# MATH597 Analysis II

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## February 25, 2022

#### Abstract

Notice that since in this course, the cross-referencing between theorems, lemmas, and propositions are quite complex and hard to keep track of, hence in this note, whenever you see a ! over =, like  $\stackrel{!}{=}$ , then that ! is clickable! It will direct you to the corresponding theorem, lemma, or proposition we're using to deduce that particular equality.

Notice that there are some proofs is **intended** left as assignments, and for completeness, I put them in Appendix A, use it in your **own risks!** You'll lose the chance to practice and really understand the materials.

Additionally, we'll use [FF99] as our main text, while using [Tao13] and [Axl19] as supplementary references.

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## Lecture 1: $\sigma$ -algebra

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### 1 Measure

**Example.** Before we start, we first see some examples.

1. Let  $X = \{a, b, c\}$ . Then

$$\mathcal{P}(X) := \{ \varnothing, \{a\}, \{b\}, \{c\}, \{a,b\}, \{a,c\}, \{b,c\}, \{a,b,c\} \},\$$

which is the *power set* of X. We see that

$$\#X = n \implies \#\mathcal{P}(X) = 2^n$$

for  $n < \infty$ .

2. If  $n = \infty$ , say  $X = \mathbb{N}$ , then

$$\mathcal{P}(\mathbb{N})$$

is an uncountable set while  $\mathbb N$  is a countable set. We can see this as follows. Consider

$$\phi \colon \mathcal{P}(\mathbb{N}) \to [0,1], \quad A \mapsto 0.a_1a_2a_3\dots \text{(base 2)},$$

where

$$a_i = \begin{cases} 1, & \text{if } i \in A \\ 0, & \text{if } i \notin A, \end{cases}$$

and for example, A can be  $A=\{2,3,6,\ldots\}\subseteq\mathbb{N}.$  Note that  $\phi$  is surjective, hence we have

$$\#\mathcal{P}(\mathbb{N}) \geq \# [0,1]$$
.

But since [0,1] is uncountable, so is  $\mathcal{P}(\mathbb{N})$ .

We like to measure the size of subsets of X. Hence, we are intriguing to define a map  $\mu$  such that

$$\mu \colon \mathcal{P}(X) \to [0, \infty]$$
.

**Example.** We first see some examples.

1. Let  $X = \{0, 1, 2\}$ . Then we want to define  $\mu \colon \mathcal{P}(X) \to [0, \infty]$ , we can have

• 
$$\mu(A) = \#A$$
. Then we have

$$-\mu(\{0,1\})=2$$

$$-\mu(\{0\}) = 1$$

•  $\mu(A) = \sum_{i \in A} 2^i$ . Then we have

$$-\mu(\{0,1\}) = 2^0 + 2^1 = 3$$

- 2. Let  $X = \{0\} \cup \mathbb{N}$ . Then we want to define  $\mu \colon \mathcal{P}(\mathbb{N}) \to [0, \infty]$ , we can have
  - $\mu(A) = \#A$ . Then we have

$$-\mu(\{2,3,4,5,\ldots\}) = \infty = \mu(\{\text{even numbers}\})$$

•  $\mu(A) = e^{-1} \sum_{i \in A} \frac{1}{i!}$ . Then we have

$$- \mu(\{0, 2, 4, 6, \ldots\}) = e^{-1} \left(1 + \frac{1}{2!} + \frac{1}{3!} + \ldots\right)$$

- $\mu(A) = \sum_{i \in A} a_i$
- 3. Let  $X = \mathbb{R}$ . Then we want to define  $\mu \colon \mathcal{P}(\mathbb{R}) \to [0, \infty]$ , we can have
  - $\mu(A) = \#A$
  - $\mu((a,b)) = b a$ .

**Problem.** Can we extend this map to all of  $\mathcal{P}(\mathbb{R})$ ?

Answer. No!

 $\bullet \ \mu((a,b)) = e^b - e^a.$ 

**Problem.** Can we extend this map to all of  $\mathcal{P}(\mathbb{R})$ ?

Answer. No!

We immediately see the problems. To extend our native measure method into  $\mathbb{R}$  is hard and will cause something counter-intuitive! Hence, rather than define measurement on *all* subsets in the power set of X, we only focus on *some* subsets. In other words, we want to define

$$\mu \colon \mathcal{P}(\mathbb{R}) \supset \mathcal{A} \to [0, \infty]$$
.

## 1.1 $\sigma$ -algebras

We start from the definition of the most fundamental element in measure theory.

<sup>1</sup> https://en.wikipedia.org/wiki/Banach-Tarski\_paradox

**Definition 1.1 (\sigma-algebra).** Let X be a set. A collection  $\mathcal{A}$  of subsets of X, i.e.,  $\mathcal{A} \subset \mathcal{P}(X)$  is called a  $\sigma$ -algebra on X if

- $\varnothing \in \mathcal{A}$ .
- $\mathcal{A}$  is closed under complements. i.e., if  $A \in \mathcal{A}$ ,  $A^c = X \setminus A \in \mathcal{A}$ .
- $\mathcal{A}$  is closed under countable unions. i.e., if  $A_i \in \mathcal{A}$ , then  $\bigcup_{i=1}^{\infty} A_i \in \mathcal{A}$ .

Remark. There are some easy properties we can immediately derive.

- $X \in \mathcal{A}$  from  $X = X \setminus \underbrace{\varnothing}_{\in \mathcal{A}}$  and  $\mathcal{A}$  is closed under complement.
- $\bigcap_{i=1}^{\infty} A_i = \left(\bigcup_{i=1}^{\infty} A_i^c\right)^c$ , namely  $\mathcal{A}$  is <u>closed under countable intersections</u>.
- $A_1 \cup A_2 \cup \ldots \cup A_n = A_1 \cup A_2 \cup \ldots \cup A_n \cup \varnothing \cup \varnothing \cup \ldots$ , hence  $\mathcal{A}$  is closed under finite unions and intersections.

An immediate definition can be given. We now define so-called Borel set.

**Definition 1.2 (Borel set).** Given a topological space X, a *Borel set* is any set in X that can be formed from open sets through the operations of countable union, countable intersection and relative complement.

#### Lecture 2: Measure

07 Jan. 11:00

**Example.** Again, we first see some examples.

- 1. Let  $\mathcal{A} = \mathcal{P}(X)$ , which is the power  $\sigma$ -algebra.
- 2. Let  $\mathcal{A} = \{\emptyset, X\}$ , which is a trivial  $\sigma$ -algebra.
- 3. Let  $B \subset X$ ,  $B \neq \emptyset$ ,  $B \neq X$ . Then we see that  $\mathcal{A} = \{\emptyset, B, B^c, X\}$  is a  $\sigma$ -algebra.

**Lemma 1.1.** Let  $\mathcal{A}_{\alpha}$ ,  $\alpha \in I$ , be a family of  $\sigma$ -algebra on X. Then

$$\bigcap_{\alpha\in I}\mathcal{A}_{\alpha}$$

is a  $\sigma$ -algebra on X.

**Remark.** Notice that I may be an uncountable intersection.

*Proof.* A simple proof can be made as follows. Firstly,  $\emptyset \in \mathcal{A}_{\alpha}$  for every  $\alpha$  clearly. Moreover, closure under complement and countable unions for every  $\mathcal{A}_{\alpha}$  implies the same must be true for  $\bigcap_{\alpha \in I} \mathcal{A}_{\alpha}$ . Hence,  $\bigcap_{\alpha \in I} \mathcal{A}_{\alpha}$  is a  $\sigma$ -algebra.

The above allows us to give the following definition.

**Definition 1.3 (Generation of**  $\sigma$ -algebra). Given  $\mathcal{E} \subset \mathcal{P}(X)$ , where  $\mathcal{E}$  is not necessarily a  $\sigma$ -algebra. Let  $\langle \mathcal{E} \rangle$  be the intersection of all  $\sigma$ -algebras on X containing  $\mathcal{E}$ , then we call  $\langle \mathcal{E} \rangle$  the  $\sigma$ -algebra generated by  $\mathcal{E}$ .

**Remark.** Clearly,  $\langle \mathcal{E} \rangle$  is the smallest  $\sigma$ -algebra containing  $\mathcal{E}$ , and it is unique. To check the uniqueness, we suppose there are two different  $\langle \mathcal{E} \rangle_1$  and  $\langle \mathcal{E} \rangle_2$  generated from  $\mathcal{E}$ . It's easy to show

$$\langle \mathcal{E} \rangle_1 \subseteq \langle \mathcal{E} \rangle_2 \,,$$

and by symmetry, they are equal.

**Example.** We see that  $\{\emptyset, B, B^c, X\} = \langle \{B\} \rangle = \langle \{B^c\} \rangle$ .

Lemma 1.2. We have

- 1. Given  $\mathcal{A}$  a  $\sigma$ -algebra,  $\mathcal{E} \subset \mathcal{A} \subset \mathcal{P}(X) \implies \langle \mathcal{E} \rangle \subset \mathcal{A}$
- 2.  $\mathcal{E} \subset \mathcal{F} \subset \mathcal{P}(X) \implies \langle \mathcal{E} \rangle \subset \langle \mathcal{F} \rangle$

*Proof.* We'll see that after proving the first claim, the second follows smoothly.

- 1. The first claim is trivial, since we know that  $\langle \mathcal{E} \rangle$  is the smallest  $\sigma$ -algebra containing  $\mathcal{E}$ , then if  $\mathcal{E} \subset \mathcal{A}$ , we clearly have  $\langle \mathcal{E} \rangle \subset \mathcal{A}$  by the definition.
- 2. The second claim is also easy. From the first claim and the definition, we have

$$\mathcal{E} \subset \mathcal{F} \subset \langle \mathcal{F} \rangle \implies \langle \mathcal{E} \rangle \subset \langle \mathcal{F} \rangle.$$

At this point, we haven't put any specific structure on X. Now we try to describe those spaces with good structure, which will give the space some nice properties.

**Definition 1.4 (Borel**  $\sigma$ -algebra). For a topological space X, the *Borel*  $\sigma$ -algebra on X, denotes as  $\mathcal{B}(X)$ , is the  $\sigma$ -algebra generated by the collection of all open sets in X.

**Example.** We see that  $\mathcal{B}(\mathbb{R})$  contains

- $\mathcal{E}_1 = \{(a, b) \mid a < b; a, b \in \mathbb{R}\}.$
- $\mathcal{E}_2 = \{ [a, b] \mid a < b; a, b \in \mathbb{R} \} \text{ since } [a, b] = \bigcap_{n=1}^{\infty} (a \frac{1}{n}, b + \frac{1}{n}).$
- $\mathcal{E}_3 = ((a, b] \mid a < b; a, b \in \mathbb{R}) \text{ since } (a, b] = \bigcap_{n=1}^{\infty} (a, b + \frac{1}{n}).$
- $\mathcal{E}_4 = ([a,b) \mid a < b; a, b \in \mathbb{R}) \text{ since } [a,b) = \bigcap_{n=1}^{\infty} (a \frac{1}{n}, b).$

- $\mathcal{E}_5 = ((a, \infty) \mid a \in \mathbb{R}) \text{ since } (a, \infty) = \bigcup_{n=1}^{\infty} (a, a+n).$
- $\mathcal{E}_6 = ([a, \infty) \mid a \in \mathbb{R}) \text{ since } [a, \infty) = \bigcup_{n=1}^{\infty} [a, a+n).$
- $\mathcal{E}_7 = ((-\infty, b) \mid b \in \mathbb{R}) \text{ since } (-\infty, b) = \bigcup_{n=1}^{\infty} (b n, b).$
- $\mathcal{E}_8 = ((-\infty, b] \mid a \in \mathbb{R}) \text{ since } (-\infty, b] = \bigcup_{n=1}^{\infty} (b n, b].$

**Proposition 1.1.**  $\mathcal{B}(\mathbb{R}) = \langle \mathcal{E}_i \rangle$  for each i = 1, ..., 8.

*Proof.* Firstly, we see that  $\mathcal{E}_i \subset \mathcal{B}(\mathbb{R}) \implies \langle \mathcal{E}_i \rangle \subset \mathcal{B}(\mathbb{R})$  by Lemma 1.2. Secondly, by definition,  $\mathcal{B}(\mathbb{R}) = \langle \mathcal{E} \rangle$  where

$$\mathcal{E} = \{ O \subseteq \mathbb{R} \mid O \text{ is open in } \mathbb{R} \}.$$

It's enough to show  $\mathcal{E} \subset \langle \mathcal{E}_i \rangle$  since if so,  $\langle \mathcal{E} \rangle \subseteq \langle \mathcal{E}_i \rangle$ , and clearly  $\langle \mathcal{E} \rangle \supseteq \langle \mathcal{E}_i \rangle = \mathcal{B}(\mathbb{R})$ , then we will have  $\langle \mathcal{E} \rangle = \langle \mathcal{E}_i \rangle$ . Let  $O \subset \mathbb{R}$  be an open set, i.e.,  $O \in \mathcal{E}$ . We claim that every open set in  $\mathbb{R}$  is a countable union of disjoint open intervals.<sup>2</sup>

Thus,

$$O = \bigcup_{j=1}^{\infty} I_j,$$

where  $I_j$  open interval with the form of  $(a, b), (-\infty, b), (a, \infty), (-\infty, \infty)$ .

