $A_n \in M$. Similarly for decreasing sequences. Hence M is a monotone class and must be equal to $\mathcal{M}$ since it contains $\mathcal{A}$.

So we know that $\mathcal{M}$ is an algebra. To show that it is an σ-algebra let A_n be given and put $\tilde{A}_n = \bigcup_{k \le n} A_n$. Then $\tilde{A}_n$ is increasing and $\bigcup_n \tilde{A}_n = \bigcup_n A_n \in \mathcal{A}$. $\square$

The typical use of this theorem is as follows: First verify some property for sets in an algebra $\mathcal{A}$. In order to show that it holds for any set in $\Sigma(\mathcal{A})$, it suffices to show that the sets for which it holds is closed under countable increasing and decreasing sequences (i.e., is a monotone class).

Now let us turn to the definition of a measure: A set X together with a σ-algebra Σ is called a **measure space**. A **measure** μ is a map $\mu : \Sigma \to [0, \infty]$ on a σ-algebra Σ such that

- $\mu(\emptyset) = 0$,
- $\mu(\bigcup_{j=1}^{\infty} A_j) = \sum_{j=1}^{\infty} \mu(A_j)$ if $A_j \cap A_k = \emptyset$ for all j, k (σ-additivity).

It is called **σ-finite** if there is a countable cover $\{X_j\}_{j=1}^{\infty}$ of X with $\mu(X_j) < \infty$ for all j. (Note that it is no restriction to assume $X_j \nearrow X$.) It is called **finite** if $\mu(X) < \infty$. The sets in Σ are called **measurable sets**.

If we replace the σ-algebra by an algebra $\mathcal{A}$, then μ is called a **premeasure**. In this case σ-additivity clearly only needs to hold for disjoint sets A_n for which $\bigcup_n A_n \in \mathcal{A}$.

Theorem 3.2. *Any measure μ satisfies the following properties:*

(i) *$A \subseteq B$ implies $\mu(A) \le \mu(B)$ (monotonicity).*

(ii) *$\mu(A_n) \to \mu(A)$ if $A_n \nearrow A$ (continuity from below).*

(iii) *$\mu(A_n) \to \mu(A)$ if $A_n \searrow A$ and $\mu(A_1) < \infty$ (continuity from above).*

Proof. The first claim is obvious. The second follows using $\tilde{A}_n = A_n \backslash A_{n-1}$ and σ-additivity. The third follows from the second using $\tilde{A}_n = A_1 \backslash A_n$ and $\mu(\tilde{A}_n) = \mu(A_1) - \mu(A_n)$. □

Example. Let $A \in \mathfrak{P}(M)$ and set $\mu(A)$ to be the number of elements of A (respectively ∞ if A is infinite). This is the so called **counting measure**.

Note that if $X = \mathbb{N}$ and $A_n = \{j \in \mathbb{N} | j \ge n\}$, then $\mu(A_n) = \infty$, but $\mu(\bigcap_n A_n) = \mu(\emptyset) = 0$ which shows that the requirement $\mu(A_1) < \infty$ in the last claim of Theorem 3.2 is not superfluous. ◇

A measure on the Borel σ-algebra is called a **Borel measure** if $\mu(C) < \infty$ for any compact set C. A Borel measures is called **outer regular** if

$$\mu(A) = \inf_{A \subseteq O, O \text{ open}} \mu(O) \tag{3.2}$$

and **inner regular** if

$$\mu(A) = \sup_{C \subseteq A, C \text{ compact}} \mu(C). \tag{3.3}$$

It is called regular if it is both outer and inner regular.

But how can we obtain some more interesting Borel measures? We will restrict ourselves to the case of $X = \mathbb{R}$ for simplicity. Then the strategy is as follows: Start with the algebra of finite unions of disjoint intervals and define μ for those sets (as the sum over the intervals). This yields a premeasure. Extend this to an *outer measure* for all subsets of $\mathbb{R}$. Show that the restriction to the Borel sets is a measure.

Let us first show how we should define μ for intervals: To every Borel measure on $\mathfrak{B}$ we can assign its distribution function

$$\mu(x) = \begin{cases} -\mu((x,0]), & x < 0 \\ 0, & x = 0 \\ \mu((0,x]), & x > 0 \end{cases} \tag{3.4}$$

which is right continuous and non-decreasing. Conversely, given a right continuous non-decreasing function $\mu : \mathbb{R} \to \mathbb{R}$ we can set

$$\mu(A) = \begin{cases} \mu(b) - \mu(a), & A = (a,b] \\ \mu(b) - \mu(a-), & A = [a,b] \\ \mu(b-) - \mu(a), & A = (a,b) \\ \mu(b-) - \mu(a-), & A = [a,b) \end{cases}, \tag{3.5}$$

where $\mu(a-) = \lim_{\varepsilon\downarrow 0} \mu(a - \varepsilon)$.

Lemma 3.3. *μ as defined in (3.5) gives rise to a unique σ-finite regular premeasure on the algebra $\mathcal{A}$ of finite unions of disjoint intervals.*

Proof. First of all, (3.5) can be extended to finite unions of disjoint intervals by summing over all intervals. It is straightforward to verify that μ is well defined (one set can be represented by different unions of intervals) and by construction additive.

To show regularity, we can assume any such union to consist of open intervals and points only. To show outer regularity replace each point $\{x\}$ by a small open interval $(x+\varepsilon, x-\varepsilon)$ and use that $\mu(\{x\}) = \lim_{\varepsilon\downarrow} \mu(x+\varepsilon) - \mu(x-\varepsilon)$. Similarly, to show inner regularity, replace each open interval (a,b) by a compact one $[a_n, b_n] \subseteq (a,b)$ and use $\mu((a,b)) = \lim_{n\to\infty} \mu(b_n) - \mu(a_n)$ if $a_n \downarrow a$ and $b_n \uparrow b$.

It remains to verify σ-additivity. We need to show

$$\mu(\bigcup_k I_k) = \sum_k \mu(I_k) \tag{3.6}$$

whenever $I_n \in \mathcal{A}$ and $I = \bigcup_k I_k \in \mathcal{A}$. Since each I_n is a finite union of intervals, we can as well assume each I_n is just one interval (just split I_n into its subintervals and note that the sum does not change by additivity). Similarly, we can assume that I is just one interval (just treat each subinterval separately).

By additivity μ is monotone and hence

$$\sum_{k=1}^{n} \mu(I_k) = \mu(\bigcup_{k=1}^{n} I_k) \le \mu(I) \tag{3.7}$$

which shows

$$\sum_{k=1}^{\infty} \mu(I_k) \le \mu(I). \tag{3.8}$$

To get the converse inequality we need to work harder.

By outer regularity we can cover each I_k by open interval J_k such that $\mu(J_k) \le \mu(I_k) + \frac{\varepsilon}{2^k}$. Suppose I is compact first. Then finitely many of the J_k, say the first n, cover I and we have

$$\mu(I) \le \mu(\bigcup_{k=1}^{n} J_k) \le \sum_{k=1}^{n} \mu(J_k) \le \sum_{k=1}^{\infty} \mu(I_k) + \varepsilon. \tag{3.9}$$

Since $\varepsilon > 0$ is arbitrary, this shows σ-additivity for compact intervals. By additivity we can always add/subtract the end points of I and hence σ-additivity holds for any bounded interval. If I is unbounded, say $I = [a, \infty)$, then given $x > 0$ we can find an n such that J_n cover at least $[0, x]$ and hence

$$\sum_{k=1}^{n} \mu(I_k) \ge \sum_{k=1}^{n} \mu(J_k) - \varepsilon \ge \mu([a, x]) - \varepsilon. \tag{3.10}$$

Since $x > a$ and $\varepsilon > 0$ are arbitrary we are done. □

This premeasure determines the corresponding measure μ uniquely (if there is one at all):

Theorem 3.4 (Uniqueness of measures)**.** *Let μ be a σ-finite premeasure on an algebra $\mathcal{A}$. Then there is at most one extension to $\Sigma(\mathcal{A})$.*

Proof. We first assume that $\mu(X) < \infty$. Suppose there is another extension $\tilde{\mu}$ and consider the set

$$S = \{A \in \Sigma(\mathcal{A}) | \mu(A) = \tilde{\mu}(A)\}. \tag{3.11}$$

I claim S is a monotone class and hence $S = \Sigma(\mathcal{A})$ since $\mathcal{A} \subseteq S$ by assumption (Theorem 3.1).

Let $A_n \nearrow A$. If $A_n \in S$ we have $\mu(A_n) = \tilde{\mu}(A_n)$ and taking limits (Theorem 3.2 (ii)) we conclude $\mu(A) = \tilde{\mu}(A)$. Next let $A_n \searrow A$ and take again limits. This finishes the finite case. To extend our result to the σ-finite case let $X_j \nearrow X$ be an increasing sequence such that $\mu(X_j) < \infty$. By the finite case $\mu(A \cap X_j) = \tilde{\mu}(A \cap X_j)$ (just restrict μ, $\tilde{\mu}$ to X_j). Hence

$$\mu(A) = \lim_{j\to\infty} \mu(A \cap X_j) = \lim_{j\to\infty} \tilde{\mu}(A \cap X_j) = \tilde{\mu}(A) \tag{3.12}$$

and we are done. □

Note that if our premeasure is regular, so will be the extension:

Lemma 3.5. *Suppose μ is a σ-finite premeasure on some algebra $\mathcal{A}$ generating the Borel sets $\mathfrak{B}$. Then outer (inner) regularity holds for all Borel sets if it holds for all sets in $\mathcal{A}$.*

Proof. We first assume that $\mu(X) < \infty$. Set

$$\mu^\circ(A) = \inf_{A\subseteq O, O \text{ open}} \mu(O) \geq \mu(A) \tag{3.13}$$

and let $M = \{A \in \mathfrak{B} | \mu^\circ(A) = \mu(A)\}$. Since by assumption M contains some algebra generating $\mathfrak{B}$ it suffices to proof that M is a monotone class.

Let $A_n \in M$ be a monotone sequence and let $O_n \supseteq A_n$ be open sets such that $\mu(O_n) \leq \mu(A_n) + \frac{1}{n}$. Then

$$\mu(A_n) \leq \mu^\circ(A_n) \leq \mu(O_n) \leq \mu(A_n) + \frac{1}{n}. \tag{3.14}$$

Now if $A_n \nearrow A$ just take limits and use continuity from below of μ. Similarly if $A_n \searrow A$.

Now let μ be arbitrary. Given A we can split it into disjoint sets A_j such that $A_j \subseteq X_j$ ($A_1 = A \cap X_1$, $A_2 = (A\backslash A_1) \cap X_2$, etc.). Let X_j be a cover with $\mu(X_j) < \infty$. By regularity, we can assume X_j open. Thus there are open (in X) sets O_j covering A_j such that $\mu(O_j) \leq \mu(A_j) + \frac{\varepsilon}{2^j}$. Then $O = \bigcup_j O_j$ is open, covers A, and satisfies

$$\mu(A) \leq \mu(O) \leq \sum_j \mu(O_j) \leq \mu(A) + \varepsilon. \tag{3.15}$$

This settles outer regularity.

Next let us turn to inner regularity. If $\mu(X) < \infty$ one can show as before that $M = \{A \in \mathfrak{B} | \mu_\circ(A) = \mu(A)\}$, where

$$\mu_\circ(A) = \sup_{C\subseteq A, C \text{ compact}} \mu(C) \leq \mu(A) \tag{3.16}$$

is a monotone class. This settles the finite case.

For the σ-finite case split again A as before. Since X_j has finite measure, there are compact subsets K_j of A_j such that $\mu(A_j) \leq \mu(K_j) + \frac{\varepsilon}{2^j}$. Now we need to distinguish two cases: If $\mu(A) = \infty$, the sum $\sum_j \mu(A_j)$ will diverge and so will $\sum_j \mu(K_j)$. Hence $\tilde{K}_n = \bigcup_{j=1}^n \subseteq A$ is compact with $\mu(\tilde{K}_n) \to \infty = \mu(A)$. If $\mu(A) < \infty$, the sum $\sum_j \mu(A_j)$ will converge and choosing n sufficiently large we will have

$$\mu(\tilde{K}_n) \leq \mu(A) \leq \mu(\tilde{K}_n) + 2\varepsilon. \tag{3.17}$$

This finishes the proof. □

So it remains to ensure that there is an extension at all. For any premeasure μ we define

$$\mu^*(A) = \inf\Big\{ \sum_{n=1}^\infty \mu(A_n) \Big| A \subseteq \bigcup_{n=1}^\infty A_n,\ A_n \in \mathcal{A} \Big\} \tag{3.18}$$

where the infimum extends over all countable covers from $\mathcal{A}$. Then the function $\mu^* : \mathfrak{P}(X) \to [0, \infty]$ is an **outer measure**, that is, it has the properties (Problem 3.2)

- $\mu^*(\emptyset) = 0$,
- $A_1 \subseteq A_2 \ \Rightarrow\ \mu^*(A_1) \le \mu^*(A_2)$, and
- $\mu^*(\bigcup_{n=1}^\infty A_n) \le \sum_{n=1}^\infty \mu^*(A_n)$ (subadditivity).

Note that $\mu^*(A) = \mu(A)$ for $A \in \mathcal{A}$ (Problem 3.3).

Theorem 3.6 (Extensions via outer measures)**.** *Let μ^* be an outer measure. Then the set Σ of all sets A satisfying the Carathéodory condition*

$$\mu^*(E) = \mu^*(A \cap E) + \mu^*(A' \cap E) \quad \forall E \subseteq X \tag{3.19}$$

(where $A' = X \backslash A$ is the complement of A) form a σ-algebra and μ^* restricted to this σ-algebra is a measure.

Proof. We first show that Σ is an algebra. It clearly contains X and is closed under complements. Let $A, B \in \Sigma$. Applying Carathéodory's condition twice finally shows

$$\begin{aligned} \mu^*(E) &= \mu^*(A \cap B \cap E) + \mu^*(A' \cap B \cap E) + \mu^*(A \cap B' \cap E) \\ &\quad + \mu^*(A' \cap B' \cap E) \\ &\ge \mu^*((A \cup B) \cap E) + \mu^*((A \cup B)' \cap E), \end{aligned} \tag{3.20}$$

where we have used De Morgan and

$$\mu^*(A \cap B \cap E) + \mu^*(A' \cap B \cap E) + \mu^*(A \cap B' \cap E) \ge \mu^*((A \cup B) \cap E) \tag{3.21}$$

which follows from subadditivity and $(A \cup B) \cap E = (A \cap B \cap E) \cup (A' \cap B \cap E) \cup (A \cap B' \cap E)$. Since the reverse inequality is just subadditivity, we conclude that Σ is an algebra.

Next, let A_n be a sequence of sets from Σ. Without restriction we can assume that they are disjoint (compare the last argument in proof of Theorem 3.1). Abbreviate $\tilde{A}_n = \bigcup_{k \le n} A_n$, $A = \bigcup_n A_n$. Then for any set E we have

$$\begin{aligned} \mu^*(\tilde{A}_n \cap E) &= \mu^*(A_n \cap \tilde{A}_n \cap E) + \mu^*(A_n' \cap \tilde{A}_n \cap E) \\ &= \mu^*(A_n \cap E) + \mu^*(\tilde{A}_{n-1} \cap E) \\ &= \ldots = \sum_{k=1}^n \mu^*(A_k \cap E). \end{aligned} \tag{3.22}$$

Using $\tilde{A}_n \in \Sigma$ and monotonicity of μ^*, we infer

$$\begin{aligned} \mu^*(E) &= \mu^*(\tilde{A}_n \cap E) + \mu^*(\tilde{A}'_n \cap E) \\ &\geq \sum_{k=1}^{n} \mu^*(A_k \cap E) + \mu^*(A' \cap E). \end{aligned} \tag{3.23}$$

Letting $n \to \infty$ and using subadditivity finally gives

$$\begin{aligned} \mu^*(E) &\geq \sum_{k=1}^{\infty} \mu^*(A_k \cap E) + \mu^*(A' \cap E) \\ &\geq \mu^*(A \cap E) + \mu^*(B' \cap E) \geq \mu^*(E) \end{aligned} \tag{3.24}$$

and we infer that Σ is a σ-algebra.

Finally, setting $E = A$ in (3.24) we have

$$\mu^*(A) = \sum_{k=1}^{\infty} \mu^*(A_k \cap A) + \mu^*(A' \cap A) = \sum_{k=1}^{\infty} \mu^*(A_k) \tag{3.25}$$

and we are done. □

Remark: The constructed measure μ is **complete**, that is, for any measurable set A of measure zero, any subset of A is again measurable (Problem 3.4).

The only remaining question is wether there are any nontrivial sets satisfying the Carathéodory condition.

Lemma 3.7. *Let μ be a premeasure on $\mathcal{A}$ and let μ^* be the associated outer measure. Then every set in $\mathcal{A}$ satisfies the Carathéodory condition.*

Proof. Let $A_n \in \mathcal{A}$ be a countable cover for E. Then for any $A \in \mathcal{A}$ we have

$$\sum_{n=1}^{\infty} \mu(A_n) = \sum_{n=1}^{\infty} \mu(A_n \cap A) + \sum_{n=1}^{\infty} \mu(A_n \cap A') \geq \mu^*(E \cap A) + \mu^*(E \cap A') \tag{3.26}$$

since $A_n \cap A \in \mathcal{A}$ is a cover for $E \cap A$ and $A_n \cap A' \in \mathcal{A}$ is a cover for $E \cap A'$. Taking the infimum we have $\mu^*(E) \geq \mu^*(E \cap A) + \mu^*(E \cap A')$ which finishes the proof. □

Thus, as a consequence we obtain:

Theorem 3.8. *For every right continuous non-decreasing function $\mu : \mathbb{R} \to \mathbb{R}$ there exists a unique regular Borel measure μ which extends (3.5). Two different functions generate the same measure if and only if they differ by a constant.*

Example. Suppose $\Theta(x) = 0$ for $x < 0$ and $\Theta(x) = 1$ for $x \geq 0$. Then we obtain the so-called **Dirac measure** at 0, which is given by $\Theta(A) = 1$ if $0 \in A$ and $\Theta(A) = 0$ if $0 \notin A$. $\diamond$

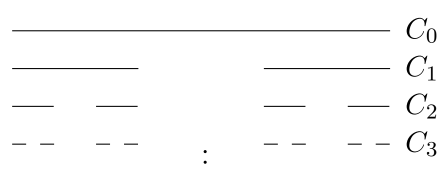
Example. Suppose $\lambda(x) = x$, then the associated measure is the ordinary **Lebesgue measure** on $\mathbb{R}$. We will abbreviate the Lebesgue measure of a Borel set A by $\lambda(A) = |A|$. $\diamond$

It can be shown that Borel measures on a separable metric space are always regular.

A set $A \in \Sigma$ is called a **support** for μ if $\mu(X \backslash A) = 0$. A property is said to hold **μ-almost everywhere** (a.e.) if the it holds on a support for μ or, equivalently, if the set where it does not hold is contained in a set of measure zero.

Example. The set of rational numbers has Lebesgue measure zero: $\lambda(\mathbb{Q}) = 0$. In fact, any single point has Lebesgue measure zero, and so has any countable union of points (by countable additivity). $\diamond$

Example. The **Cantor set** is an example of a closed uncountable set of Lebesgue measure zero. It is constructed as follows: Start with $C_0 = [0, 1]$ and remove the middle third to obtain $C_1 = [0, \frac{1}{3}] \cup [\frac{2}{3}, 1]$. Next, again remove the middle third's of the remaining sets to obtain $C_2 = [0, \frac{1}{9}] \cup [\frac{2}{9}, \frac{1}{3}] \cup [\frac{2}{3}, \frac{7}{9}] \cup [\frac{8}{9}, 1]$.



Proceeding like this we obtain a sequence of nesting sets C_n and the limit $C = \bigcap_n C_n$ is the Cantor set. Since C_n is compact, so is C. Moreover, C_n consists of 2^n intervals of length 3^{-n}, and thus its Lebesgue measure is $\lambda(C_n) = (2/3)^n$. In particular, $\lambda(C) = \lim_{n\to\infty} \lambda(C_n) = 0$. Using the ternary expansion it is extremely simple to describe: C is the set of all $x \in [0, 1]$ whose ternary expansion contains no one's, which shows that C is uncountable (why?). It has some further interesting properties: it is totally disconnected (i.e., it contains no subintervals) and perfect (it has no isolated points). $\diamond$

Problem 3.1 (Vitali set)**.** *Call two numbers $x, y \in [0, 1)$ equivalent if $x - y$ is rational. Construct the set V by choosing one representative from each equivalence class. Show that V cannot be measurable with respect to any finite translation invariant measure on $[0, 1)$. (Hint: How can you build up $[0, 1)$ from V?)*

Problem 3.2. *Show that μ^* defined in (3.18) is an outer measure. (Hint for the last property: Take a cover $\{B_{nk}\}_{k=1}^\infty$ for A_n such that $\mu^*(A_n) = \frac{\varepsilon}{2^n} + \sum_{k=1}^\infty \mu(B_{nk})$ and note that $\{B_{nk}\}_{n,k=1}^\infty$ is a cover for $\bigcup_n A_n$.)*

Problem 3.3. *Show that μ^* defined in (3.18) extends μ. (Hint: For the cover A_n it is no restriction to assume $A_n \cap A_m = \emptyset$ and $A_n \subseteq A$.)*

Problem 3.4. *Show that the measure constructed in Theorem 3.6 is complete.*

3.2. Measurable functions

The Riemann integral works by splitting the x coordinate into small intervals and approximating $f(x)$ on each interval by its minimum and maximum. The problem with this approach is that the difference between maximum and minimum will only tend to zero (as the intervals get smaller) if $f(x)$ is sufficiently nice. To avoid this problem we can force the difference to go to zero by considering, instead of an interval, the set of x for which $f(x)$ lies between two given numbers $a < b$. Now we need the size of the set of these x, that is, the size of the preimage $f^{-1}((a,b))$. For this to work, preimages of intervals must be measurable.

