# MAT4400: Notes on Linear analysis (Proofs excluded)

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### 1 $\sigma$ -Algebras (Ch. 3 in [Schilling(2017)])

**Definition 1.1** ( $\sigma$ -Algebra). A family  $\mathscr{A}$  of subsets of X with:

- (i)  $X \in \mathcal{A}$ ,
- (ii)  $A \in \mathcal{A} \Rightarrow A^c \in \mathcal{A}$ ,
- (iii)  $(A_n)_{n\in\mathbb{N}}\in\mathscr{A}\Rightarrow\bigcup_{n\in\mathbb{N}}$

Theorem 1.2 (and Definition).

- (i) The intersection of arbitrarily many  $\sigma$ -algebras in X is againg a  $\sigma$ -algebra in X.
- (ii) For every system of sets  $p \subset \mathcal{P}(X)$  there exists a smallest $\sigma$ -algebra containingp. This is the  $\sigma$ -algebra generated by p, denoted  $\sigma(p)$ , and  $\sigma(p)$  is called its generator.

**Definition 1.3** (Borel). The  $\sigma$ -algebra  $\sigma(\mathcal{O})$  generated by the open sets  $\mathcal{O} = \mathcal{O}_{\mathbb{R}^n}$  of  $\mathbb{R}^n$  is called **Borel**  $\sigma$ -algebra, and its members are called **Borel sets** or **Borel measurable sets**.

# 3 Uniqueness of Measures (Ch. 5 in [Schilling(2017)])

**Lemma 3.1.** A Dynkin system D is a  $\sigma$ -algebra iff it is stable under finite intersections, i.e.  $A, B \in D \Rightarrow A \cap B \in D$ .

**Theorem 3.2** (Dynkin). Assume X is a set, S is a collection of subsets of X closed under finite intersections, that is, if  $A, B \in S \Rightarrow A \cap B \in S$ . Then  $D(S) = \sigma(S)$ .

**Theorem 3.3** (uniqueness of measures). Let (X, B) be a measurable space, and  $S \subset P(X)$  be the generator of B, i.e.  $B = \sigma(S)$ . If S satisfies the following conditions:

- 1. S is stable under finite intersections ( $\cap$ -stable), i.e.  $A, C \in S \Rightarrow A \cap C \in S$ .
- 2. There exists an exhausting sequence  $(G_n)_{N\in\mathbb{N}}\subset with\ G_N\uparrow X$ . Assume also that there are two measures  $\mu,\nu$  satisfying:
- 3.  $\mu(A) = \nu(A), \forall A \in S$ .
- 4.  $\mu(G_n) = \nu(G_n) < \infty$ .

Then  $\mu = \nu$ .

Proof (outline). Define

$$D_n := \{ A \in B : \mu(G_n \cap A) = \nu(G_n \cap A) \ (< \infty) \},$$

and show that it is a Dynkin system. Then, use the fact that S is  $\cap$ -stable and Theorem 3.2 to argue that  $D(S) = \sigma(S)... \rightarrow ... B = D_n$ .

### 4 Existence of Measures (Ch. 6 in [Schilling(2017)])

**Theorem 4.1** (Carathéodory). Let  $S \subset P(X)$  be a semi-ring and  $\mu: S \to [0, \infty)$  a premeasure. Then  $\mu$  has an extension to a measure  $\mu^*$  on  $\sigma(S)$ , i.e. that  $\mu(s) = \mu^*(s)$ ,  $\forall s \in \sigma(S)$ . Also, if S contains an exhausting sequence,

Also, if S contains an exhausting sequence,  $S_n \uparrow X$ , s.t.  $\mu(S_n) < \infty$ , then the extension is unique.

Outline of proof: Firstly, let us define an outer measure.

**Definition 4.2** (Outer measure). An outer measure is a function  $\mu^*: P(X) \to [0, \infty)$  with the following properties:

- 1.  $\mu^*(\emptyset) = 0$ ,
- $2. \ A \subset B \Rightarrow u^*(A) \leq \mu^*(B),$

3. 
$$\mu^* \left( \bigcup_{n \in \mathbb{N}} A_n \right) \le \sum_{n \in \mathbb{N}} \mu^* (A_n),$$

and define for each  $A \subset X$  the family of countable *S-coverings*:

$$C(A) := \left\{ (S_n)_{n \in \mathbb{N}} \subset S : \bigcup_{n \in \mathbb{N}} S_n \supset A \right\},\,$$

and the set function

$$\mu^*(A) := \inf \left\{ \sum_{n \in \mathbb{N}} \mu(S_n) : (S_n)_{n \in \mathbb{N}} \in C(A) \right\}.$$

Step 1: Claim:  $\mu^*(A)$  is an outer measure.

Proof.

- 1.  $C(\emptyset) = \{\text{any sequence in } S \text{ containing } \emptyset\}$  $\Rightarrow \mu^*(\emptyset) = 0.$
- 2. Assume  $A \subset B$ . Then  $C(A) \subset C(B)$  $\Rightarrow \mu^*(A) < \mu^*B$ .
- 3. If  $\mu^*(A_n) = \infty$  for some n, then there is nothing to prove. Thus, assume  $\mu^*(A_n) < \infty \ \forall n$ . Fix  $\epsilon > 0$ , and for every n choose  $A_{n_k} \in S$  s.t.

$$A_n \subset \bigcup_{k \in \mathbb{N}} A_{n_k},$$
$$\sum_{k \in \mathbb{N}} \mu^*(A_{n_k}) < \mu^*(A_n) + \frac{\epsilon}{2^n}.$$

Then

$$\bigcup_{n\in\mathbb{N}} A_n \subset \bigcup_{k\in\mathbb{N}} \bigcup_{n\in\mathbb{N}} A_{n_k},$$

so

$$\mu^* \left( \bigcup_{n \in \mathbb{N}} A_n \right) \le \sum_{n \in \mathbb{N}} \sum_{k \in \mathbb{N}} \mu \left( A_{n_k} \right)$$
$$< \sum_{n \in \mathbb{N}} \left( \mu^* (A_n) + \frac{\epsilon}{2^n} \right)$$
$$= \sum_{n \in \mathbb{N}} \mu^* (A_n) + \epsilon.$$

As  $\epsilon$  was arbitrarily, we get that

$$\mu^* \left( \bigcup_{n \in \mathbb{N}} A_n \right) \le \sum_{n \in \mathbb{N}} \mu^* (A_n),$$

so  $\mu^*$  fulfills all the conditions for being an outer measure.

**Step 2:** Showing that  $\mu^*$  extends  $\mu$ , i.e.  $\mu^*(s) = \mu(s) \ \forall s \in S$ .

Step 3: Define  $\mu^*$ -measurable sets

$$\Sigma^* := \left\{ A \subset X : \mu^*(Q) = \mu^*(Q \cap A) + \mu^*(Q \backslash A) \right.$$
 
$$\forall \ Q \subset X \right\}$$

**Step 4:** Show that  $\mu|_{\Sigma^*}$  is a measure. In particular,  $\mu|_{\sigma(S)}$  is a measure which extends  $\mu$ .

### 5 Measurable Mappings (Ch. 7 in [Schilling(2017)])

We consider maps  $T: X \to X'$  between two measurable spaces  $(X, \mathcal{A})$  and  $(X', \mathcal{A}')$  which respects the measurable structurs, the  $\sigma$ -algbras on X and X'. These maps are useful as we can transport a measure  $\mu$ , defined on  $(X, \mathcal{A})$ , to  $(X', \mathcal{A}')$ .

**Definition 5.1.** Let  $(X, \mathcal{A}), (X', \mathcal{A}')$  b measurable spaces. A map  $T: X \to X'$  is called  $\mathcal{A}/\mathcal{A}'$ -measurable if the pre-imag of every measurable set is a measurable set:

$$T^{-1}(A') \in \mathcal{A}, \quad \forall A' \in \mathcal{A}'.$$

- A  $\mathcal{B}(\mathbb{R}^n)/\mathcal{B}(\mathbb{R}^m)$  measurable map is often called a Borel map.
- The notation  $T:(X,\mathcal{A})\to (X',\mathcal{A}')$  is often used to indicate measurability of the map T.

**Lemma 5.2.** Let (X, A), (x', A') be measurable spaces and let  $A' = \sigma(G')$ . Then  $T: X \to X'$  is A/A'-measurable iff  $T^{-1}(G') \subset A$ , i.e. if

$$T^{-1}(G') \in \mathcal{A}, \ \forall G' \in \mathcal{G}'.$$

**Theorem 5.3.** Let  $(X_i, \mathcal{A}_i)$ , i = 1, 2, 3, be measurable spaces and  $T: X_1 \to X_2$ ,  $S: X_2 \to X_3$  be  $\mathcal{A}_1/\mathcal{A}_2$  and  $\mathcal{A}_2/\mathcal{A}_3$ -measurable maps respectivly. Then  $S \circ T: X_1 \to X_3$  is  $\mathcal{A}_1/\mathcal{A}_3$ -measurable.