For example,  $\mathcal{E}_1$  is trivially true, and

$$(a,b) = \bigcup_{n=1}^{\infty} \underbrace{\left[a + \frac{1}{n}, b - \frac{1}{n}\right]}_{\in \mathcal{E}_2}$$

shows the case for  $\mathcal{E}_2$  and

$$(a,\infty) = \bigcup_{k=1}^{\infty} (a, a+k)$$

shows the case for  $\mathcal{E}_5$ . It's now straightforward to check open intervals are in  $\langle \mathcal{E}_i \rangle$  for every i.

Now, to put a structure on a space, we define the following.

Definition 1.5 (Measurable space, A-measurable set). A measurable space or Borel space is a tuple of a set X and a  $\sigma$ -algebra A on X, denoted by (X, A).

Furthermore, for every  $E \in \mathcal{A}$  is called an  $\mathcal{A}$ -measurable set.

<sup>&</sup>lt;sup>2</sup>https://math.stackexchange.com/questions/318299/any-open-subset-of-bbb-r-is-a-countable-union-of-disjoint-open-intervals

#### 1.2 Measures

With the definition of measurable space, we now can refine our measure function  $\mu$  as follows.

**Definition 1.6 (Measure, Measure space).** Given a measurable space on  $(X, \mathcal{A})$ , a measure is a function  $\mu$  such that

$$\mu \colon \mathcal{A} \to [0, \infty]$$

with

1.  $\mu(\emptyset) = 0$ 

2.  $\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mu(A_i)$  if  $A_1, A_2, \ldots \in \mathcal{A}$  are **disjoint**. We call this Countable additivity.

We denote  $(X, \mathcal{A}, \mu)$  a measure space.

**Notation.** We denote  $[0, \infty] := [0, \infty) \cup \{\infty\}$ .

**Remark.** The motivation of why we only want *countable additivity* but not uncountable additivity can be seen by the following example. We'll consider the most intuitive measure on  $\mathbb{R}, \mathcal{B}(\mathbb{R})$ .

Since we have

$$(0,1] = (1/2,1] \cup (1/4,1/2] \cup (1/8,1/4] \cup \dots$$

and also

$$(0,1] = \bigcup_{x \in (0,1]} \{x\}.$$

Specifically, in the first case, we are claiming that

$$1 = \underbrace{\frac{1}{2}}_{\mu((\frac{1}{2},1])} + \underbrace{\frac{1}{4}}_{\mu((\frac{1}{4},\frac{1}{2}])} + \underbrace{\frac{1}{8}}_{\mu((\frac{1}{8},\frac{1}{4}])} + \dots;$$

while in the second case, we are claiming that

$$1 = \sum_{x \in (0,1]} 0$$

since  $\mu(x) = 0$  for  $x \in \mathbb{R}$ , which is clearly not what we want.

**Example.** We see some examples.

- 1. For any (X, A), we let  $\mu(A) := \#A$ . This is called *counting measure*.
- 2. Let  $x_0 \in X$ . For any (X, A), the Dirac measure at  $x_0$  is

$$\mu(A) = \begin{cases} 1, & \text{if } x_0 \in A; \\ 0, & \text{if } x_0 \notin A. \end{cases}$$

3. For  $(\mathbb{N}, \mathcal{P}(\mathbb{N}))$ ,

$$\mu(A) = \sum_{i \in A} a_i,$$

where  $a_1, a_2, ... \in [0, \infty)$ .

#### Lecture 3: Construct a Measure

10 Jan. 11:00

**Note.** If  $A, B \in \mathcal{A}$  and  $A \subset B$ , then

$$\mu(B \setminus A) + \mu(A) = \mu(B) \implies \mu(B \setminus A) = \mu(B) - \mu(A) \text{ if } \mu(A) < \infty.$$

**Theorem 1.1.** Given  $(X, \mathcal{A}, \mu)$  be a measure space.

1. (monotonicity)

$$A, B \in \mathcal{A}, A \subset B \implies \mu(A) \le \mu(B).$$

2. (countable subadditivity)

$$A_1, A_2, \ldots \in \mathcal{A} \implies \mu\left(\bigcup_{i=1}^{\infty} A_i\right) \leq \sum_{i=1}^{\infty} \mu(A_i).$$

3. (continuity from below/ monotone convergence theorem (MCT) for sets)

$$\begin{cases} A_1, A_2, \dots \in \mathcal{A} \\ A_1 \subset A_2 \subset A_3 \subset \dots \end{cases} \implies \mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \lim_{n \to \infty} \mu(A_n).$$

4. (continuity from above)

$$\begin{cases} A_1, A_2, \dots \in \mathcal{A} \\ A_1 \supset A_2 \supset A_3 \supset \dots \implies \mu \left(\bigcap_{i=1}^{\infty} A_i\right) = \lim_{n \to \infty} \mu(A_n). \\ \mu(A_1) < \infty \end{cases}$$

*Proof.* We prove this theorem one by one.

1. Since  $A \subset B$ , hence we have

$$\mu(B) = \mu\left(\underbrace{(B \setminus A)}_{\text{disjoint}} \cup \underline{A}\right) \stackrel{!}{=} \underbrace{\mu(B \setminus A)}_{>0} + \mu(A) \ge \mu(A).$$

2. This should be trivial from countable additivity with the fact that  $\mu(A) \ge 0$  for all A.

DIY!

3. Let  $B_1 = A_1$ ,  $B_i = A_i \setminus A_{i-1}$  for  $i \geq 2$ , then

$$\bigcup_{i=1}^{\infty} A_i = \bigcup_{i=1}^{\infty} B_i$$

1 MEASURE

are a disjoint union and  $B_i \in \mathcal{A}$ , hence we see that

$$\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mu(B_i) = \lim_{n \to \infty} \sum_{i=1}^{n} \mu(B_i).$$

With 
$$\mu\left(\bigcup_{i=1}^n B_i\right) = \mu(A_n)$$
, we have

$$\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \lim_{n \to \infty} \sum_{i=1}^{n} \mu(B_i) = \lim_{n \to \infty} \mu\left(\bigcup_{i=1}^{n} B_i\right) = \lim_{n \to \infty} \mu(A_n).$$

4. Let  $E_i = A_1 \setminus A_i \implies E_i \in \mathcal{A}, E_1 \subset E_2 \subset \dots$  We then have

$$\bigcup_{i=1}^{\infty} E_i = \bigcup_{i=1}^{\infty} (A_1 \setminus A_i) = A_1 \setminus \left(\bigcap_{i=1}^{\infty} A_i\right),$$

which implies

$$\bigcap_{i=1}^{\infty} A_i = A_1 \setminus \left(\bigcup_{i=1}^{\infty} E_i\right) \implies \mu\left(\bigcap_{i=1}^{\infty} A_i\right) = \mu(A_1) - \mu\left(\bigcup_{i=1}^{\infty} E_i\right)$$

since  $\mu\left(\bigcup_{i=1}^{\infty} E_i\right) \leq \mu(A_1) < \infty$ . Then from continuity from below, we further have

$$\mu\left(\bigcap_{i=1}^{\infty} A_i\right) = \mu(A_1) - \lim_{n \to \infty} \mu(E_n) = \mu(A_1) - \lim_{n \to \infty} (\mu(A_1) - \mu(A_n)).$$

From monotonicity, we see that  $\mu(A_n) \leq \mu(A_1) < \infty$ , hence we can split the limit and further get

$$\mu\left(\bigcap_{i=1}^{\infty} A_i\right) = \mu(A_1) - \mu(A_1) + \lim_{n \to \infty} \mu(A_n) = \lim_{n \to \infty} \mu(A_n).$$

**Example.** Given  $(\mathbb{N}, \mathcal{P}(\mathbb{N}), \text{ counting measure})$ . Then we see

- $A_n = \{n, n+1, n+2, \ldots\} \implies \mu(A_n) = \infty$
- $A_1 \supset A_2 \supset A_3 \supset \dots$

$$\bullet \bigcap_{i=1}^{\infty} A_i = \emptyset \implies \mu \left( \bigcap_{i=1}^{\infty} A_i \right) = 0$$

**Remark.** We see that in this case, since  $\mu(A_1) \not< \infty$ , hence continuity from above doesn't hold.

We now try to characterize some properties of a measure space.

Definition 1.7 ( $\mu$ -null,  $\mu$ -subnull, Complete measure space). Given  $(X, \mathcal{A}, \mu)$ ,

- $A \subset X$  is a  $\mu$ -null set if  $A \in \mathcal{A}$  and  $\mu(A) = 0$ .
- $A \subset X$  is a  $\mu$ -subnull set if  $\exists \mu$ -null set B such that  $A \subset B$ .
- $(X, \mathcal{A}, \mu)$  is a *complete* measure space if every  $\mu$ -subnull set is  $\mathcal{A}$ -measurable.

**Note.** We see that for a  $\mu$ -subnull set, it's not necessary  $\mathcal{A}$ -measurable.

There are some useful terminologies we'll use later relating to  $\mu$ -null.

**Definition 1.8 (Almost everywhere).** Given  $(X, \mathcal{A}, \mu)$ , a statement P(x),  $x \in X$  holds  $\mu$ -almost everywhere (a.e.) if the set

$$\{x \in X : P(x) \text{ does not hold}\}\$$

is  $\mu$ -null.

It's always pleasurable working with finite rather than infinite, hence we give the following definition.

Definition 1.9 (Finite measure). Given  $(X, A, \mu)$ 

- $\mu$  is a finite measure if  $\mu(X) < \infty$ .
- $\mu$  is a  $\sigma$ -finite measure if  $X = \bigcup_{n=1}^{\infty} X_n, X_n \in \mathcal{A}, \mu(X_n) < \infty$ .

**Exercise.** Every measure space can be **completed**. Namely, we can always find a bigger  $\sigma$ -algebra to complete the space.

#### 1.3 Outer Measures

We start by giving a definition.

**Definition 1.10 (Outer measure).** An outer measure on X is a map

$$\mu^* \colon \mathcal{P}(X) \to [0, \infty]$$

such that

- $\mu^*(\emptyset) = 0$
- (monotonicity)  $\mu^*(A) \leq \mu^*(B)$  if  $A \subset B$
- (countable subadditivity)  $\mu^* \left( \bigcup_{i=1}^{\infty} A_i \right) \leq \sum_{i=1}^{\infty} \mu^*(A_i)$  for every  $A_i \subset X$ .

**Example.** For  $A \subset \mathbb{R}$ ,

$$\mu^*(A) = \inf \left\{ \sum_{i=1}^{\infty} (b_i - a_i) \colon \bigcup_{i=1}^{\infty} (a_i, b_i) \supset A \right\}$$

is an outer measure due to the Proposition 1.2 we're going to show.

**Remark.** We see that an outer measure need not be a measure. Check Definition 1.6.

**Proposition 1.2.** Let  $\mathcal{E} \subset \mathcal{P}(X)$  such that  $\emptyset, X \in \mathcal{E}$ . Let

$$\rho \colon \mathcal{E} \to [0, \infty]$$

such that  $\rho(\emptyset) = 0$ . Then

$$\mu^*(A) := \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) : \bigvee_{i \in \mathbb{N}} E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset A \right\}$$

is an outer measure on X.

**Note.** Recall the Tonelli's Theorem<sup>3</sup> for series:

If  $a_{ij} \in [0, \infty], \forall i, j \in \mathbb{N}$ , then

$$\sum_{(i,j)\in\mathbb{N}^2} a_{ij} = \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} a_{ij} = \sum_{j=1}^{\infty} \sum_{i=1}^{\infty} a_{ij}.$$

Specifically, in [Tao13] Theorem 0.0.2.

## Lecture 4: Carathéodory extension Theorem

12 Jan. 11:00

As previously seen. We now prove the Proposition 1.2.

*Proof.* We need to prove

- $\mu^*$  is well-defined. i.e., inf is taken over a non-empty set. This is trivial since  $X \in \mathcal{E}$  and  $X \supset A$  for any  $A \in \mathcal{E}$ .
- $\mu^*(\varnothing) = 0$ . Since  $\varnothing \in \mathcal{E}$  and

$$\mu^*(\varnothing) = \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) : \forall E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset \varnothing \right\} = 0$$

since  $\rho(\varnothing) = 0$  for all i and further, by Squeeze Theorem<sup>4</sup>, we see that  $\lim_{n \to \infty} \sum_{i=1}^{n} \rho(\varnothing) = 0$ .