A function $f : X \to \mathbb{R}^n$ is called **measurable** if $f^{-1}(A) \in \Sigma$ for every $A \in \mathfrak{B}^n$. A complex-valued function is called measurable if both its real and imaginary parts are. Clearly it suffices to check this condition for every set A in a collection of sets which generate $\mathfrak{B}^n$, say for all open intervals or even for open intervals which are products of (a, ∞):

Lemma 3.9. *A function $f : X \to \mathbb{R}^n$ is measurable if and only if*

$$f^{-1}(I) \in \Sigma \qquad \forall\, I = \prod_{j=1}^n (a_j, \infty). \tag{3.27}$$

In particular, a function $f : X \to \mathbb{R}^n$ is measurable if and only if every component is measurable.

Proof. All you have to use is $f^{-1}(\mathbb{R}^n \setminus A) = X \setminus f^{-1}(A)$, $f^{-1}(\bigcup_j A_j) = \bigcup_j f^1(A_j)$ and the fact that any open set is a countable union of open intervals. □

If Σ is the Borel σ-algebra, we will call a measurable function also **Borel function**. Note that, in particular,

Lemma 3.10. *Any continuous function is measurable and the composition of two measurable functions is again measurable.*

Moreover, sometimes it is also convenient to allow $\pm\infty$ as possible values for f, that is, functions $f : X \to \overline{\mathbb{R}}$, $\overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, \infty\}$. In this case $A \subseteq \overline{\mathbb{R}}$ is called Borel if $A \cap \mathbb{R}$ is.

The set of all measurable functions forms an algebra.

Lemma 3.11. *Suppose $f, g : X \to \mathbb{R}$ are measurable functions. Then the sum $f + g$ and the product fg is measurable.*

Proof. Note that addition and multiplication are continuous functions from $\mathbb{R}^2 \to \mathbb{R}$ and hence the claim follows from the previous lemma. □

Moreover, the set of all measurable functions is closed under all important limiting operations.

Lemma 3.12. *Suppose $f_n : X \to \mathbb{R}$ is a sequence of measurable functions, then*

$$\inf_{n\in\mathbb{N}} f_n, \quad \sup_{n\in\mathbb{N}} f_n, \quad \liminf_{n\to\infty} f_n, \quad \limsup_{n\to\infty} f_n \tag{3.28}$$

are measurable as well.

Proof. It suffices to proof that $\sup f_n$ is measurable since the rest follows from $\inf f_n = -\sup(-f_n)$, $\liminf f_n = \sup_k \inf_{n\geq k} f_n$, and $\limsup f_n = \inf_k \sup_{n\geq k} f_n$. But $(\sup f_n)^{-1}((a,\infty)) = \bigcup_n f_n^{-1}((a,\infty))$ and we are done. □

It follows that if f and g are measurable functions, so are $\min(f,g)$, $\max(f,g)$, $|f| = \max(f,-f)$, $f^{\pm} = \max(\pm f, 0)$.

3.3. Integration — Sum me up Henri

Now we can define the integral for measurable functions as follows. A measurable function $s : X \to \mathbb{R}$ is called **simple** if its range is finite $s(X) = \{\alpha_j\}_{j=1}^p$, that is, if

$$s = \sum_{j=1}^{p} \alpha_j \, \chi_{A_j}, \qquad A_j = s^{-1}(\alpha_j) \in \Sigma. \tag{3.29}$$

Here χ_A is the **characteristic function** of A, that is, $\chi_A(x) = 1$ if $x \in A$ and $\chi_A(x) = 0$ else.

For a positive simple function we define its **integral** as

$$\int_A s \, d\mu = \sum_{j=1}^{n} \alpha_j \, \mu(A_j \cap A). \tag{3.30}$$

Here we use the convention $0 \cdot \infty = 0$.

Lemma 3.13. *The integral has the following properties:*

(i) $\int_A s\,d\mu = \int_X \chi_A\, s\,d\mu$.
(ii) $\int_{\bigcup_{j=1}^\infty A_j} s\,d\mu = \sum_{j=1}^\infty \int_{A_j} s\,d\mu$.
(iii) $\int_A \alpha\, s\,d\mu = \alpha \int_A s\,d\mu$.
(iv) $\int_A (s+t)d\mu = \int_A s\,d\mu + \int_A t\,d\mu$.
(v) $A \subseteq B \;\Rightarrow\; \int_A s\,d\mu \le \int_B s\,d\mu$.
(vi) $s \le t \;\Rightarrow\; \int_A s\,d\mu \le \int_A t\,d\mu$.

Proof. (i) is clear from the definition. (ii) follows from σ-additivity of μ. (iii) is obvious. (iv) Let $s = \sum_j \alpha_j\,\chi_{A_j}$, $t = \sum_j \beta_j\,\chi_{B_j}$ and abbreviate $C_{jk} = (A_j \cap B_k) \cap A$. Then

$$\begin{aligned}\int_A (s+t)d\mu &= \sum_{j,k} \int_{C_{jk}} (s+t)d\mu = \sum_{j,k} (\alpha_j + \beta_k)\mu(C_{jk})\\ &= \sum_{j,k} \left(\int_{C_{jk}} s\,d\mu + \int_{C_{jk}} t\,d\mu \right) = \int_A s\,d\mu + \int_A t\,d\mu \quad (3.31)\end{aligned}$$

(v) follows from monotonicity of μ. (vi) follows using $t - s \ge 0$ and arguing as in (iii). □

Our next task is to extend this definition to arbitrary positive functions by

$$\int_A f\,d\mu = \sup_{s \le f} \int_A s\,d\mu, \qquad (3.32)$$

where the supremum is taken over all simple functions $s \le f$. Note that, except for possibly (ii) and (iv), Lemma 3.13 still holds for this extension.

Theorem 3.14 (monotone convergence)**.** *Let f_n be a monotone non-decreasing sequence of positive measurable functions, $f_n \nearrow f$. Then*

$$\int_A f_n\,d\mu \to \int_A f\,d\mu. \qquad (3.33)$$

Proof. By property (v) $\int_A f_n\,d\mu$ is monotone and converges to some number α. By $f_n \le f$ and again (v) we have

$$\alpha \le \int_A f\,d\mu. \qquad (3.34)$$

To show the converse let s be simple such that $s \le f$ and let $\theta \in (0,1)$. Put $A_n = \{x \in A | f_n(x) \ge \theta s(x)\}$ and note $A_n \nearrow X$ (show this). Then

$$\int_A f_n\,d\mu \ge \int_{A_n} f_n\,d\mu \ge \theta \int_{A_n} s\,d\mu. \qquad (3.35)$$

Letting $n \to \infty$ we see

$$\alpha \geq \theta \int_A s\, d\mu. \tag{3.36}$$

Since this is valid for any $\theta < 1$, it still holds for $\theta = 1$. Finally, since $s \leq f$ is arbitrary, the claim follows. □

In particular

$$\int_A f\, d\mu = \lim_{n\to\infty} \int_A s_n\, d\mu, \tag{3.37}$$

for any monotone sequence $s_n \nearrow f$ of simple functions. Note that there is always such a sequence, for example,

$$s_n(x) = \sum_{k=0}^{n^2} \frac{k}{n} \chi_{f^{-1}(A_k)}(x), \quad A_k = [\frac{k}{n}, \frac{k+1}{n^2}),\ A_{n^2} = [n, \infty). \tag{3.38}$$

By construction s_n converges uniformly if f is bounded, since $s_n(x) = n$ if $f(x) = \infty$ and $f(x) - s_n(x) < \frac{1}{n}$ if $f(x) < n + 1$.

Now what about the missing items (ii) and (iv) from Lemma 3.13? Since limits can be spread over sums, the extension is linear (i.e., item (iv) holds) and (ii) also follows directly from the monotone convergence theorem. We even have the following result:

Lemma 3.15. *If $f \geq 0$ is measurable, then $d\nu = f\, d\mu$ defined via*

$$\nu(A) = \int_A f\, d\mu \tag{3.39}$$

is a measure such that

$$\int g\, d\nu = \int g f\, d\mu. \tag{3.40}$$

Proof. As already mentioned, additivity of μ is equivalent to linearity of the integral and σ-additivity follows from the monotone convergence theorem

$$\nu(\bigcup_{n=1}^{\infty} A_n) = \int (\sum_{n=1}^{\infty} \chi_{A_n}) f\, d\mu = \sum_{n=1}^{\infty} \int \chi_{A_n} f\, d\mu = \sum_{n=1}^{\infty} \nu(A_n). \tag{3.41}$$

The second claim holds for simple functions and hence for all functions by construction of the integral. □

If f_n is not necessarily monotone we have at least

Theorem 3.16 (Fatou's Lemma)**.** *If f_n is a sequence of nonnegative measurable function, then*

$$\int_A \liminf_{n\to\infty} f_n\, d\mu \leq \liminf_{n\to\infty} \int_A f_n\, d\mu, \tag{3.42}$$

Proof. Set $g_n = \inf_{k\geq n} f_k$. Then $g_n \leq f_n$ implying

$$\int_A g_n \, d\mu \leq \int_A f_n \, d\mu. \tag{3.43}$$

Now take the lim inf on both sides and note that by the monotone convergence theorem

$$\liminf_{n\to\infty} \int_A g_n \, d\mu = \lim_{n\to\infty} \int_A g_n \, d\mu = \int_A \lim_{n\to\infty} g_n \, d\mu = \int_A \liminf_{n\to\infty} f_n \, d\mu, \tag{3.44}$$

proving the claim. □

If the integral is finite for both the positive and negative part $f^{\pm}$ of an arbitrary measurable function f, we call f **integrable** and set

$$\int_A f \, d\mu = \int_A f^+ d\mu - \int_A f^- d\mu. \tag{3.45}$$

The set of all integrable functions is denoted by $\mathcal{L}^1(X, d\mu)$.

Lemma 3.17. *Lemma 3.13 holds for integrable functions s, t.*

Similarly, we handle the case where f is complex-valued by calling f integrable if both the real and imaginary part are and setting

$$\int_A f \, d\mu = \int_A \mathrm{Re}(f) d\mu + \mathrm{i} \int_A \mathrm{Im}(f) d\mu. \tag{3.46}$$

Clearly f is integrable if and only if $|f|$ is.

Lemma 3.18. *For any integrable functions f, g we have*

$$|\int_A f \, d\mu| \leq \int_A |f| \, d\mu \tag{3.47}$$

and (triangle inequality)

$$\int_A |f+g| \, d\mu \leq \int_A |f| \, d\mu + \int_A |g| \, d\mu. \tag{3.48}$$

Proof. Put $\alpha = \frac{z^*}{|z|}$, where $z = \int_A f \, d\mu$ (without restriction $z \neq 0$). Then

$$|\int_A f \, d\mu| = \alpha \int_A f \, d\mu = \int_A \alpha \, f \, d\mu = \int_A \mathrm{Re}(\alpha \, f) \, d\mu \leq \int_A |f| \, d\mu. \tag{3.49}$$

proving the first claim. The second follows from $|f+g| \leq |f| + |g|$. □

In addition, our integral is well behaved with respect to limiting operations.

Theorem 3.19 (dominated convergence)**.** *Let f_n be a convergent sequence of measurable functions and set $f = \lim_{n\to\infty} f_n$. Suppose there is an integrable function g such that $|f_n| \le g$. Then f is integrable and*

$$\lim_{n\to\infty} \int f_n d\mu = \int f d\mu. \tag{3.50}$$

Proof. The real and imaginary parts satisfy the same assumptions and so do the positive and negative parts. Hence it suffices to prove the case where f_n and f are nonnegative.

By Fatou's lemma

$$\liminf_{n\to\infty} \int_A f_n d\mu \ge \int_A f\, d\mu \tag{3.51}$$

and

$$\liminf_{n\to\infty} \int_A (g - f_n) d\mu \ge \int_A (g - f) d\mu. \tag{3.52}$$

Subtracting $\int_A g\, d\mu$ on both sides of the last inequality finishes the proof since $\liminf(-f_n) = -\limsup f_n$. □

Remark: Since sets of measure zero do not contribute to the value of the integral, it clearly suffices if the requirements of the dominated convergence theorem are satisfied almost everywhere (with respect to μ).

Note that the existence of g is crucial, as the example $f_n(x) = \frac{1}{n}\chi_{[-n,n]}(x)$ on $\mathbb{R}$ with Lebesgue measure shows.

Example. If $\mu(x) = \sum_n \alpha_n \Theta(x - x_n)$ is a sum of Dirac measures $\Theta(x)$ centered at $x = 0$, then

$$\int f(x) d\mu(x) = \sum_n \alpha_n f(x_n). \tag{3.53}$$

Hence our integral contains sums as special cases. ◇

Problem 3.5. *Show that the set $B(X)$ of bounded measurable functions is a Banach space. Show that the set $S(X)$ of simple functions is dense in $B(X)$. Show that the integral is a bounded linear functional on $B(X)$. (Hence Theorem 1.25 could be used to extend the integral from simple to bounded measurable functions.)*

Problem 3.6. *Show that the dominated convergence theorem implies (under the same assumptions)*

$$\lim_{n\to\infty} \int |f_n - f| d\mu = 0. \tag{3.54}$$

Problem 3.7. *Suppose $y \mapsto f(x,y)$ is measurable for every x and $x \mapsto f(x,y)$ is continuous for every y. Show that*

$$F(x) = \int_A f(x,y)\, d\mu(y) \tag{3.55}$$

is continuous if there is an integrable function $g(y)$ such that $|f(x,y)| \le g(y)$.

Problem 3.8. *Suppose $y \mapsto f(x,y)$ is measurable for fixed x and $x \mapsto f(x,y)$ is differentiable for fixed y. Show that*

$$F(x) = \int_A f(x,y)\, d\mu(y) \tag{3.56}$$

is differentiable if there is an integrable function $g(y)$ such that $|\frac{\partial}{\partial x} f(x,y)| \le g(y)$. Moreover, $x \mapsto \frac{\partial}{\partial x} f(x,y)$ is measurable and

$$F'(x) = \int_A \frac{\partial}{\partial x} f(x,y)\, d\mu(y) \tag{3.57}$$

in this case.

3.4. Product measures

Let μ_1 and μ_2 be two measures on Σ_1 and Σ_2, respectively. Let $\Sigma_1 \otimes \Sigma_2$ be the σ-algebra generated by **rectangles** of the form $A_1 \times A_2$.

Example. Let $\mathfrak{B}$ be the Borel sets in $\mathbb{R}$ then $\mathfrak{B}^2 = \mathfrak{B} \otimes \mathfrak{B}$ are the Borel sets in $\mathbb{R}^2$ (since the rectangles are a basis for the product topology). ⋄

Any set in $\Sigma_1 \otimes \Sigma_2$ has the **section property**, that is,

Lemma 3.20. *Suppose $A \in \Sigma_1 \otimes \Sigma_2$ then its sections*

$$A_1(x_2) = \{x_1 | (x_1, x_2) \in A\} \quad \text{and} \quad A_2(x_1) = \{x_2 | (x_1, x_2) \in A\} \tag{3.58}$$

are measurable.

Proof. Denote all sets $A \in \Sigma_1 \otimes \Sigma_2$ in with the property that $A_1(x_2) \in \Sigma_1$ by S. Clearly all rectangles are in S and it suffices to show that S is a σ-algebra. Moreover, if $A \in S$, then $(A')_1(x_2) = (A_1(x_2))' \in \Sigma_2$ and thus S is closed under complements. Similarly, if $A_n \in S$, then $(\bigcup_n A_n)_1(x_2) = \bigcup_n (A_n)_1(x_2)$ shows that S is closed under countable unions. □

This implies that if f is a measurable function on $X_1 \times X_2$, then $f(., x_2)$ is measurable on X_1 for every x_2 and $f(x_1, .)$ is measurable on X_2 for every x_1 (observe $A_1(x_2) = \{x_1 | f(x_1, x_2) \in B\}$, where $A = \{(x_1, x_2) | f(x_1, x_2) \in B\}$). In fact, this is even equivalent since $\chi_{A_1(x_2)}(x_1) = \chi_{A_2(x_1)}(x_2) = \chi_A(x_1, x_2)$.

Given two measures μ_1 on Σ_1 and μ_2 on Σ_2 we now want to construct the **product measure**, $\mu_1 \otimes \mu_2$ on $\Sigma_1 \otimes \Sigma_2$ such that

$$\mu_1 \otimes \mu_2(A_1 \times A_2) = \mu_1(A_1)\mu_2(A_2), \quad A_j \in \Sigma_j,\ j = 1, 2. \tag{3.59}$$

Theorem 3.21. *Let μ_1 and μ_2 be two σ-finite measures on Σ_1 and Σ_2, respectively. Let $A \in \Sigma_1 \otimes \Sigma_2$. Then $\mu_2(A_2(x_1))$ and $\mu_1(A_1(x_2))$ are measurable and*

$$\int_{X_1} \mu_2(A_2(x_1)) d\mu_1(x_1) = \int_{X_2} \mu_1(A_1(x_2)) d\mu_2(x_2). \tag{3.60}$$

Proof. Let S be the set of all subsets for which our claim holds. Note that S contains at least all rectangles. It even contains the algebra of finite disjoint unions of rectangles. Thus it suffices to show that S is a monotone class. If μ_1 and μ_2 are finite, this follows from continuity from above and below of measures. The case if μ_1 and μ_2 are σ-finite can be handled as in Theorem 3.4. □

Hence we can define

$$\mu_1 \otimes \mu_2(A) = \int_{X_1} \mu_2(A_2(x_1)) d\mu_1(x_1) = \int_{X_2} \mu_1(A_1(x_2)) d\mu_2(x_2) \tag{3.61}$$

or equivalently

$$\begin{aligned} \mu_1 \otimes \mu_2(A) &= \int_{X_1} \left(\int_{X_2} \chi_A(x_1, x_2) d\mu_2(x_2) \right) d\mu_1(x_1) \\ &= \int_{X_2} \left(\int_{X_1} \chi_A(x_1, x_2) d\mu_1(x_1) \right) d\mu_2(x_2). \end{aligned} \tag{3.62}$$

Additivity of $\mu_1 \otimes \mu_2$ follows from the monotone convergence theorem.

Note that (3.59) uniquely defines $\mu_1 \otimes \mu_2$ as a σ-finite premeasure on the algebra of finite disjoint unions of rectangles. Hence by Theorem 3.4 it is the only measure on $\Sigma_1 \otimes \Sigma_2$ satisfying (3.59).

Finally we have:

Theorem 3.22 (Fubini)**.** *Let f be a measurable function on $X_1 \times X_2$ and let μ_1, μ_2 be σ-finite measures on X_1, X_2, respectively.*

(i) *If $f \ge 0$ then $\int f(., x_2) d\mu_2(x_2)$ and $\int f(x_1, .) d\mu_1(x_1)$ are both measurable and*

$$\begin{aligned} \iint f(x_1, x_2) d\mu_1 \otimes \mu_2(x_1, x_2) &= \int \left(\int f(x_1, x_2) d\mu_1(x_1) \right) d\mu_2(x_2) \\ &= \int \left(\int f(x_1, x_2) d\mu_2(x_2) \right) d\mu_1(x_1). \end{aligned} \tag{3.63}$$

(ii) *If f is complex then*

$$\int |f(x_1, x_2)| d\mu_1(x_1) \in \mathcal{L}^1(X_2, d\mu_2) \tag{3.64}$$

respectively

$$\int |f(x_1, x_2)| d\mu_2(x_2) \in \mathcal{L}^1(X_1, d\mu_1) \tag{3.65}$$

if and only if $f \in \mathcal{L}^1(X_1 \times X_2, d\mu_1 \otimes d\mu_2)$. In this case (3.63) holds.

Proof. By Theorem 3.21 the claim holds for simple functions. Now (i) follows from the monotone convergence theorem and (ii) from the dominated convergence theorem. □

In particular, if $f(x_1, x_2)$ is either nonnegative or integrable, then the order of integration can be interchanged.

Lemma 3.23. *If μ_1 and μ_2 are σ-finite regular Borel measures with, so is $\mu_1 \otimes \mu_2$.*

Proof. Regularity holds for every rectangle and hence also for the algebra of finite disjoint unions of rectangles. Thus the claim follows from Lemma 3.5. □

Note that we can iterate this procedure.