Corollary 5.4. Every continuous map betwen metric spaces is a Borel map.

**Definition 5.5.** (and lemma) Let  $(T_i)_{i \in I}$ ,  $T_I : X \to X_i$ , be arbitrarily many mappings from the same space X into measurable spaces

 $(X_i, \mathcal{A}_i)$ . The smallest  $\sigma$ -algebra on X that makes all  $T_i$  simultaneously measurable is

$$\sigma(T_i: i \in I) := \sigma\left(\bigcup_{i \in I} T_i^{-1}(\mathcal{A}_i)\right)$$

**Corollary 5.6.** A function  $f:(X,\mathcal{B})\to\mathbb{R}$  is measurable if  $f((a,+\infty))\in\mathcal{B}, \ \forall a\in\mathbb{R}.$ 

Corollary 5.7. Assume  $(X, \mathcal{B})$  is a measurable space, (Y, d) is a metric space, and  $(f_n : (X, \mathcal{B}) \to Y)_{n=1}^{\infty}$  is a sequence of measurable maps. Assume this sequence of images  $(f_n(x))_{n=1}^{\infty}$  is convergent in  $Y \ \forall x \in X$ . Define

$$f: X \to Y$$
, by  $f(x) = \lim_{n \to \infty} f_n(x)$ .

Then f is measurable.

**Theorem 5.8.** Let (X, A), (X', A') be measurable spaces and  $T: X \to X'$  be an A/A'-measurable map. For every measurable  $\mu$  on (X, A),

$$\mu'(A') := \mu(T^{-1}(A')), \quad A' \in \mathcal{A}',$$

defines a measure on (X', A').

**Definition 5.9.** The measure  $\mu'(\cdot)$  in the above theorem is called the push forward or image measure of  $\mu$  under T and it is denoted as  $T(\mu)(\cdot)$ ,  $T_{*\mu}(\cdot)$  or  $\mu \circ T^{-1}(\cdot)$ .

**Theorem 5.10.** If  $T \in \mathbb{R}^{n \times n}$  is an orthogonal matrix, then  $\lambda^n = T(\lambda^n)$ .

**Theorem 5.11.** Let  $S \in \mathbb{R}^{n \times n}$  be an invertible matrix. Then

$$S(\lambda^n) = |\det s^{-1}|\lambda^n = |\det S|^{-1}\lambda^n.$$

Corollary 5.12. Lebesgue measure is invariant under motions:  $\lambda^n = M(\lambda^n)$  for all motions M in  $\mathbb{R}^n$ . In particular, congruent sets have the same measure. Two sets of points are called congruent if, and only if, one can be transformed into the other by an isometry

### Measurable Functions (Ch. 8 in [Schilling(2017)])

A measurable function is a measurable map  $u: X \to \mathbb{R}$  from some measurable space  $(X, \mathscr{A})$  to  $(\mathbb{R}, \mathscr{B}(\mathbb{R}^1))$ . They play central roles in the theory of integration.

We recall that  $u:X\to\mathbb{R}$  is  $\mathscr{A}/\mathscr{B}(\mathbb{R}^1)$ -measurable if

$$u^{-1}(B) \in \mathscr{A}, \ \forall B \in \mathscr{B}(\mathbb{R}^1).$$

Moreover from a lemma from chapter 7, we actually only need to show that

 $u^{-1}(G) \in \mathscr{A}, \ \forall G \in \mathcal{G} \text{ where } \mathcal{G} \text{ generates } \mathscr{B}(\mathbb{R}^1).$ 

#### Proposition 5.13.

- 1 If  $f, g: (X, \mathcal{B}) \to \mathbb{C}$  are measurable, then the function f + g,  $f \cdot g$ , cf,  $(c \in \mathbb{C})$  are measurable.
- 2 If  $b: \mathbb{C} \to \mathbb{C}$  is Borel and  $b: (\mathbb{C}, \mathscr{B}) \to \mathbb{C}$  is measurable, then  $b \circ f$  is measurable.
- 3 If  $f(x) = \lim_{n \to \infty} f_n(x)$ ,  $x \in X$  and  $f_n$  are measurable, then f is measurable.
- 4 If  $X = \bigcup_{n=1}^{\infty} A_n$ ,  $(A_n \in \mathcal{B})$ ,  $f|_{A_n} : (A_n, \mathcal{B}_{A_n}) \to \mathbb{C}$  is measurable  $\forall n$ , then f is measurable.

**Definition 5.14.** Given a measurable space  $(X, \mathcal{B})$ , a measurable function  $f: (X, \mathcal{B}) \to \mathbb{C}$  is called simple if

$$f(x) = \sum_{k=1}^{N} c_k \mathbb{1}_{A_k}(x),$$

for some  $c_k \in \mathbb{C}$ ,  $A_k \in \mathcal{B}$ , where  $\mathbb{1}$  is the characteristic function,

$$\mathbb{1}_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A. \end{cases}$$

The representation of simple function is **not** unique. We denote the standard representation of f by

$$f(x) = \sum_{n=0}^{N} z_n \mathbb{1}_{B_n}(x),$$

for  $N \in \mathbb{N}$ ,  $z_n \in \mathbb{R}$ ,  $B_n \in \mathcal{A}$ , and

$$X = \bigcup_{n=1}^{N} B_n,$$

for  $B_n \cap B_m = \emptyset$ ,  $n \neq m$ . The set of simple functions is denoted  $\mathcal{E}(\mathscr{A})$  of  $\mathcal{E}$ .

**Definition 5.15.** Assume  $\mu$  is a measure on  $(X, \mathcal{B})$ . Given a *positive* simple function

$$f = \sum_{k=1}^{N} c_k \mathbb{1}_{A_k}, \quad (c_k \ge 0).$$

We define

$$\int_X f d\mu = \sum_{k=1}^n c_k \mu(A_k) \in [0, +\infty].$$

We also denote this by  $I_{\mu}(f)$ .

**Lemma 5.16.** This is well defined, that is,  $\int_x f d\mu$  does not depend on the presentation of the simple function f.

**Properties 5.17.** For every positive simple function

1 
$$\int_X cf d\mu = c \int_X f d\mu$$
, for only  $c \ge 0$ 

$$2 \int_X (f+g)d\mu = \int_X f d\mu + \int_X g d\mu.$$

**Corollary 5.18.** If  $f \ge g \ge 0$  are simple functions, then

$$\int_{\mathbf{Y}} f d\mu \ge \int_{\mathbf{Y}} g d\mu.$$

**Definition 5.19.** If  $f: X \to [0, +\infty)$  is measurable, then we define

$$\int_X f d\mu = \sup \left\{ \int_X g d\mu : f \ge g \ge 0, \ g \text{ is simple} \right\}$$

**Remark.** This means that any measurable function can be approximated by simple functions.

**Properties 5.20.** Measurable functions like this have the following properties

$$1 \int_X cf d\mu = c \int_X f d\mu, \quad \forall c \ge 0.$$

- 2 If  $f \geq g \geq 0$ , then  $\int_X f d\mu \geq \int_X g d\mu$  for any measurable g, f.
- 3 If  $f \ge 0$  is simple, then  $\int_X f d\mu$  is the same value as obtained before.

To advance in measure theory we consider measurable functions

$$f: X \to [0, +\infty].$$

Measurability is understood w.r.t the  $\sigma$ -algebra  $\mathcal{B}([0,+\infty])$  generated by  $\mathcal{B}([0,+\infty))$  and  $\{+\infty\}$ . In other words,  $A \subset [0,+\infty] \in B([0,+\infty])$  iff  $A \cap [0,+\infty) \in \mathcal{B}([0,+\infty))$ .

**Remark.** Hence  $f: X \to [0, +\infty]$  is measurable iff  $f^{-1}(A)$  is measurable  $\forall A \in \mathcal{B}([0, +\infty))$ .

**Definition 5.21.** For measurable functions  $f_X \to [0, +\infty]$ , we define

$$\int_X f d\mu = \sup \left\{ \int_x g d\mu \ : \ f \ge g \ge 0 \ : \ g \text{ is simple} \right\} \in [0, +\infty].$$

#### Theorem 5.22. Monotone convergence theorem

Assume  $(X, \mathcal{B}, \mu)$  is a measure space,  $(f)_{n=1}^{\infty}$  is an increasing sequence of measurable positive functions  $f_n: X \to [0, +\infty]$ . Define  $f(x) = \lim_{n \to \infty} f_n(x)$ . Then f is measurable and

$$\int_{Y} f d\mu = \lim_{n \to \infty} \int_{Y} f_n d\mu.$$

**Theorem 5.23.** Assume  $(X, \mathcal{B})$  is a measurable space and  $f: X \to [0, +\infty]$  is measurable. Then there are simple functions  $g_n$ , s.t.

$$0 \le g_1 \le g_2 \le \dots$$
,  $g_n(x) \to f(x)$ ,  $\forall x \in X$ .