<sup>3</sup>https://en.wikipedia.org/wiki/Fubini%27s\_theorem

<sup>4</sup>https://en.wikipedia.org/wiki/Squeeze\_theorem

•  $A \subset B \implies \mu^*(A) \leq \mu^*(B)$ . We simply show this by contradiction. Suppose  $A \subset B$  and  $\mu^*(A) > \mu^*(B)$ , then by definition of  $\mu^*$ , we have

$$\mu^*(A) = \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) \colon \bigvee_{i \in \mathbb{N}} E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset A \right\}$$
$$> \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) \colon \bigvee_{i \in \mathbb{N}} E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset B \right\} = \mu^*(B).$$

Now, let  $B =: (B \setminus A) \cup A$ , then we have

$$\mu^*(A) = \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) \colon \forall E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset A \right\}$$
$$> \inf \left\{ \sum_{i=1}^{\infty} \rho(E_i) \colon \forall E_i \in \mathcal{E}, \bigcup_{i=1}^{\infty} E_i \supset (B \setminus A) \cup A \right\} = \mu^*(B).$$

Now, since  $B \setminus A \supseteq \emptyset$ , then this inequality can't hold, hence a contradiction  $\oint$ .

• Countable subadditivity. Let  $A_1, A_2, \ldots \in X$ . If one of  $\mu^*(A_n) = \infty$ , then result holds. So we may assume  $\mu^*(A_n) < \infty$  for all  $n \in \mathbb{N}$ . Now, fix any  $\epsilon > 0$ , we will show that

$$\mu^* \left( \bigcup_{n=1}^{\infty} A_n \right) \le \sum_{n=1}^{\infty} \mu^* (A_n) + \epsilon.$$

For each  $n \in \mathbb{N}$ ,  $\exists E_{n,1}, E_{n,2}, \ldots \in \mathcal{E}$  such that

$$\bigcup_{k=1}^{\infty} E_{n,k} \supset A_n$$

and

$$\mu^*(A_n) + \frac{\epsilon}{2^n} > \sum_{k=1}^{\infty} \rho(E_{n,k}).$$

Then we see that

$$\bigcup_{k=1}^{\infty} A_n \subset \bigcup_{n=1}^{\infty} \bigcup_{k=1}^{\infty} E_{k,n} = \bigcup_{(n,k) \in \mathbb{N}^2} E_{k,n},$$

which implies

$$\mu^* \left( \bigcup_{n=1}^{\infty} A_n \right) \le \sum_{(n,k) \in \mathbb{N}^2} \rho \left( E_{k,n} \right) \stackrel{!}{=} \sum_{n=1}^{\infty} \sum_{k=1}^{\infty} \rho(E_{k,n}) \le \sum_{n=1}^{\infty} \left( \mu^*(A_n) + \frac{\epsilon}{2^n} \right)$$

from the inequality just derived. Now, since the last term is just

$$\sum_{n=1}^{\infty} \left( \mu^*(A_n) + \frac{\epsilon}{2^n} \right) = \sum_{n=1}^{\infty} \mu^*(A_n) + \epsilon,$$

<sup>&</sup>lt;sup>5</sup>This is an important trick!!

hence we finally have

$$\mu^* \left( \bigcup_{n=1}^{\infty} A_n \right) \le \sum_{n=1}^{\infty} \mu^*(A_n) + \epsilon$$

for arbitrarily small fixed  $\epsilon > 0$ , hence the subadditivity is proved.

**Definition 1.11 (Carathéodory measurable).** Let  $\mu^*$  be an outer measure on X. We say  $A \subset X$  is Carathéodory measurable (C-measurable) with respect to  $\mu^*$  if

$$\forall E \subset X, \ \mu^*(E) = \mu^* (E \cap A) + \mu^* (E \setminus A).$$

**Lemma 1.3.** Let  $\mu^*$  be an outer measure on X. Suppose  $B_1, \ldots, B_N$  are disjoint C-measurable sets. Then,

$$\forall E \subset X, \ \mu^* \left( E \cap \left( \bigcup_{i=1}^N B_i \right) \right) = \sum_{i=1}^N \mu^* \left( E \cap B_i \right).$$

*Proof.* Since we have

$$\mu^* \left( E \cap \left( \bigcup_{i=1}^N B_i \right) \right) = \mu^* \left( E' \cap B_1 \right) + \mu^* \left( E' \setminus B_1 \right)^6$$

$$= \mu^* \left( E \cap \left( \bigcup_{i=1}^N B_i \cap B_1 \right) \right) + \mu^* \left( E \cap \left( \bigcup_{i=1}^N B_i \right) \cap B_1^c \right)$$

$$= \mu^* (E \cap B_1) + \mu^* \left( E \cap \left( \bigcup_{i=2}^N B_i \right) \right)$$

where the equality comes from the fact that  $B_1$  is C-measurable and disjoint from  $B_i$ ,  $i \neq 1$ . Then, we simply iterate this argument and have the result.

**Remark.** This implies that if we restrict an outer measure on a C-measurable set, then it becomes finite additive.

Theorem 1.2 (Carathéodory extension Theorem). Let  $\mu^*$  be an outer measure on X. Let  $\mathcal{A}$  be the collection of C-measurable sets (with respect to  $\mu^*$ ). Then,

- 1.  $\mathcal{A}$  is a  $\sigma$ -algebra on X.
- 2.  $\mu = \mu^*|_{\mathcal{A}}$  is a measure on  $(X, \mathcal{A})$ .
- 3.  $(X, \mathcal{A}, \mu)$  is a complete measure space.

<sup>&</sup>lt;sup>6</sup>Here,  $E' := E \cap \left(\bigcup_{i=1}^{N} B_i\right)$  for the simplicity of notation.

*Proof.* We divide the proof in several steps.

- 1. We show  $\mathcal{A}$  is a  $\sigma$ -algebra by showing
  - (a)  $\varnothing \in \mathcal{A}$ . To show this, we simply check that  $\varnothing$  is C-measurable. We see that

$$\label{eq:multiple} \underset{E\subset X}{\forall}\ \mu^*(E) = \mu^*(E\cap\varnothing) + \mu^*(E\setminus\varnothing) = \mu^*(E),$$

which just shows  $\emptyset \in \mathcal{A}$ .

(b)  $\mathcal{A}$  closed under complements. This is equivalent to say that if A is C-measurable, so is  $A^c$ . We see that if A is C-measurable, then for every  $E \subset X$ ,

$$\mu^*(E) = \mu^*(E \cap A) + \mu^*(E \setminus A).$$

Observing that  $E \cap A = E \setminus A^c$  and  $E \setminus A = E \cap A^c$ , hence

$$\mu^*(E) = \mu^*(E \setminus A^c) + \mu^*(E \cap A^c).$$

We immediately see that above implies  $A^c \in \mathcal{A}$ .

(c)  $\mathcal{A}$  closed under countable unions.

**Note.** To show  $\mathcal{A}$  closed under countable unions, we show that  $\mathcal{A}$  is closed under:

finite unions  $\stackrel{\text{then}}{\Longrightarrow}$  countable disjoint unions  $\stackrel{\text{then}}{\Longrightarrow}$  countable unions.

ullet We show  $\mathcal A$  is closed under finite unions.

Claim. 
$$A, B \in \mathcal{A} \implies A \cup B \in \mathcal{A}$$
.

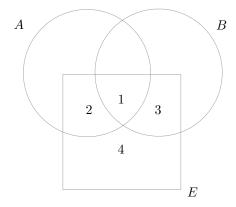
Fix  $E \subset X$  arbitrary. We need to show that

$$\mu^*(E) = \mu^*(E \cap (A \cup B)) + \mu^*(E \setminus (A \cup B)),$$

i.e.,

$$\mu^*(1 \cup 2 \cup 3 \cup 4) = \mu^*(1 \cup 2 \cup 3) + \mu^*(4)$$

given  $A, B \in \mathcal{A}$ .



- Since A is C-measurable,
  - \*  $\mu^*(1 \cup 2 \cup 3 \cup 4) = \mu^*(1 \cup 2) + \mu^*(3 \cup 4)$
  - \*  $\mu^*(1 \cup 2 \cup 3) = \mu^*(1 \cup 2) + \mu^*(3)$
- Since B is C-measurable,

\* 
$$\mu^*(3 \cup 4) = \mu^*(3) + \mu^*(4)$$

Hence, we have

$$\mu^*(1 \cup 2 \cup 3 \cup 4) = \mu^*(1 \cup 2) + \mu^*(3 \cup 4)$$
$$= \mu^*(1 \cup 2) + \mu^*(3) + \mu^*(4)$$
$$= \mu^*(1 \cup 2 \cup 3) + \mu^*(4).$$

 $\bullet$  We show  $\mathcal{A}$  is closed under countable disjoint unions.

Let  $A_1, A_2, \ldots \in \mathcal{A}$  and <u>disjoint</u>. Fix  $E \subset X$  arbitrary. Since  $\mu^*$  is countably subadditive,

$$\mu^*(E) \le \mu^* \left( E \cap \bigcup_{i=1}^{\infty} A_i \right) + \mu^* \left( E \setminus \bigcup_{i=1}^{\infty} A_i \right),$$

hence we only need to show another way around.

Fix  $N \in \mathbb{N}$ , we have  $\bigcup_{n=1}^{N} A_n \in \mathcal{A}$  since N is finite, and

$$\mu^{*}(E) = \mu^{*} \left( E \cap \left( \bigcup_{n=1}^{N} A_{n} \right) \right) + \mu^{*} \left( E \setminus \left( \bigcup_{n=1}^{N} A_{n} \right) \right)$$

$$\geq \sum_{n=1}^{N} \mu^{*}(E \cap A_{n}) + \mu^{*} \left( E \setminus \bigcup_{n=1}^{\infty} A_{n} \right).$$

$$\stackrel{!}{=} \mu^{*} \left( E \cap \left( \bigcup_{n=1}^{N} A_{n} \right) \right) \xrightarrow{\leq \mu^{*} \left( E \setminus \left( \bigcup_{n=1}^{N} A_{n} \right) \right)}.$$

Now, take  $N \to \infty$  then we are done.

ullet We show  ${\mathcal A}$  is closed under countable unions.

DIY

The proof will be continued...

### Lecture 5: Hahn-Kolmogorov Theorem

14 Jan. 11:00

Firstly, we see a stronger version of Lemma 1.3 we have seen before.

**Lemma 1.4.** Let  $\mu^*$  be an outer measure on X. Suppose  $B_1, B_2, \ldots$  are disjoint C-measurable sets. Then,

$$\forall E \subset X, \ \mu^* \left( E \cap \left( \bigcup_{i=1}^{\infty} B_i \right) \right) = \sum_{i=1}^{\infty} \mu^* \left( E \cap B_i \right).$$

Proof.

$$\sum_{n=1}^{\infty} \mu^*(E \cap B_i) \ge \mu^* \left( E \cap \bigcup_{n=1}^{\infty} B_n \right) \ge \mu^* \left( E \cap \left( \bigcup_{n=1}^{N} B_n \right) \right) \stackrel{!}{=} \sum_{n=1}^{N} \mu^* \left( E \cap B_n \right).$$

Now, we just take  $N \to \infty$  (or note that  $N \in \mathbb{N}$  is arbitrary, we then get the result according to Squeeze Theorem<sup>7</sup>).

Let's continue the proof of Theorem 1.2.

- 2. Since from Definition 1.6, we need to show
  - $\mu(\varnothing) = 0$ . This means that we need to show  $\mu^*|_{\mathcal{A}}(\varnothing) = 0$ . Since  $\varnothing \in \mathcal{A}$  and  $\mu^*$  is an outer measure, hence from the property of outer measure, it clearly holds.
  - Countable additivity of  $\mu^*$  on  $\mathcal{A}$  follows from the Lemma 1.4 with E=X
- 3. The proof is given in Theorem A.1.

## 1.4 Hahn-Kolmogorov Theorem

We see that we can start with any collection of open sets  $\mathcal{E}$  and any  $\rho$  such that it assigns measure on  $\mathcal{E}$ , then it induces an outer measure by Proposition 1.2, finally complete the outer measure by Theorem 1.2.

Specifically, we have

$$(\mathcal{E}, \rho) \xrightarrow{\operatorname{Proposition } 1.2} (\mathcal{P}(X), \mu^*) \xrightarrow{\operatorname{Theorem } 1.2} (\mathcal{A}, \mu)$$

To introduce this concept, we see that we can start with a more general definition compared to  $\sigma$ -algebra we are working on till now.

**Definition 1.12 (Algebra).** Let X be a set. A collection  $\mathcal{A}$  of subsets of X, i.e.,  $\mathcal{A} \subset \mathcal{P}(X)$  is called an *algebra on* X if

- $\varnothing \in \mathcal{A}$ .
- $\mathcal{A}$  is closed under complements. i.e., if  $A \in \mathcal{A}$ ,  $A^c = X \setminus A \in \mathcal{A}$ .
- $\mathcal{A}$  is closed under **finite** unions. i.e., if  $A_i \in \mathcal{A}$ , then  $\bigcup_{i=1}^n A_i \in \mathcal{A}$  for  $n < \infty$ .

**Remark.** The only difference between an algebra and a  $\sigma$ -algebra is whether they closed under **countable** unions in the definition.

Now, we can look at a more general setup compared to an outer measure.