Lemma 3.24. *Suppose μ_j, $j = 1, 2, 3$ are σ-finite measures. Then*

$$(\mu_1 \otimes \mu_2) \otimes \mu_3 = \mu_1 \otimes (\mu_2 \otimes \mu_3). \tag{3.66}$$

Proof. First of all note that $(\Sigma_1 \otimes \Sigma_2) \otimes \Sigma_3 = \Sigma_1 \otimes (\Sigma_2 \otimes \Sigma_3)$ is the sigma algebra generated by the cuboids $A_1 \times A_2 \times A_3$ in $X_1 \times X_2 \times X_3$. Moreover, since

$$\begin{aligned}((\mu_1 \otimes \mu_2) \otimes \mu_3)(A_1 \times A_2 \times A_3) &= \mu_1(A_1)\mu_2(A_2)\mu_3(A_3) \\ &= (\mu_1 \otimes (\mu_2 \otimes \mu_3))(A_1 \times A_2 \times A_3)\end{aligned} \tag{3.67}$$

the two measures coincide on the algebra of finite disjoint unions of cuboids. Hence they coincide everywhere by Theorem 3.4. □

Example. If λ is Lebesgue measure on $\mathbb{R}$, then $\lambda^n = \lambda \otimes \cdots \otimes \lambda$ is Lebesgue measure on $\mathbb{R}^n$. Since λ is regular, so is λ^n. ⋄

Chapter 4

The Lebesgue spaces L^p

4.1. Functions almost everywhere

We fix some measure space (X, Σ, μ) and define the L^p norm by

$$\|f\|_p = \left(\int_X |f|^p \, d\mu \right)^{1/p}, \qquad 1 \le p \tag{4.1}$$

and denote by $\mathcal{L}^p(X, d\mu)$ the set of all complex valued measurable functions for which $\|f\|_p$ is finite. First of all note that $\mathcal{L}^p(X, d\mu)$ is a linear space, since $|f+g|^p \le 2^p \max(|f|,|g|)^p \le 2^p \max(|f|^p, |g|^p) \le 2^p(|f|^p + |g|^p)$. Of course our hope is that $\mathcal{L}^p(X, d\mu)$ is a Banach space. However, there is a small technical problem (recall that a property is said to hold almost everywhere if the set where it fails to hold is contained in a set of measure zero):

Lemma 4.1. *Let f be measurable, then*

$$\int_X |f|^p \, d\mu = 0 \tag{4.2}$$

if and only if $f(x) = 0$ almost everywhere with respect to μ.

Proof. Observe that we have $A = \{x | f(x) \ne 0\} = \bigcup_n A_n$, where $A_n = \{x | \, |f(x)| > \frac{1}{n}\}$. If $\int |f|^p d\mu = 0$ we must have $\mu(A_n) = 0$ for every n and hence $\mu(A) = \lim_{n\to\infty} \mu(A_n) = 0$. The converse is obvious. □

Note that the proof also shows that if f is not 0 almost everywhere, there is an $\varepsilon > 0$ such that $\mu(\{x | \, |f(x)| \ge \varepsilon\}) > 0$.

Example. Let λ be the Lebesgue measure on $\mathbb{R}$. Then the characteristic function of the rationals $\chi_{\mathbb{Q}}$ is zero a.e. (with respect to λ). Let Θ be the Dirac measure centered at 0, then $f(x) = 0$ a.e. (with respect to Θ) if and only if $f(0) = 0$. ⋄

Thus $\|f\|_p = 0$ only implies $f(x) = 0$ for almost every x, but not for all! Hence $\|.\|_p$ is not a norm on $\mathcal{L}^p(X, d\mu)$. The way out of this misery is to identify functions which are equal almost everywhere: Let

$$\mathcal{N}(X, d\mu) = \{f | f(x) = 0 \ \mu\text{-almost everywhere}\}. \tag{4.3}$$

Then $\mathcal{N}(X, d\mu)$ is a linear subspace of $\mathcal{L}^p(X, d\mu)$ and we can consider the quotient space

$$L^p(X, d\mu) = \mathcal{L}^p(X, d\mu)/\mathcal{N}(X, d\mu). \tag{4.4}$$

If $d\mu$ is the Lebesgue measure on $X \subseteq \mathbb{R}^n$ we simply write $L^p(X)$. Observe that $\|f\|_p$ is well defined on $L^p(X, d\mu)$.

Even though the elements of $L^p(X, d\mu)$ are strictly speaking equivalence classes of functions, we will still call them functions for notational convenience. However, note that for $f \in L^p(X, d\mu)$ the value $f(x)$ is not well defined (unless there is a continuous representative and different continuous functions are in different equivalence classes, e.g., in the case of Lebesgue measure).

With this modification we are back in business since $L^p(X, d\mu)$ turns out to be a Banach space. We will show this in the following sections.

But before that let us also define $L^\infty(X, d\mu)$. It should be the set of bounded measurable functions $B(X)$ together with the sup norm. The only problem is that if we want to identify functions equal almost everywhere, the supremum is no longer independent of the equivalence class. The solution is the **essential supremum**

$$\|f\|_\infty = \inf\{C \,|\, \mu(\{x| \,|f(x)| > C\}) = 0\}. \tag{4.5}$$

That is, C is an essential bound if $|f(x)| \le C$ almost everywhere and the essential supremum is the infimum over all essential bounds.

Example. If λ is the Lebesgue measure, then the essential sup of $\chi_{\mathbb{Q}}$ with respect to λ is 0. If Θ is the Dirac measure centered at 0, then the essential sup of $\chi_{\mathbb{Q}}$ with respect to Θ is 1 (since $\chi_{\mathbb{Q}}(0) = 1$, and $x = 0$ is the only point which counts for Θ). ⋄

As before we set

$$L^\infty(X, d\mu) = B(X)/\mathcal{N}(X, d\mu) \tag{4.6}$$

and observe that $\|f\|_\infty$ is independent of the equivalence class.

If you wonder where the ∞ comes from, have a look at Problem 4.1.

Problem 4.1. *Suppose $\mu(X) < \infty$. Show that*

$$\lim_{p\to\infty} \|f\|_p = \|f\|_\infty \tag{4.7}$$

for any bounded measurable function.

4.2. Jensen ≤ Hölder ≤ Minkowski

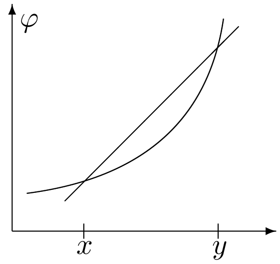
As a preparation for proving that L^p is a Banach space, we will need Hölder's inequality, which plays a central role in the theory of L^p spaces. In particular, it will imply Minkowski's inequality, which is just the triangle inequality for L^p. Our proof is based on Jensen's inequality and emphasizes the connection with convexity. In fact, the triangle inequality just states that a norm is convex:

$$\|\lambda f + (1-\lambda)g\| \le \lambda\|f\| + (1-\lambda)\|g\|, \quad \lambda \in (0,1). \tag{4.8}$$

Recall that a real function φ defined on an open interval $I = (a,b)$ is called **convex** if

$$\varphi((1-\lambda)x + \lambda y) \le (1-\lambda)\varphi(x) + \lambda\varphi(y), \quad \lambda \in (0,1) \tag{4.9}$$

that is, on (x,y) the graph of $\varphi(x)$ lies below or on the line connecting $(x,\varphi(x))$ and $(y,\varphi(y))$:



If the inequality is strict, then φ is called **strictly convex**. It is not hard to see (use $z = (1-\lambda)x + \lambda y$) that the definition implies

$$\frac{\varphi(z)-\varphi(x)}{z-x} \le \frac{\varphi(y)-\varphi(x)}{y-x} \le \frac{\varphi(y)-\varphi(z)}{y-z}, \quad x < z < y. \tag{4.10}$$

Lemma 4.2. *Let $\varphi : (a,b) \to \mathbb{R}$ be convex. Then*

(i) *φ is continuous.*

(ii) *The left/right derivatives $\varphi'_\pm(x) = \lim_{\varepsilon\downarrow 0} \frac{\varphi(x\pm\varepsilon)-\varphi(x)}{\pm\varepsilon}$ exist and are monotone nondecreasing. Moreover, φ' exists except at a countable number of points.*

(iii) *$\varphi(y) \ge \varphi(x) + \alpha(y-x)$ for any α with $\varphi'_-(x) \le \alpha \le \varphi'_+(x)$.*

Proof. Abbreviate $D(x,y) = \frac{\varphi(y)-\varphi(x)}{y-x}$ and observe that (4.10) implies

$$D(x,z) \le D(y,z) \qquad \text{for } x < y. \tag{4.11}$$

Hence $\varphi'_\pm(x)$ exist and we have $\varphi'_-(x) \le \varphi'_+(x) \le \varphi'_-(y) \le \varphi'_+(y)$ for $x < y$. So (ii) follows. Next

$$\varphi'_+(x) \le D(y,x) \le \varphi'_-(y) \tag{4.12}$$

shows $\varphi(y) \ge \varphi(x) + \varphi'_\pm(x)(y-x)$ if $\pm(y-x) > 0$ and proves (iii). Moreover, $|D(y,x)| \le \varphi'_-(z)$ for $x, y < z$ proves (i). □

Remark: It is not hard to see that $\varphi \in C^1$ is convex if and only if $\varphi'(x)$ is monotone nondecreasing (e.g., $\varphi'' \ge 0$ if $\varphi \in C^2$).

With these preparations out of the way we can show

Theorem 4.3 (Jensen's inequality)**.** *Let $\varphi : (a,b) \to \mathbb{R}$ be convex ($a = -\infty$ or $b = \infty$ being allowed). Suppose μ is a finite measure satisfying $\mu(X) = 1$ and $f \in \mathcal{L}^1(X, d\mu)$ with $a < f(x) < b$. Then the negative part of $\varphi \circ f$ is integrable and*

$$\varphi\Big(\int_X f\, d\mu\Big) \le \int_X (\varphi \circ f)\, d\mu. \tag{4.13}$$

For $\varphi \ge 0$ and $f \ge 0$ the requirement that f is integrable can be dropped if $\varphi(b)$ is understood as $\lim_{x\to b} \varphi(x)$ (note that φ is either always decreasing or eventually increasing).

Proof. By (iii) of the previous lemma we have

$$\varphi(f(x)) \ge \varphi(I) + \alpha(f(x) - I), \qquad I = \int_X f\, d\mu \in (a,b). \tag{4.14}$$

This shows that the negative part of $\varphi \circ f$ is integrable and integrating over X finishes the proof in the case $f \in \mathcal{L}^1$. If $f \ge 0$ the claim holds at least if X is replaced by $X_n = \{x \in X | f(x) \le n\}$. Since $X_n \nearrow X$ we can now let $n \to \infty$. □

Observe that if φ is strictly convex, then equality can only occur if f is constant.

Now we are ready to prove

Theorem 4.4 (Hölder's inequality)**.** *Let p and q be* **dual** *indices, that is,*

$$\frac{1}{p} + \frac{1}{q} = 1 \tag{4.15}$$

with $1 \le p \le \infty$. If $f \in L^p(X, d\mu)$ and $g \in L^q(X, d\mu)$ then $fg \in L^1(X, d\mu)$ and

$$\|f\, g\|_1 \le \|f\|_p \|g\|_q. \tag{4.16}$$

Proof. The case $p = 1, q = \infty$ (respectively $p = \infty, q = 1$) follows directly from the properties of the integral and hence it remains to consider $1 < p, q < \infty$.

First of all it is no restriction to assume $\|g\|_q = 1$. Let $A = \{x| \, |g(x)| > 0\}$, then (note $(1-q)p = -q$)

$$\|f\,g\|_1^p = \Big| \int_A |f|\,|g|^{1-q}|g|^q d\mu\Big|^p \le \int_A (|f|\,|g|^{1-q})^p |g|^q d\mu = \int_A |f|^p d\mu \le \|f\|_p^p, \tag{4.17}$$

where we have used Jensen's inequality with $\varphi(x) = |x|^p$ applied to the function $h = |f|\,|g|^{1-q}$ and measure $d\nu = |g|^q d\mu$ (note $\nu(X) = \int |g|^q d\mu = \|g\|_q^q = 1$). □

As a consequence we also get

Theorem 4.5 (Minkowski's inequality)**.** *Let $f, g \in L^p(X, d\mu)$, then*

$$\|f + g\|_p \le \|f\|_p + \|g\|_p. \tag{4.18}$$

Proof. Since the cases $p = 1, \infty$ are straightforward, we only consider $1 < p < \infty$. Using $|f+g|^p \le |f|\,|f+g|^{p-1} + |g|\,|f+g|^{p-1}$ we obtain from Hölder's inequality (note $(p-1)q = p$)

$$\begin{aligned} \|f + g\|_p^p &\le \|f\|_p \|(f+g)^{p-1}\|_q + \|g\|_p \|(f+g)^{p-1}\|_q \\ &= (\|f\|_p + \|g\|_p)\|(f+g)\|_p^{p-1}. \end{aligned} \tag{4.19}$$

□

This shows that $L^p(X, d\mu)$ is a normed linear space.

Problem 4.2. *Prove*

$$\prod_{k=1}^n x_k^{\alpha_k} \le \sum_{k=1}^n \alpha_k x_k, \qquad \textit{if} \quad \sum_{k=1}^n \alpha_k = 1, \tag{4.20}$$

for $\alpha_k > 0$, $x_k > 0$. (Hint: Take a sum of Dirac-measures and use that the exponential function is convex.)

Problem 4.3. *Show the following generalization of Hölder's inequality*

$$\|f\,g\|_r \le \|f\|_p \|g\|_q, \tag{4.21}$$

where $\frac{1}{p} + \frac{1}{q} = \frac{1}{r}$.

Problem 4.4 (Lyapunov inequality)**.** *Let $0 < \theta < 1$. Show that if $f \in L^{p_1} \cap L^{p_2}$, then $f \in L^p$ and*

$$\|f\|_p \le \|f\|_{p_1}^{\theta} \|f\|_{p_2}^{1-\theta}, \tag{4.22}$$

where $\frac{1}{p} = \frac{\theta}{p_1} + \frac{1-\theta}{p_2}$.

4.3. Nothing missing in L^p

Finally it remains to show that $L^p(X, d\mu)$ is complete.

Theorem 4.6. *The space $L^p(X, d\mu)$ is a Banach space.*

Proof. Suppose f_n is a Cauchy sequence. It suffices to show that some subsequence converges (show this). Hence we can drop some terms such that

$$\|f_{n+1} - f_n\|_p \leq \frac{1}{2^n}. \tag{4.23}$$

Now consider $g_n = f_n - f_{n-1}$ (set $f_0 = 0$). Then

$$G(x) = \sum_{k=1}^{\infty} |g_k(x)| \tag{4.24}$$

is in L^p. This follows from

$$\Big\| \sum_{k=1}^{n} |g_k| \Big\|_p \leq \sum_{k=1}^{n} \|g_k(x)\|_p \leq \|f_1\|_p + \frac{1}{2} \tag{4.25}$$

using the monotone convergence theorem. In particular, $G(x) < \infty$ almost everywhere and the sum

$$\sum_{n=1}^{\infty} g_n(x) = \lim_{n\to\infty} f_n(x) \tag{4.26}$$

is absolutely convergent for those x. Now let $f(x)$ be this limit. Since $|f(x) - f_n(x)|^p$ converges to zero almost everywhere and $|f(x) - f_n(x)|^p \leq 2^p G(x)^p \in L^1$, dominated convergence shows $\|f - f_n\|_p \to 0$. □

In particular, in the proof of the last theorem we have seen:

Corollary 4.7. *If $\|f_n - f\|_p \to 0$ then there is a subsequence which converges pointwise almost everywhere.*

It even turns out that L^p is separable.

Lemma 4.8. *Suppose X is a second countable topological space (i.e., it has a countable basis) and μ is a regular Borel measure. Then $L^p(X, d\mu)$, $1 \leq p < \infty$ is separable.*

Proof. The set of all characteristic functions $\chi_A(x)$ with $A \in \Sigma$ and $\mu(A) < \infty$, is total by construction of the integral. Now our strategy is as follows: Using outer regularity we can restrict A to open sets and using the existence of a countable base, we can restrict A to open sets from this base.

Fix A. By outer regularity, there is a decreasing sequence of open sets O_n such that $\mu(O_n) \to \mu(A)$. Since $\mu(A) < \infty$ it is no restriction to assume $\mu(O_n) < \infty$, and thus $\mu(O_n \backslash A) = \mu(O_n) - \mu(A) \to 0$. Now dominated convergence implies $\|\chi_A - \chi_{O_n}\|_p \to 0$. Thus the set of all characteristic functions $\chi_O(x)$ with O open and $\mu(O) < \infty$, is total. Finally let $\mathcal{B}$ be a countable basis for the topology. Then, every open set O can be written as $O = \bigcup_{j=1}^{\infty} \tilde{O}_j$ with $\tilde{O}_j \in \mathcal{B}$. Moreover, by considering the set of all finite unions of elements from $\mathcal{B}$ it is no restriction to assume $\bigcup_{j=1}^{n} \tilde{O}_j \in \mathcal{B}$. Hence there is an increasing sequence $\tilde{O}_n \nearrow O$ with $\tilde{O}_n \in \mathcal{B}$. By monotone convergence, $\|\chi_O - \chi_{\tilde{O}_n}\|_p \to 0$ and hence the set of all characteristic functions $\chi_{\tilde{O}}$ with $\tilde{O} \in \mathcal{B}$ is total. □

To finish this chapter, let us show that continuous functions are dense in L^p.

Theorem 4.9. *Let X be a locally compact metric space and let μ be a σ-finite regular Borel measure. Then the set $C_c(X)$ of continuous functions with compact support is dense in $L^p(X, d\mu)$, $1 \le p < \infty$.*

Proof. As in the previous proof the set of all characteristic functions $\chi_K(x)$ with K compact is total (using inner regularity). Hence it suffices to show that $\chi_K(x)$ can be approximated by continuous functions. By outer regularity there is an open set $O \supset K$ such that $\mu(O \backslash K) \le \varepsilon$. By Urysohn's lemma (Lemma 1.11) there is a continuous function f_ε which is one on K and 0 outside O. Since

$$\int_X |\chi_K - f_\varepsilon|^p d\mu = \int_{O \backslash K} |f_\varepsilon|^p d\mu \le \mu(O \backslash K) \le \varepsilon \tag{4.27}$$

we have $\|f_\varepsilon - \chi_K\| \to 0$ and we are done. □

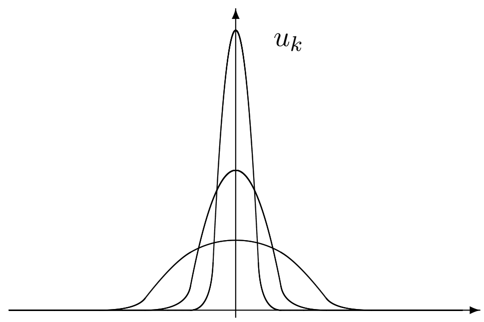
If X is some subset of $\mathbb{R}^n$ we can do even better.

A nonnegative function $u \in C_c^\infty(\mathbb{R}^n)$ is called a **mollifier** if

$$\int_{\mathbb{R}^n} u(x) dx = 1 \tag{4.28}$$

The standard mollifier is $u(x) = \exp(\frac{1}{|x|^2 - 1})$ for $|x| < 1$ and $u(x) = 0$ else.

If we scale a mollifier according to $u_k(x) = k^n u(k\,x)$ such that its mass is preserved ($\|u_k\|_1 = 1$) and it concentrates more and more around the origin



we have the following result (Problem 4.6):

Lemma 4.10. *Let u be a mollifier in $\mathbb{R}^n$ and set $u_k(x) = k^n u(k\,x)$. Then for any (uniformly) continuous function $f : \mathbb{R}^n \to \mathbb{C}$ we have that*

$$f_k(x) = \int_{\mathbb{R}^n} u_k(x-y) f(y) dy \tag{4.29}$$

is in $C^\infty(\mathbb{R}^n)$ and converges to f (uniformly).

Now we are ready to prove

Theorem 4.11. *If $X \subseteq \mathbb{R}^n$ and μ is a Borel measure, then the set $C_c^\infty(X)$ of all smooth functions with compact support is dense in $L^p(X, d\mu)$, $1 \le p < \infty$.*

Proof. By our previous result it suffices to show that any continuous function $f(x)$ with compact support can be approximated by smooth ones. By setting $f(x) = 0$ for $x \notin X$, it is no restriction to assume $X = \mathbb{R}^n$. Now choose a mollifier u and observe that f_k has compact support (since f has). Moreover, since f has compact support it is uniformly continuous and $f_k \to f$ uniformly. But this implies $f_k \to f$ in L^p. □

Problem 4.5. *Let μ be a finite measure. Show that*

$$d(A, B) = \mu(A \Delta B), \qquad A\Delta B = (A \cup B) \backslash (A \cap B) \tag{4.30}$$

is a metric on Σ if we identify sets of measure zero. Show that if $\mathcal{A}$ is an algebra, then it is dense in $\Sigma(\mathcal{A})$. (Hint: Show that the sets which can be approximated by sets in $\mathcal{A}$ form a monotone class.)

Problem 4.6. *Prove Lemma 4.10. (Hint: To show that f_k is smooth use Problem 3.7 and 3.8.)*

Chapter 5

The main theorems about Banach spaces

5.1. The Baire theorem and its consequences

Recall that the interior of a set is the largest open subset (that is, the union of all open subsets). A set is called **nowhere dense** if its closure has empty interior. The key to several important theorems about Banach spaces is the observation that a Banach space cannot be the countable union of nowhere dense sets.

Theorem 5.1 (Baire category theorem)**.** *Let X be a complete metric space, then X cannot be the countable union of nowhere dense sets.*

Proof. Suppose $X = \bigcup_{n=1}^{\infty} X_n$. We can assume that the sets X_n are closed and none of them contains a ball, that is, $X \backslash X_n$ is open and nonempty for every n. We will construct a Cauchy sequence x_n which stays away from all X_n.