Moreover, if f is bounded, we can choose  $g_n$  s.t. the convergence is uniform, that is,

$$\lim_{n \to \infty} \sup_{x \in X} |g_n(x) - f(x)| = 0.$$

### 6 Integration of measurable functions

(Ch. 9 in [Schilling(2017)])

Through this chapter  $(X, \mathscr{A}, \mu)$  will be some measure space. Recall that  $\mathcal{M}^+(\mathscr{A})$   $[\mathcal{M}^+_{\mathbb{R}}(\mathscr{A})]$  are the  $\mathscr{A}$ -measurable positive functions and  $\mathcal{E}(\mathscr{A})$   $[\mathcal{E}^+_{\mathbb{R}}(\mathscr{A})]$  are the positive and simple functions.

The fundamental idea of *Integration* is to measure the area between the graph of the function and the abscissa. For positive simple functions  $f \in \mathcal{E}^+(\mathscr{A})$  in standard representation, this is done easily

if 
$$f = \sum_{i=0}^{M} y_i \mathbb{1}_{A_i} \in \mathcal{E}^+(\mathscr{A})$$
 then  $\sum_{i=0}^{M} y_i \mu(A_i)$ 

would be the  $\mu$ -area enclosed by the graph and the abscissa. We note that the representation of f should not impact the integral of f.

**Lemma 6.1.** Let  $\sum_{i=0}^{M} y_i \mathbb{1}_{A_i} = \sum_{k=0}^{N} z_k \mathbb{1}_{B_k}$  be two standard representations of the same function  $f \in \mathcal{E}^+(\mathscr{A})$ . Then

$$\sum_{i=0}^{M} y_i \mu(A_i) = \sum_{k=0}^{N} z_k \mu(B_k).$$
 (2)

**Definition 6.2.** Let  $f = \sum_{i=0}^{M} y_i \mathbb{1}_{A_i} \in \mathcal{E}^+(\mathscr{A})$  be a simple function in standard representation. Then the number

$$I_{\mu}(f) = \sum_{i=0}^{M} y_i \mu(A_i) \in [0, \infty]$$
 (3)

(which is independent of the representation of f) is called the  $\mu$ -integral of f.

**Proposition 6.3.** Let  $f, g \in \mathcal{E}^+(\mathscr{A})$ . Then

- (i)  $I_{\mu}(\mathbb{1}_A) = \mu(A) \quad \forall A \in \mathscr{A}.$
- (ii)  $I_{\mu}(\lambda f) = \lambda I_{\mu}(f) \quad \forall \lambda \geq 0.$
- (iii)  $I_{\mu}(f+g) = I_{\mu}(f) + I_{\mu}(g)$ .
- (iv)  $f \leq g \Rightarrow I_{\mu}(f) \leq I_{\mu}(g)$ .

In theorem 8.8 we saw that we could for every  $u \in \mathcal{M}^+(\mathscr{A})$  write it as an increasing limit of simple functions. By corollary 8.10, the suprema of simple functions are again measurable, so that

$$u \in \mathcal{M}^+(\mathscr{A}) \Leftrightarrow u = \sup_{n \in \mathbb{N}} f_n, f \in \mathcal{E}^+(\mathscr{A}),$$
  
$$f_n \leq f_{n+1} \leq \dots$$

We will use this to "inscribe" simple functions (which we know how to integrate) below the graph of a positive measurable function u and exhaust the  $\mu$ -area below u.

**Definition 6.4.** Let  $(X, \mathscr{A}, \mu)$  be a measure space. The  $(\mu)$ -integral of a positive function  $u \in \mathcal{M}^{+}_{\mathbb{D}}(\mathscr{A})$  is given by

$$\int ud\mu = \sup \left\{ I_{\mu}(g) : g \le u, \ g \in \mathcal{E}^{+}(\mathscr{A}) \right\},\,$$

with  $\int u d\mu \in [0, +\infty]$ . If we need to emphasize the *integration variable*, we write  $\int u(x)\mu(dx)$ . The key observation is that the integral  $\int \dots d\mu$  extends  $I_{\mu}$ .

**Lemma 6.5.** For all  $f \in \mathcal{E}^+(\mathscr{A})$  we have  $\int f d\mu = I_{\mu}(f)$ .

The next theorem is one of many convergence theorems. It shows that we could have defined 4 using any increasing sequence  $f_n \uparrow u$  of simple functions  $f_n \in \mathcal{E}^+(\mathscr{A})$ .

**Theorem 6.6.** (<u>Beppo Levi</u>) Let  $(X, \mathcal{A}, \mu)$  be a measure space. For an increasing sequence of functions  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{M}^+_{\mathbb{R}}(\mathcal{A}),\ 0\leq u_n\leq u_{n+1}\leq\ldots$ , we have for the supremum  $u=\sup_{n\in\mathbb{N}}u_n\in\mathcal{M}^+_{\mathbb{R}}(\mathcal{A})$  and

$$\int \sup_{n \in \mathbb{N}} u_n d\mu = \sup_{n \in \mathbb{N}} \int u_n d\mu. \tag{5}$$

Note we can write  $\lim_{n\to\infty}$  instead of  $\sup_{n\in\mathbb{N}}$  as the supremum of an increasing sequence is its limit. Moreover, this theorem holds in  $[0, +\infty]$ , so the case  $+\infty = +\infty$  is possible.

Corollary 6.7. Let  $u \in \mathcal{M}_{\bar{\mathbb{R}}}^+(\mathscr{A})$ . Then

$$\int u d\mu = \lim_{n \to \infty} \int f_n d\mu$$

holds for every sequence  $(f_n)_{n\in\mathbb{N}}\subset\mathcal{E}^+(\mathscr{A})$  with  $\lim_{n\to\infty}f_n=u$ .

**Proposition 6.8.** (of integral) Let  $u, v \in \mathcal{M}^+_{\bar{\mathbb{R}}}(\mathscr{A})$ . Then

- (i)  $\int \mathbb{1}_A d\mu = \mu(A) \quad \forall A \in \mathscr{A}.$
- (ii)  $\int \alpha u d\mu = \alpha \int u d\mu \quad \forall \alpha \geq 0.$
- (iii)  $\int u + v d\mu = \int u d\mu + \int v d\mu$ .
- (iv)  $u < v \Rightarrow \int u d\mu < \int v d\mu$ .

Corollary 6.9. Let  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{M}_{\mathbb{R}}^+(\mathscr{A})$ . Then  $\sum_{n=1}^{\infty}u_n$  is measurable and we have

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

(including the possibility  $+\infty = +\infty$ .)

**Theorem 6.10.** (<u>Fatou</u>) Let  $(u_n)_{n\in\mathbb{N}}\subset \mathcal{M}^+_{\mathbb{R}}(\mathscr{A})$  be a sequence of positive measurable functions. Then  $u=\liminf_{n\to\infty}u_n$  is measurable and

$$\int \liminf_{n \to \infty} u_n d\mu = \liminf_{n \to \infty} \int u_n d\mu \qquad (6)$$

### 7 Integrals of Measurable Functions

(Ch. 10 in [Schilling(2017)])

We have defined our integral for positive measurable functions, i.e. functions in  $\mathcal{M}^+(\mathscr{A})$ . To extend our integral to not only functions in  $\mathcal{M}^+(\mathscr{A})$  we first notice that

$$u \in \mathcal{M}_{\overline{\mathbb{R}}}(\mathscr{A}) \Leftrightarrow u = u^+ - u^-, \ u^+, u^- \in \mathcal{M}_{\overline{\mathbb{R}}}^+,$$

i.e. that every measurable function can be written as a sum of **positive** measurable functions.

**Definition 7.1** ( $\mu$ -integrable). A function  $u: X \to \overline{\mathbb{R}}$  on  $(X, \mathscr{A}, \mu)$  is  $\mu$ -integrable, if it is  $\mathscr{A}/\mathscr{B}(\overline{\mathbb{R}})$ -measurable and if  $\int u^+ d\mu$ ,  $\int u^- d\mu < \infty$  (recall the definition for the integral of positive measurable functions). Then

$$\int ud\mu := \int u^+ d\mu - \int u^- d\mu \in (-\infty, \infty)$$

is the  $(\mu$ -)integral of u. We write  $\mathcal{L}^1(\mu)$  for the set of all real-valued  $\mu$ -integrable functions <sup>1</sup>.