<sup>&</sup>lt;sup>7</sup>https://en.wikipedia.org/wiki/Squeeze\_theorem

**Definition 1.13 (Pre-measure).** Let  $A_0$  be an algebra on X. We say

$$\mu_0 \colon \mathcal{A}_0 \to [0, \infty]$$

is a pre-measure if

- 1.  $\mu_0(\emptyset) = 0$
- 2. (finite additivity)  $\mu_0\left(\bigcup_{i=1}^n A_i\right) = \sum_{i=1}^n \mu_0(A_i)$  if  $A_1, \ldots, A_n \in \mathcal{A}_0$  are disjoint.
- 3. (countable additivity within the algebra) If  $A \in \mathcal{A}_0$  and  $A = \bigcup_{n=1}^{\infty} A_n$ ,  $A_n \in \mathcal{A}_0$ , disjoint, then

$$\mu_0(A) = \sum_{n=1}^{\infty} \mu_0(A_n).$$

**Lemma 1.5.**  $(1) + (3) \implies (2)$  in Definition 1.13.

*Proof.* It's easy to see that since  $\mu_0$  is monotone.

Theorem 1.3 (Hahn-Kolmogorov Theorem). Let  $\mu_0$  be a pre-measure on algebra  $\mathcal{A}_0$  on X. Let  $\mu^*$  be the outer measure induced by  $(\mathcal{A}_0, \mu_0)$  in Proposition 1.2. Let  $\mathcal{A}$  and  $\mu$  be the Carathéodory  $\sigma$ -algebra and measure for  $\mu^*$ , then  $(\mathcal{A}, \mu)$  extends  $(\mathcal{A}_0, \mu_0)$ . i.e.,

$$\mathcal{A} \supset \mathcal{A}_0, \quad \mu|_{\mathcal{A}_0} = \mu_0.$$

*Proof.* We prove this theorem in two parts.

• We first show  $A \supset A_0$ . Let  $A \in A_0$ , we want to show  $A \in A$ , i.e., A is C-measurable, i.e.,

$$\forall E \subset X \ \mu^*(E) = \mu^*(E \cap A) + \mu^*(E \cap A^c).$$

We first fix an  $E \subset X$ . From countable subadditivity of  $\mu^*$ , we have

$$\mu^*(E) \le \mu^*(E \cap A) + \mu^*(E \cap A^c).$$

Hence, we only need to show another direction. If  $\mu^*(E) = \infty$ , then  $\mu^*(E) = \infty \ge \mu^*(E \cap A) + \mu^*(E \cap A^c)$  clearly. So, assume  $\mu^*(E) < \infty$ .

Fix  $\epsilon > 0$ . By the Proposition 1.2 of  $\mu^*$ ,  $\exists B_1, B_2, \ldots \in \mathcal{A}_0$ ,  $\bigcup_{n=1}^{\infty} B_n \supset E$  such that

$$\mu^*(E) + \epsilon \stackrel{!}{\geq} \sum_{n=1}^{\infty} \mu_0(B_n) = \sum_{n=1}^{\infty} \left( \mu_0(\underbrace{B_n \cap A}_{\in \mathcal{A}_0}) + \mu_0(\underbrace{B_n \cap A^c}_{\in \mathcal{A}_0}) \right)$$

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by the finite additivity of  $\mu_0$ . Note that

$$\begin{cases} \bigcup_{n=1}^{\infty} (B_n \cap A) \supset E \cap A \\ \bigcup_{n=1}^{\infty} (B_n \cap A^c) \subset E \cap A^c \end{cases} \Longrightarrow \mu^*(E) + \epsilon \ge \mu^*(E \cap A) + \mu^*(E \cap A^c)$$

since

$$\mu^*(E \cap A) \le \mu^* \left( \bigcup_{n=1}^{\infty} (B_n \cap A) \right) \le \sum_{n=1}^{\infty} \mu^*(B_n \cap A)$$

and

$$\mu^*(E \cap A^c) \le \mu^* \left( \bigcup_{n=1}^{\infty} (B_n \cap A^c) \right) \le \sum_{n=1}^{\infty} \mu^*(B_n \cap A^c).$$

We then see that for any  $\epsilon > 0$ , the inequality

$$\mu^*(E) + \epsilon > \mu^*(E \cap A) + \mu^*(E \cap A^c)$$

holds, hence so does

$$\mu^*(E) \ge \mu^*(E \cap A) + \mu^*(E \cap A^c),$$

which implies  $A \supset A_0$ .

The proof will be continued...

## Lecture 6: Hahn-Kolmogorov Theorem and Extension.

18 Jan. 11:00

Let's continue the proof of Theorem 1.3.

• Let  $A \in \mathcal{A}_0$ , we want to show that

$$\mu(A) = \mu_0(A).$$

- Firstly, let

$$B_i = \begin{cases} A, & \text{if } i = 1\\ \varnothing, & \text{if } i \ge 2 \end{cases} \in \mathcal{A}_0,$$

hence  $\bigcup_{i=1}^{\infty} B_i = A$ , then we see that

$$\mu^*(A) \le \sum_{i=1}^{\infty} \mu_0(B_i) = \mu_0(A)$$

from the definition of  $\mu^*$  and countable additivity within the algebra of  $\mu_0$ .

– Secondly, let  $B_i \in \mathcal{A}_0$ ,  $\bigcup_{i=1}^{\infty} B_i \supset A$  be arbitrary. Let  $C_1 = A \cap B_1 \in \mathcal{A}_0$ ,  $C_i = A \cap B_i \setminus \left(\bigcup_{j=1}^{i-1} B_j\right) \in \mathcal{A}_0$  for  $i \geq 2$  since the operations are finite. Then we see

$$A = \bigcup_{i=1}^{\infty} C_i \in \mathcal{A}_0$$

are disjoint countable unions, by countable additivity within the algebra, we therefore have

$$\mu_0(A) = \sum_{i=1}^{\infty} \mu_0(C_i) \le \sum_{i=1}^{\infty} \mu_0(B_i) \implies \mu_0(A) \le \mu^*(A)$$

by taking the infimum from the definition of  $\mu^*$ .

Combine these two inequality, we see that

$$\mu^*(A) = \mu_0(A),$$

for every  $A \in \mathcal{A}_0$ , which implies

$$\mu(A) = \mu_0(A)$$

for every  $A \in \mathcal{A}_0$  from Theorem 1.2, where we extend  $\mu^*$  to  $\mu$  respect to  $\mathcal{A}_0$ .

**Definition 1.14 (HK extension).**  $(A, \mu)$  obtained from Theorem 1.3 is the *Hahn-Kolmogorov extensions* of  $(A_0, \mu_0)$ .

We can show the uniqueness of HK extension.

Theorem 1.4 (Uniqueness of HK extension). Let  $\mathcal{A}_0$  be an algebra on X,  $\mu_0$  be a pre-measure on  $\mathcal{A}_0$ . Let  $(\mathcal{A}, \mu)$  be the HK extension of  $(\mathcal{A}_0, \mu_0)$ . Let  $(\mathcal{A}', \mu')$  be another extension of  $(\mathcal{A}_0, \mu_0)$ . Then if  $\mu_0$  is  $\sigma$ -finite,  $\mu = \mu'$  on  $\mathcal{A} \cap \mathcal{A}'$ .

**Note.** Notice that  $A_0 \subset A$ , A' since they both extend  $A_0$ .

*Proof.* Let  $A \in \mathcal{A} \cap \mathcal{A}'$ , we need to show

$$\underbrace{\mu(A)}_{\mu^*(A)} = \mu'(A).$$

Firstly, it's easy to show that  $\mu^*(A) \ge \mu'(A)$  by choosing the arbitrary cover of A and using the definition of  $\mu^*$ .

Secondly, we will show that  $\mu(A) \leq \mu'(A)$ .

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• Assume  $\mu(A) < \infty$ , and fix  $\epsilon > 0$ . Then there exists  $B_i \in \mathcal{A}_0$  with  $B := \bigcup_{i=1}^{\infty} B_i \supset A$  such that

$$\mu(A) + \epsilon = \mu^*(A) + \epsilon \stackrel{!}{\geq} \sum_{i=1}^{\infty} \mu_0(B_i) \stackrel{!}{=} \sum_{i=1}^{\infty} \mu(B_i) \geq \mu\left(\bigcup_{i=1}^{\infty} B_i\right) = \mu(B).$$

This implies that

$$\mu(B \setminus A) = \mu(B) - \mu(A) \le \epsilon$$

where the first equality comes from  $A \subset B$  and  $\mu(A) < \infty$ . On the other hand,

$$\mu(B) = \lim_{N \to \infty} \mu\left(\bigcup_{i=1}^{N} B_i\right) = \lim_{N \to \infty} \mu'\left(\bigcup_{i=1}^{N} B_i\right) = \mu'(B),$$

hence,

$$\mu(A) \le \mu(B) = \mu'(B) = \mu'(A) + \mu'(B \setminus A) \stackrel{9}{\le} \mu'(A) + \mu(B \setminus A) \le \mu'(A) + \epsilon$$
 for arbitrary  $\epsilon$ , so we conclude  $\mu(A) \le \mu'(A)$ .

• Assume  $\mu(A) = \infty$ . Since  $\mu_0$  is  $\sigma$ -finite, so we know  $X = \bigcup_{n=1}^{\infty} X_n$  for some  $X_n \in \mathcal{A}_0$  such that

$$\mu_0(X_n) < \infty.$$

Replacing  $X_n$  by  $X_1 \cup \ldots \cup X_n \in A_0$ , we may assume that

$$X_1 \subset X_2 \subset \dots$$

Then,

$$\bigvee_{n \in \mathbb{N}} \mu(A \cap X_n) < \infty \stackrel{!}{\Longrightarrow} \mu(A \cap X_n) \le \mu'(A \cap X_n).$$

From the continuity of measure, we then have

$$\mu(A) = \lim_{n \to \infty} \mu(A \cap X_n) \le \lim_{n \to \infty} \mu'(A \cap X_n) = \mu'(A).$$

 $<sup>^8\</sup>mu = \mu' \text{ on } \mathcal{A}_0$ 

<sup>&</sup>lt;sup>9</sup>From the first part.

Corollary 1.1. Let  $\mu_0$  be a pre-measure on algebra  $\mathcal{A}_0$  on X. Suppose  $\mu_0$  is  $\sigma$ -finite, then

 $\exists!$  measure  $\mu$  on  $\langle \mathcal{A}_0 \rangle$  that extends  $\mathcal{A}_0$ .

Furthermore,

- The completion of  $(X, \langle A_0 \rangle, \mu)$  is the HK extension of  $(A_0, \mu_0)$ .

$$\mu(A) = \inf \left\{ \sum_{i=1}^{\infty} \mu_0(B_i) \mid B_i \in \mathcal{A}_0, \, \forall \bigcup_{i=1}^{\infty} B_i \supset A \right\}$$

for all  $A \in \overline{\langle \mathcal{A}_0 \rangle}$ .

## Lecture 7: Borel Measures

21 Jan. 11:00

#### 1.5 Borel Measures on $\mathbb{R}$

We first introduce so-called distribution function.

**Definition 1.15 (Distribution function).** An increasing a function

$$F: \mathbb{R} \to \mathbb{R}$$

and right-continuous. F is then a distribution function.

<sup>a</sup>Here, increasing means  $F(x) \leq F(y)$  for x < y.

**Example.** Here are some examples of right-continuous functions.

- 1. F(x) = x.
- 2.  $F(x) = e^x$ .
- 3. Define

$$F(x) = \begin{cases} 1, & \text{if } x \ge 0 \\ 0, & \text{if } x < 0. \end{cases}$$

4. Let  $\mathbb{Q} := \{r_1, r_2, \ldots\}$ . Define

$$F_n(x) = \begin{cases} 1, & \text{if } x \ge r_n \\ 0, & \text{if } x < r_n, \end{cases}$$

and

$$F(x) := \sum_{n=1}^{\infty} \frac{F_n(x)}{2^n}.$$

Then F is a distribution function (hence right-continuous). This is shown in Lemma A.1.

Note. If F is increasing, and

$$F(\infty)\coloneqq \lim_{x\nearrow\infty} F(x), \quad F(-\infty)\coloneqq \lim_{x\searrow\infty} F(x)$$

exist in  $[-\infty, \infty]$ .

In probability theory, cumulative distribution function (CDF) is a distribution function with  $F(\infty) = 1$ ,  $F(-\infty) = 0$ .

Now, we can define a *Borel measure* on  $(X, \mathcal{B}(\mathbb{R}))$ .

**Definition 1.16 (Borel messure).** A Borel measure is any measure  $\mu$  defined on the  $\sigma$ -algebra of Borel sets.

**Definition 1.17 (Locally finite).** Let X be a Hausdorff topological space,  $\mu$  on  $(X, \mathcal{B}(X))$  is called *locally finite* if  $\mu(K) < \infty$  for every compact set  $K \subset X$ .

**Note.** Some authors will require a Borel measure equipped with the locally finite property. But formally, this is not so common.

**Lemma 1.6.** Let  $\mu$  be a locally finite Borel measure on  $\mathbb{R}$ , then

$$F_{\mu}(x) = \begin{cases} \mu((0, x]), & \text{if } x > 0 \\ 0, & \text{if } x = 0 \\ -\mu((x, 0]), & \text{if } x < 0 \end{cases}$$

is a distribution function.