Since $X \backslash X_1$ is open and nonempty there is a closed ball $B_{r_1}(x_1) \subseteq X \backslash X_1$. Reducing r_1 a little, we can even assume $\overline{B_{r_1}(x_1)} \subseteq X \backslash X_1$. Moreover, since X_2 cannot contain $B_{r_1}(x_1)$ there is some $x_2 \in B_{r_1}(x_1)$ that is not in X_2. Since $B_{r_1}(x_1) \cap (X \backslash X_2)$ is open there is a closed ball $\overline{B_{r_2}(x_2)} \subseteq B_{r_1}(x_1) \cap (X \backslash X_2)$. Proceeding by induction we obtain a sequence of balls such that

$$\overline{B_{r_n}(x_n)} \subseteq B_{r_{n-1}}(x_{n-1}) \cap (X \backslash X_n). \tag{5.1}$$

Now observe that in every step we can choose r_n as small as we please, hence without loss of generality $r_n \to 0$. Since by construction $x_n \in \overline{B_{r_N}(x_N)}$ for $n \geq N$, we conclude that x_n is Cauchy and converges to some point $x \in X$.

But $x \in \overline{B_{r_n}(x_n)} \subseteq X \backslash X_n$ for every n, contradicting our assumption that the X_n cover X. □

(Sets which can be written as countable union of nowhere dense sets are called of first category. All other sets are second category. Hence the name category theorem.)

In other words, if $X_n \subseteq X$ is a sequence of closed subsets which cover X, at least one X_n contains a ball of radius $\varepsilon > 0$.

There is a reformulation which is also worthwhile noting:

Corollary 5.2. *Let X be a complete metric space, then any countable intersection of open dense sets is again dense.*

Proof. Let O_n be open dense sets whose intersection is not dense. Then this intersection must be missing some ball B_ε. The closure of this ball will lie in $\bigcup_n X_n$, where $X_n = X \backslash O_n$ are closed and nowhere dense. But $\overline{B_\varepsilon}$ is a complete metric space, a contradiction. □

Now we come to the first important consequence, the **uniform boundedness principle**.

Theorem 5.3 (Banach-Steinhaus)**.** *Let X be a Banach space and Y some normed linear space. Let $\{A_\alpha\} \subseteq \mathfrak{L}(X,Y)$ be a family of bounded operators. Suppose $\|A_\alpha x\| \le C(x)$ is bounded for every fixed $x \in X$, then $\|A_\alpha\| \le C$ is uniformly bounded.*

Proof. Let

$$X_n = \{x| \, \|A_\alpha x\| \le n \text{ for all } \alpha\} = \bigcap_\alpha \{x| \, \|A_\alpha x\| \le n\}, \qquad (5.2)$$

then $\bigcup_n X_n = X$ by assumption. Moreover, by continuity of A_α and the norm, each X_n is an intersection of closed sets and hence closed. By Baire's theorem at least one contains a ball of positive radius: $B_\varepsilon(x_0) \subset X_n$. Now observe

$$\|A_\alpha y\| \le \|A_\alpha(y + x_0)\| + \|A_\alpha x_0\| \le n + \|A_\alpha x_0\| \qquad (5.3)$$

for $\|y\| < \varepsilon$. Setting $y = \varepsilon \frac{x}{\|x\|}$ we obtain

$$\|A_\alpha x\| \le \frac{n + C(x_0)}{\varepsilon} \|x\| \qquad (5.4)$$

for any x. □

The next application is

Theorem 5.4 (open mapping)**.** *Let $A \in \mathfrak{L}(X,Y)$ be a bounded linear operator from one Banach space onto another. Then A is open (i.e., maps open sets to open sets).*

Proof. Denote by $B_r^X(x) \subseteq X$ the open ball with radius r centered at x and let $B_r^X = B_r^X(0)$. Similarly for $B_r^Y(y)$. By scaling and translating balls (using linearity of A), it suffices to prove $B_\varepsilon^Y \subseteq A(B_1^X)$ for some $\varepsilon > 0$. Since A is surjective we have

$$Y = \bigcup_{n=1}^{\infty} A(B_n^X) \tag{5.5}$$

and the Baire theorem implies that for some n, $\overline{A(B_n^X)}$ contains a ball $B_\varepsilon^Y(y)$. Without restriction $n = 1$ (just scale the balls). Since $-\overline{A(B_1^X)} = \overline{A(-B_1^X)} = \overline{A(B_1^X)}$ we see $B_\varepsilon^Y(-y) \subseteq \overline{A(B_1^X)}$ and by convexity of $\overline{A(B_1^X)}$ we also have $B_\varepsilon^Y \subseteq \overline{A(B_1^X)}$.

So we have $B_\varepsilon^Y \subseteq \overline{A(B_1^X)}$, but we would need $B_\varepsilon^Y \subseteq A(B_1^X)$. To complete the proof we will show $\overline{A(B_1^X)} \subseteq A(B_2^X)$ which implies $B_{\varepsilon/2}^Y \subseteq A(B_1^X)$.

For every $y \in \overline{A(B_1^X)}$ we can choose some sequence $y_n \in A(B_1^X)$ with $y_n \to y$. Moreover, there even is some $x_n \in B_1^X$ with $y_n = A(x_n)$. However x_n might not converge, so we need to argue more carefully and ensure convergence along the way: start with $x_1 \in B_1^X$ such that $y - Ax_1 \in B_{\varepsilon/2}^Y$. Scaling the relation $B_\varepsilon^Y \subset \overline{A(B_1^X)}$ we have $B_{\varepsilon/2}^Y \subset \overline{A(B_{1/2}^X)}$ and hence we can choose $x_2 \in B_{1/2}^X$ such that $(y - Ax_1) - Ax_2 \in B_{\varepsilon/4}^Y \subset \overline{A(B_{1/4}^X)}$. Proceeding like this we obtain a sequence of points $x_n \in B_{2^{1-n}}^X$ such that

$$y - \sum_{k=1}^{n} Ax_k \in B_{\varepsilon 2^{-n}}^Y. \tag{5.6}$$

By

$$\|\sum_{k=1}^{n} x_k\| \le \sum_{k=1}^{n} \frac{2}{2^n} \tag{5.7}$$

the limit $x = \sum_{k=1}^{\infty} x_k$ exists and satisfies $\|x\| \le 2$. Hence $y = Ax \in A(B_2^X)$ as desired. □

Remark: The requirement that A is onto is crucial. In fact, the converse is also true: If A is open, then the image of the unit ball contains again some ball $B_\varepsilon^Y \subseteq A(B_1^X)$. Hence by scaling $B_{r\varepsilon}^Y \subseteq A(B_r^X)$ and letting $r \to \infty$ we see that A is onto: $Y = A(X)$.

As an immediate consequence we get the inverse mapping theorem:

Theorem 5.5 (inverse mapping)**.** *Let $A \in \mathfrak{L}(X,Y)$ be a bounded linear bijection between Banach spaces. Then A^{-1} is continuous.*

Another important consequence is the closed graph theorem. The **graph** of an operator A is just

$$\Gamma(A) = \{(x, Ax) | x \in \mathfrak{D}(A)\}. \tag{5.8}$$

If A is linear, the graph is a subspace of the Banach space $X \oplus Y$ (provided X and Y are Banach spaces), which is just the cartesian product together with the norm

$$\|(x, y)\|_{X\oplus Y} = \|x\|_X + \|y\|_Y \tag{5.9}$$

(check this). Note that $(x_n, y_n) \to (x, y)$ if and only if $x_n \to x$ and $y_n \to y$.

Theorem 5.6 (closed graph)**.** *Let $A : X \to Y$ be a linear map from a Banach space X to another Banach space Y. Then A is bounded if and only its graph is closed.*

Proof. If $\Gamma(A)$ is closed, then it is again a Banach space. Now the projection $\pi_1(x, Ax) = x$ onto the first component is a continuous bijection onto X. So by the inverse mapping theorem its inverse π_1^{-1} is again continuous, and so is $A = \pi_2 \circ \pi_1^{-1}$, where $\pi_2(x, Ax) = Ax$ is the projection onto the second component. The converse is easy. □

Remark: The crucial fact here is that A is defined on all of X!

Operators whose graphs are closed are called **closed operators**. Being closed is the next option you have once an operator turns out to be unbounded. If A is closed, then $x_n \to x$ does not guarantee you that Ax_n converges (like continuity would), but it at least guarantees that if Ax_n converges, it converges to the right thing, namely Ax:

- A bounded: $x_n \to x$ implies $Ax_n \to Ax$.
- A closed: $x_n \to x$ and $Ax_n \to y$ implies $y = Ax$.

If an operator is not closed, you can try to take the closure of its graph, to obtain a closed operator. If A is bounded this always works (which is just the contents of Theorem 1.25). However, in general, the closure of the graph might not be the graph of an operator (we might pick up points $(x, y_{1,2}) \in \overline{\Gamma(A)}$ with $y_1 \neq y_2$). If this works, and $\overline{\Gamma(A)}$ is the graph of some operator $\overline{A}$, then A is called **closable** and $\overline{A}$ is called the **closure** of A.

The closed graph theorem tells us that closed linear operators can be defined on all of X if and only if they are bounded. So if we have an unbounded operator we cannot have both! That is, if we want our operator to be at least closed, we have to live with domains. This is the reason why in quantum mechanics most operators are defined on domains. In fact, there is another important property which does not allow unbounded operators to be defined on the entire space:

Theorem 5.7 (Helling-Toeplitz). *Let $A : \mathfrak{H} \to \mathfrak{H}$ be a linear operator on some Hilbert space $\mathfrak{H}$. If A is symmetric, that is $\langle g, Af\rangle = \langle Ag, f\rangle$, $f, g \in \mathfrak{H}$, then A is bounded.*

Proof. It suffices to prove that A is closed. In fact, $f_n \to f$ and $Af_n \to g$ implies

$$\langle h, g\rangle = \lim_{n\to\infty} \langle h, Af_n\rangle = \lim_{n\to\infty} \langle Ah, f_n\rangle = \langle Ah, f\rangle = \langle h, Af\rangle \tag{5.10}$$

for any $h \in \mathfrak{H}$. Hence $Af = g$. □

Problem 5.1. *Show that the differential operator $A = \frac{d}{dx}$ defined on $\mathfrak{D}(A) = C^1[0,1] \subset C[0,1]$ (sup norm) is a closed operator. (Compare also Problem 1.10.)*

5.2. The Hahn-Banach theorem and its consequences

Let X be a Banach space. Recall that we have called the set of all bounded linear functionals the dual space X^* (which is again a Banach space by Theorem 1.23).

Example. Consider the Banach space $\ell^p(\mathbb{N})$ of all sequences $x = (x_j)_{j=1}^\infty$ for which the norm

$$\|x\|_p = \left(\sum_{n\in\mathbb{N}} |x_n|^p\right)^{1/p} \tag{5.11}$$

is finite. Then, by Hölder's inequality, every $y \in \ell^q(\mathbb{N})$ gives rise to a bounded linear functional

$$l_y(x) = \sum_{n\in\mathbb{N}} y_n x_n \tag{5.12}$$

whose norm is $\|l_y\| = \|y\|_q$ (Problem 5.2). But can every element of $\ell^p(\mathbb{N})^*$ be written in this form?

Suppose $p = 1$ and choose $l \in \ell^1(\mathbb{N})^*$. Now define

$$y_n = l(\delta^n), \tag{5.13}$$

where $\delta^n_n = 1$ and $\delta^n_m = 0$, $n \neq m$. Then

$$|y_n| = |l(\delta^n)| \le \|l\| \, \|\delta^n\|_1 = \|l\| \tag{5.14}$$

shows $\|y\|_\infty \le \|l\|$, that is, $y \in \ell^\infty(\mathbb{N})$. By construction $l(x) = l_y(x)$ for every $x \in \operatorname{span}\{\delta^n\}$. By continuity of ℓ it even holds for $x \in \overline{\operatorname{span}\{\delta^n\}} = \ell^1(\mathbb{N})$. Hence the map $y \mapsto l_y$ is an isomorphism, that is, $\ell^1(\mathbb{N})^* \cong \ell^\infty(\mathbb{N})$. A similar argument shows $\ell^p(\mathbb{N})^* \cong \ell^q(\mathbb{N})$, $1 \le p < \infty$ (Problem 5.3). One usually identifies $\ell^p(\mathbb{N})^*$ with $\ell^q(\mathbb{N})$ using this canonical isomorphism and simply writes $\ell^p(\mathbb{N})^* = \ell^q(\mathbb{N})$. In the case $p = \infty$ this is not true, as we will see soon. ◇

It turns out that many questions are easier to handle after applying a linear functional $\ell \in X^*$. For example, suppose $x(t)$ is a function $\mathbb{R} \to X$ (or $\mathbb{C} \to X$), then $\ell(x(t))$ is a function $\mathbb{R} \to \mathbb{C}$ (respectively $\mathbb{C} \to \mathbb{C}$) for any $\ell \in X^*$. So to investigate $\ell(x(t))$ we have all tools from real/complex analysis at our disposal. But how do we translate this information back to $x(t)$? Suppose we have $\ell(x(t)) = \ell(y(t))$ for all $\ell \in X^*$. Can we conclude $x(t) = y(t)$? The answer is yes and will follow from the Hahn-Banach theorem.

We first prove the real version from which the complex one then follows easily.

Theorem 5.8 (Hahn-Banach, real version)**.** *Let X be a real vector space and $\varphi : X \to \mathbb{R}$ a convex function (i.e., $\varphi(\lambda x + (1-\lambda)y) \le \lambda\varphi(x) + (1-\lambda)\varphi(y)$ for $\lambda \in (0,1)$).*

If ℓ is a linear functional defined on some subspace $Y \subset X$ which satisfies $\ell(y) \le \varphi(y)$, $y \in Y$, then there is an extension $\overline{\ell}$ to all of X satisfying $\overline{\ell}(x) \le \varphi(x)$, $x \in X$.

Proof. Let us first try to extend ℓ in just one direction: Take $x \notin Y$ and set $\tilde{Y} = \operatorname{span}\{x, Y\}$. If there is an extension $\tilde{\ell}$ to $\tilde{Y}$ it must clearly satisfy

$$\tilde{\ell}(y + \alpha x) = \ell(y) + \alpha\tilde{\ell}(x). \tag{5.15}$$

So all we need to do is to choose $\tilde{\ell}(x)$ such that $\tilde{\ell}(y + \alpha x) \le \varphi(y + \alpha x)$. But this is equivalent to

$$\sup_{\alpha>0, y\in Y} \frac{\varphi(y - \alpha x) - \ell(y)}{-\alpha} \le \tilde{\ell}(x) \le \inf_{\alpha>0, y\in Y} \frac{\varphi(y + \alpha x) - \ell(y)}{\alpha} \tag{5.16}$$

and is hence only possible if

$$\frac{\varphi(y_1 - \alpha_1 x) - \ell(y_1)}{-\alpha_1} \le \frac{\varphi(y_2 + \alpha_2 x) - \ell(y_2)}{\alpha_2} \tag{5.17}$$

for any $\alpha_1, \alpha_2 > 0$ and $y_1, y_2 \in Y$. Rearranging this last equations we see that we need to show

$$\alpha_2\ell(y_1) + \alpha_1\ell(y_2) \le \alpha_2\varphi(y_1 - \alpha_1 x) + \alpha_1\varphi(y_2 + \alpha_2 x). \tag{5.18}$$

Starting with the left hand side we have

$$\begin{aligned} \alpha_2\ell(y_1) + \alpha_1\ell(y_2) &= (\alpha_1+\alpha_2)\ell\left(\lambda(y_1 - \alpha_1 x) + (1-\lambda)(y_2 + \alpha_2 x)\right) \\ &\le (\alpha_1+\alpha_2)\varphi\left(\lambda(y_1 - \alpha_1 x) + (1-\lambda)(y_2 + \alpha_2 x)\right) \\ &\le \alpha_2\varphi(y_1 - \alpha_1 x) + \alpha_1\varphi(y_2 + \alpha_2 x), \end{aligned} \tag{5.19}$$

where $\lambda = \frac{\alpha_2}{\alpha_1+\alpha_2}$. Hence one dimension works.

To finish the proof we appeal to Zorn's lemma: Let E be the collection of all extensions $\tilde{\ell}$ satisfying $\tilde{\ell}(x) \le \varphi(x)$. Then E can be partially ordered by inclusion (with respect to the domain) and every linear chain has an upper bound (defined on the union of all domains). Hence there is a maximal element $\overline{\ell}$ by Zorn's lemma. This element is defined on X, since if it were not, we could extend it as before contradicting maximality. □

Theorem 5.9 (Hahn-Banach, complex version)**.** *Let X be a complex vector space and $\varphi : X \to \mathbb{R}$ a convex function satisfying $\varphi(\alpha x) \le \varphi(x)$ if $|\alpha| = 1$.*

If ℓ is a linear functional defined on some subspace $Y \subset X$ which satisfies $|\ell(y)| \le \varphi(y)$, $y \in Y$, then there is an extension $\overline{\ell}$ to all of X satisfying $|\overline{\ell}(x)| \le \varphi(x)$, $x \in X$.

Proof. Set $\ell_r = \mathrm{Re}(\ell)$ and observe

$$\ell(x) = \ell_r(x) - \mathrm{i}\ell_r(\mathrm{i}x). \tag{5.20}$$

By our previous theorem, there is a real linear extension $\overline{\ell}_r$ satisfying $\overline{\ell}_r(x) \le \varphi(x)$. Now set $\overline{\ell}(x) = \overline{\ell}_r(x) - \mathrm{i}\overline{\ell}_r(\mathrm{i}x)$. Then $\overline{\ell}(x)$ is real linear and by $\overline{\ell}(\mathrm{i}x) = \overline{\ell}_r(\mathrm{i}x) + \mathrm{i}\overline{\ell}_r(x) = \mathrm{i}\overline{\ell}(x)$ also complex linear. To show $|\overline{\ell}(x)| \le \varphi(x)$ we abbreviate $\alpha = \frac{\overline{\ell}(x)^*}{|\overline{\ell}(x)|}$ and use

$$|\overline{\ell}(x)| = \alpha\overline{\ell}(x) = \overline{\ell}(\alpha x) = \overline{\ell}_r(\alpha x) \le \varphi(\alpha x) \le \varphi(x), \tag{5.21}$$

which finishes the proof. □

With the choice $\varphi(x) = c\|x\|$ we obtain:

Corollary 5.10. *Let X be a Banach space and let ℓ be a bounded linear functional defined on some subspace $Y \subseteq X$. Then there is an extension $\overline{\ell} \in X^*$ preserving the norm.*

Moreover, we can now easily prove our anticipated result

Corollary 5.11. *Suppose $\ell(x) = 0$ for all ℓ in some total subset $Y \subseteq X^*$. Then $x = 0$.*

Proof. Clearly if $\ell(x) = 0$ holds for all ℓ in some total subset, this holds for all $\ell \in X^*$. If $x \ne 0$ we can construct a bounded linear functional by setting $\ell(x) = 1$ and extending it to X^* using the previous corollary. But this contradicts our assumption. □

Example. Let us return to our example $\ell^\infty(\mathbb{N})$. Let $c(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ be the subspace of convergent sequences. Set

$$l(x) = \lim_{n\to\infty} x_n, \qquad x \in c(\mathbb{N}), \tag{5.22}$$

then l is bounded since

$$|l(x)| = \lim_{n\to\infty} |x_n| \le \|x\|_\infty. \tag{5.23}$$

Hence we can extend it to $\ell^\infty(\mathbb{N})$ by Corollary 5.10. Then $l(x)$ cannot be written as $l(x) = l_y(x)$ for some $y \in \ell^1(\mathbb{N})$ (as in (5.12)) since $y_n = l(\delta^n) = 0$ shows $y = 0$ and hence $\ell_y = 0$. The problem is that $\overline{\operatorname{span}\{\delta^n\}} = c_0(\mathbb{N}) \ne \ell^\infty(\mathbb{N})$, where $c_0(\mathbb{N})$ is the subspace of sequences converging to 0.