**Theorem 7.2.** Let  $u \in \mathcal{M}_{\overline{\mathbb{R}}}(\mathscr{A})$ , then the following conditions are equivalent:

- (i)  $u \in \mathcal{L}^{\frac{1}{\mathbb{R}}}(\mu)$ .
- (ii)  $u^+, u^- \in \mathcal{L}^1_{\overline{\mathbb{R}}}(\mu)$ .
- (iii)  $|u| \in \mathcal{L}^{\underline{1}}_{\overline{\mathbb{R}}}(\mu)$ .
- (iv)  $\exists w \in \mathcal{L}^1_{\mathbb{R}}(\mu) \text{ with } w \geq 0 \text{ s.t. } |u| \leq w.$

**Theorem 7.3** (Properties of the  $\mu$ -integral). The  $\mu$ -integral is: **homogeneous**, additive, and:

- (i)  $\min\{u, v\}$ ,  $\max\{u, v\} \in \mathcal{L}^1_{\overline{\mathbb{R}}}(\mu)$  (lattice property)
- (ii)  $u \le v \Rightarrow \int u d\mu \le \int v d\mu$  (monotone)
- (iii)  $\left| \int u d\mu \right| \le \int |u| d\mu$  (triangle inequality)

**Remark.** If  $u(x) \pm v(x)$  is defined in  $\overline{\mathbb{R}}$  for all  $x \in X$  then we can exclude  $\infty - \infty$  and the theorem above just says that the integral is linear:

$$\int (au+bv)d\mu = a\int ud\mu + b\int vd\mu.$$

This is always true for real-valued  $u, v \in \mathcal{L}^1(\mu) = \mathcal{L}^1_{\mathbb{R}}(\mu)$ , making  $\mathcal{L}^1(\mu)$  a vector space with addition and scalar multiplication defined by

$$(u+v)(x) := u(x) + v(x), (a \cdot u)(x) := a \cdot u(x),$$

and

$$\int ...d\mu : \mathcal{L}^1(\mu) \to \mathbb{R}, \ u \mapsto \int u d\mu,$$

is a positive linear functional.

#### 8 Null sets and the Almost Everywhere

(Ch. 11 in [Schilling(2017)])

**Definition 8.1.** A  $(\mu$ -)null set  $N \in \mathcal{N}_{\mu}$  is a measurable set  $N \in \mathscr{A}$  satisfying

$$N \in \mathcal{N}_{\mu} \iff N \in \mathscr{A} \text{ and } \mu(N) = 0.$$

This can be used generally about a 'statement' or 'property', but we will be interested in questions like 'when is u(x) equal to v(x)', and we answer this by saying

 $u=v\ a.e.\Leftrightarrow \{x:u(x)\neq v(x)\}\$ is (contained in) a  $\mu$ -null set, i.e.

$$u = v \quad \mu$$
-a.e.  $\Leftrightarrow \mu(\lbrace x : u(x) \neq v(x) \rbrace) = 0.$ 

The last phrasing should of course include that the set  $\{x : u(x) \neq v(x)\}$  is in  $\mathscr{A}$ .

**Theorem 8.2.** Let  $u \in \mathcal{M}_{\overline{\mathbb{D}}}(\mathscr{A})$ , then:

- (i)  $\int |u| d\mu = 0 \Leftrightarrow |u| = 0 \text{ a.e.} \Leftrightarrow \mu \{u \neq 0\} = 0,$
- (ii)  $\mathbb{1}_N u \in \mathcal{L}^{\underline{1}}_{\mathbb{R}}(\mu) \ \forall \ N \in \mathcal{N}_{\mu},$
- (iii)  $\int_{N} u d\mu = 0$ .
- (i) is really useful, later we will define  $\mathcal{L}^p$  and the  $||\cdot||_p$ -(semi)norm. Then (i) means that if we have a sequence  $u_n$  converging to u in the  $||\cdot||_p$ -norm then  $u_n(x) = u(x)$  a.e.

Corollary 8.3. Let  $u = v \mu$ -a.e. Then

(i) 
$$u, v \ge 0 \Rightarrow \int u d\mu = \int v d\mu$$
,

<sup>&</sup>lt;sup>1</sup>In words, we extend our integral to positive measurable functions by noticing that we can write every measurable function as a sum of positive measurable functions, something that we do know how to integrate. We don't want to run into the problem of  $\infty - \infty$ , thus we require the integral of the positive and negative parts to both (separately) be less than infinity.

(ii) 
$$u \in \mathcal{L}^{1}_{\mathbb{R}}(\mu) \Rightarrow v \in \mathcal{L}^{1}_{\mathbb{R}}(\mu)$$
 and  $\int u d\mu = \int v d\mu$ .

Corollary 8.4. If  $u \in \mathcal{M}_{\overline{\mathbb{R}}}(\mathscr{A})$ ,  $v \in \mathcal{L}^{1}_{\overline{\mathbb{R}}}(\mu)$  and  $v \geq 0$  then

$$|u| \le v \text{ a.e. } \Rightarrow u \in \mathcal{L}^{\frac{1}{\mathbb{R}}}(\mu).$$

**Proposition 8.5** (Markow inequality). For all  $u \in \mathcal{L}^1_{\mathbb{R}}(\mu)$ ,  $A \in \mathscr{A}$  and c > 0

$$u\left(\{|u|\geq c\}\cap A\right)\leq \frac{1}{c}\int_A|u|d\mu,$$

if A = X, then (obviosly)

$$u\{|u| \ge c\} \le \frac{1}{c} \int |u| d\mu.$$

#### Completions of measure spaces

**Definition 8.6.** A measure space  $(X, \mathcal{B}, \mu)$  is called **complete** if whenever  $A \in \mathcal{B}$  and  $\mu(A) = 0$ , we have  $B \in \mathcal{B} \ \forall B \subset A$ .

**Remark.** Any measure space can be completed as follows:

Let  $\bar{\mathscr{B}}$  be the  $\sigma$ -algebra generated by  $\mathscr{B}$  and all sets  $B \subset X$  s.t. there exists  $A \in \mathscr{B}$  with  $B \subset A$  and  $\mu(A) = 0$ .

**Proposition 8.7.** The  $\sigma$ -algebra  $\bar{\mathscr{B}}$  can also be described as follows:

$$\bar{\mathscr{B}} := \left\{ B \subset X : A_1 \subset B \subset A_2 \right.$$

for some  $A_1, A_2 \in \mathcal{B}$  with  $\mu(A_2 \backslash A_1) = 0$ ,

with  $B, A_1, A_2$  as above, we define

$$\bar{\mu} := \mu(A_1) = \mu(A_2)$$

Then  $(X, \overline{\mathscr{B}}, \overline{\mu})$  is a complete measure space.

**Definition 8.8.** If  $\mu$  is a Borel measure on a **metric** space (X, d), then the completion  $\overline{\mathscr{B}}(X)$  of the Borel  $\sigma$ -algebra with respect to  $\mu$  is called the  $\sigma$ -algebra of  $\mu$ -measurable sets.

Remark. For  $\mu = \lambda_n$  on  $\mathbb{R}^n$  we talk about the  $\sigma$ -algebra of **Lebesgue measurable sets**. Instead of  $\bar{\lambda_n}$  we still write  $\lambda_n$  and call it the **Lebesgue measure**. A function  $f: \mathbb{R}^n \to \mathbb{C}$ , measurable w.r.t. the  $\sigma$ -algebra of Lebesgue measurable sets is called the **Lebesgue measur**able. The following result shows that any Lebesgue measurable function coincides with a Borel function a.e.

**Proposition 8.9.** Assume  $(X, \mathcal{B}, \mu)$  is a measure space and consider its completion  $(X, \bar{\mathcal{B}}, \bar{\mu})$ . Assume  $f: X \to \mathbb{C}$  is  $\bar{\mathcal{B}}$ -measurable. Then there is a  $\mathcal{B}$ -measurable function  $q: X \to \mathbb{C}$  s.t.  $f = q \bar{\mu}$ -a.e.

### 9 Convergence Theorems and Their Applications (Ch. 12 in [Schilling(2017)])

- To interchange limits and integrals in **Riemann integrals** one typically has to assume uniform convergence. ;- The set of Riemann integrable functions is somewhat limited, see theorem 9.4

**Theorem 9.1** (Generalization of Beppo Levi, monotone convergence).

(i) Let  $(u_n)_{n\in\mathbb{N}} \subset \mathcal{L}^1(\mu)$  be s.t.  $u_1 \leq u_2 \leq ...$ with limit  $u := \sup_{n\in\mathbb{N}} u_n = \lim_{n\to\infty} u_n$ . Then  $u \in \mathcal{L}^1(\mu)$  iff

$$\sup_{n\in\mathbb{N}}\int u_n d\mu < +\infty,$$

in which case

$$\sup_{n\in\mathbb{N}}\int u_n d\mu = \int \sup_{n\in\mathbb{N}} u_n d\mu.$$

(ii) Same thing only with a decreasing sequence ...  $> -\infty$  in which case

$$\inf_{n\in\mathbb{N}}\int u_n d\mu = \int \inf_{n\in\mathbb{N}} u_n d\mu.$$

**Theorem 9.2** (Lebesgue; dominated convergence). Let  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{L}^1(\mu)$  s.t.