*Proof.* To show  $F_{\mu}$  is increasing, consider x < y such that

$$F_{\mu}(x) \leq F_{\mu}(y)$$

by considering

• x > 0: Then  $F_{\mu}(x) = \mu((0, x])$  and

$$F_{\mu}(y) = \mu((0, y]) = \mu((0, x] \cup (x, y]) \ge \mu((0, x]) = F_{\mu}(x).$$

• x = 0: Then  $F_{\mu}(x) = 0$  and

$$F_{\mu}(y) = \mu((0, y]) \ge 0 = F_{\mu}(0)$$

since y > 0.

• x < 0: Follows the same argument with x > 0.

<sup>&</sup>lt;sup>10</sup>There are distributions [FF99] Ch9., but these are different from distribution functions.

Now, we need to show  $F_{\mu}$  is right-continuous. Firstly, assume that  $x \geq 0$ , then we see that

$$F_{\mu}(x) = \mu((0, x]) = \mu((0, x^{+}])$$

from the fact that a measure is right-continuous.<sup>11</sup> Now, if  $x \leq 0$ , the same argument follows since multiplying -1 will not change the fact that a measure is continuous.

#### Definition 1.18 (Half intervals). We call

$$\varnothing$$
,  $(a, b]$ ,  $(a, \infty)$ ,  $(-\infty, b]$ ,  $(-\infty, \infty)$ 

half-intervals.

**Lemma 1.7.** Let  $\mathcal{H}$  be the collection of finite disjoint unions of half-intervals. Then,  $\mathcal{H}$  is an algebra on  $\mathbb{R}$ .

Proof. We see that

- $\emptyset \in \mathcal{H}$ . Clearly.
- $\bullet$  To show  $\mathcal{H}$  is closed under complements, we have

$$- \varnothing^c = \mathbb{R} = (-\infty, \infty) \in \mathcal{H}.$$

$$-(a,b]^c = (-\infty, a] \cup (a, \infty) \in \mathcal{H}^{12}$$

$$- (a, \infty)^c = (-\infty, a] \in \mathcal{H}.$$

$$-(-\infty,b]^c = (b,\infty) \in \mathcal{H}.$$

$$- (-\infty, \infty)^c = \emptyset \in \mathcal{H}.$$

•  $\mathcal{H}$  is closed under finite unions, clearly.

<sup>&</sup>lt;sup>11</sup>Actually, a measure is always continuous.

 $<sup>^{12}\</sup>mathrm{Since}$  it's a two disjoint union of half intervals.

Proposition 1.3 (Distribution function defines a pre-measure). Let  $F: \mathbb{R} \to \mathbb{R}$  be a distribution function. For a half interval I, define

$$\ell(I) := \ell_F(I) = \begin{cases} 0, & \text{if } I = \emptyset; \\ F(b) - F(a), & \text{if } I = (a, b]; \\ F(\infty) - F(a), & \text{if } I = (a, \infty]; \\ F(b) - F(-\infty), & \text{if } I = (-\infty, b]; \\ F(\infty) - F(-\infty), & \text{if } I = (-\infty, \infty). \end{cases}$$

Define  $\mu_0 := \mu_{0,F}$  as

$$\mu_{0,F} \colon \mathcal{H} \to [0,\infty]$$

by

$$\mu_0(A) = \sum_{k=1}^{N} \ell(I_k) \text{ if } A = \bigcup_{k=1}^{N} I_k,$$

where A is a finite disjoint union of half intervals  $I_1, \ldots, I_N$ . Then,  $\mu_0$  is a pre-measure on  $\mathcal{H}$ .

*Proof.* We see that

- 1.  $\mu_0$  is well-defined.
- 2.  $\mu_0(\emptyset) = 0$ .
- 3.  $\mu_0$  is finite additive.
- 4.  $\mu_0$  is countable additive within  $\mathcal{H}$ .

Suppose  $A \in \mathcal{H}$  where  $A = \bigcup_{i=1}^{\infty} A_i$  is a countable disjoint union. It is enough to consider the case that A = I,  $A_k = I_k$  are all half-intervals.<sup>13</sup> Focus on the case I = (a, b]. Let

$$(a,b] = \bigcup_{n=1}^{\infty} (a_n, b_n],$$

which is a disjoint union. Then we only need to check

$$F(b) - F(a) = \sum_{n=1}^{\infty} (F(b_n) - F(a_n)).$$

• Since  $(a, b] \supset \bigcup_{n=1}^{N} (a_n, b_n]$  for any fixed  $N \in \mathbb{N}$ , hence

$$\bigvee_{N \in \mathbb{N}} F(b) - F(a) \ge \sum_{n=1}^{N} \left( F(b_n) - F(a_n) \right).$$

<sup>13</sup>Since  $\mathcal{H}$  is only a collection of *finite* disjoint half intervals, hence after considering A=I, we can apply the same argument iteratively and stop in finite steps. Formally, we can consider  $H \in \mathcal{H}$ ,  $H = \bigcup_{i=1}^{\infty} A^i$ , where  $A^i$  being a half interval. Then by the above argument, we have  $A^i = I^i$  and so on.

By letting  $N \to \infty$ , we have

$$F(b) - F(a) \ge \sum_{n=1}^{\infty} (F(b_n) - F(a_n)).$$

• Fix  $\epsilon > 0$ . Since F is right-continuous,  $\exists a' > a$  such that

$$F(a') - F(a) < \epsilon$$
.

For each  $n \in \mathbb{N}$ ,  $\exists b'_n > b_n$  such that

$$F(b_n') - F(b_n) < \frac{\epsilon}{2^n}.$$

Then, we have

$$[a',b] \subset \bigcup_{n=1}^{\infty} (a_n,b'_n),$$

hence

$$\underset{N\in\mathbb{N}}{\exists} [a',b] \subset \bigcup_{n=1}^{N} (a_n,b'_n),^{14}$$

which is only finitely many unions now. In this case, we have

$$F(b) - F(a') \le \sum_{n=1}^{N} F(b'_n) - F(a_n).$$

Finally, we see that

$$F(b) - F(a) \le F(b) - F(a') + \epsilon$$

$$\le \sum_{n=1}^{\infty} (F(b'_n) - F(a_n)) + \epsilon$$

$$\le \sum_{n=1}^{\infty} \left( F(b_n) - F(a_n) + \frac{\epsilon}{2^n} \right) + \epsilon$$

$$= \sum_{n=1}^{\infty} (F(b_n) - F(a_n)) + 2\epsilon$$

for any fixed  $\epsilon > 0$ , hence

$$F(b) - F(a) \le \sum_{n=1}^{\infty} (F(b_n) - F(a_n)).$$

Combine these two inequalities, we have

$$F(b) - F(a) = \sum_{n=1}^{\infty} (F(b_n) - F(a_n))$$

as we desired.

<sup>&</sup>lt;sup>14</sup>This essentially follows from the fact that open sets are closed under countable unions, hence the equality will not hold, even after taking the limit.

**Remark.** It's again the  $\frac{\epsilon}{2^n}$  trick we saw before!

## Lecture 8: Lebesgue-Stieltjes Measure on $\mathbb{R}$

24 Jan. 11:00

To classify all measures, we now see this last theorem to complete the task.

#### Theorem 1.5 (Locally finite Borel measures on $\mathbb{R}$ ). We have

1.  $F: \mathbb{R} \to \mathbb{R}$  a distribution function, then there exists a **unique** locally finite Borel measure  $\mu_F$  on  $\mathbb{R}$  satisfying

$$\mu_F((a,b]) = F(b) - F(a)$$

for every a < b.

2. Suppose  $F, G: \mathbb{R} \to \mathbb{R}$  are distribution functions. Then,

$$\mu_F = \mu_G$$

on  $\mathcal{B}(\mathbb{R})$  if and only if F - G is a constant function.

Proof. ■ HW.

**Remark.** Theorem 1.5 simply states that given a distribution function, if we restrict our attention on locally finite measures on  $\mathbb{R}$  following our usual convention, then it defines the measure on  $\mathcal{B}(\mathbb{R})$  uniquely up to a *constant shift*.

## 1.6 Lebesgue-Stieltjes Measure on $\mathbb{R}$

We see that

F distribution function  $\stackrel{!}{\Longrightarrow}$   $\mu_F$  on Carathéodory  $\sigma$ -algebra  $\mathcal{A}_{\mu_F} \supset \mathcal{B}(\mathbb{R})$ .

Furthermore, we have

$$(\mathcal{A}_{\mu_F}, \mu_F) = \overline{(\mathcal{B}(\mathbb{R}), \mu_F)}.$$

Definition 1.19 (Lebesgue-Stieltjes measure). Given a distribution function F, we say  $\mu_F$  on  $\mathcal{A}_{\mu_F}$  is called the *Lebesgue-Stieltjes measure* corresponding to F.

Definition 1.20 (Lebesgue measure, Lebesgue  $\sigma$ -algebra). From Definition 1.19, if F(x) = x, then the induced  $(\mathcal{A}_{\mu_F}, \mu_F)$  is denoted as  $(\mathcal{L}, m)$ , where  $\mathcal{L}$  is called *Lebesgue*  $\sigma$ -algebra, and m is called *Lebesgue* measure.

**Remark.** Recall that  $\mathcal{L}$  is induced by Theorem 1.2, namely given m, for all  $A \subset \mathbb{R}$ , we have

$$\mathcal{L} := \left\{ A \subset \mathbb{R} \mid \bigvee_{E \subset \mathbb{R}} m(A) = m(A \cap E) + m(A \setminus E) \right\}$$

**Note.** We see that since F is right-continuous and increasing, hence

$$F(x^{-}) \le F(x) = F(x^{+}).^{15}$$

**Example.** We first see some examples.

- 1.  $\mu_F((a,b]) = F(b) F(a)$ . Then
  - $\mu_F(\{a\}) = F(a) F(a^-)$
  - $\mu_F([a,b]) = F(b) F(a^-)$
  - $\mu_F((a,b)) = F(b^-) F(a)$
- 2. We define

$$F(x) = \begin{cases} 1, & \text{if } x \ge 0; \\ 0, & \text{if } x < 0. \end{cases}$$

Then

- $\mu_F(\{0\}) = 1$
- $\mu_F(\mathbb{R}) = 1$
- $\mu_F(\mathbb{R}\setminus\{0\})=0$ . This is easy to see since  $\mathbb{R}\setminus\{0\}=(-\infty,0)\cup(0,\infty)$ , hence

$$\mu_F(\mathbb{R} \setminus \{0\}) = \mu_F((-\infty, 0) \cup (0, \infty))$$

$$= \underbrace{\mu_F((-\infty, 0))}_{0-0} + \underbrace{\mu_F((0, \infty))}_{1-1} = 0.$$

We call that  $\mu_F$  is the *Dirac measure* at 0.

3. Denote  $\mathbb{Q} = \{r_1, r_2, \ldots\}$ , and we define

$$F(x) = \sum_{n=1}^{\infty} \frac{F_n(x)}{2^n} \text{ where } F_n(x) = \begin{cases} 1, & \text{if } x \ge r_n; \\ 0, & \text{if } x < r. \end{cases}$$

Then

- $\mu_F(\lbrace r_i \rbrace) > 0$  for all  $r_i \in \mathbb{Q}$ .
- $\mu_F(\mathbb{R} \setminus \mathbb{Q}) = 0$ .

<sup>&</sup>lt;sup>15</sup>Some text will use x- and x+ instead of  $x^-$  and  $x^+$ , respectively. <sup>16</sup>It follows from  $F(0^-) - F(-\infty) = 0 - 0 = 0$ .

<sup>&</sup>lt;sup>17</sup>It follows from  $F(\infty) - F(0) = 1 - 1 = 0$ .

This is shown in Lemma A.2.

- 4. If F is continuous at a, then  $\mu_F(\{a\}) = 0$ .
- 5. F(x) = x, then recall that we denote  $\mu_F := m$ , and we have

• 
$$m((a,b]) = m((a,b)) = m([a,b]) = b - a$$
.

- 6.  $F(x) = e^x$ 
  - $\mu_F((a,b]) = \mu_F((a,b)) = e^b e^a$ .

Remark. We see that the first two examples are discrete measures.

**Example (Middle thirds Cantor set).** Let  $C := \bigcap_{n=1}^{\infty} K_n$ , where we have

$$K_0 := [0, 1]$$

$$K_1 := K_0 \setminus \left(\frac{1}{3}, \frac{2}{3}\right)$$

$$K_2 := K_1 \setminus \left(\frac{1}{9}, \frac{2}{9}\right) \cup \left(\frac{7}{9}, \frac{8}{9}\right)$$

$$\vdots$$

$$K_n := K_{n-1} \setminus \bigcup_{k=1}^{3^n - 1} \left(\frac{3k+1}{3^{n+1}}, \frac{3^{k+2}}{3^{n+1}}\right).$$

We see that C is uncountable and with m(C) = 0. And observe that  $x \in C$  if and only if  $x = \sum_{n=1}^{\infty} \frac{a_n}{3}$  for some  $a_n \in \{0, 2\}$ . Hence, we can instead formulate  $K_n$  by

$$K_n = \bigcup_{\substack{a_i \in \{0,2\}\\1 \le i \le n}} \left[ \sum_{i=1}^{\infty} \frac{a_i}{3^i}, \sum_{i=1}^{\infty} \frac{a_i}{3^i} + \frac{1}{3^n} \right].$$

Figure 1: The top line corresponds to  $K_0$ , and then  $K_1$ , etc.