Moreover, there is also no other way to identify $\ell^\infty(\mathbb{N})^*$ with $\ell^1(\mathbb{N})$, since $\ell^1(\mathbb{N})$ is separable whereas $\ell^\infty(\mathbb{N})$ is not. This will follow from Lemma 5.15 (iii) below. ⋄

Another useful consequence is

Corollary 5.12. *Let $Y \subseteq X$ be a subspace of a normed linear space and let $x_0 \in X\backslash\overline{Y}$. Then there exists an $\ell \in X^*$ such that (i) $\ell(y) = 0$, $y \in Y$, (ii) $\ell(x_0) = \operatorname{dist}(x_0, Y)$, and (iii) $\|\ell\| = 1$.*

Proof. Replacing Y by $\overline{Y}$ we see that it is no restriction to assume that Y is closed. (Note that $x_0 \in X\backslash\overline{Y}$ if and only if $\operatorname{dist}(x_0, Y) > 0$.) Let $\tilde{Y} = \operatorname{span}\{x_0, Y\}$ and define

$$\ell(y + \alpha x_0) = \alpha \operatorname{dist}(x_0, Y). \tag{5.24}$$

By construction ℓ is linear on $\tilde{Y}$ and satisfies (i) and (ii). Moreover, by $\operatorname{dist}(x_0, Y) \le \|x_0 - \frac{-y}{\alpha}\|$ for any $y \in Y$ we have

$$|\ell(y + \alpha x_0)| = |\alpha| \operatorname{dist}(x_0, Y) \le \|y + \alpha x_0\|, \quad y \in Y. \tag{5.25}$$

Hence $\|\ell\| \le 1$ and there is an extension to X^* by Corollary 5.10. To see that the norm is in fact equal to one, take a sequence $y_n \in Y$ such that $\operatorname{dist}(x_0, Y) \ge (1 - \frac{1}{n})\|x_0 + y_n\|$. Then

$$|\ell(y_n + x_0)| = \operatorname{dist}(x_0, Y) \ge (1 - \frac{1}{n})\|y_n + x_0\| \tag{5.26}$$

establishing (iii). □

A straightforward consequence of the last corollary is also worthwhile noting:

Corollary 5.13. *Let $Y \subseteq X$ be a subspace of a normed linear space. Then $\ell(x) = 0$ for every $\ell \in X^*$ which vanishes on Y if and only if $x \in \overline{Y}$.*

If we take the **bidual** (or **double dual**) X^{**}, then the Hahn-Banach theorem tells us, that X can be identified with a subspace of X^{**}. In fact, consider the linear map $J : X \to X^{**}$ defined by $J(x)(\ell) = \ell(x)$ (i.e., $J(x)$ is evaluation at x). Then

Theorem 5.14. *Let X be a Banach space, then $J : X \to X^{**}$ is isometric (norm preserving).*

Proof. Fix $x_0 \in X$. By $|J(x_0)(\ell)| = |\ell(x_0)| \le \|\ell\|_* \|x_0\|$ we have at least $\|J(x_0)\|_{**} \le \|x_0\|$. Next, by Hahn-Banach there is a linear functional ℓ_0 with norm $\|\ell_0\|_* = 1$ such that $\ell_0(x_0) = \|x_0\|$. Hence $|J(x_0)(\ell_0)| = |\ell_0(x_0)| = \|x_0\|$ shows $\|J(x_0)\|_{**} = \|x_0\|$. □

Thus $J : X \to X^{**}$ is an isometric embedding. In many cases we even have $J(X) = X^{**}$ and X is called **reflexive** in this case.

Example. The Banach spaces $\ell^p(\mathbb{N})$ with $1 < p < \infty$ are reflexive: If we identify $\ell^p(\mathbb{N})^*$ with $\ell^q(\mathbb{N})$ and choose $z \in \ell^p(\mathbb{N})^{**}$, then

$$z(y) = \sum_{n \in \mathbb{N}} y_n x_n = y(x) \tag{5.27}$$

for some $x \in \ell^p(\mathbb{N})$, that is, $z = J(x)$. (Warning: It does not suffice to just argue $\ell^p(\mathbb{N})^{**} \cong \ell^q(\mathbb{N})^* \cong \ell^p(\mathbb{N})$.)

However, ℓ^1 is not since $\ell^1(\mathbb{N})^* \cong \ell^\infty(\mathbb{N})$ but $\ell^\infty(\mathbb{N})^* \not\cong \ell^1(\mathbb{N})$ as noted earlier. ⋄

Example. By the same argument (using the Riesz lemma), every Hilbert space is reflexive. ⋄

Lemma 5.15. *Let X be a Banach space.*

(i) *If X is reflexive, so is every closed subspace.*
(ii) *X is reflexive if and only if X^* is.*
(iii) *If X^* is separable, so is X.*

Proof. (i) Let Y be a closed subspace, then

$$\begin{array}{ccc} X & \xrightarrow{J_X} & X^{**} \\ j \uparrow & & \uparrow j_{**} \\ Y & \xrightarrow[J_Y]{} & Y^{**} \end{array}$$

commutes, where j is the natural inclusion and $j_{**}(y'')$, $y'' \in Y^{**}$, is defined via $(j_{**}(y''))(\ell) = y''(\ell|_Y)$. In fact, we have $j_{**}(J_Y(y))(\ell) = J_Y(y)(\ell|_Y) = \ell(y) = J_X(y)(\ell)$. Moreover, since J_X is surjective, for every $y'' \in Y^{**}$ there is an $x \in X$ such that $j_{**}(y'') = J_X(x)$. Since $j_{**}(y'')(\ell) = y''(\ell|_Y)$ vanishes on all $\ell \in X^*$ which vanish on Y, so does $\ell(x) = J_X(x)(\ell) = j_{**}(y'')(\ell)$ and thus $x \in Y$ by Corollary 5.13. That is, $j_{**}(Y^{**}) = J_X(Y)$ and $J_Y = j \circ J_X \circ j_{**}^{-1}$ is surjective.

(ii) Suppose X is reflexive, then the two maps

$$(J_X)_*: \begin{array}{ccc} X^* & \to & X^{***} \\ x' & \mapsto & x' \circ J_X^{-1} \end{array} \qquad (J_X)^*: \begin{array}{ccc} X^{***} & \to & X^* \\ x''' & \mapsto & x''' \circ J_X \end{array}$$

are inverse of each other. Moreover, fix $x'' \in X^{**}$ and let $x = J_X^{-1}(x'')$. Then $J_{X^*}(x')(x'') = x''(x') = J(x)(x') = x'(x) = x'(J_X^{-1}(x''))$, that is $J_{X^*} = (J_X)_*$ respectively $(J_{X^*})^{-1} = (J_X)^*$, which shows X^* reflexive if X reflexive. To see the converse, observe that X^* reflexive implies X^{**} reflexive and hence $J_X(X) \cong X$ is reflexive by (i).

(iii) Let $\{\ell_n\}_{n=1}^\infty$ be a dense set in X^*. Then we can choose $x_n \in X$ such that $\|x_n\| = 1$ and $\ell_n(x_n) \geq \|\ell_n\|/2$. We will show that $\{x_n\}_{n=1}^\infty$ is total in X. If it were not, we could find some $x \in X \backslash \overline{\text{span}\{x_n\}_{n=1}^\infty}$ and hence there is a functional $\ell \in X^*$ as in Corollary 5.12. Choose a subsequence $\ell_{n_k} \to \ell$, then

$$\|\ell - \ell_{n_k}\| \geq |(\ell - \ell_{n_k})(x_{n_k})| = |\ell_{n_k}(x_{n_k})| \geq \|\ell_{n_k}\|/2 \tag{5.28}$$

which implies $\ell_{n_k} \to 0$ and contradicts $\|\ell\| = 1$. □

If X is reflexive, then the converse of (iii) is also true (since $X \cong X^{**}$ separable implies X^* separable), but in general this fails as the example $\ell^1(\mathbb{N})^* = \ell^\infty(\mathbb{N})$ shows.

Problem 5.2. *Show that* $\|l_y\| = \|y\|_q$*, where* $l_y \in \ell^p(\mathbb{N})^*$ *as defined in (5.12). (Hint: Choose* $x \in \ell^p$ *such that* $|x| = |y|^{q/p}$ *and* $xy = |y|^q$*.)*

Problem 5.3. *Show that every* $l \in \ell^p(\mathbb{N})^*$, $1 \leq p < \infty$*, can be written as*

$$l(x) = \sum_{n\in\mathbb{N}} y_n x_n \tag{5.29}$$

with some $y \in \ell^q(\mathbb{N})$*.*

Problem 5.4. *Let* $c_0(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ *be the subspace of sequences which converge to* 0*, and* $c(\mathbb{N}) \subset \ell^\infty(\mathbb{N})$ *the subspace of convergent sequences.*

(i) *Show that* $c_0(\mathbb{N})$*,* $c(\mathbb{N})$ *are both Banach spaces and that* $c(\mathbb{N}) = \text{span}\{c_0(\mathbb{N}), e\}$*, where* $e = (1, 1, 1, \dots) \in c(\mathbb{N})$*.*

(ii) *Show that every* $l \in c_0(\mathbb{N})^*$ *can be written as*

$$l(x) = \sum_{n\in\mathbb{N}} y_n x_n \tag{5.30}$$

with some $y \in \ell^1(\mathbb{N})$ *which satisfies* $\|y\|_1 = \|\ell\|$*.*

(iii) *Show that every* $l \in c(\mathbb{N})^*$ *can be written as*

$$l(x) = \sum_{n\in\mathbb{N}} y_n x_n + y_0 \lim_{n\to\infty} x_n \tag{5.31}$$

with some $y \in \ell^1(\mathbb{N})$ *which satisfies* $|y_0| + \|y\|_1 = \|\ell\|$*.*

Problem 5.5. *Let $\{x_n\} \subset X$ be a total set of vectors and suppose the complex numbers c_n satisfy $|c_n| \le c\|x_n\|$. Is there a bounded linear functional $\ell \in X^*$ with $\ell(x_n) = c_n$ and $\|\ell\| \le c$? (Hint: Consider e.g. $X = \ell^2(\mathbb{Z})$.)*

5.3. Weak convergence

In the last section we have seen that $\ell(x) = 0$ for all $\ell \in X^*$ implies $x = 0$. Now what about convergence? Does $\ell(x_n) \to \ell(x)$ for every $\ell \in X^*$ imply $x_n \to x$? Unfortunately the answer is no:

Example. Let u_n be an orthonormal set in some Hilbert space. Then $\langle g, u_n\rangle \to 0$ for every g since these are just the expansion coefficients of g which are in ℓ^2 by Bessel's inequality. Since by the Riesz lemma (Theorem 2.8), every bounded linear functional is of this form, we have $\ell(u_n) \to 0$ for every bounded linear functional. (Clearly u_n does not converge to 0, since $\|u_n\| = 1$.) $\diamond$

If $\ell(x_n) \to \ell(x)$ for every $\ell \in X^*$ we say that x_n **converges weakly** to x and write

$$\operatorname*{w-lim}_{n\to\infty} x_n = x \quad \text{or} \quad x_n \rightharpoonup x. \tag{5.32}$$

Clearly $x_n \to x$ implies $x_n \rightharpoonup x$ and hence this notion of convergence is indeed weaker. Moreover, the weak limit is unique, since $\ell(x_n) \to \ell(x)$ and $\ell(x_n) \to \ell(\tilde{x})$ implies $\ell(x - \tilde{x}) = 0$. A sequence x_n is called **weak Cauchy sequence** if $\ell(x_n)$ is Cauchy (i.e. converges) for every $\ell \in X^*$.

Lemma 5.16. *Let X be a Banach space.*

(i) *$x_n \rightharpoonup x$ implies $\|x\| \le \liminf \|x_n\|$.*
(ii) *Every weak Cauchy sequence x_n is bounded: $\|x_n\| \le C$.*
(iii) *If X is reflexive, then every weak Cauchy sequence converges weakly.*

Proof. (i) Choose $\ell \in X^*$ such that $\ell(x) = \|x\|$ (for the limit x) and $\|\ell\| = 1$. Then

$$\|x\| = \ell(x) = \liminf \ell(x_n) \le \liminf \|x_n\|. \tag{5.33}$$

(ii) For every ℓ we have that $|J(x_n)(\ell)| = |\ell(x_n)| \le C(\ell)$ is bounded. Hence by the uniform boundedness principle we have $\|x_n\| = \|J(x_n)\| \le C$.
(iii) If x_n is a weak Cauchy sequence, then $\ell(x_n)$ converges and we can define $j(\ell) = \lim \ell(x_n)$. By construction j is a linear functional on X^*. Moreover, by (ii) we have $|j(\ell)| \le \sup \|\ell(x_n)\| \le \|\ell\| \sup \|x_n\| \le c\|\ell\|$ which shows $j \in X^{**}$. Since X is reflexive, $j = J(x)$ for some $x \in X$ and by construction $\ell(x_n) \to J(x)(\ell) = \ell(x)$, that is, $x_n \rightharpoonup x$. $\square$

Remark: One can equip X with the weakest topology for which all $\ell \in X^*$ remain continuous. This topology is called the **weak topology** and it is given by taking all finite intersections of inverse images of open sets as a base. By construction, a sequence will converge in the weak topology if and only if it converges weakly. By Corollary 5.12 the weak topology is Hausdorff, but it will not be metrizable in general. In particular, sequences do not suffice to describe this topology.

In a Hilbert space there is also a simple criterion for a weakly convergent sequence to converge in norm.

Lemma 5.17. *Let $\mathfrak{H}$ be a Hilbert space and let $f_n \rightharpoonup f$. Then $f_n \to f$ if and only if* $\limsup \|f_n\| \le \|f\|$.

Proof. By (i) of the previous lemma we have $\lim \|f_n\| = \|f\|$ and hence

$$\|f - f_n\|^2 = \|f\|^2 - 2\mathrm{Re}(\langle f, f_n\rangle) + \|f_n\|^2 \to 0. \tag{5.34}$$

The converse is straightforward. □

Now we come to the main reason why weakly convergent sequences are of interest: A typical approach for solving a given equation is as follows:

(i) Construct a sequence x_n of approximating solutions.
(ii) Use a compactness argument to extract a convergent subsequence.
(iii) Show that the limit solves the equation.

In a finite dimensional vector space the most important compactness criterion is boundedness (Heine-Borel theorem). In infinite dimensions this breaks down:

Theorem 5.18. *The closed unit ball in X is compact if and only if X is finite dimensional.*

For the proof we will need

Lemma 5.19. *Let X be a normed linear space and $Y \subset X$ some subspace. If $\overline{Y} \ne X$, then for every $\varepsilon \in (0,1)$ there exists an x_ε with $\|x_\varepsilon\| = 1$ and*

$$\inf_{y\in Y} \|x_\varepsilon - y\| \ge 1 - \varepsilon. \tag{5.35}$$

Proof. Abbreviate $d = \mathrm{dist}(x, Y) > 0$ and choose $y_\varepsilon \in Y$ such that $\|x - y_\varepsilon\| \le \frac{d}{1-\varepsilon}$. Set

$$x_\varepsilon = \frac{x - y_\varepsilon}{\|x - y_\varepsilon\|}. \tag{5.36}$$

Then x_ε is the vector we look for since

$$\begin{aligned} \|x_\varepsilon - y\| &= \frac{1}{\|x - y_\varepsilon\|}\|x - (y_\varepsilon + \|x - y_\varepsilon\| y)\| \\ &\ge \frac{d}{\|x - y_\varepsilon\|} \ge 1 - \varepsilon \end{aligned} \tag{5.37}$$

as required. □

Proof. (of Theorem 5.18) If X is finite dimensional, then X is isomorphic to $\mathbb{C}^n$ and the closed unit ball is compact by the Heine-Borel theorem.

Conversely, suppose X is infinite dimensional and abbreviate $S^1 = \{x \in X| \, \|x\| = 1\}$. Choose $x_1 \in S^1$ and set $Y_1 = \text{span}\{x_1\}$. Then, by the lemma there is an $x_2 \in S^1$ such that $\|x_2 - x_1\| \ge \frac{1}{2}$. Setting $Y_2 = \text{span}\{x_1, x_2\}$ and invoking again our lemma, there is an $x_3 \in S^1$ such that $\|x_3 - x_j\| \ge \frac{1}{2}$ for $j = 1, 2$. Proceeding by induction, we obtain a sequence $x_n \in S^1$ such that $\|x_n - x_m\| \ge \frac{1}{2}$ for $m \ne n$. In particular, this sequence cannot have any convergent subsequence. (Recall that in a metric space compactness and sequential compactness are equivalent.) □

If we are willing to treat convergence for weak convergence, the situation looks much brighter!

Theorem 5.20. *Let X be a reflexive Banach space, then every bounded sequence has a weakly convergent subsequence.*

Proof. Let x_n be some bounded sequence and consider $Y = \overline{\text{span}\{x_n\}}$. Then Y is reflexive by Lemma 5.15 (i). Moreover, by construction Y is separable and so is Y^* by the remark after Lemma 5.15.

Let ℓ_k be a dense set in Y^*, then by the usual diagonal sequence argument we can find a subsequence x_{n_m} such that $\ell_k(x_{n_m})$ converges for every k. Denote this subsequence again by x_n for notational simplicity. Then,

$$\begin{aligned} \|\ell(x_n) - \ell(x_m)\| &\le \|\ell(x_n) - \ell_k(x_n)\| + \|\ell_k(x_n) - \ell_k(x_m)\| \\ &\quad + \|\ell_k(x_m) - \ell(x_m)\| \\ &\le 2C\|\ell - \ell_k\| + \|\ell_k(x_n) - \ell_k(x_m)\| \end{aligned} \tag{5.38}$$

shows that $\ell(x_n)$ converges for every $\ell \in \overline{\text{span}\{\ell_k\}} = Y^*$. Thus there is a limit by Lemma 5.16 (iii). □

Note that this theorem breaks down if X is not reflexive.

Example. Let $X = L^1(\mathbb{R})$. Every bounded φ gives rise to a linear functional

$$\ell_\varphi(f) = \int f(x)\varphi(x)\, dx \tag{5.39}$$

in $L^1(\mathbb{R})^*$. Take some nonnegative u_1 with compact support, $\|u_1\|_1 = 1$, and set $u_k(x) = k u_1(k\, x)$. Then we have

$$\int u_k(x)\varphi(x)\, dx \to \varphi(0). \tag{5.40}$$

(see Problem 4.6) and if $u_{k_j} \rightharpoonup u$ we conclude

$$\int u(x)\varphi(x)\,dx = \varphi(0) \tag{5.41}$$

for every continuous φ with compact support. In particular choosing $\varphi_k(x) = \max(0, 1-k|x|)$ we infer from the dominated convergence theorem

$$1 = \int u(x)\varphi_k(x)\,dx \to \int u(x)\chi_{\{0\}}(x)\,dx = 0, \tag{5.42}$$

a contradiction.

In fact, u_k *converges* to the Dirac measure centered at 0, which is not in $L^1(\mathbb{R})$. ◇

Finally, let me remark that similar concepts can be introduced for operators. This is of particular importance for the case of unbounded operators, where convergence in the operator norm makes no sense at all.

A sequence of operators A_n is said to **converge strongly** to A,

$$\operatorname*{s-lim}_{n\to\infty} A_n = A \quad :\Leftrightarrow \quad A_n x \to Ax \quad \forall x \in \mathfrak{D}(A) \subseteq \mathfrak{D}(A_n). \tag{5.43}$$

It is said to **converge weakly** to A,

$$\operatorname*{w-lim}_{n\to\infty} A_n = A \quad :\Leftrightarrow \quad A_n x \rightharpoonup Ax \quad \forall x \in \mathfrak{D}(A) \subseteq \mathfrak{D}(A_n). \tag{5.44}$$

Clearly norm convergence implies strong convergence and strong convergence implies weak convergence.

Example. Consider the operator $S_n \in \mathfrak{L}(\ell^p(\mathbb{N}))$ which shifts a sequence n places to the left

$$S_n\,(x_1, x_2, \dots) = (x_{n+1}, x_{n+2}, \dots). \tag{5.45}$$

and the operator $S_n^* \in \mathfrak{L}(\ell^p(\mathbb{N}))$ which shifts a sequence n places to the right and fills up the first n places with zeros

$$S_n^*\,(x_1, x_2, \dots) = (\underbrace{0, \dots, 0}_{n \text{ places}}, x_1, x_2, \dots). \tag{5.46}$$

Then S_n converges to zero strongly but not in norm and S_n^* converges weakly to zero but not strongly. ◇

Lemma 5.21. *Suppose A_n is a sequence of bounded operators.*

(i) $\operatorname{s-lim}_{n\to\infty} A_n = A$ *implies* $\|A\| \le \liminf \|A_n\|$.

(ii) *Every strong Cauchy sequence A_n is bounded: $\|A_n\| \le C$.*

(iii) *If $A_n y \to Ay$ for y in a dense set and $\|A_n\| \le C$, then* $\operatorname{s-lim}_{n\to\infty} A_n = A$.

The same result holds if strong convergence is replaced by weak convergence.

Proof. (i) and (ii) follow as in Lemma 5.16 (i).
(iii) Just use

$$\begin{aligned}\|A_n x - Ax\| &\leq \|A_n x - A_n y\| + \|A_n y - Ay\| + \|Ay - Ax\| \\ &\leq 2C\|x - y\| + \|A_n y - Ay\| \qquad (5.47)\end{aligned}$$

and choose y in the dense subspace such that $\|x-y\| \leq \frac{\varepsilon}{4C}$ and n large such that $\|A_n y - Ay\| \leq \frac{\varepsilon}{2}$.

The case of weak convergence is left as an exercise. □

For an application of this lemma see Problem 5.9.

Remark: For a sequence of linear functionals ℓ_n, strong convergence is also called **weak-$*$** convergence. That is, the weak-$*$ limit of ℓ_n is ℓ if

$$\ell_n(x) \to \ell(x) \qquad \forall x \in X. \qquad (5.48)$$

Note that this is not the same as weak convergence on X^*, since ℓ is the weak limit of ℓ_n if

$$j(\ell_n) \to j(\ell) \qquad \forall j \in X^{**}, \qquad (5.49)$$

whereas for the weak-$*$ limit this is only required for $j \in J(X) \subseteq X^{**}$ (recall $J(x)(\ell) = \ell(x)$). So the weak topology on X^* is the weakest topology for which all $j \in X^{**}$ remain continuous and the weak-$*$ topology on X^* is the weakest topology for which all $j \in J(X)$ remain continuous. In particular, the weak-$*$ topology is weaker than the weak topology and both are equal if X is reflexive.