(a) 
$$|u_n|(x) \le w(x), w \in \mathcal{L}^1(\mu),$$

(b) 
$$u(x) = \lim_{n \to \infty} u_n(x)$$
 exists in  $\mathbb{R}$ ,

then  $u \in \mathcal{L}^1(\mu)$  and we have

(i) 
$$\lim_{n\to\infty} \int |u_n-u| d\mu=0;$$

(ii) 
$$\lim_{n\to\infty} \int u_n d\mu = \int \lim_{n\to\infty} u_n d\mu = \int u d\mu$$
;

### Application 2: Riemann v Lebesgue Integration

Consider only  $(X, \mathcal{A}, \mu) = (\mathbb{R}, \mathcal{B}(\mathbb{R}), \lambda)$ .

**Theorem 9.3.** Let  $u : [a,b] \to \mathbb{R}$  be a measurable and Riemann integrable function. Then

$$u \in \mathcal{L}^1(\lambda) \ and \int_{[a,b]} u d\lambda = \int_a^b u(x) dx.$$
 (7)

**Theorem 9.4.** Let  $u:[a,b] \to \mathbb{R}$  be a bounded function, it is Riemann integrable **iff** the points in (a,b) where u is discontinuous are a (subset of) Borel measurable null set.

#### Improper Riemann Integrals

- The Lebesgue integral extends the (proper) Riemann integral. However, there is a further extension of the Riemann integral which cannot be captured by Lebesgue's theory. u is Lebesgue integrable  $iff \ |u|$  ha finite Lebesgue integral. From Lebesgue integral does not respect sign-changes and cancellations. However, the following  $improper\ Riemann\ integral\ does$ :

$$(R)\int_{0}^{\infty} u(x)dx := \lim_{n \to \infty} (R)\int_{0}^{a} u(x)dx.$$
 (8)

Corollary 9.5. Let  $u:[0,\infty)\to\mathbb{R}$  be a measurable, Riemann integrable function for every interval  $[0,N],\ N\in\mathbb{N}$ . Then  $u\in\mathcal{L}^1[0,\infty)$  iff

$$\lim_{N \to \infty} (R) \int_{0}^{N} |u(x)| dx < \infty.$$
 (9)

In this case,  $(R) \int_0^\infty u(x) dx = \int_{[0,\infty)} u d\lambda$ 

**Proposition 9.6** (appearing as example 12.13 in Schilling). Let  $f_{\alpha}(x) := x^{\alpha}, x > 0$  and  $\alpha \in \mathbb{R}$ . Then

(i) 
$$f(\alpha) \in \mathcal{L}^1(0,1) \Leftrightarrow \alpha > -1$$
.

(ii) 
$$f(\alpha) \in \mathcal{L}^1[1,\infty) \Leftrightarrow \alpha < -1$$
.

# 10 Regularity of measures (Append H in [Schilling(2017)])

We let (X, d) be a metric space and denote by  $\mathcal{O}$  the open, by  $\mathcal{C}$  the closed and  $\mathscr{B}(X) = \sigma(\mathcal{O})$  the Borel set of X.

**Definition 10.1.** A measure  $\mu$  on  $(X, d, \mathcal{B}(X))$  is called outer regular, if

$$\mu(B) = \inf \{ \mu(U) \mid B \subset U, U \text{ open} \}$$
 (10)

and inner regular, if  $\mu(K) < \infty$  for all compact sets  $K \subset X$  and

$$\mu(U) = \sup \left\{ \mu(K) \mid K \subset U, \ K \text{ compact} \right\}. \tag{11}$$

A measure which is both inner and outer regular is called **regular**. We write  $\mathfrak{m}_r^+(X)$  for the family of regular measures on  $(X, \mathcal{B}(X))$ .

**Remark.** The space X is called  $\sigma$ -compact if there is a sequence of compact sets  $K_n \uparrow X$ . A typical example of such a space is a locally compact, separable metric space.

**Theorem 10.2.** Let (X,d) be a metric space. Every finite measure  $\mu$  on  $(X, \mathcal{B}(X))$  is outer regular. If X is  $\sigma$ -compact, then  $\mu$  is also inner regular, hence regular.

**Theorem 10.3.** Let (X,d) be a metric space and  $\mu$  be a measure on (X,B(X)) such that  $\mu(K) < \infty$  for all compact sets  $K \subset X$ .

- 1 If X is  $\sigma$ -compact, then  $\mu$  is inner regular.
- 2 If there exists a sequence  $G_n \in \mathcal{O}$ ,  $G_n \uparrow X$  such that  $\mu(G_n) < \infty$ , then  $\mu$  is outer regular.

# 11 The Function Spaces $\mathcal{L}^p$ (Ch. 13 in [Schilling(2017)])

Assume V is a vector space over  $\mathbb{K} \in \{\mathbb{C}, \mathbb{R}\}$ .

**Definition 11.1.** A seminorn on V is a map  $p:V\to [0,+\infty)$  s.t.

(1) 
$$p(cx) = |c|p(x) \ \forall x \in V, \forall c \in \mathbb{K}.$$

(2) 
$$p(x + y) \le p(x) + p(y) \ \forall x, y \in V$$
. triangle inequality.

A seminorm is called a norm if we also have

$$p(x) = 0 \iff x = 0.$$

A norm is commonly denoted ||x||, and a vectorspace equipped with a norm is called a **normed space**.

**Definition 11.2.** Assume (X, d) is a measure space. Fix  $1 \le p \le \infty$ . For every measurable function  $f: X \to \mathbb{C}$  we define the following

$$||f||_p = \left(\int_X |f|^p d\mu\right)^{1/p} \in [0, +\infty].$$
 (12)

We can see that  $||cf||_p = |c|||f||_p \ \forall c \in \mathbb{C}$ .

Notice that by Theorem 8.2(i) we have that  $||f||_p = 0 \Rightarrow f = 0$  a.e. Consider for example  $\lim_{n\to\infty} ||f_n-f||_p=0$ , then we can find a subsequence s.t.  $\lim_{k\to\infty} |f_{n(k)} - f| = 0$  a.e., i.e.  $\lim_{k\to\infty} f_{n(k)} = f$  a.e.

#### Lemma 11.3.

$$||f+g||_p \le ||f||_p + ||g||_p.$$
 (13)

**Definition 11.4.** We define

$$\mathcal{L}^p(X, d\mu) = \{ f : X \to \mathbb{C} \mid f \text{ is measurable and } \}$$

This is a vectorspace with seminorm  $f \mapsto ||f||_p$ . And in general this is not a normed space, since  $||f||_p = 0 \iff f = 0 \text{ a.e.}$ 

Generally, if p is a seminorm on a vectorspace V, then

$$V_0 = \{ x \in V \mid p(x) = 0 \} \tag{14}$$

which is a subspace of V. Then we consider the quotient/factor space  $V/V_0$ .

**Definition 11.5.** For  $x, y \in V$ , define

$$x \sim y \iff x - y \in V_0.$$
 (15)

This is an equivalence relation on V. The representation class of V is defined by [x] or  $x+V_0$ .

Then  $V/V_0$  is equals the set of equivalence classes. We can show that it is a normed space.

$$[x] + [y] = [x+y] \;\;,\;\; c[x] = [cx] \;\;,\;\; ||[x]|| = p(x).$$

Applying this to  $\mathcal{L}^p(X,d\mu)$  we get the normed space

$$L^{p}(X, d\mu) := \mathcal{L}^{p}(X, d\mu) / \mathcal{N} = \mathcal{L}^{p}(X, d\mu) /_{\sim}.$$
(16)

Where  $\mathcal{N}$  is the space of measurable functions f s.t. f = 0 a.e. The equivalence relation  $\sim$  is defined by

$$u \sim v \iff \{u \neq v\} \in \mathcal{N}_{\mu} \iff \mu \{u \neq v\} = 0,$$

and so  $L^p(X, d\mu)$  consists of all equivalence classes  $[u]_p = \{v \in \mathcal{L}^p | u \sim v\}$ . So for every  $u \in$  $L^p$  there is no  $v \in L^p$  such that  $\mu\{u \neq v\} \neq 0$ .