The proof of m(C) = 0 is given in Lemma A.3.

#### 1.6.1 Cantor Function

Consider F as follows. We define a function F to be 0 to the left of 0, and 1 to the right of 1. Then, define F to be  $\frac{1}{2}$  on  $\left(\frac{1}{3},\frac{2}{3}\right)$ ,  $\frac{1}{4}$  on  $\left(\frac{1}{9},\frac{2}{9}\right)$ ,  $\frac{3}{4}$  on  $\left(\frac{7}{9},\frac{8}{9}\right)$  and so on. This is so-called *Cantor Function*. We can show F is continuous and increasing, which makes F a distribution function. Also, we see that the measure this F induced is called *Cantor measure*.



Figure 2: Cantor Function (Devil's Staircase).

We see that F is *continuous* and increasing. Furthermore,

| Cantor Measure $\mu_F$  |        | Lebesgue Measure $m$   |
|---|--------|--|
| $\mu_F(\mathbb{R} \setminus C) = 0$<br>$\mu_F(C) = 1$<br>$\mu_F(\{a\}) = 0$ | $\iff$ | $m(\mathbb{R} \setminus C) = \infty > 0$<br>m(C) = 0<br>$m(\{a\}) = 0$ |

**Remark.**  $\mu_F$  and m are said to be singular to each other.

## 1.7 Regularity Properties of Lebesgue-Stieltjes Measures

We first see a lemma.

**Lemma 1.8.** Let  $\mu$  be Lebesgue-Stieltjes measure on  $\mathbb{R}$ . Then we have

$$\mu(A) \stackrel{!}{=} \inf \left\{ \sum_{i=1}^{\infty} \mu((a_i, b_i)) \mid \bigcup_{i=1}^{\infty} (a_i, b_i) \supset A \right\}$$
$$= \inf \left\{ \sum_{i=1}^{\infty} \mu((a_i, b_i)) \mid \bigcup_{i=1}^{\infty} (a_i, b_i) \supset A \right\}$$

for every  $A \in \mathcal{A}_{\mu}$ 

*Proof.* The second equality follows from the continuity of the measure.

Remark. This is similar to

$$(a,b) = \bigcup_{n=1}^{\infty} (a,b-1/n], \quad (a,b] = \bigcap_{n=1}^{\infty} (a,b+1/n].$$

## Lecture 9: Properties of Lebesgue-Stieltjes measure

26 Jan. 11:00

As previously seen. Let  $X \subset [0, \infty]$ . Recall that

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• Finite supremum.

$$\alpha = \sup X < \infty \iff \begin{cases} \forall & \alpha \geq x \\ \forall & \exists \\ \epsilon > 0 & x \in X \end{cases}$$

• Infinite supremum.

$$\alpha = \sup X = \infty \iff \bigvee_{L>0} \underset{x \in X}{\exists} x \ge L.$$

This should be useful latter on.

Theorem 1.6 (Regularity). Let  $\mu$  be Lebesgue-Stieltjes measure. Then, for all  $A \in \mathcal{A}_{\mu}$ ,

- 1. (outer regularity)  $\mu(A) = \inf \{ \mu(O) \mid O \supset A, O \text{ is open} \}$
- 2. (inner regularity)  $\mu(A) = \sup{\{\mu(K) \mid K \subset A, K \text{ is compact}\}}$

*Proof.* We check them separately.

•

DIY

- 2. Let  $s := \sup\{\mu(K) \mid K \subset A, K \text{ is compact}\}$ , then by monotonicity, we have  $\mu(A) \geq s$ . To show the other direction, we consider
  - A is a bounded set.

Then  $\overline{A} \in \mathcal{B}(\mathbb{R}) \subset \mathcal{A}_{\mu}$ ,  $\overline{A}$  is also bounded  $\Longrightarrow \mu(\overline{A}) < \infty$ . Fix  $\epsilon > 0$ , then by outer regularity, there exists an open  $O \supset \overline{A} \setminus A$ , and  $\mu(O) - \mu(\overline{A} \setminus A) = \mu(O \setminus (\overline{A} \setminus A)) \le \epsilon$ . Let  $K := \underbrace{A \setminus O}_{K \subset A} = \underbrace{\overline{A} \setminus O}_{\text{compact}}$ , we

show that

$$\mu(K) \ge \mu(A) - \epsilon$$
.

DIY

• A is an unbounded set with  $\mu(A) < \infty$ .

Let 
$$A = \bigcup_{n=1}^{\infty} A_n$$
,  $A_n = A \cap [-n, n]$  where  $A_1 \subset A_2 \subset ...$ , then

$$\lim_{n\to\infty}\mu(A_n)=\mu(A)<\infty.$$

• A is an unbounded set with  $\mu(A) = \infty$ .

We can show that

$$\lim_{n \to \infty} \mu(A_n) = \mu(A) = \infty.$$

Fix L > 0, then  $\exists N$  such that  $\mu(A_N) \geq L$ .

**Definition 1.21** ( $G_{\delta}$ -set,  $F_{\sigma}$ -set). Let X be a topological space. Then

- A  $G_{\delta}$ -set is  $G = \bigcap_{i=1}^{\infty} O_i$ ,  $O_i$  open.
- A  $F_{\sigma}$ -set is  $F = \bigcup_{i=1}^{\infty} F_i$ ,  $F_i$  closed.

**Theorem 1.7.** Let  $\mu$  be a Lebesgue-Stieltjes measure. Then  $TFAE^a$ :

- 1.  $A \in \mathcal{A}_{\mu}$
- 2.  $A = G \setminus M$ , G is a  $G_{\delta}$ -set, M is a  $\mu$ -null set.
- 3.  $A = F \setminus N$ , F is a  $F_{\sigma}$ -set, N is a  $\mu$ -null set.

<sup>a</sup> TFAE: The following are equivalent.

*Proof.* We see that  $(2.) \implies (1.)$  and  $(3.) \implies (1.)$  are clear.

- $\bullet$  (1.)  $\Longrightarrow$  (3.)
  - Assume  $\mu(A) < \infty$ . From the inner regularity, we have

 $\forall n \in \mathbb{N} \exists \text{compact } K_n \subset A \text{ such that } \mu(K_n) + \frac{1}{n} \geq \mu(A).$ 

Let  $F = \bigcup_{n=1}^{\infty} K_n$ , then  $N = A \setminus F$  is  $\mu$ -null.

Check!

– Assume  $\mu(A) = \infty$ . Let  $A = \bigcup_{k \in \mathbb{Z}} A_k$ ,  $A_k = A \cap (k, k+1]$ . From what we have just shown above,

$$\forall k \in \mathbb{Z} \ A_k = F_k \cup N_k, \ A = \underbrace{\left(\bigcup_k F_k\right)}_{F_{\sigma}\text{-set}} \cup \underbrace{\left(\bigcup_k N_k\right)}_{\mu\text{-null}}.$$

•  $(1.) \implies (2.)$ 

We see that

$$A^c = F \cup N, \quad A = F^c \cap N^c = F^c \setminus N.$$

**Proposition 1.4.** Let  $\mu$  be a Lebesgue-Stieltjes measure, and  $A \in \mathcal{A}_{\mu}$ ,  $\mu(A) < \infty$ . Then we have

$$\forall \epsilon > 0 \ \exists I = \bigcup_{i=1}^{N(\epsilon)} I_i$$

disjoint open intervals such that  $\mu(A \triangle I) \leq \epsilon$ .

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*Proof.* Using outer regularity and the fact that every open set is  $\bigcup_{i=1}^{\infty} I_i$ , where  $I_i$  are disjoint open intervals.

We now see some properties of Lebesgue measure.

**Theorem 1.8.** Let  $A \in \mathcal{L}$ , then we have  $A + s \in \mathcal{L}$ ,  $rA \in \mathcal{L}$  for all  $r, s \in \mathbb{R}$ . i.e.,

$$m(A+s) = m(A), \quad m(rA) = |r| \cdot m(A).$$

Proof.

 $\vdash$  DI

**Example.** We now see some examples.

1. Let  $\mathbb{Q} =: \{r_i\}_{i=1}^{\infty}$  which is dense in  $\mathbb{R}$ . Let  $\epsilon > 0$ , and

$$O = \bigcup_{i=1}^{\infty} \left( r_i - \frac{\epsilon}{2^i}, r_i + \frac{\epsilon}{2^i} \right).$$

We see that O is open and dense<sup>18</sup> in  $\mathbb{R}$ . But we see

$$m(O) \le \sum_{i=1}^{\infty} \frac{2\epsilon}{2^i} = 2\epsilon.$$

Furthermore,  $\partial O = \overline{O} \setminus O$ ,  $m(\partial O) = \infty$ 

- 2. There exists uncountable set A with m(A) = 0.
- 3. There exists A with m(A) > 0 but A contains no non-empty open intervals.
- 4. There exists  $A \notin \mathcal{L}$ . e.g. Vitali set.<sup>19</sup>
- 5. There exists  $A \in \mathcal{L} \setminus \mathcal{B}(\mathbb{R})$ .

## Lecture 10: Integration

26 Jan. 11:00

## 2 Integration

#### 2.1 Measurable Function

We start with a definition.

**Definition 2.1 (Measurable function).** Suppose  $(X, \mathcal{A}), (Y, \mathcal{B})$  are measurable spaces. Then we say  $f: X \to Y$  is  $(\mathcal{A}, \mathcal{B})$ -measurable if

$$\underset{B\in\mathcal{B}}{\forall} f^{-1}(B)\in\mathcal{A}.$$

**Remark.** If A and B are given, we'll sometimes say f is measurable if it'll not cause any confusions.

<sup>18</sup>https://en.wikipedia.org/wiki/Dense\_set

<sup>19</sup>https://en.wikipedia.org/wiki/Vitali\_set

**Lemma 2.1.** Given two measurable spaces  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$ , and suppose  $\mathcal{B} = \langle \mathcal{E} \rangle$  for some  $\mathcal{E} \subset Y$ . Then,

$$f \colon X \to Y \text{ is } (\mathcal{A}, \mathcal{B})\text{-measurable} \iff \bigvee_{E \in \mathcal{E}} f^{-1}(E) \in \mathcal{A}.$$

*Proof.* We see that the *only if* part ( $\Longrightarrow$ ) is clear. On the other direction, we consider the following. Let  $\mathcal{D} = \{E \subset Y \mid f^{-1}(E) \in \mathcal{A}\}$ , then

- $\mathcal{E} \subset \mathcal{D}$  by assumption
- $\mathcal{D}$  is a  $\sigma$ -algebra

Check!

hence, we see that  $\langle \mathcal{E} \rangle = \mathcal{B} \subset \mathcal{D}$  from Lemma 1.2. The result then follows from the definition of  $(\mathcal{A}, \mathcal{B})$ -measurable.

Note. Recall that

- $f^{-1}(E^c) = f^{-1}(E)^c$
- $\bullet \ f^{-1}\left(\bigcup_{\alpha}E_{\alpha}\right)=\bigcup_{\alpha}f^{-1}(E_{\alpha})$

**Definition 2.2** ( $\mathcal{A}$ -measurable). Let  $(X, \mathcal{A})$  be a measurable space. Then,

$$\begin{array}{l} f\colon X\to\mathbb{R}\\ f\colon X\to\overline{\mathbb{R}}\\ f\colon X\to\mathbb{C} \end{array} \text{ is } \mathcal{A}\text{-}\textit{measurable} \text{ if } \begin{cases} f\text{ is } (\mathcal{A},\mathcal{B}(\mathbb{R}))\text{-}\text{measurable}\\ f\text{ is } (\mathcal{A},\mathcal{B}(\overline{\mathbb{R}}))\text{-}\text{measurable}\\ \Re f,\Im f\colon X\to\mathbb{R} \text{ are } \mathcal{A}\text{-}\text{measurable}. \end{cases}$$

Notation. Notice that

- $\overline{\mathbb{R}} = [-\infty, \infty]$
- $\mathcal{B}(\overline{\mathbb{R}}) = \{ E \subset \overline{\mathbb{R}} \mid E \cap \mathbb{R} \in \mathcal{B}(\mathbb{R}) \}.$
- $\Re f$  is the real part of f, while  $\Im f$  is the imaginary part of f.

Example. We see that

- $A = P(X) \implies$  Every function is A-measurable.
- $\mathcal{A} = \{\emptyset, X\} \implies$  The only  $\mathcal{A}$ -measurable functions are constant functions.