With this notation it is also possible to slightly generalize Theorem 5.20 (Problem 5.10):

Theorem 5.22. *Suppose X is separable. Then every bounded sequence $\ell_n \in X^*$ has a weak-$*$ convergent subsequence.*

Example. Let us return to the example after Theorem 5.20. Consider the Banach space of bounded contionuous functions $X = C(\mathbb{R})$. Using $\ell_f(\varphi) = \int \varphi f\, dx$ we can regard $L^1(\mathbb{R})$ as a subspace of X^*. Then the Dirac measure centered at 0 is also in X^* and it is the weak-$*$ limit of the sequence u_k. ⋄

Problem 5.6. *Suppose $\ell_n \to \ell$ in X^* and $x_n \rightharpoonup x$ in X. Then $\ell_n(x_n) \to \ell(x)$.*

Problem 5.7. *Show that if $\{\ell_j\} \subseteq X^*$ is some total set, then $x_n \rightharpoonup x$ if and only if x_n is bounded and $\ell_j(x_n) \to \ell_j(x)$ for all j. Show that this is wrong without the boundedness assumption (Hint: Take e.g. $X = \ell^2(\mathbb{N})$).*

Problem 5.8 (Convolution). *Show that for $f \in L^1(\mathbb{R}^n)$ and $g \in L^p(\mathbb{R}^n)$, the* **convolution**

$$(g * f)(x) = \int_{\mathbb{R}^n} g(x-y)f(y)dy = \int_{\mathbb{R}^n} g(y)f(x-y)dy \tag{5.50}$$

is in $L^p(\mathbb{R}^n)$ and satisfies **Young's inequality**

$$\|f * g\|_p \le \|f\|_1 \|g\|_p. \tag{5.51}$$

(Hint: Without restriction $\|f\|_1 = 1$. Now use Jensen and Fubini.)

Problem 5.9 (Smoothing). *Suppose $f \in L^p(\mathbb{R}^n)$. Show that f_k defined as in (4.29) converges to f in L^p. (Hint: Use Lemma 5.21 and Young's inequality.)*

Problem 5.10. *Prove Theorem 5.22*

Chapter 6

The dual of L^p

6.1. Decomposition of measures

Let μ, ν be two measures on a measure space (X, Σ). They are called **mutually singular** (in symbols $\mu \perp \nu$) if they are supported on disjoint sets. That is, there is a measurable set N such that $\mu(N) = 0$ and $\nu(X \backslash N) = 0$.

Example. Let λ be the Lebesgue measure and Θ the Dirac measure (centered at 0), then $\lambda \perp \Theta$: Just take $N = \{0\}$, then $\lambda(\{0\}) = 0$ and $\Theta(\mathbb{R}\backslash\{0\}) = 0$. ⋄

On the other hand, ν is called **absolutely continuous** with respect to μ (in symbols $\nu \ll \mu$) if $\mu(A) = 0$ implies $\nu(A) = 0$.

Example. The prototypical example is the measure $d\nu = f\, d\mu$ (compare Lemma 3.15). Indeed $\mu(A) = 0$ implies

$$\nu(A) = \int_A f\, d\mu = 0 \tag{6.1}$$

and shows that ν is absolutely continuous with respect to μ. In fact, we will show below that every absolutely continuous measure is of this form. ⋄

The two main results will follow as simple consequence of the following result:

Theorem 6.1. *Let μ, ν be σ-finite measures. Then there exists a unique (a.e.) nonnegative function f and a set N of μ measure zero, such that*

$$\nu(A) = \nu(A \cap N) + \int_A f\, d\mu. \tag{6.2}$$

Proof. We first assume μ, ν to be finite measures. Let $\alpha = \mu + \nu$ and consider the Hilbert space $L^2(X, d\alpha)$. Then

$$\ell(h) = \int_X h\, d\nu \tag{6.3}$$

is a bounded linear functional by Cauchy-Schwarz:

$$\begin{aligned} |\ell(h)|^2 &= \left| \int_X 1 \cdot h\, d\nu \right|^2 \le \left(\int |1|^2\, d\nu \right) \left(\int |h|^2 d\nu \right) \\ &\le \nu(X) \left(\int |h|^2 d\alpha \right) = \nu(X) \|h\|^2. \end{aligned} \tag{6.4}$$

Hence by the Riesz lemma (Theorem 2.8) there exists an $g \in L^2(X, d\alpha)$ such that

$$\ell(h) = \int_X hg\, d\alpha. \tag{6.5}$$

By construction

$$\nu(A) = \int \chi_A\, d\nu = \int \chi_A g\, d\alpha = \int_A g\, d\alpha. \tag{6.6}$$

In particular, g must be positive a.e. (take A the set where g is negative). Furthermore, let $N = \{x | g(x) \ge 1\}$, then

$$\nu(N) = \int_N g\, d\alpha \ge \alpha(N) = \mu(N) + \nu(N), \tag{6.7}$$

which shows $\mu(N) = 0$. Now set

$$f = \frac{g}{1-g} \chi_{N'}, \qquad N' = X \backslash N. \tag{6.8}$$

Then, since (6.6) implies $d\nu = g\, d\alpha$ respectively $d\mu = (1-g) d\alpha$, we have

$$\begin{aligned} \int_A f d\mu &= \int \chi_A \frac{g}{1-g} \chi_{N'}\, d\mu \\ &= \int \chi_{A \cap N'} g\, d\alpha \\ &= \nu(A \cap N') \end{aligned} \tag{6.9}$$

as desired. Clearly f is unique, since if there is a second function $\tilde{f}$, then $\int_A (f - \tilde{f}) d\mu = 0$ for every A shows $f - \tilde{f} = 0$ a.e..

To see the σ-finite case, observe that $X_n \nearrow X$, $\mu(X_n) < \infty$ and $Y_n \nearrow X$, $\nu(Y_n) < \infty$ implies $X_n \cap Y_n \nearrow X$ and $\alpha(X_n \cap Y_n) < \infty$. Hence when restricted to $X_n \cap Y_n$ we have sets N_n and functions f_n. Now take $N = \bigcup N_n$ and choose f such that $f|_{X_n} = f_n$ (this is possible since $f_{n+1}|_{X_n} = f_n$ a.e.). Then $\mu(N) = 0$ and

$$\nu(A \cap N') = \lim_{n \to \infty} \nu(A \cap (X_n \backslash N)) = \lim_{n \to \infty} \int_{A \cap X_n} f\, d\mu = \int_A f\, d\mu, \tag{6.10}$$

which finishes the proof. □

Now the anticipated results follow with no effort:

Theorem 6.2 (Lebesgue decomposition). *Let μ, ν be two σ-finite measures on a measure space (X, Σ). Then ν can be uniquely decomposed as $\nu = \nu_{ac} + \nu_{sing}$, where ν_{ac} and ν_{sing} are mutually singular and ν_{ac} is absolutely continuous with respect to μ.*

Proof. Taking $\nu_{sing}(A) = \nu(A \cap N)$ and $d\nu_{ac} = f\, d\mu$ there is at least one such decomposition. To show uniqueness, let ν be finite first. If there is another one $\nu = \tilde{\nu}_{ac} + \tilde{\nu}_{sing}$, then let $\tilde{N}$ be such that $\mu(\tilde{N}) = 0$ and $\tilde{\nu}_{sing}(\tilde{N}') = 0$. Then $\tilde{\nu}_{sing}(A) - \nu_{sing}(A) = \int_A (\tilde{f} - f) d\mu$. In particular, $\int_{A \cap N' \cap \tilde{N}'} (\tilde{f} - f) d\mu = 0$ and hence $\tilde{f} = f$ a.e. away from $N \cup \tilde{N}$. Since $\mu(N \cup \tilde{N}) = 0$, we have $\tilde{f} = f$ a.e. and hence $\tilde{\nu}_{ac} = \nu_{ac}$ as well as $\tilde{\nu}_{sing} = \nu - \tilde{\nu}_{ac} = \nu - \nu_{ac} = \nu_{sing}$. The σ-finite case follows as usual. □

Theorem 6.3 (Radon-Nikodym). *Let μ, ν be two σ-finite measures on a measure space (X, Σ). Then ν is absolutely continuous with respect to μ if and only if there is a positive measurable function f such that*

$$\nu(A) = \int_A f\, d\mu \tag{6.11}$$

for every $A \in \Sigma$. The function f is determined uniquely a.e. with respect to μ and is called the **Radon-Nikodym derivative** $\frac{d\nu}{d\mu}$ *of ν with respect to μ.*

Proof. Just observe that in this case $\nu(A \cap N) = 0$ for every A, that is $\nu_{sing} = 0$. □

Problem 6.1. *Let μ is a Borel measure on $\mathfrak{B}$ and suppose its distribution function $\mu(x)$ is differentiable. Show that the Radon-Nikodym derivative equals the ordinary derivative $\mu'(x)$.*

Problem 6.2. *Suppose μ and ν are inner regular measures. Show that $\nu \ll \mu$ if and only if $\mu(C) = 0$ implies $\nu(C) = 0$ for every compact set C.*

Problem 6.3 (Chain rule). *Show that $\nu \ll \mu$ is a transitive relation. In particular, if $\omega \ll \nu \ll \mu$ show that*

$$\frac{d\omega}{d\mu} = \frac{d\omega}{d\nu}\frac{d\nu}{d\mu}.$$

Problem 6.4. *Suppose $\nu \ll \mu$. Show that*

$$\frac{d\omega}{d\mu} d\mu = \frac{d\omega}{d\nu} d\nu + d\zeta,$$

where ζ is a positive measure which is singular with respect to ν. Show that $\zeta = 0$ if and only if $\mu \ll \nu$.

6.2. Complex measures

Let (X, Σ) be some measure space. A map $\nu : \Sigma \to \mathbb{C}$ is called a **complex measure** if

$$\nu(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \nu(A_n), \qquad A_n \cap A_m = \emptyset, \quad n \neq m. \tag{6.12}$$

Note that a positive measure is a complex measure only if it is finite (the value ∞ is not allowed for complex measures). Moreover, the definition implies that the sum is independent of the order of the sets A_j, hence the sum must be absolutely convergent.

Example. Let μ be a positive measure. For every $f \in L^1(X, d\mu)$ we have that $f\, d\mu$ is a complex measure (compare the proof of Lemma 3.15 and use dominated in place of monotone convergence). In fact, we will show that every complex measure is of this form. ◇

The **total variation** of a measure is defined as

$$|\nu|(A) = \sup\Big\{ \sum_{n=1}^{\infty} |\nu(A_n)| \Big| A_n \in \Sigma \text{ disjoint},\ A = \bigcup_{n=1}^{\infty} A_n \Big\}. \tag{6.13}$$

Theorem 6.4. *The total variation is a positive measure.*

Proof. Suppose $A = \bigcup_{n=1}^{\infty} A_n$. We need to show $|\nu|(A) = \sum_{n=1}^{\infty} |\nu|(A_n)$ for disjoint sets A_n.

Let $B_{n,k}$ be disjoint sets such that

$$|\nu|(A_n) \leq \sum_{k=1}^{\infty} |\nu(B_{n,k})| + \frac{\varepsilon}{2^n}. \tag{6.14}$$

Then

$$\sum_{n=1}^{\infty} |\nu|(A_n) \leq \sum_{n,k=1}^{\infty} |\nu(B_{n,k})| + \varepsilon \leq |\nu|(A) + \varepsilon \tag{6.15}$$

since $\bigcup_{n,k=1}^{\infty} B_{n,k} = A$. Letting $\varepsilon \to 0$ shows $|\nu|(A) \geq \sum_{n=1}^{\infty} |\nu|(A_n)$.

Conversely, if $A = \bigcup_{n=1}^{\infty} B_n$, then

$$\begin{aligned} \sum_{k=1}^{\infty} |\nu(B_k)| &= \sum_{k=1}^{\infty} \Big| \sum_{n=1}^{\infty} \nu(B_k \cap A_n) \Big| \leq \sum_{k,n=1}^{\infty} |\nu(B_k \cap A_n)| \\ &= \sum_{n=1}^{\infty} \sum_{k=1}^{\infty} |\nu(B_k \cap A_n)| \leq \sum_{n=1}^{\infty} |\nu|(A_n). \end{aligned} \tag{6.16}$$

Taking the supremum shows $|\nu|(A) \leq \sum_{n=1}^{\infty} |\nu|(A_n)$. □

Theorem 6.5. *The total variation $|\nu|$ of a complex measure ν is a finite measure.*

Proof. Splitting ν into its real and imaginary part, it is no restriction to assume that ν is real-valued since $|\nu|(A) \le |\mathrm{Re}(\nu)|(A) + |\mathrm{Im}(\nu)|(A)$.

The idea is as follows: Suppose we can split any given set A with $|\nu|(A) = \infty$ into two subsets B and $A\backslash B$ such that $|\nu(B)| \ge 1$ and $|\nu|(A\backslash B) = \infty$. Then we can construct a sequence B_n of disjoint sets with $|\nu(B_n)| \ge 1$ for which

$$\sum_{n=1}^{\infty} \nu(B_n) \tag{6.17}$$

diverges (the terms of a convergent series must converge to zero). But σ-additivity requires that the sum converges to $\nu(\bigcup_n B_n)$, a contradiction.

It remains to show existence of this splitting. Let A with $|\nu|(A) = \infty$ be given. Then there are disjoint sets A_j such that

$$\sum_{j=1}^{n} |\nu(A_j)| \ge 2(1 + |\nu(A)|). \tag{6.18}$$

Now let $A_\pm = \bigcup\{A_j | \pm\nu(A_j) > 0\}$. Then for one of them we have $|\nu(A_\sigma)| \ge 1 + |\nu(A)|$ and hence

$$|\nu(A\backslash A_\sigma)| = |\nu(A) - \nu(A_\sigma)| \ge |\nu(A_\sigma)| - |\nu(A)| \ge 1. \tag{6.19}$$

Moreover, by $|\nu|(A) = |\nu|(A_\sigma) + |\nu|(A\backslash A_\sigma)$ either A_σ or $A\backslash A_\sigma$ must have infinite $|\nu|$ measure. □

Note that this implies that every complex measure ν can be written as a linear combination of four positive measures. In fact, first we can split ν into its real and imaginary part

$$\nu = \nu_r + \mathrm{i}\nu_i, \qquad \nu_r(A) = \mathrm{Re}(\nu(A)),\ \nu_i(A) = \mathrm{Im}(\nu(A)). \tag{6.20}$$

Second we can split any real (also called signed) measure according to

$$\nu = \nu_+ - \nu_-, \qquad \nu_\pm(A) = \frac{|\nu|(A) \pm \nu(A)}{2}. \tag{6.21}$$

This splitting is also known as **Hahn decomposition** of a signed measure.

If μ is a positive and ν a complex measure we say that ν is absolutely continuous with respect to μ if $\mu(A) = 0$ implies $\nu(A) = 0$.

Lemma 6.6. *If μ is a positive and ν a complex measure then $\nu \ll \mu$ if and only if $|\nu| \ll \mu$.*

Proof. If $\nu \ll \mu$, then $\mu(A) = 0$ implies $\mu(B) = 0$ for every $B \subseteq A$ and hence $|\mu|(A) = 0$. Conversely, if $|\nu| \ll \mu$, then $\mu(A) = 0$ implies $|\nu(A)| \le |\nu|(A) = 0$. □

Now we can prove the complex version of the Radon-Nikodym theorem:

Theorem 6.7 (complex Radon-Nikodym)**.** *Let (X, Σ) be a measure space, μ a positive σ-finite measure and ν a complex measure which is absolutely continuous with respect to μ. Then there is a unique $f \in L^1(X, d\mu)$ such that*

$$\nu(A) = \int_A f\, d\mu. \tag{6.22}$$

Proof. By treating the real and imaginary part separately it is no restriction to assume that ν is real-valued. Let $\nu = \nu_+ - \nu_-$ be its Hahn decomposition, then both ν_+ and ν_- are absolutely continuous with respect to μ and by the Radon-Nikodym theorem there are functions $f_\pm$ such that $d\nu_\pm = f_\pm d\mu$. By construction

$$\int_X f_\pm d\mu = \nu_\pm(X) \le |\nu|(X) < \infty, \tag{6.23}$$

which shows $f = f_+ - f_- \in L^1(X, d\mu)$. Moreover, $d\nu = d\nu_+ - d\nu_- = f\, d\mu$ as required. □

In this case the total variation of $d\nu = f\, d\mu$ is just $d|\nu| = |f| d\mu$:

Lemma 6.8. *Suppose $d\nu = f\, d\mu$, where μ is a positive measure and $f \in L^1(X, d\mu)$. Then*

$$|\nu|(A) = \int_A |f| d\mu. \tag{6.24}$$

Proof. If A_n are disjoint sets and $A = \bigcup_n A_n$ we have

$$\sum_n |\nu(A_n)| = \sum_n \Big| \int_{A_n} f\, d\mu \Big| \le \sum_n \int_{A_n} |f| d\mu = \int_A |f| d\mu. \tag{6.25}$$

Hence $|\nu|(A) \le \int_A |f| d\mu$. To show the converse define

$$A_k^n = \{x | \frac{k-1}{n} \le \frac{\arg(f(x))}{2\pi} < \frac{k}{n}\}, \quad 1 \le k \le n. \tag{6.26}$$

Then the simple functions

$$s_n(x) = \sum_{k=1}^n \mathrm{e}^{-2\pi\mathrm{i}\frac{k-1}{n}} \chi_{A_k^n}(x) \tag{6.27}$$

converge to $f(x)^*/|f(x)|$ pointwise and hence

$$\lim_{n\to\infty} \int_A s_n f\, d\mu = \int_A |f| d\mu \tag{6.28}$$

by dominated convergence. Moreover,

$$\Big| \int_A s_n f\, d\mu \Big| \le \sum_{k=1}^n \Big| \int_{A_k^n} f\, d\mu \Big| \le \sum_{k=1}^n |\nu(A_k^n)| \le |\nu|(A) \tag{6.29}$$

shows $\int_A |f| d\mu \le |\nu|(A)$. □

As a consequence we obtain (Problem 6.5):

Corollary 6.9. *If ν is a complex measure, then $d\nu = h\, d|\nu|$, where $|h| = 1$.*

In particular, note that

$$\Big| \int_A f\, d\nu \Big| \le \|f\|_\infty |\nu|(A). \tag{6.30}$$

Problem 6.5. *Prove Corollary 6.9 (Hint: Use the complex Radon-Nikodym theorem to get existence of f. Then show that $1 - |f|$ vanishes a.e.).*

6.3. The dual of L^p, $p < \infty$

After these preparations we are able to compute the dual of L^p for $p < \infty$.

Theorem 6.10. *Consider $L^p(X, d\mu)$ for some σ-finite measure. Then the map $g \in L^q \to \ell_g \in (L^p)^*$ given by*

$$\ell_g(f) = \int_X g f\, d\mu \tag{6.31}$$

is isometric. Moreover, for $1 \le p < \infty$ it is also surjective.

Proof. Given $g \in L^q$ it follows from Hölder's inequality that ℓ_g is a bounded linear functional with $\|\ell_g\| \le \|g\|_q$. That in fact $\|\ell_g\| = \|g\|_q$ can be shown as in the discrete case (compare Problem 5.2).

To show that this map is surjective, first suppose $\mu(X) < \infty$ and choose some $\ell \in (L^p)^*$. Since $\|\chi_A\|_p = \mu(A)^{1/p}$, we have $\chi_A \in L^p$ for every $A \in \Sigma$ and we can define

$$\nu(A) = \ell(\chi_A). \tag{6.32}$$

Suppose $A = \bigcup_{j=1}^\infty A_j$. Then, by dominated convergence, $\| \sum_{j=1}^n \chi_{A_j} - \chi_A\|_p \to 0$ (this is false for $p = \infty$!) and hence

$$\nu(A) = \ell(\sum_{j=1}^\infty \chi_{A_j}) = \sum_{j=1}^\infty \ell(\chi_{A_j}) = \sum_{j=1}^\infty \nu(A_j). \tag{6.33}$$

Thus ν is a complex measure. Moreover, $\mu(A) = 0$ implies $\chi_A = 0$ in L^p and hence $\nu(A) = \ell(\chi_A) = 0$. Thus ν is absolutely continuous with respect to μ and by the complex Radon-Nikodym theorem $d\nu = g\, d\mu$ for some $g \in L^1(X, d\mu)$. In particular, we have

$$\ell(f) = \int_X f g\, d\mu \tag{6.34}$$

for every simple function f. Clearly, the simple functions are dense in L^p, but since we only know $g \in L^1$ we cannot control the integral. So suppose f is bounded and pick a sequence of simple function f_n converging to f. Without restriction we can assume that f_n converges also pointwise and $\|f_n\|_\infty \le$

$\|f\|_\infty$. Hence by dominated convergence $\ell(f) = \lim \ell(f_n) = \lim \int_X f_n g\, d\mu = \int_X fg\, d\mu$. Thus equality holds for every bounded function.

Next let $A_n = \{x | 0 < |g| < n\}$. Then, if $1 < p$,

$$\|\chi_{A_n} g\|_q^q = \int_{A_n} \frac{|g|^q}{g} g\, d\mu = \ell(\chi_{A_n} \frac{|g|^q}{g}) \le \|\ell\| \|\chi_{A_n} \frac{|g|^q}{g}\|_p^{1/p} = \|\ell\| \|\chi_{A_n} g\|_q^{q/p} \tag{6.35}$$

and hence

$$\|\chi_{A_n} g\|_q \le \|\ell\|. \tag{6.36}$$

Letting $n \to \infty$ shows $g \in L^q$. If $p = 1$, let $A_n = \{x | |g| \ge \|\ell\| + \frac{1}{n}\}$, then

$$(\|\ell\| + \frac{1}{n})\mu(A_n) \le \int_X \chi_{A_n} |g|\, d\mu \le \|\ell\| \mu(A_n), \tag{6.37}$$

which shows $\mu(A_n) = 0$ and hence $\|g\|_\infty \le \|\ell\|$, that is $g \in L^\infty$. This finishes the proof for finite μ.