We will further continue to denote the norm by  $||\cdot||_p$ , and we will normally **not** distinguish between  $f \in \mathcal{L}^p(X, d\mu)$  and the vector in  $L^p(X, d\mu)$  that f defines.

**Definition 11.6.** A normed space  $(X, ||\cdot||)$  is called a Banach space if V is complete w.r.t the metric d(x, y) = ||x - y||.

**Theorem 11.7.** If  $(X, \mathcal{B}, \mu)$  is a measure space,  $1 \le p \le \infty$ , then  $L^p(X, d\mu)$  is a Banach space.

**Definition 11.8.** A measurable function f:  $X \to \mathbb{C}$  is called **essentially bounded** if there is  $c \geq 0$  s.t.

$$\mathcal{L}^{p}(X, d\mu) = \{ f : X \to \mathbb{C} \mid f \text{ is measurable and } ||f||_{p} < \infty \}. \mu(\{x : |f(x)| > c\}) = 0. \tag{17}$$

That is  $|f| \leq c$  a.e. The smallest such c is called the essential supremum of f and is denoted by  $||f||_{\infty}$ .

Definition 11.9.

 $\mathcal{L}^{\infty}(X, d\mu) = \{ f : X \to \mathbb{C} \mid f \text{ is measurable and } ||f||_{\infty} < \infty \}.$ 

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

Where by the previous definition these spaces become the spaces of all essentially bounded func-

**Theorem 11.10.** If  $(X, \mathcal{B}, \mu)$  is a  $\sigma$ -finite measure space, then  $L^{\infty}(X, d\mu)$  is a Banach space.

#### Convergence in $\mathcal{L}^p$ and completeness

**Lemma 11.11.** For any sequence  $(u_n)_{n\in\mathbb{N}}\subset$  $\mathcal{L}^p(\mu), p \in [1, \infty), \text{ of positive functions } u_n \geq 0$ we have

$$\left| \left| \sum_{n=1}^{\infty} u_n \right| \right|_p \le \sum_{n=1}^{\infty} ||u_n||_p.$$

**Theorem 11.12** (Riesz-Fischer). The spaces  $\mathcal{L}^p(\mu), p \in [1, \infty), \text{ are complete, i.e.}$ Cauchy sequence  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{L}^p(\mu)$  converges to some limit  $u \in \mathcal{L}^p(\mu)$ 

Corollary 11.13. Let  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{L}^p(\mu), p\in$  $[1,\infty)$  with  $\mathcal{L}^p - \lim_{n\to\infty} u_n = u$ . there exists a subsequence  $(u_{n_k})_{k\in\mathbb{N}}$  s.t.  $\lim_{k\to\infty} u_{n_k}(x) = u(x)$  holds for almost every **Theorem 11.14.** Let  $(u_n)_{n\in\mathbb{N}}\subset \mathcal{L}^p(\mu), p\in [1,\infty)$ , be a sequence of functions s.t.  $|u_n|\leq w \ \forall n\in\mathbb{N}$  and some  $w\in \mathcal{L}^p(\mu)$ . If  $u(x)=\lim_{n\to\infty}u_n(x)$  exists for (almost) every  $x\in X$ , then

$$u \in \mathcal{L}^p$$
 and  $\lim_{n \to \infty} ||u - u_n||_p = 0.$ 

**Theorem 11.15** (F. Riesz (convergence theorem)). Let  $(u_n)_{n\in\mathbb{N}}\subset\mathcal{L}^p(\mu), p\in[1,\infty)$ , be a sequence s.t.  $\lim_{n\to\infty}u_n(x)=u(x)$  for almost every  $x\in X$  and some  $u\in\mathcal{L}^p(\mu)$ . Then

$$\lim_{n \to \infty} ||u_n - u||_p = 0 \Longleftrightarrow \lim_{n \to \infty} ||u_n||_p = ||u||_p.$$

### 12 Dense and Determining Sets (Ch. 17 in [Schilling(2017)])

**Definition 12.1** (Dense Sets). A set  $\mathcal{D} \subset \mathcal{L}^p(\mu), p \in [0, \infty]$ , is called *dense* if for every  $u \in \mathcal{L}^p(\mu)$  there exist a sequence  $(f_n)_{n \in \mathbb{N}} \subset \mathcal{D}$  s.t.  $\lim_{n \to \infty} ||u - f_n||_p = 0$ .

**Theorem 12.2.** Assume X, d is a metric space and  $\mu$  is a Borel measure that is finite on every ball  $1 \le p < \infty$ . Then the space of bounded continuous functions with bounded support is dense in  $\mathcal{L}^p(X, d\mu)$ . Where bounded support means that f vanishes outside some ball.

**Theorem 12.3.** Assume (X,d) is a separable locally compact metric space and  $\mu$  is a Borel Measure on X s.t.  $\mu(K) < \infty \ \forall \ compact \ K \subset K$ . Then the space  $C_c(X)$  of continuous compactly supported functions is dense in  $\mathcal{L}^p(X,d\mu)$ .

Recall that the support of a function f is  $\operatorname{supp}(f) = \{x \in X : f(x) \neq 0\}$ , closed support is the closure of  $\operatorname{supp}(f)$  (i.e. boundary points are included), often just written as  $\operatorname{supp}(f)$ , and a function is said to have compact support if  $\operatorname{supp}(f)$  is compact.

In particular, either theorem shows that if  $\mu$  is a Borel measure on  $\mathbb{R}^n$  s.t. the measure of every ball is finite, then  $C_c(\mathbb{R}^n)$  is dense in  $\mathcal{L}^p(\mathbb{R}^n, d\mu)$ ,  $1 \leq p < \infty$ . Later we will see that even  $C_c^{\infty}(\mathbb{R}^n)$  is dense in  $L^p(\mathbb{R}^n, d\mu)$ .

**Remark.** These results do not extend to  $p = \infty$  in general.

For  $\mu = \lambda_n$  we write simply  $\mathcal{L}^p(\mathbb{R}^n)$ .

**Remark.** Theorem 17.8 in the book is WRONG. For example,  $X = \mathbb{Q}$  with the usual metric is  $\sigma$ -compact, supports nonzero finite measure, but  $C_c(\mathbb{Q}) = 0$ .

### Modes of Convergence (mixture of ex. 11.12 and ch. 22 p. 258-261.)

Assume  $(X, \mathcal{B}, \mu)$  is a measure space. Given measurable functions  $f_n, f: X \to \mathbb{C}$ , recall that

$$f_n \to f$$
 a.e.

means that  $f_n(x) \xrightarrow[n \to \infty]{} f(x)$  for all x outside a set of measure zero.

**Theorem 12.4** (Egorov). Assume  $\mu(X) < \infty$  and  $f_n \to f$  a.e. Then,  $\forall \epsilon > 0$ , there exists  $X_{\epsilon} \in \mathcal{B}$  s.t.  $\mu(X_{\epsilon}) < \epsilon$  and  $f_n \to f$  uniformly on  $X \setminus X_{\epsilon}$ .

In addition to pointwise and uniform convergence we also consider the following:

 $f_n \to f$  in the *p-th mean* if  $||f_n - f||_p \xrightarrow[n \to \infty]{} 0$ . For p = 1 we say in mean, for p = 2 we say in quadratic mean.

 $f_n \to f$  in measure if  $\forall \epsilon > 0$  we have

$$\mu\left(\left\{x:\left|f_n(x)-f(x)\right|\geq\epsilon\right\}\right)\xrightarrow[n\to\infty]{}0.$$

**Theorem 12.5** (Lemma 22.4 in the book?). Assume  $(X, \mathcal{B}, d\mu)$  is a measure space,  $1 \leq p < \infty$ ,  $f_n, f: X \to \mathbb{C}$  are measurable functions. Then

- (i) If  $f_n \to f$  in the p-th mean, then  $f_n \to f$  in measure.
- (ii) If  $f_n \to f$  in measure, then there is a subsequence  $(f_{n_k})_{k=1}^{\infty}$  s.t.  $f_{n_k} \to f$  a.e.
- (iii) If  $f_n \to f$  a.e. and  $\mu(X) < \infty$ , then  $f_n \to f$  in measure.

In particular, if  $f_n \to f$  in the p-th mean, then  $f_{n_k} \to f$  a.e. for a subsequence  $(f_{n_k})_k$ .

# 13 Abstract Hilbert Spaces (Ch. 26 in [Schilling(2017)])

Assume  $\mathcal{H}$  is a vector space over  $\mathbb{C}$ .