**Definition 2.3 (Lebesgue measurable).** A Lebesgue measurable function f is a measurable function

$$f: (\mathbb{R}, \mathcal{L}) \to (\mathbb{C}, \mathcal{B}(\mathbb{C})).$$

**Lemma 2.2.** Given  $f: X \to \mathbb{R}$ , TFAE.

- 1. f is A-measurable
- 2.  $\forall a \in \mathbb{R}, f^{-1}((a, \infty)) \in \mathcal{A}$
- 3.  $\forall a \in \mathbb{R}, f^{-1}([a, \infty)) \in \mathcal{A}$
- 4.  $\forall a \in \mathbb{R}, f^{-1}((-\infty, a)) \in \mathcal{A}$
- 5.  $\forall a \in \mathbb{R}, f^{-1}((-\infty, a]) \in \mathcal{A}$

*Proof.* The result follows from Lemma 2.1 we just saw.

Remark (Operations preserve A-measurability). Given  $f, g: X \to \mathbb{R}$  and is A-measurable, then

1.  $\phi: \mathbb{R} \to \mathbb{R}$ ,  $\mathcal{A}$ -measurable<sup>20</sup>, then

$$\phi \circ f \colon X \to \mathbb{R}$$

is A-measurable.

- 2. -f, 3f,  $f^2$ , |f| are all  $\mathcal{A}$ -measurable, and  $\frac{1}{f}$  is  $\mathcal{A}$ -measurable if  $f(x) \neq 0, \forall x \in X$ .
- 3. f + g is  $\mathcal{A}$ -measurable. We see this from

$$(f+g)^{-1}((a,\infty)) = \bigcup_{r \in \mathbb{Q}} (f^{-1}((r,\infty)) \cap g^{-1}((a-r,\infty)))$$

with Lemma 2.2.

4.  $f \cdot g$  is  $\mathcal{A}$ -measurable. We see this from

$$f(x)g(x) = \frac{1}{2} \left( (f(x) + g(x))^2 - f(x)^2 - g(x)^2 \right).$$

5. We see that

$$(f \vee g)(x) := \max\{f(x), g(x)\}\$$
and  $(f \wedge g)(x) := \min\{f(x), g(x)\}\$ 

are A-measurable.

6. Let  $f_n : X \to \overline{\mathbb{R}}$  be A-measurable. Then

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

are A-measurable.

*Proof.* Consider  $\sup_{n\in\mathbb{N}} f_n =: g$ , then

$$g^{-1}((a,\infty]) = \bigcup_{n \in \mathbb{N}} f_n^{-1}((a,\infty])$$

for  $\sup_n f_n(x) = g(x) > a$ . A similar argument can prove the case of checking  $f_n$ .

And notice that  $\limsup_{n\to\infty} f_n = \inf_{k\in\mathbb{N}} \sup_{n\geq k} f_n$ , then the similar argument also proves this case.

- 7. If  $\lim_{n\to\infty} f_n(x)$  converges for every  $x\in X$ , then f is  $\mathcal{A}$ -measurable.
- 8. If  $f: \mathbb{R} \to \mathbb{R}$  is continuous
  - $\implies f$  is Borel measurable
  - $\implies f$  is Lebesgue measurable

since the preimage of an open set of a continuous function is open, then we consider  $f^{-1}((a,\infty))$ .

**Definition 2.4 (Support).** The *support* of function  $f: X \to \overline{\mathbb{R}}$  is

$$\operatorname{supp} f := \{ x \in X \mid f(x) \neq 0 \}.$$

**Definition 2.5 (Positive and Negative part).** For  $f: X \to \overline{\mathbb{R}}$ , let  $f^+ := f \vee 0$  and  $f^- := (-f) \vee 0$ , where we call  $f^+$  the positive part of f while  $f^-$  the negative part of f.

ai.e., 
$$f^+(x) = \max\{f(x), 0\}, \quad f^-(x) = \max\{-f(x), 0\}$$

**Remark.** If  $\operatorname{supp} f^+ \cap \operatorname{supp} f^- = \emptyset$  and  $f(x) = f^+(x) - f^-(x)$ , then f is  $\mathcal{A}$ -measurable  $\iff f^+, f^-$  are  $\mathcal{A}$ -measurable.

**Definition 2.6 (Characteristic (Indicator) function).** For  $E \subset X$ , the *characteristic (indicator) function* of E is

$$\mathcal{X}_E(x) = \mathbb{1}_E(x) = \begin{cases} 1, & \text{if } x \in E; \\ 0, & \text{if } x \in E^c. \end{cases}$$

**Remark.** We see that  $\mathbb{1}_E$  is A-measurable  $\iff E \in A$ .

**Definition 2.7 (Simple function).** Let  $(X, \mathcal{A})$  be a measurable space. Then a *simple function*  $\phi \colon X \to \mathbb{C}$  that is  $\mathcal{A}$ -measurable and takes only finitely many values.

Remark. We see that if

$$\phi(X) = \{c_1, \dots, c_N\},\$$

<sup>&</sup>lt;sup>20</sup>In this case, we also call it *Borel measurable*.

then

$$E_i = \phi^{-1}(\{c_i\}) \in \mathcal{A} \implies \phi = \sum_{i=1}^N \underbrace{c_i}_{\neq \pm \infty} \mathbb{1}_{\underbrace{E_i}}.$$

#### Lecture 11: Integration of nonnegative functions

31 Jan. 11:00

As previously seen. For a simple function  $\phi$ ,  $c_i$  can actually be in  $\mathbb{C}$ .

**Theorem 2.1.** Given a measurable space  $(X, \mathcal{A})$  and let  $f: X \to [0, \infty]$ , the followings are equivalent.

- 1. f is a  $\mathcal{A}$ -measurable function.
- 2. There exists simple functions  $0 \le \phi_1(x) \le \phi_2(x) \le \ldots \le f(x)$  such that

$$\forall \lim_{x \in X} \lim_{n \to \infty} \phi_n(x) = f(x)$$

i.e., f is a pointwise upward limit of simple functions.

*Proof.* We'll prove both directions.

- It's clear that (2.)  $\implies$  (1.) from the fact that  $f(x) = \sup_n \phi_n(x)$  and the remark.
- We want to show that  $(1.) \implies (2.)$ . Assume f is  $\mathcal{A}$ -measurable, and fix  $n \in \mathbb{N}$ .

Let 
$$F_n = f^{-1}([2^n, \infty]) \in \mathcal{A}$$
. Also, for  $0 \le k \le 2^{2n} - 1$ ,  $E_{n,k} = f^{-1}(\left[\frac{k}{2^n}, \frac{k+1}{2^n}\right]) \in \mathcal{A}$ .

Then, define  $\phi_n$  be

$$\phi_n = \sum_{k=0}^{2^{2n}-1} \frac{k}{2^n} \mathbb{1}_{E_{n,k}} + 2^n \mathbb{1}_{F_n},$$

we have

$$-0 \le \phi_1(x) \le \phi_2(x) \le \ldots \le f(x)$$
 for every  $x \in X$ 

$$- \forall x \in X \setminus F_n$$
, we have  $0 \le f(x) - \phi_n(x) \le \frac{1}{2^n}$ 

Furthermore, we see that

$$F_1 \supset F_2 \supset \dots, \quad \bigcap_{n=1}^{\infty} F_n = f^{-1}(\{\infty\}),$$

then

$$-x \in f^{-1}([0,\infty]) = X \setminus \bigcap_{n=1}^{\infty} F_n \implies \lim_{n \to \infty} \phi_n(x) = f(x)$$
$$-x \in f^{-1}(\{\infty\}) = \bigcap_{n=1}^{\infty} F_n \implies f_n(x) \ge 2^n \implies \lim_{n \to \infty} \phi_n(x) = \infty = f(x)$$

Corollary 2.1. If f is bounded on a set  $A \subset \mathbb{R}$ , i.e.,  $\exists L > 0$  such that

$$\forall_{x \in A} |f(x)| \le L,$$

then there exists a sequence of simple functions  $\{\phi_n\}$  such that  $\phi_n \to f$  uniformly on A.

Proof.

DIY

Corollary 2.2. If  $f: X \to \mathbb{C}$  is a measurable function if and only if there exists simple functions  $\phi_n: X \to \mathbb{C}$  such that

$$0 \le |\phi_1(x)| \le |\phi_2(x)| \le \ldots \le |f(x)|$$

with

$$\forall_{x \in X} \lim_{n \to \infty} \phi_n(x) = f(x).$$

Proof.

DIY

# 2.2 Integration of Nonnegative Functions

We start with our first definition about integral.

**Definition 2.8 (Integration of nonnegative function).** Let  $(X, \mathcal{A}, \mu)$  be a measure space, and  $\phi \colon X \to [0, \infty]$  such that

$$\phi = \sum_{i=1}^{N} c_i \mathbb{1}_{E_i}$$

be a simple function. Define

$$\int \phi = \int \phi \, \mathrm{d}\mu = \int_X \phi \, \mathrm{d}\mu = \sum_{i=1}^N c_i \mu(E_i).$$

Furthermore, for  $A \in \mathcal{A}$ ,

$$\int_A \phi = \int_A \phi \, \mathrm{d}\mu = \int \phi \mathbb{1}_A \, \mathrm{d}\mu.$$

Note. Note that

• In the expression  $\sum_{i=1}^{N} c_i \mu(E_i)$ , we're using the convention  $0 \cdot \infty = 0$ .

• The function  $\phi \mathbb{1}_A$  is also a simple function since both  $\phi$  and  $\mathbb{1}_A$  are simple function.

**Proposition 2.1.** Suppose we have  $\phi, \psi \geq 0$  be two simple functions. Then,

- Definition 2.8 is well-defined.
- $\int c\phi = c \int \phi \text{ for } c \in [0, \infty).$
- $\int \phi + \psi = \int \phi + \int \psi$ .
- $\phi(x) \ge \psi(x)$  for all  $x \implies \int \phi \ge \int \psi$ .
- $\nu(A) = \int_A \phi \, d\mu$  is a measure on (X, A).

Proof.

-(DI)

Definition 2.9 (Generatlization of Integration of nonnegative function). Given  $(X, \mathcal{A}, \mu)$  with  $f: X \to [0, \infty]$  be  $\mathcal{A}$ -measurable. Define

$$\int f = \int f \, \mathrm{d}\mu = \sup \left\{ \int \phi \colon 0 \le \phi \le f \text{ such that } \phi \text{ is simple} \right\}.$$

Note. Note that

- If f is a simple function, the Definition 2.8 and Definition 2.9 of  $\int f$  are the same.
- $\int cf = c \int f$  for  $c \in [0, \infty)$ .
- If  $f \ge g \ge 0 \implies \int f \ge \int g$ .
- But  $\int f + g = \int f + \int g$  is not trivial.

Theorem 2.2 (Monotone Convergence Theorem (MCT)). Given  $(X, \mathcal{A}, \mu)$  be a measure space. Then if

- $f_n: X \to [0, \infty]$  be  $\mathcal{A}$ -measurable for every  $n \in \mathbb{N}$ ;
- $0 \le f_1(x) \le f_2(x) \le \dots$  for every  $x \in X$ ;
- $\lim_{n\to\infty} f_n(x) = f(x)$  for every  $x \in X$ ,

we have

$$\lim_{n \to \infty} \int f_n = \int f.$$

*Proof.* Note that if  $\lim_{n\to\infty}\int f_n$  exists, then it's equal to  $\sup_n\int f_n$ .

Then

• 
$$f_n \le f \implies \int f_n \le \int f \implies \lim_{n \to \infty} \int f_n \le \int f$$
.

• Fix a simple function  $0 \le \phi \le f$ , then it's enough to show  $\lim_{n \to \infty} \int f_n \ge \int \phi$ .

We first fix  $\alpha = (0,1)$ , then it's also enough to show

$$\lim_{n \to \infty} \int f_n \ge \alpha \int \phi.$$

Let  $A_n := \{x \in X \mid f_n(x) \ge \alpha \phi(x)\}$ , then since  $f_n$  is measurable,

$$-A_n \in \mathcal{A}$$

$$-A_1 \subset A_2 \subset A_3 \subset \dots$$

$$-\bigcup_{n=1}^{\infty} A_n = X \underline{\hspace{1cm}}$$

Check!

We then have

$$\int f_n \ge \int f_n \mathbb{1}_{A_n} \ge \int \alpha \phi \mathbb{1}_{A_n} = \alpha \int_{A_n} \phi = \alpha \nu(A_n)$$

where  $\nu(A) = \int_A \phi$  is a measure. This implies

$$\lim_{n \to \infty} \int f_n \ge \alpha \lim_{n \to \infty} \nu(A_n) \stackrel{21}{=} \alpha \nu(X) = \alpha \int \phi.$$

Corollary 2.3 (Linearity of nonnegative integral). Let  $f, g \ge 0$  be measurable, then

$$\int f + g = \int f + \int g.$$

*Proof.* There exists simple functions  $\phi_n$  and  $\psi_n$  such that

- $0 \le \phi_1 \le \phi_2 \le \dots$  and  $\phi_n \to f$  pointwise
- $0 \le \psi_1 \le \psi_2 \le \dots$  and  $\psi_n \to g$  pointwise

Then,

$$\int (f+g) \stackrel{!}{=} \lim_{n \to \infty} \int (\phi_n + \psi_n) = \lim_{n \to \infty} \int \phi_n + \int \psi_n \stackrel{!}{=} \int f + \int g.$$

## Lecture 12: Fatou's Lemma

2 Feb. 11:00

We start with a useful corollary.