If μ is σ-finite, let $X_n \nearrow X$ with $\mu(X_n) < \infty$. Then for every n there is some g_n on X_n and by uniqueness of g_n we must have $g_n = g_m$ on $X_n \cap X_m$. Hence there is some g and by $\|g_n\| \le \|\ell\|$ independent of n, we have $g \in L^q$. □

6.4. The dual of L^∞ and the Riesz representation theorem

In the last section we have computed the dual space of L^p for $p < \infty$. Now we want to investigate the case $p = \infty$. Recall that we already know that the dual of L^∞ is much larger than L^1 since it cannot be separable in general.

Example. Let ν be a complex measure. Then

$$\ell_\nu(f) = \int_X f d\nu \tag{6.38}$$

is a bounded linear functional on $B(X)$ (the Banach space of bounded measurable functions) with norm

$$\|\ell_\nu\| = |\nu|(X) \tag{6.39}$$

by (6.30). If ν is absolutely continuous with respect to μ, then it will even be a bounded linear functional on $L^\infty(X, d\mu)$ since the integral will be independent of the representative in this case. ◇

So the dual of $B(X)$ contains all complex measures. However, this is still not all of $B(X)^*$. In fact, it turns out that it suffices to require only finite additivity for ν.

Let (X, Σ) be a measure space. A complex content ν is a map $\nu : \Sigma \to \mathbb{C}$ such that (finite additivity)

$$\nu(\bigcup_{k=1}^n A_k) = \sum_{k=1}^n \nu(A_k), \qquad A_j \cap A_k = \emptyset,\ j \neq k. \tag{6.40}$$

Given a content ν we can define the corresponding integral for simple functions $s(x) = \sum_{k=1}^n \alpha_k \chi_{A_k}$ as usual

$$\int_A s\, d\nu = \sum_{k=1}^n \alpha_k \nu(A_k \cap A). \tag{6.41}$$

As in the proof of Lemma 3.13 one shows that the integral is linear. Moreover,

$$|\int_A s\, d\nu| \leq |\nu|(A)\, \|s\|_\infty, \tag{6.42}$$

where

$$|\nu|(A) = \sup\Big\{ \sum_{k=1}^n |\mu(A_k)| \Big| A_k \in \Sigma \text{ disjoint},\ A = \bigcup_{k=1}^n A_k \Big\}. \tag{6.43}$$

(Note that this definition agrees with the one for complex measures.) Hence this integral can be extended to all of $B(X)$ by Theorem 1.25 (compare Problem 3.5). However, note that our convergence theorems (monotone convergence, dominated convergence) will no longer hold in this case (unless ν happens to be a measure).

In particular, every complex content gives rise to a bounded linear functional on $B(X)$ and the converse also holds:

Theorem 6.11. *Every bounded linear functional $\ell \in B(X)^*$ is of the form*

$$\ell(f) = \int_X f\, d\nu \tag{6.44}$$

for some unique complex content ν and $\|\ell\| = |\nu|(X)$.

Proof. Let $\ell \in B(X)^*$ be given. If there is a content ν at all it is uniquely determined by $\nu(A) = \ell(\chi_A)$. Using this as definition for ν, we see that finite additivity follows from linearity of ℓ. Moreover, (6.44) holds for characteristic functions. Since the characteristic functions are total, (6.44) holds everywhere by continuity. □

Remark: To obtain the dual of $L^\infty(X, d\mu)$ from this you just need to restrict to those linear functionals which vanish on $\mathcal{N}(X, d\mu)$, that is, those whose content is *absolutely continuous* with respect to μ (note that the Radon-Nikodym theorem does not hold unless the content is a measure).

Example. Consider $B(\mathbb{R})$ and define

$$\ell(f) = \lim_{\varepsilon\downarrow 0}(\lambda f(-\varepsilon) + (1-\lambda) f(\varepsilon)), \qquad \lambda \in [-1,1], \tag{6.45}$$

for f in the subspace of bounded measurable functions which have left and right limits at 0. Since $\|\ell\| = 1$ we can extend it to all of $B(\mathbb{R})$ using the Hahn-Banach theorem. Then the corresponding content ν is no measure:

$$\lambda = \nu([-1,0)) = \nu(\bigcup_{n=1}^{\infty}[-\frac{1}{n}, -\frac{1}{n+1})) \neq \sum_{n=1}^{\infty}\nu([-\frac{1}{n}, -\frac{1}{n+1})) = 0 \tag{6.46}$$

Observe that the corresponding distribution function (defined as in (3.4)) is non-decreasing but not right continuous! If we render the distribution function right continuous, we get the Dirac measure (centered at 0). In addition, the Dirac measure has the same integral at least for continuous functions! ⋄

Theorem 6.12 (Riesz representation)**.** *Let $I \subseteq \mathbb{R}$ be a compact interval. Every bounded linear functional $\ell \in C(I)^*$ is of the form*

$$\ell(f) = \int_X f\, d\nu \tag{6.47}$$

for some unique complex Borel measure ν and $\|\ell\| = |\nu|(X)$.

Proof. Without restriction $I = [0,1]$. Extending ℓ to a bounded linear functional $\bar{\ell} \in B(I)^*$ we have a corresponding content ν. Splitting this content into real and imaginary part we see that it is no restriction to assume that ν is real. Moreover, the same proof as in the case of measures shows that $|\nu|$ is a positive content and splitting ν into $\nu_\pm = (|\nu| \pm \nu)/2$ it is no restriction to assume ν is positive.

Now the idea is as follows: Define a distribution function for ν. By finite additivity of ν it will be non-decreasing, but it might not be right-continuous. However, right-continuity is needed to use Theorem 3.8. So why not change the distribution function at each jump such that it becomes right continuous? This is fine if we can show that this does not alter the value of the integral of continuous functions.

Let $f \in C(I)$ be given. Fix points $a \le x_0^n < x_1^n < \dots x_n^n \le b$ such that $x_0^n \to a$, $x_n^n \to b$, and $\sup_k |x_{k-1}^n - x_k^n| \to 0$ as $n \to \infty$. Then the sequence of simple functions

$$f_n(x) = f(x_0^n)\chi_{[x_0^n, x_1^n)} + f(x_1^n)\chi_{[x_1^n, x_2^n)} + \dots + f(x_{n-1}^n)\chi_{[x_{n-1}^n, x_n^n]}. \tag{6.48}$$

converges uniformly to f by continuity of f. Moreover,

$$\int_I f\, d\nu = \lim_{n\to\infty}\int_I f_n\, d\nu = \lim_{n\to\infty}\sum_{k=1}^{n} f(x_{k-1}^n)(\nu(x_k) - \nu(x_{k-1})), \tag{6.49}$$

where $\nu(x) = \nu([0,x))$, $\nu(1) = \nu([0,1])$, and the points x_k^n are chosen to stay away from all discontinuities of $\nu(x)$. Since ν is monotone, there are at most countably many discontinuities and this is possible. In particular, we can change $\nu(x)$ at its discontinuities such that it becomes right continuous without changing the value of the integral (for continuous functions). Now Theorem 3.8 ensures existence of a corresponding measure. □

Note that ν will be a positive measure if ℓ is a **positive functional**, that is, $\ell(f) \geq 0$ whenever $f \geq 0$.

Problem 6.6 (Weak convergence of measures)**.** *A sequence of measures ν_n are said to converge weakly to a measure ν if*

$$\int_X f d\nu_n \to \int_X f d\nu, \qquad f \in C(I). \tag{6.50}$$

Show that every bounded sequnce of measures has a weakly convergent subsequence. Show that the limit ν is a positive measure if all ν_n are. (Hint: Compare this definition to the definition of weak-$*$ convergence in Section 5.3.)

Chapter 7

Bounded linear operators

7.1. Banach algebras

In this section we want to have a closer look at the set of bounded linear operators $\mathfrak{L}(X)$ from a Banach space X into itself. We already know that they form a Banach space, however, in this case we even have a multiplication given by composition. Clearly this multiplication satisfies

$$(A+B)C = AC + BC, \quad A(B+C) = AB + AC, \quad A, B, C \in \mathfrak{L}(X) \tag{7.1}$$

and

$$(AB)C = A(BC), \qquad \alpha\,(AB) = (\alpha A)B = A\,(\alpha B), \quad \alpha \in \mathbb{C}. \tag{7.2}$$

Moreover, we have (Problem 7.1)

$$\|AB\| \le \|A\|\|B\|. \tag{7.3}$$

However, note that our multiplication is not commutative (unless X is one dimensional). We even have an identity, the identity operator $\mathbb{I}$.

A Banach space X together with a multiplication satisfying the above requirements is called a **Banach algebra**. In particular, note that (7.3) ensures that multiplication is continuous (Problem 7.2). An element $e \in X$ satisfying

$$ex = xe = x, \qquad \forall x \in X \tag{7.4}$$

is called **identity** (show that e is unique) and we will assume $\|e\| = 1$ in this case.

Example. The continuous functions $C(I)$ over some compact interval form a commutative Banach algebra with identity 1. ⋄

Example. The space $L^1(\mathbb{R})$ together with the convolution

$$(g * f)(x) = \int_{\mathbb{R}^n} g(x-y) f(y) dy = \int_{\mathbb{R}^n} g(y) f(x-y) dy \tag{7.5}$$

is a commutative Banach algebra (Problem 7.3) without identity. ⋄

Let X be a Banach algebra with identity e. Then $x \in X$ is called **invertible** if there is some $y \in X$ such that

$$xy = yx = e. \tag{7.6}$$

In this case y is called the inverse of x and is denoted by x^{-1}. It is straightforward to show that the inverse is unique (if one exists at all) and that

$$(xy)^{-1} = y^{-1} x^{-1}. \tag{7.7}$$

Example. Let $X = \mathfrak{L}(\ell^1(\mathbb{N}))$ and let $S^{\pm}$ be defined via

$$S^- x_n = \begin{cases} 0 & n = 1 \\ x_{n-1} & n > 1 \end{cases}, \quad S^+ x_n = x_{n+1} \tag{7.8}$$

(i.e., S^- shifts each sequence one place right (filling up the first place with a 0) and S^+ shifts one place left (dropping the first place)). Then $S^+ S^- = \mathbb{I}$ but $S^- S^+ \neq \mathbb{I}$. So you really need to check both $xy = e$ and $yx = e$ in general. ⋄

Lemma 7.1. *Let X be a Banach algebra with identity e. Suppose $\|x\| < 1$. Then $e - x$ is invertible and*

$$(e - x)^{-1} = \sum_{n=0}^{\infty} x^n. \tag{7.9}$$

Proof. Since $\|x\| < 1$ the series converges and

$$(e - x) \sum_{n=0}^{\infty} x^n = \sum_{n=0}^{\infty} x^n - \sum_{n=1}^{\infty} x^n = e \tag{7.10}$$

respectively

$$\left(\sum_{n=0}^{\infty} x^n \right) (e - x) = \sum_{n=0}^{\infty} x^n - \sum_{n=1}^{\infty} x^n = e. \tag{7.11}$$

□

Corollary 7.2. *Suppose x is invertible and $\|y x^{-1}\| < 1$ or $\|x^{-1} y\| < 1$, then*

$$(x - y)^{-1} = \sum_{n=0}^{\infty} (x^{-1} y)^n x^{-1} = \sum_{n=0}^{\infty} x^{-1} (y x^{-1})^n. \tag{7.12}$$

Proof. Just observe $x - y = x(e - x^{-1} y) = (e - y x^{-1}) x$. □

The **resolvent set** is defined as

$$\rho(x) = \{\alpha \in \mathbb{C} | \exists (x-\alpha)^{-1}\} \subseteq \mathbb{C} \tag{7.13}$$

and its complement is called the **spectrum**

$$\sigma(x) = \mathbb{C} \backslash \rho(x). \tag{7.14}$$

It is important to observe that the fact that the inverse has to exist as an element of X. That is, if X are bounded linear operators, it does not suffice that $x - \alpha$ is bijective, the inverse must also be bounded!

Example. If $X = \mathrm{GL}(n) = \mathfrak{L}(\mathbb{C}^n)$ is the space of n by n matrices, then the spectrum is just the set of eigenvalues. ⋄

Example. If $X = C(I)$, then the spectrum of a function $x \in C(I)$ is just its range, $\sigma(x) = x(I)$. ⋄

The map $\alpha \mapsto (x-\alpha)^{-1}$ is called the **resolvent** of $x \in X$. By (7.12) we have

$$(x-\alpha)^{-1} = \sum_{n=0}^{\infty} (\alpha - \alpha_0)^n (x - \alpha_0)^{-n-1}, \quad |\alpha - \alpha_0| < \|\alpha_0 - x\|, \tag{7.15}$$

which shows that $(x-\alpha)^{-1}$ has a convergent power series with coefficients in X around every point $\alpha_0 \in \rho(x)$. As in the case of coefficients in $\mathbb{C}$, such functions will be called **analytic**. In particular, $\ell((x-\alpha)^{-1})$ is a complex valued analytic function for every $\ell \in X^*$ and we can apply well-known results from complex analysis:

Theorem 7.3. *For every $x \in X$, the spectrum $\sigma(x)$ is compact, nonempty and satisfies*

$$\sigma(x) \subseteq \{\alpha | \, |\alpha| \leq \|x\|\}. \tag{7.16}$$

Proof. Equation (7.15) already shows that $\rho(x)$ is open. Hence $\sigma(x)$ is closed. Moreover, $x - \alpha = -\alpha(e - \frac{1}{\alpha}x)$ together with Lemma 7.1 shows

$$(x-\alpha)^{-1} = -\frac{1}{\alpha} \sum_{n=0}^{\infty} \left(\frac{x}{\alpha}\right)^n, \quad |\alpha| > \|x\|, \tag{7.17}$$

which implies $\sigma(x) \subseteq \{\alpha | \, |\alpha| \leq \|x\|\}$ is bounded and thus compact. Moreover, taking norms shows

$$\|(x-\alpha)^{-1}\| \leq \frac{1}{|\alpha| - \|x\|}, \quad |\alpha| > \|x\|, \tag{7.18}$$

which implies $(x-\alpha)^{-1} \to 0$ as $\alpha \to \infty$. In particular, if $\sigma(x)$ is empty, then $\ell((x-\alpha)^{-1})$ is an entire analytic function which vanishes at infinity. By Liouville's theorem we must have $\ell((x-\alpha)^{-1}) = 0$ in this case, and so $(x-\alpha)^{-1} = 0$, which is impossible. □

As another simple consequence we obtain:

Theorem 7.4. *Suppose X is a Banach algebra in which every element except 0 is invertible. Then X is isomorphic to $\mathbb{C}$.*

Proof. Pick $x \in X$ and $\alpha \in \sigma(x)$, then $x-\alpha$ is not invertible and hence $x-\alpha=0$, that is $x=\alpha$. Thus every element is a multiple of the identity. □

Theorem 7.5 (Spectral mapping)**.** *For any polynomial p and $x \in X$ we have*

$$\sigma(p(x)) = p(\sigma(x)). \tag{7.19}$$

Proof. Fix $\alpha_0 \in \mathbb{C}$ and observe

$$p(x) - p(\alpha_0) = (x-\alpha_0) q_0(x). \tag{7.20}$$

If $p(\alpha_0) \notin \sigma(p(x))$ we have

$$(x-\alpha_0)^{-1} = q_0(x)((x-\alpha_0)q_0(x))^{-1} = ((x-\alpha_0)q_0(x))^{-1} q_0(x) \tag{7.21}$$

(check this — since $q_0(x)$ commutes with $(x-\alpha_0)q_0(x)$ it also commutes with its inverse). Hence $\alpha_0 \notin \sigma(x)$.

Conversely, let $\alpha_0 \in \sigma(p(x))$, then

$$p(x) - \alpha_0 = a(x-\lambda_1)\cdots(x-\lambda_n) \tag{7.22}$$

and at least one $\lambda_j \in \sigma(x)$ since otherwise the right hand side would be invertible. But then $p(\lambda_j) = \alpha_0$, that is, $\alpha_0 \in p(\sigma(x))$. □

Next let us look at the convergence radius of the **Neumann series** for the resolvent

$$(x-\alpha)^{-1} = -\frac{1}{\alpha}\sum_{n=0}^{\infty}\left(\frac{x}{\alpha}\right)^n \tag{7.23}$$

encountered in the proof of Theorem 7.3 (which is just the Laurent expansion around infinity).

The number

$$r(x) = \sup_{\alpha\in\sigma(x)} |\alpha| \tag{7.24}$$

is called the **spectral radius** of x. Note that by (7.16) we have

$$r(x) \le \|x\|. \tag{7.25}$$

Theorem 7.6. *The spectral radius satisfies*

$$r(x) = \inf_{n\in\mathbb{N}} \|x^n\|^{1/n} = \lim_{n\to\infty} \|x^n\|^{1/n}. \tag{7.26}$$

Proof. By spectral mapping we have $r(x)^n = r(x^n) \le \|x^n\|$ and hence

$$r(x) \le \inf \|x^n\|^{1/n}. \tag{7.27}$$

Conversely, fix $\ell \in X^*$, and consider

$$\ell((x-\alpha)^{-1}) = -\frac{1}{\alpha}\sum_{n=0}^{\infty}\frac{1}{\alpha^n}\ell(x^n). \tag{7.28}$$

Then $\ell((x-\alpha)^{-1})$ is analytic in $|\alpha| > r(x)$ and hence (7.28) converges absolutely for $|\alpha| > r(x)$ by a well-known result from complex analysis. Hence for fixed α with $|\alpha| > r(x)$, $\ell(x^n/\alpha^n)$ converges to zero for any $\ell \in X^*$. Since any weakly convergent sequence is bounded we have

$$\frac{\|x^n\|}{|\alpha|^n} \le C(\alpha) \tag{7.29}$$

and thus

$$\limsup_{n\to\infty} \|x^n\|^{1/n} \le \limsup_{n\to\infty} C(\alpha)^{1/n}|\alpha| = |\alpha|. \tag{7.30}$$

Since this holds for any $|\alpha| > r(x)$ we have

$$r(x) \le \inf \|x^n\|^{1/n} \le \liminf_{n\to\infty} \|x^n\|^{1/n} \le \limsup_{n\to\infty} \|x^n\|^{1/n} \le r(x), \tag{7.31}$$

which finishes the proof. □

To end this section let us look at two examples illustrating these ideas.

Example. Let $X = \mathrm{GL}(2)$ be the space of two by two matrices and consider

$$x = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}. \tag{7.32}$$

Then $x^2 = 0$ and consequently $r(x) = 0$. This is not surprising, since x has the only eigenvalue 0. The same is true for any nilpotent matrix. ⋄

Example. Consider the linear Volterra integral operator

$$K(x)(t) = \int_0^t k(t,s)x(s)ds, \qquad x \in C([0,1]), \tag{7.33}$$

then, using induction, it is not hard to verify

$$|K^n(x)(t)| \le \frac{\|k\|_\infty^n t^n}{n!}\|x\|_\infty \tag{7.34}$$

and thus $r(K) = 0$. Hence for every $\lambda \in \mathbb{C}$ and every $y \in C(I)$, the equation

$$x - \lambda K\, x = y \tag{7.35}$$

has a unique solution given by

$$x = (\mathbb{I} - \lambda K)^{-1} y = \sum_{n=0}^{\infty} \lambda^n K^n\, y. \tag{7.36}$$

⋄

Problem 7.1. *Show that* $\|AB\| \le \|A\|\|B\|$ *for every* $A, B \in \mathfrak{L}(X)$.

Problem 7.2. *Show that the multiplication in a Banach algebra* X *is continuous:* $x_n \to x$ *and* $y_n \to y$ *implies* $x_n y_n \to xy$.

Problem 7.3. *Show that* $L^1(\mathbb{R})$ *with convolution as multiplication is a commutative Banach algebra without identity (Hint: Problem 5.8).*

7.2. The C^* algebra of operators and the spectral theorem

We start by introducing a conjugation for operators on a Hilbert space $\mathfrak{H}$. Let $A = \mathfrak{L}(\mathfrak{H})$, then the **adjoint operator** is defined via

$$\langle f, A^* g\rangle = \langle Af, g\rangle \tag{7.37}$$

(compare Corollary 2.9).

Example. If $\mathfrak{H} = \mathbb{C}^n$ and $A = (a_{jk})_{1\le j,k\le n}$, then $A^* = (a_{kj}^*)_{1\le j,k\le n}$. ⋄

Lemma 7.7. *Let* $A, B \in \mathfrak{L}(\mathfrak{H})$, *then*

(i) $(A+B)^* = A^* + B^*$, $\quad (\alpha A)^* = \alpha^* A^*$,

(ii) $A^{**} = A$,

(iii) $(AB)^* = B^* A^*$,

(iv) $\|A\|^2 = \|A^* A\|$.