**Definition 13.1.** A pre-inner product on  $\mathcal{H}$  is a map  $(\cdot, \cdot) : \mathcal{H} \times \mathcal{H} \to \mathbb{C}$  which is

(i) Sesquilinear: linear in the first variable and antilinear in the second:

$$(\alpha u + \beta v, w) = \alpha(u, w) + \beta(v, w),$$
  

$$(w, \alpha u + \beta v) = \bar{\alpha}(w, u) + \bar{\beta}(w, v), \ u, v, w \in H \text{ and } \alpha, \beta \in \mathbb{C}.$$

- (ii) Hermitian:  $(u, v) = \overline{(u, v)}$ .
- (iii) Positive semidefinite: (u, v) > 0.

It is called an **inner product**, or a scalar product, if instead of (iii) the map is positive definite; (u, v) > 0. This definition also works for  $\mathbb{R}$  instead of  $\mathbb{C}$ .

Cauchy-Schwartz inequality If  $(\cdot, \cdot)$  is a pre-inner product, then  $|(u, v)| \leq (u, u)^{1/2} (v, v)^{1/2}$ .

**Corollary 13.2.** Assume we have a seminorm  $||u|| := (u, u)^{1/2}$ . It is a norm iff  $(\cdot, \cdot)$  is an inner product.

**Definition 13.3** (Hilbert space). A Hilbert space is a complex vector space  $\mathcal{H}$  with an inner product  $(\cdot, \cdot)$  s.t.  $\mathcal{H}$  is complete with respect to the norm  $||u|| = (u, u)^{1/2}$ .

- 1. The norm on a Hilbert space is determined by the inner product, but the inner product can also be recovered by the norm by the polarization identity:  $(u, v) = \frac{||u+v||^2 ||u-v||^2}{4} + i \frac{||u+iv||^2 ||u-iv||^2}{4}$ .
- 2.  $Parallelogram\ law:\ ||u+v||^2+||u-v||^2=2||u||^2+2||v||^2.$
- A norm on a vector space is given by an inner product iff it satisfies the parallelogram law, and then the scalar product is uniquely determined by the polarization identity.

Recall that a subset  $\mathcal{C}$  of a vector space is called convex if

$$u, w \in \mathcal{C} \to tu + (1-t)w \in \mathcal{C} \ \forall t \in (0,1).$$

The following is one of the key properties of the Hilbert space

**Theorem 13.4** (projection theorem). Assume  $\mathcal{H}$  is a Hilbert space and  $\mathcal{C} \subset H$  is a closed convex subset. Then for every  $u \in H$  there is a unique  $u_0 \in \mathcal{C}$  (minimizer) s.t.

$$||u - u_0|| = d(u, C) (= \inf_{x \in C} ||u - x||).$$

# 14 Orthogonal Projections (Ch. 26 in [Schilling(2017)])

For a Hilbert space  $\mathcal{H}$  and a subset  $A \subset H$ , let

$$A^{\perp} := \{ x \in H : x \perp y \ \forall y \in A \} \,,$$

where  $x \perp y$  means that (x, y) = 0.  $A^{\perp}$  is a closed subspace of  $\mathcal{H}$ .

**Proposition 14.1.** Assume  $\mathcal{H}_0$  is a closed subspace of a Hilbert space  $\mathcal{H}$ . Then every  $u \in H$  uniquely decomposes as

$$u = u_0 + u_1, \text{ with } u_0 \in H \text{ and } u_1 \in \mathcal{H}_0^{\perp}.$$

Moreover,  $||u - u_0|| = d(u, \mathcal{H}_0)$  and  $||u||^2 = ||u_0||^2 + ||u_1||^2$ .

For a closed subspace  $\mathcal{H}_0 \subset \mathcal{H}$ , consider the map  $P: H \to \mathcal{H}_0$  s.t.  $Pu \in \mathcal{H}_0$  is the unique element satisfying  $u - Pu = H_0^{\perp}$ . The operator P is linear. It is also contractive, meaning that  $||Pu|| \leq ||u||$ , since  $||u||^2 = ||Pu||^2 + ||u - Pu||^2$ . It is called the orthogonal projection onto  $\mathcal{H}_0$ .

If  $\mathcal{H}_0$  is finite dimensional with an orthonormal basis  $u_1, ..., u_n$  then

$$Pu = \sum_{k=1}^{n} (u, u_k) u_k.$$

Orthonormal bases can be defined for arbitrary Hilbert spaces.

**Definition 14.2** (orthonormal system). An orthonormal system in  $\mathcal{H}$  is a collection of vectors  $u_i \in H$   $(i \in I)$ s.t.

$$(u_i, u_j) = \delta_{ij} \ \forall i, j \in I.$$

It is called an *orthonormal basis* if span $\{u_i\}_{i\in I}$  denotes the linear span of  $\{u_i\}_{i\in I}$ , the space of finite linear combinations of the vectors  $u_i$ .

**Definition 14.3.** A Hilbert space  $\mathcal{H}$  is said to be *separable* if  $\mathcal{H}$  contains a countable dense subset  $G \subset \mathcal{H}$ .

**Theorem 14.4.** Every Hilbert space  $\mathcal{H}$  has an orthonormal basis. If  $\mathcal{H}$  is separable, then there is a countable orthonormal basis.

**Proposition 14.5.** Assume  $\{u_i\}_{i\in I}$  is an orthonormal system in a Hilbert space H. Take  $u \in \mathcal{H}$ . Then

- (i) Bessel's inequality:  $\sum_{i \in I} |(u, u_i)|^2 \le ||u||^2$ , in particular,  $\{i : (u, u_i) \ne 0\}$  is countable.
- (ii) Parseval's identity: If  $\{u_i\}_{i\in I}$  is an orthonormal basis, then  $\sum_{i\in I} |(u,u_i)|^2 = ||u||^2.$

If  $(u_i)_{i\in I}$  is an orthonormal basis, then the numbers  $(u, u_i)$  are called the **Fourier coefficients** of u with respect to  $(u_i)_{i\in I}$ . The Parseval identity then suggests that u is determined by its Fourier coefficients. This is true, and even more, we have:

**Proposition 14.6.** Assume  $(u_i)_{i\in I}$  is an orthonormal basis in a Hilbert space  $\mathcal{H}$ . Then for every vector  $(c_i)_{i\in I} \in l^2(I)$  there is a unique vector  $u \in \mathcal{H}$  with Fourier coefficients  $c_i$ , and we write

$$u = \sum_{i \in I} c_i u_i.$$

**Remark.** Equivalently, the element  $u = \sum_{i \in I} c_i u_i$  can be described as the unique element in  $\mathcal{H}$  s.t.  $\forall \epsilon > 0$  there is a finite  $F_0 \subset I$  s.t.  $||u - \sum_{i \in F} c_i u_i|| < \epsilon \ \forall$  finite  $F \supset F_0$ .

**Corollary 14.7.** We have a linear isomorphism  $U: l^2(I) \xrightarrow{\sim} \mathcal{H}$ ,  $U((c_i)_{i \in I}) = \sum_{i \in I} c_i u_i$ . By Parseval's identity this isomorphism is isometric, that is,  $||Ux|| = ||x|| \ \forall x \in l^2(I)$ . By the polarization identity this is equivalent to

$$(Ux, Uy) = (x, y) \ \forall x, y \in l^2(I).$$

Therefor U is unitary.

Corollary 14.8. Up to a unitary isomorphism, there is only one infinite dimensional separable Hilbert space, namely,  $l^2$ .

### 15 Dual spaces (Ch. 26 in [Schilling(2017)])

**Lemma 15.1.** Assume V is a normed space over  $K = \mathbb{R}$  or  $K = \mathbb{C}$ . Consider a linear functional  $f: V \to K$ . The following are equivalent (TFAE):

- (i) f is continuous;
- (ii) f is continuous at 0;
- (iii) There is a  $c \ge 0$  s.t.  $|f(x)| \le c||x|| \ \forall x \in V$ .

If (i)-(iii) are satisfied, then f is called a bounded linear functional. The constant c in (iii) is denoted by ||f||. We have  $||f|| = \sup_{x \neq 0} \frac{|f(x)|}{||x||} = \sup_{||x|| < 1} |f(x)| = \sup_{||x|| < 1} |f(x)|$ .

**Proposition 15.2.** For every normed vector space V over  $K = \mathbb{R}$  or  $K = \mathbb{C}$ , the bounded linear functionals on V form a Banach space  $V^*$ .

**Remark.** The sequence  $\{||f_n - f_m||\}_{m=1}^{\infty}$  actually converges, since

$$|||f_n - f_m||| \le ||f_m - f_n||.$$

When we study/use normed spaces, it is often important to understand the dual spaces. For Hilbert spaces this is particularly easy:

**Theorem 15.3** (Riesz). Assume  $\mathcal{H}$  is a Hilbert space. Then every  $f \in \mathcal{H}^*$  has the form

$$f(x) = (x, y),$$

for a uniquely defined  $y \in \mathcal{H}$ . Moreover, we have ||f|| = ||y||.

For every Hilbert space  $\mathcal{H}$  we can define the conjugate Hilbert space  $\bar{\mathcal{H}}$ , which has its elements as the symbols  $\bar{x}$  for  $x \in \mathcal{H}$ , with the linear structure and inner product defined by

$$\bar{x} + \bar{y} = \overline{x + y}, c \cdot \bar{x} = \overline{cx}, (\bar{x}, \bar{y}) = \overline{(x, y)} = (y, x).$$

**Corollary 15.4.** For every Hilbert space  $\mathcal{H}$ , we have an isometric isomorphism  $\overline{\mathcal{H}} \xrightarrow{\sim} \mathcal{H}^*$ ,  $\overline{x} \mapsto (\cdot, x)$ .

### 16 Hahn-Banach Theorem (Ch. 4.2 in [Teschl(2010)])

**Theorem 16.1** (Hahn-Banach). Assume V is a real vector space,  $V_0 \subset V$  a subspace,  $e: V \to \mathbb{R}$  a convex function and  $f: V_0 \to \mathbb{R}$  a linear functional s.t.  $f \leq e$  on  $V_0$ . Then f can be extended to a linear functional F on V s.t.  $F \leq e$ .

**Theorem 16.2** (Hahn-Banach). Assume V is a real or complex vector space, p a seminorm on  $V_0$ ,  $V_0 \subset$ , and f a linear functional on  $V_0$  s.t.

$$|f(x)| \le p(x) \ \forall x \in V_0.$$

Then f can be extended to a linear functional F on V s.t.  $|F(x)| \le p(x) \ \forall x \in V$ .

Corollary 16.3. Assume V is a normed real or complex vector space,  $V_0 \subset V$  and  $f \in V_0^*$ . Then there is a  $F \in V^*$  s.t.

$$F|_{V_0}f$$
 and  $||F|| = ||f||$ .

**Corollary 16.4.** Assume V is a normed space and  $x \in V, x \neq 0$ . Then there is a  $F \in V^*$  s.t. ||F|| = 1 and F(x) = ||x||.

Such an F is called a *supporting functional* at x.

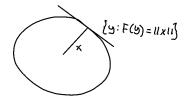


Figure 1: Tangent space?

If V is a normed vector space, then every  $x \in X$  defines a bounded linear functional on  $V^*$  by

$$V^* \ni f \mapsto f(x)$$
.

As  $|f(x)| \leq ||f|| \cdot ||x||$ , this functional has norm  $\leq ||x||$ . By using a supporting functional at x, we actually see that we get norm ||x||. Thus, we have an isometric embedding  $V \subset V^{**} := (V^*)^*$ . We can therefor see view V as a subspace of  $V^{**}$ .

**Definition 16.5.** A normed space V is called reflexive if  $V^{**} = V$ .

**Remark.** This is stronger than requiring  $V \cong V^{**}$ .

**Remark.** Every Hilbert space  $\mathcal{H}$  is reflexive. Indeed,  $\mathcal{H}^* = \bar{\mathcal{H}}$ . By Riesz' theorem every bounded linear functional f on  $\bar{\mathcal{H}}$  has the form

$$f(\bar{x}) = (\bar{x}, \bar{y}) = (y, x),$$

for some  $y \in \mathcal{H}$ , which exactly means that f = y in  $\mathcal{H}^{**}$ .

As we will see later, the spaces  $\mathcal{L}^p(X, d\mu)$ , with  $\mu$   $\sigma$ -finite and 1 , are reflexive. $The spaces <math>\mathcal{L}'(X, d\mu)$  and  $\mathcal{L}^{\infty}(X, \mu)$  are usually not reflexive.

### 17 Radon-Nikodym Theorem (Ch. 20 in [Schilling(2017)])

Assume  $(X, \mathcal{B}, \mu)$  is a measure space. Are there other measures on  $(X, \mathcal{B})$ ?

**Example 17.1.** Take a measurable function  $f: X \to [0, +\infty]$  and define

$$\nu(A) = \int_A f d\mu \text{ for } A \in \mathcal{B}.$$

This is a measure by the monotone convergence theorem. We write  $d\nu = f d\mu$ .

**Proposition 17.2.** Assume  $(X, \mathcal{B})$  is a measurable space,  $\mu$  and  $\nu$  are  $\sigma$ -finite measures on  $(X, \mathcal{B})$ . Then there exist  $N \in \mathcal{B}$  and a measurable  $f: X \to [0, +\infty]$  s.t.  $\mu(N) = 0$  and  $\nu(A) = \nu(A \cap N) + \int_A f d\mu \ \forall A \in \mathcal{B}$ .

When can we discard the term  $\nu(A \cap N)$ ?

**Definition 17.3.** Given measure  $\mu$  and  $\nu$  on  $X, \mathcal{B}$ , we say that  $\nu$  is absolutely continuous with respect to  $\mu$  and write  $\nu << \mu$ , if  $\nu(A) = 0$  whenever  $A \in \mathcal{B}, \mu(A) = 0$ .

**Lemma 17.4.** Assume  $\mu$  and  $\nu$  are measures on  $(X, \mathcal{B})$ ,  $\nu(X) < \infty$ . Then  $\nu << \mu$  iff  $\forall \epsilon > 0 \exists \delta > 0$  s.t. if  $A \in \mathcal{B}$ ,  $\mu(A) < \delta$ , then  $\nu(A) < \epsilon$ .

*Proof.* " $\Rightarrow$ ": obvious. " $\Leftarrow$ ": Assume this is not true. Then, there is a  $\epsilon > 0$  s.t.  $\forall \delta > 0$  we can find  $A \in \mathcal{B}$  satisfying  $\mu(A) < \delta$ ,  $\nu(A) \geq \epsilon$ . Let  $A_n$  be such a set A for  $\delta = 1/2^n$ . Put  $A = \bigcap_{n \in \mathbb{N}} \bigcup_{k=n} A_k$ . Then

$$\mu(A) \le \lim_{n \to \infty} \mu\left(\bigcup_{k=n}^{\infty} A_k\right) \le \lim_{n \to \infty} \sum_{k=n}^{\infty} \mu(A_k)$$
$$\le \lim_{n \to \infty} \sum_{k=n}^{\infty} \frac{1}{2^k} = \lim_{n \to \infty} \frac{1}{2^{n-1}} = 0.$$

As  $\nu(X) < \infty$ , we also have

$$\nu(A) = \lim_{n \to \infty} \nu\left(\bigcup_{k=n}^{\infty} A_k\right) \ge \epsilon.$$

This contradicts the assumption  $\nu \ll \mu$ .

**Remark.** The result is not true for infinite  $\nu$ .

**Theorem 17.5** (Radon-Nikodym). Assume  $\mu$  and  $\nu$  are  $\sigma$ -finite measures on a measurable space  $(X, \mathcal{B}), \ \nu << \mu$ . Then there is a measurable function  $f: X \to [0, +\infty)$  s.t.  $d\nu = f d\mu$  (that is,  $\nu(A) \int_A f d\mu$ ). If  $\tilde{f}$  is another function with the same properties, then  $f = \tilde{f} \ \mu - a.e.$ 

The function is called the Radon-Nikodym derivative at  $\nu$  w.r.t.  $\mu$  and is denoted by  $\frac{d\nu}{d\mu}$ .

**Example 17.6.** Consider a real-valued function  $f \in C'[a,b]$  s.t.  $f'(t) > 0 \, \forall \, t \in [a,b]$ . Let c = f(a), d = f(b). We know that for every Riemann integrable function g on [c,d] we have

$$\int_{c}^{d} g(f)dt = \int_{a}^{b} g(f(t))f'(t)dt.$$

Equivalently,

$$\int_{a}^{d} g \circ g^{-1} dt = \int_{a}^{b} g f'(t) dt. \tag{18}$$

Denote by  $\lambda_{[a,b]}$ ,  $\lambda_{[c,d]}$  the Lebesgue measure restricted to the Borel subsets of [a,b] and [c,d], respectively. Then eq. 18 implies that

$$d\left((f^{-1})_*\lambda_{[c,d]}\right) = f'd\lambda_{[a,b]},$$

since the integration of  $g = \mathbb{1}_{[\alpha,\beta]}$  gives the same results for any interval  $[\alpha,\beta] \subset [a,b]$  and since a finite Borel measure on [a,b] is determined by its values on such intervals. Thus, the recurrence is  $(f^{-1})_*\lambda_{[c,d]} << \lambda_{[a,b]}$  and

$$\frac{d\left((f^{-1})_*\lambda_{[c,d]}\right)}{d\lambda_{[a,b]}} = f'.$$

#### References

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