Proof. (i) is obvious. (ii) follows from $\langle f, A^{**}g\rangle = \langle A^* f, g\rangle = \langle f, Ag\rangle$. (iii) follows from $\langle f, (AB)g\rangle = \langle A^* f, Bg\rangle = \langle B^* A^* f, g\rangle$. (iv) follows from

$$\begin{aligned}\|A^*A\| &= \sup_{\|f\|=\|g\|=1} |\langle f, A^*Ag\rangle| = \sup_{\|f\|=\|g\|=1} |\langle Af, Ag\rangle| \\ &= \sup_{\|f\|=1} \|Af\|^2 = \|A\|^2, \end{aligned} \tag{7.38}$$

where we have used $\|f\| = \sup_{\|g\|=1} |\langle g, f\rangle|$ (compare Theorem 1.17). □

In general, a Banach algebra X together with an **involution**

$$(x+y)^* = x^* + y^*, \quad (\alpha x)^* = \alpha^* x^*, \quad x^{**} = x, \quad (xy)^* = y^* x^*, \tag{7.39}$$

satisfying

$$\|x\|^2 = \|x^* x\| \tag{7.40}$$

is called a C^* **algebra**. Any subalgebra which is also closed under involution, is called a $*$-algebra. Note that (7.40) implies $\|x\|^2 = \|x^*x\| \le \|x\|\|x^*\|$ and hence $\|x\| \le \|x^*\|$. By (ii) we also have $\|x^*\| \le \|x^{**}\| = \|x\|$ and hence

$$\|x\| = \|x^*\|, \qquad \|x\|^2 = \|x^* x\| = \|x x^*\|. \tag{7.41}$$

Example. The continuous function $C(I)$ together with complex conjugation form a commutative C^* algebra. ◇

If X has an identity e, we clearly have $e^* = e$ and $(x^{-1})^* = (x^*)^{-1}$ (show this). We will always assume that we have an identity. In particular,

$$\sigma(x^*) = \sigma(x)^*. \tag{7.42}$$

If X is a C^* algebra, then $x \in X$ is called **normal** if $x^* x = x x^*$, **self-adjoint** if $x^* = x$, and **unitary** if $x^* = x^{-1}$. Clearly both self-adjoint and unitary elements are normal.

Lemma 7.8. *If $x \in X$ is normal, then $\|x^2\| = \|x\|^2$ and $r(x) = \|x\|$.*

Proof. Using (7.40) twice we have

$$\|x^2\| = \|(x^2)^*(x^2)\|^{1/2} = \|(xx^*)^*(xx^*)\|^{1/2} = \|x^* x\| = \|x\|^2 \tag{7.43}$$

and hence $r(x) = \lim_{k\to\infty} \|x^{2^k}\|^{1/2^k} = \|x\|$. □

Lemma 7.9. *If x is self-adjoint, then $\sigma(x) \subseteq \mathbb{R}$.*

Proof. Suppose $\alpha + \mathrm{i}\beta \in \sigma(x)$. Then

$$\alpha^2 + (\beta+\lambda)^2 \le \|x + \mathrm{i}\lambda\|^2 = \|(x+\mathrm{i}\lambda)(x-\mathrm{i}\lambda)\| = \|(x^2+\lambda^2)\| \le \|x\|^2 + \lambda^2 \tag{7.44}$$

and hence $\alpha^2 + \beta^2 + 2\beta\lambda \le \|x\|^2$ which gives a contradiction if we let $|\lambda| \to \infty$ unless $\beta = 0$. □

Given $x \in X$ we can consider the C^* algebra $C^*(x)$ (with identity) generated by x. If x is normal, $C^*(x)$ is commutative and isomorphic to $C(\sigma(x))$ (the continuous functions on the spectrum).

Theorem 7.10 (Spectral theorem)**.** *If X is a C^* algebra and x is self-adjoint, then there is an isometric isomorphism $\Phi : C(\sigma(x)) \to C^*(x)$ such that $f(t) = t$ maps to $\Phi(t) = x$ and $f(t) = 1$ maps to $\Phi(1) = e$.*

Moreover, for every $f \in C(\sigma(x))$ we have

$$\sigma(f(x)) = f(\sigma(x)), \tag{7.45}$$

where $f(x) = \Phi(f(t))$.

Proof. First of all, Φ is well-defined for polynomials. Moreover, by spectral mapping we have

$$\|p(x)\| = r(p(x)) = \sup_{\alpha\in\sigma(p(x))} |\alpha| = \sup_{\alpha\in\sigma(x)} |p(\alpha)| = \|p\|_\infty \tag{7.46}$$

for any polynomial p. Hence Φ is isometric and uniquely extends to a map on all of $C(\sigma(x))$ since the polynomials are dense by the Stone–Weierstraß theorem (see the next section). □

In particular this last theorem tells us that we have a functional calculus for self-adjoint operators, that is, if $A \in \mathfrak{L}(\mathfrak{H})$ is self-adjoint, then $f(A)$ is well defined for every $f \in C(\sigma(A))$. If f is given by a power series, $f(A)$ defined via Φ coincides with $f(A)$ defined via its power series. Using the Riesz representation theorem we get another formulation in terms of spectral measures:

Theorem 7.11. *Let $\mathfrak{H}$ be a Hilbert space, and let $A \in \mathfrak{L}(\mathfrak{H})$ be self-adjoint. For every $u, v \in \mathfrak{H}$ there is a corresponding complex Borel measure $\mu_{u,v}$ (the* **spectral measure***) such that*

$$\langle u, f(A)v\rangle = \int_{\sigma(A)} f(t) d\mu_{u,v}(t), \qquad f \in C(\sigma(A)). \tag{7.47}$$

We have

$$\mu_{u,v_1+v_2} = \mu_{u,v_1} + \mu_{u,v_2}, \quad \mu_{u,\alpha v} = \alpha\mu_{u,v}, \quad \mu_{v,u} = \mu_{u,v}^* \tag{7.48}$$

and $|\mu_{u,v}|(\sigma(A)) \le \|u\|\|v\|$. Furthermore, $\mu_u = \mu_{u,u}$ is a positive Borel measure with $\mu_u(\sigma(A)) = \|u\|^2$.

Proof. Consider the continuous functions on $I = [-\|A\|, \|A\|]$ and note that any $f \in C(I)$ gives rise to some $f \in C(\sigma(A))$ by restricting its domain. Clearly $\ell_{u,v}(f) = \langle u, f(A)v\rangle$ is a bounded linear functional and the existence of a corresponding measure $\mu_{u,v}$ with $|\mu_{u,v}|(I) = \|\ell_{u,v}\| \le \|u\|\|v\|$ follows from Theorem 6.12. Since $\ell_{u,v}(f)$ depends only on the value of f on $\sigma(A) \subseteq I$, $\mu_{u,v}$ is supported on $\sigma(A)$.

Moreover, if $f \ge 0$ we have $\ell_u(f) = \langle u, f(A)u\rangle = \langle f(A)^{1/2}u, f(A)^{1/2}u\rangle = \|f(A)^{1/2}u\|^2 \ge 0$ and hence ℓ_u is positive and the corresponding measure μ_u is positive. The rest follows from the properties of the scalar product. □

It is often convenient to regard $\mu_{u,v}$ as a complex measure on $\mathbb{R}$ by using $\mu_{u,v}(\Omega) = \mu_{u,v}(\Omega \cap \sigma(A))$. If we do this, we can also consider f as a function on $\mathbb{R}$. However, note that $f(A)$ depends only on the values of f on $\sigma(A)$!

Note that the last theorem can be used to define $f(A)$ for any bounded measurable function $f \in B(\sigma(A))$ via Corollary 2.9 and extend the functional calculus from continuous to measurable functions:

Theorem 7.12 (Spectral theorem)**.** *If $\mathfrak{H}$ is a Hilbert space and $A \in \mathfrak{L}(\mathfrak{H})$ is self-adjoint, then there is an homomorphism $\Phi : B(\sigma(x)) \to \mathfrak{L}(\mathfrak{H})$ given by*

$$\langle u, f(A)v\rangle = \int_{\sigma(A)} f(t) d\mu_{u,v}(t), \qquad f \in B(\sigma(A)). \tag{7.49}$$

Moreover, if $f_n(t) \to f(t)$ pointwise and $\sup_n \|f_n\|_\infty$ is bounded, then $f_n(A)u \to f(A)u$ for every $u \in \mathfrak{H}$.

Proof. The map Φ is well-defined linear operator by Corollary 2.9 since we have

$$\Big|\int_{\sigma(A)} f(t) d\mu_{u,v}(t)\Big| \le \|f\|_\infty |\mu_{u,v}|(\sigma(A)) \le \|f\|_\infty \|u\|\|v\| \tag{7.50}$$

and (7.48). Next, observe that $\Phi(f)^* = \Phi(f^*)$ and $\Phi(fg) = \Phi(f)\Phi(g)$ holds at least for continuous functions. To obtain it for arbitrary bounded functions, choose a (bounded) sequence f_n converging to f in $L^2(\sigma(A), d\mu_u)$ and observe

$$\|(f_n(A) - f(A))u\|^2 = \int |f_n(t) - f(t)|^2 d\mu_u(t) \tag{7.51}$$

(use $\|h(A)u\|^2 = \langle h(A)u, h(A)u\rangle = \langle u, h(A)^* h(A)u\rangle$). Thus $f_n(A)u \to f(A)u$ and for bounded g we also have that $(gf_n)(A)u \to (gf)(A)u$ and $g(A)f_n(A)u \to g(A)f(A)u$. This establishes the case where f is bounded and g is continuous. Similarly, approximating g removes the continuity requirement from g.

The last claim follows since $f_n \to f$ in L^2 by dominated convergence in this case. □

In particular, given a self-adjoint operator A we can define the **spectral projections**

$$P_A(\Omega) = \chi_\Omega(A), \qquad \Omega \in \mathfrak{B}(\mathbb{R}). \tag{7.52}$$

They are **orthogonal projections**, that is $P^2 = P$ and $P^* = P$.

Lemma 7.13. *Suppose P is an orthogonal projection, then $\mathfrak{H}$ decomposes in an orthogonal sum*

$$\mathfrak{H} = \mathrm{Ker}(P) \oplus \mathrm{Ran}(P) \tag{7.53}$$

and $\mathrm{Ker}(P) = (\mathbb{I} - P)\mathfrak{H}$, $\mathrm{Ran}(P) = P\mathfrak{H}$.

Proof. Clearly, every $u \in \mathfrak{H}$ can be written as $u = (\mathbb{I} - P)u + Pu$ and

$$\langle (\mathbb{I} - P)u, Pu\rangle = \langle P(\mathbb{I} - P)u, u\rangle = \langle (P - P^2)u, u\rangle = 0 \tag{7.54}$$

shows $\mathfrak{H} = (\mathbb{I} - P)\mathfrak{H} \oplus P\mathfrak{H}$. Moreover, $P(\mathbb{I} - P)u = 0$ shows $(\mathbb{I} - P)\mathfrak{H} \subseteq \mathrm{Ker}(P)$ and if $u \in \mathrm{Ker}(P)$ then $u = (\mathbb{I} - P)u \in (\mathbb{I} - P)\mathfrak{H}$ shows $\mathrm{Ker}(P) \subseteq (\mathbb{I} - P)\mathfrak{H}$. □

In addition, the spectral projections satisfy

$$P_A(\mathbb{R}) = \mathbb{I}, \qquad P_A(\bigcup_{n=1}^\infty \Omega_n)u = \sum_{n=1}^\infty P_A(\Omega_n)u, \quad u \in \mathfrak{H}. \tag{7.55}$$

Such a family of projections is called a **projection valued measure** and

$$P_A(t) = P_A((-\infty, t]) \tag{7.56}$$

is called a **resolution of the identity**. Note that we have

$$\mu_{u,v}(\Omega) = \langle u, P_A(\Omega)v\rangle. \tag{7.57}$$

Using them we can define an operator valued integral as usual such that

$$A = \int t\, dP_A(t). \tag{7.58}$$

In particular, if $P_A(\{\alpha\}) \neq 0$, then α is an eigenvalue and $\mathrm{Ran}(P_A(\{\alpha\}))$ is the corresponding eigenspace since

$$AP_A(\{\alpha\}) = \alpha P_A(\{\alpha\}). \tag{7.59}$$

The fact that eigenspaces to different eigenvalues are orthogonal now generalizes to

Lemma 7.14. *Suppose* $\Omega_1 \cap \Omega_2 = \emptyset$, *then*

$$\mathrm{Ran}(P_A(\Omega_1)) \perp \mathrm{Ran}(P_A(\Omega_2)). \tag{7.60}$$

Proof. Clearly $\chi_{\Omega_1}\chi_{\Omega_2} = \chi_{\Omega_1\cap\Omega_2}$ and hence

$$P_A(\Omega_1)P_A(\Omega_2) = P_A(\Omega_1 \cap \Omega_2). \tag{7.61}$$

Now if $\Omega_1 \cap \Omega_2 = \emptyset$, then

$$\langle P_A(\Omega_1)u, P_A(\Omega_2)v\rangle = \langle u, P_A(\Omega_1)P_A(\Omega_2)v\rangle = \langle u, P_A(\emptyset)v\rangle = 0, \tag{7.62}$$

which shows that the ranges are orthogonal to each other. □

Example. Let $A \in \mathrm{GL}(n)$ be some symmetric matrix and let $\alpha_1, \dots, \alpha_m$ be its (distinct) eigenvalues. Then

$$A = \sum_{j=1}^{m} \alpha_j P_A(\{\alpha_j\}), \tag{7.63}$$

where $P_A(\{\alpha_j\})$ is the projection onto the eigenspace corresponding to the eigenvalue α_j. ◇

Problem 7.4. *Let* $A \in \mathfrak{L}(\mathfrak{H})$. *Show that* A *is normal if and only if*

$$\|Au\| = \|A^*u\|, \quad \forall u \in \mathfrak{H}. \tag{7.64}$$

(Hint: Problem 1.6)

Problem 7.5. *Show that an orthogonal projection* $P \neq 0$ *has norm one.*

7.3. The Stone–Weierstraß theorem

In the last section we have seen that the C^* algebra of continuous functions $C(K)$ over some compact set plays a crucial role. Hence it is important to be able to identify dense sets:

Theorem 7.15 (Stone–Weierstraß, real version)**.** *Suppose K is a compact set and let $C(K,\mathbb{R})$ be the Banach algebra of continuous functions (with the sup norm).*

If $F \subset C(K,\mathbb{R})$ contains the identity 1 and separates points (i.e., for every $x_1 \neq x_2$ there is some function $f \in F$ such that $f(x_1) \neq f(x_2)$), then the algebra generated by F is dense.

Proof. Denote by A the algebra generated by F. Note that if $f \in \overline{A}$, we have $|f| \in \overline{A}$: By the Weierstraß approximation theorem (Theorem 1.14) there is a polynomial $p_n(t)$ such that $\big||f| - p_n(t)\big| < \frac{1}{n}$ and hence $p_n(f) \to |f|$.

In particular, if f, g in $\overline{A}$, we also have

$$\max\{f,g\} = \frac{(f+g)+|f-g|}{2}, \quad \min\{f,g\} = \frac{(f+g)-|f-g|}{2} \tag{7.65}$$

in $\overline{A}$.

Now fix $f \in C(K,\mathbb{R})$. We need to find some $f_\varepsilon \in \overline{A}$ with $\|f - f_\varepsilon\|_\infty < \varepsilon$.

First of all, since A separates points, observe that for given $y, z \in K$ there is a function $f_{y,z} \in A$ such that $f_{y,z}(y) = f(y)$ and $f_{y,z}(z) = f(z)$ (show this). Next, for every $y \in K$ there is a neighborhood $U(y)$ such that

$$f_{y,z}(x) > f(x) - \varepsilon, \quad x \in U(y) \tag{7.66}$$

and since K is compact, finitely many, say $U(y_1), \dots U(y_j)$, cover K. Then

$$f_z = \max\{f_{y_1,z}, \dots, f_{y_j,z}\} \in \overline{A} \tag{7.67}$$

and satisfies $f_z > f - \varepsilon$ by construction. Since $f_z(z) = f(z)$ for every $z \in K$ there is a neighborhood $V(z)$ such that

$$f_z(x) < f(x) + \varepsilon, \quad x \in V(z) \tag{7.68}$$

and a corresponding finite cover $V(z_1), \dots V(z_k)$. Now

$$f_\varepsilon = \min\{f_{z_1}, \dots, f_{z_k}\} \in \overline{A} \tag{7.69}$$

satisfies $f_\varepsilon < f + \varepsilon$. Since $f - \varepsilon < f_{z_l} < f_\varepsilon$, we have found a required function. □

Theorem 7.16 (Stone–Weierstraß)**.** *Suppose K is a compact set and let $C(K)$ be the C^* algebra of continuous functions (with the sup norm).*

If $F \subset C(K)$ contains the identity 1 and separates points, then the $*$-algebra generated by F is dense.

Proof. Just observe that $\tilde{F} = \{\mathrm{Re}(f), \mathrm{Im}(f) | f \in F\}$ satisfies the assumption of the real version. Hence any real-valued continuous functions can be approximated by elements from $\tilde{F}$, in particular this holds for the real and imaginary part for any given complex-valued function. □

Note that the additional requirement of being a $*$-algebra, that is, closed under complex conjugation, is crucial: The functions which are holomorphic on the unit ball and continuous on the boundary separate points, but they are not dense (since the uniform limit of holomorphic functions is again holomorphic).

Corollary 7.17. *Suppose K is a compact set and let $C(K)$ be the C^* algebra of continuous functions (with the sup norm).*

If $F \subset C(K)$ separates points, then the closure of the $*$-algebra generated by F is either $C(K)$ or $\{f \in C(K) | f(t_0) = 0\}$ for some $t_0 \in K$.

Proof. There are two possibilities, either all $f \in F$ vanish at one point $t_0 \in K$ (there can be at most one such point since F separates points) or there is no such point. If there is no such point we can proceed as in the proof of the Stone–Weierstraß theorem to show that the identity can be approximated by elements in $\overline{A}$ (note that to show $|f| \in \overline{A}$ if $f \in \overline{A}$ we do not need the identity, since p_n can be chosen to contain no constant term). If there is such a t_0, the identity is clearly missing from $\overline{A}$. However, adding the identity to $\overline{A}$ we get $\overline{A} + \mathbb{C} = C(K)$ and it is easy to see that $\overline{A} = \{f \in C(K) | f(t_0) = 0\}$. □

Problem 7.6. *Show that the $*$-algebra generated by $f_z(t) = \frac{1}{t-z}$, $z \in \mathbb{C}$, is dense in the C^* algebra $C_\infty(\mathbb{R})$ of continuous functions vanishing at infinity.*

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Glossary of notations

$B(X)$	. . . Banach space of bounded measurable functions.
$\mathbb{C}$	. . . the set of complex numbers
$\mathfrak{C}(\mathfrak{H})$	. . . set of compact operators, 32.
$C(U)$	. . . set of continuous functions from U to $\mathbb{C}$.
$C(U,V)$	. . . set of continuous functions from U to V.
$C_c^\infty(U,V)$	. . . set of compactly supported smooth functions
$\chi_\Omega(.)$	. . . characteristic function of the set Ω
dim	. . . dimension of a linear space
$\mathrm{dist}(x,Y)$	$= \inf_{y\in Y} \|x-y\|$, distance between x and Y
$\mathfrak{D}(.)$	. . . domain of an operator
e	. . . exponential function, $\mathrm{e}^z = \exp(z)$
hull(.)	. . . convex hull
$\mathfrak{H}$	. . . a Hilbert space
i	. . . complex unity, $\mathrm{i}^2 = -1$
Im(.)	. . . imaginary part of a complex number
inf	. . . infimum
$\mathrm{Ker}(A)$	. . . kernel of an operator A, 22
$\mathfrak{L}(X,Y)$	. . . set of all bounded linear operators from X to Y, 23
$\mathfrak{L}(X)$	$= \mathfrak{L}(X,X)$
$L^p(X,d\mu)$	. . . Lebesgue space of p integrable functions, 60
$L^\infty(X,d\mu)$	. . . Lebesgue space of bounded functions, 60
max	. . . maximum
$\mathbb{N}$	. . . the set of positive integers
$\mathbb{N}_0$	$= \mathbb{N} \cup \{0\}$
$\mathbb{Q}$	. . . the set of rational numbers

$\mathbb{R}$	... the set of real numbers
$\mathrm{Ran}(A)$	... range of an operator A, 23
$\mathrm{Re}(.)$	... real part of a complex number
sup	... supremum
supp	... support of a function
$\mathrm{span}(M)$	... set of finite linear combinations from M, 15
$\mathbb{Z}$	... the set of integers
$\mathbb{I}$	... identity operator
$\sqrt{z}$	... square root of z with branch cut along $(-\infty, 0)$
z^*	... complex conjugation
$\Vert.\Vert$	... norm
$\Vert.\Vert_p$	... norm in the Banach space L^p
$\langle ., .. \rangle$	... scalar product in $\mathfrak{H}$
$\oplus$	... orthogonal sum of linear spaces or operators, 31
∂	... gradient
∂_α	... partial derivative
$M^\perp$	... orthogonal complement, 29
(λ_1, λ_2)	$= \{\lambda \in \mathbb{R} \,\vert\, \lambda_1 < \lambda < \lambda_2\}$, open interval
$[\lambda_1, \lambda_2]$	$= \{\lambda \in \mathbb{R} \,\vert\, \lambda_1 \leq \lambda \leq \lambda_2\}$, closed interval

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