 $<sup>^{21}</sup>$ This follows from the continuity of measure from below

Corollary 2.4 (Tonelli's theorem for nonnegative series and integrals). Given  $g_n \geq 0$  for every  $n \in \mathbb{N}$  and let  $g_n$  be measurable, then

$$\int \sum_{n=1}^{\infty} g_n = \sum_{n=1}^{\infty} \int g_n.$$

**Remark.** Recall that we have seen two series case before. We'll later see two integrals cases.

*Proof.* Let  $f_N := \sum_{n=1}^N g_n$  such that  $\lim_{N \to \infty} f_N \sum_{n=1}^\infty g_n =: f$ , then since  $g_n \ge 0$ , we have  $0 \le f_1 \le f_2 \le \dots$  with

$$\lim_{N \to \infty} f_N(x) = \sum_{n=1}^{\infty} g_n(x).$$

By Theorem 2.2, we have

$$\lim_{N \to \infty} \int \underbrace{\sum_{n=1}^{N} g_n}_{f_N} = \int \underbrace{\sum_{n=1}^{\infty} g_n}_{f}.$$

Now, since the terms in the limit on the left-hand side is just a finite sum, by Corollary 2.3, we have

$$\underbrace{\lim_{N \to \infty} \sum_{n=1}^{N} \int g_n = \lim_{N \to \infty} \int \sum_{n=1}^{N} g_n = \int \sum_{n=1}^{\infty} g_n,}_{n=1}$$

hence

$$\int \sum_{n=1}^{\infty} g_n = \sum_{n=1}^{\infty} \int g_n.$$

Theorem 2.3 (Fatou's Lemma). Suppose  $f_n \ge 0$  and measurable, then

$$\int \liminf_{n \to \infty} f_n \le \liminf_{n \to \infty} \int f_n.$$

Remark. Recall that

$$\liminf_{n \to \infty} f_n := \lim_{k \to \infty} \inf_{n \ge k} f_n = \sup_{k \in \mathbb{N}} \inf_{n \ge k} f_n$$

and

$$\exists \lim_{n \to \infty} a_n \iff \limsup_{n \to \infty} a_n = \liminf_{n \to \infty} a_n.$$

2 INTEGRATION

*Proof.* Let  $g_k = \inf_{n \geq k} f_n$ , then  $g_k$  is measurable and  $0 \leq g_1 \leq g_2 \leq \ldots$  Now, from Theorem 2.2, we have

$$\int \lim_{k \to \infty} g_k = \lim_{k \to \infty} \int g_k.$$

Notice that the left-hand side is just  $\int \liminf_{n \to \infty} f_n$ , while the right-hand side is just  $\lim_{k \to \infty} \int \inf_{n \ge k} f_n$ , i.e.,

$$\int \liminf_{n \to \infty} f_n = \lim_{k \to \infty} \int \inf_{n \ge k} f_n.$$

We see that we want to take the inf outside the integral on the right-hand side. Observe that

$$\label{eq:definition} \bigvee_{m \geq k} \inf_{n \geq k} f_n \leq f_m \implies \bigvee_{m \geq k} \int \inf_{n \geq k} f \leq \int f_m \implies \int \inf_{n \geq k} f_n \leq \inf_{m \geq k} \int f_m.$$

Then, we have

$$\int \liminf_{n \to \infty} f_n = \lim_{k \to \infty} \int \inf_{n > k} f_n \le \lim_{k \to \infty} \inf_{m > k} \int f_m = \liminf_{m \to \infty} \int f_m.$$

**Example.** Given  $(\mathbb{R}, \mathcal{L}, m)$ .

- 1. Escape to horizontal infinity. Let  $f_n := \mathbb{1}_{(n,n+1)}$ . We immediately see that
  - $f_n \to 0$  pointwise
  - $\int f_n = 1$  for every n
  - $\int f = 0$

From Theorem 2.3, we have a strict inequality

$$0 = \int \liminf_{n \to \infty} f_n, \liminf_{n \to \infty} \int f_n = 1.$$

- 2. Escape to width infinity. Let  $f_n := \frac{1}{n} \mathbb{1}_{(0,n)}$ .
- 3. Escape to vertical infinity. Let  $f_n := n \mathbb{1}_{(0,\frac{1}{n})}$ .

Lemma 2.3 (Markov's inequiality). Let  $f \ge 0$  be measurable. Then

$$\bigvee_{c \in (0,\infty)} \mu\left(\left\{x \mid f(x) \ge c\right\}\right) \le \frac{1}{c} \int f.$$

*Proof.* Denote  $\{x \mid f(x) \geq c\} =: E$ , then

$$f(x) \ge c \mathbb{1}_E(x) \implies \int f \ge c \int \mathbb{1}_E = c \cdot \mu(E).$$

2 INTEGRATION

**Remark.** Notice that  $E = f^{-1}([c, \infty])$ , hence E is measurable.

**Proposition 2.2.** Let  $f \ge 0$  be measurable. Then,

$$\int f = 0 \iff f = 0 \text{ a.e.}.$$

i.e.,

$$\int f \, \mathrm{d}\mu = 0 \iff \mu(A) = 0$$

where  $A = \{x \mid f(x) > 0\} = f^{-1}((0, \infty]).$ 

*Proof.* Firstly, assume that  $f = \phi$  is a simple function. We may write

$$\phi = \sum_{i=1}^{N} c_i \mathbb{1}_{E_i}$$

where  $E_i$  are disjoint and  $c_i \in (0, \infty)$ . Then,

$$\int \phi = \sum_{i=1}^{N} c_i \mu(E_i) = 0$$

$$\iff \mu(E_1) = \dots = \mu(E_N) = 0$$

$$\iff \mu(A) = 0, \ A = \bigcup_{i=1}^{N} E_i.$$

Now, assume that f is a general function where  $f \geq 0$  is the only constraint.

1. Assume  $\mu(A)=0$  (i.e., f=0 a.e.). Let  $0\leq \phi \leq f,$  where  $\phi$  is simple. Then

$$\bigvee_{x \in A^c} \phi(x) = 0$$

since f(x) = 0,  $\forall x \in A^c$ . This implies that  $\phi = 0$  a.e. since  $\mu(A) = 0$ , so  $\int \phi = 0$ . We then have

$$\int f = 0$$

from Definition 2.9.

- 2. Assume  $\int f = 0$ . Let  $A_n = f^{-1}\left(\left[\frac{1}{n}, \infty\right]\right)$ . Then we see that
  - $A_1 \subset A_2 \subset \dots$

$$\bullet \bigcup_{n=1}^{\infty} A_n = f^{-1} \left( \bigcup_{n=1}^{\infty} \left[ \frac{1}{n}, \infty \right] \right) = f^{-1}((0, \infty)) = A.$$

We then have

$$\mu(A_n) = \mu\left(\left\{x \mid f(x) \ge \frac{1}{n}\right\}\right) \stackrel{!}{\le} n \int f = 0,$$

which further implies

$$\mu(A) = \lim_{n \to \infty} \mu(A_n) = 0$$

from the continuity of measure from below.

Corollary 2.5. If  $f, g \ge 0$  are both measurable and f = g a.e., then

$$\int f = \int g.$$

*Proof.* Let  $A = \{x \mid f(x) \neq g(x)\}^{22}$ . Then by assumption,  $\mu(A) = 0$ , hence

$$f \mathbb{1}_A = 0$$
 a.e.,  $g \mathbb{1}_A = 0$  a.e..

This further implies that

$$\begin{split} \int f &= \int f(\mathbbm{1}_A + \mathbbm{1}_{A^c}) \\ &\stackrel{!}{=} \int f \mathbbm{1}_A + \int f \mathbbm{1}_{A^c} \\ &= \int f \mathbbm{1}_{A^c} = \int g \mathbbm{1}_{A^c} \\ &= \int g \mathbbm{1}_{A^c} + \int g \mathbbm{1}_A = \int g. \end{split}$$

Corollary 2.6. Let  $f_n \geq 0$  be measurable. Then

1.

$$\begin{cases}
0 \le f_1 \le f_2 \le \dots \le f \text{ a.e.} \\
\lim_{n \to \infty} f_n = f \text{ a.e.}
\end{cases} \implies \lim_{n \to \infty} \int f_n = \int f.$$

2.  $\lim_{n \to \infty} f_n = f$  a.e.  $\Longrightarrow \int f \le \liminf_{n \to \infty} \int f_n$ .

Proof.

 $\blacksquare \sqcap DIY$ 

**Remark.** Almost all the theorems we've proved can be replaced by theorems dealing with almost everywhere condition.

## Lecture 13: Integration of Complex Functions

4 Feb. 11:00

# 2.3 Integration of Complex Functions

As usual, we start with a definition.

 $<sup>^{22}</sup>A$  is measurable indeed.

**Definition 2.10 (Integrable).** Let  $(X, \mathcal{A}, \mu)$  be a measure space and let  $f: X \to \overline{\mathbb{R}}$  and  $g: X \to \mathbb{C}$  be measurable.<sup>a</sup>

Then f, g are called *integrable* if  $\int |f| < \infty$  and  $\int |g| < \infty$ , and we define

$$\int f = \int f^{+} - \int f^{-}, \quad \int g = \int \Re g + i \int \Im g.$$

Furthermore, for  $f \colon X \to \overline{\mathbb{R}}$ , we define

$$\int f = \begin{cases} \infty, & \text{if } \int f^+ = \infty, \int f^- < \infty; \\ -\infty, & \text{if } \int f^+ < \infty, \int f^- = \infty. \end{cases}$$

 $^a \text{Recall}$  that for a complex-valued function like g, this means that both  $\Re g$  and  $\Im g$  are measurable.

We now see a lemma.

**Lemma 2.4.** Let  $f, g: X \to \overline{\mathbb{R}}$  or  $\mathbb{C}$  integrable. Assume that f(x) + g(x) is well-defined for all  $x \in X$ .

Then we have

- 1. f + g, cf for all  $c \in \mathbb{C}$  are integrable.
- 2.  $\int f + g = \int f + \int g$ . This is not trivial since  $(f+g)^+ \neq f^+ + g^+$ .
- 3.  $\left| \int f \right| \leq \int |f|$ .

<sup>a</sup>That is, we never see  $\infty + (-\infty)$  or  $(-\infty) + \infty$ .

Proof. Check [FF99] page 53.

**Lemma 2.5.** Let  $(X, \mathcal{A}, \mu)$  be a measure space and let f be an integrable function on X. Then

- 1. f is finite a.e., i.e.,  $\{x \in X \mid |f(x)| = \infty\}$  is a null set.
- 2. The set  $\{x \in X \mid f(x) \neq 0\}$  is  $\sigma$ -finite.

Proof.

HW 5 - Q8 by Lemma 2.3 **Proposition 2.3.** Let  $(X, \mathcal{A}, \mu)$  be a measure space, then

1. If h is integrable on X, then

$$\bigvee_{E\in\mathcal{A}}\int_{E}h=0\iff\int|h|=0\iff h=0\ a.e.$$

2. If f, g are integrable on X, then

$$\underset{E\in\mathcal{A}}{\forall}\ \int_{E}f=\int_{E}g\iff f=g\ \textit{a.e.}$$

*Proof.* We prove this one by one.

1. We see that the second equivalence is done in Proposition 2.2, hence we prove the first equivalence only. Since we have

$$\int |h| = 0 \implies \left| \int_E h \right| \leq \int_E |h| \leq \int |h| = 0,$$

which shows one implication. Now assume that  $\int_E h = 0$  for all  $E \in \mathcal{A}$ , then we can write h as

$$h = u + iv = (u^{+} - u^{-}) + i(v^{+} - v^{-}).$$

Let  $B := \{x \in X \mid u^+(x) > 0\}$ , then by assumption, we have

$$0 = \int_{B} h = \Re \int_{B} h = \int_{B} u = \int_{B} u^{+} = \int_{B} u^{+} + \int_{B^{c}} u^{+} = \int u^{+},$$

hence  $u^+ = 0$  almost everywhere. Similarly, we have  $u^-, v^+, v^-$  are all zero almost everywhere. This gives us that h is zero almost everywhere as desired.

-

DIY

Theorem 2.4 (Dominated Convergence Theorem). Let  $(X, \mathcal{A}, \mu)$  be a measure space, and

- Let  $f_n$  be integrable on X.
- $\lim_{n\to\infty} f_n(x) = f(x)$  almost everywhere.
- There is a  $g: X \to [0, \infty]$  such that g is integrable and

$$\bigvee_{n \in \mathbb{N}} |f_n(x)| \le g(x) \text{ a.e.}$$

Then we have

$$\lim_{n \to \infty} \int f_n = \int f = \int \lim_{n \to \infty} f_n.$$

*Proof.* Let F be the countable union of null set on which the three conditions may fail. Then we see that after modifying the definition of  $f_n$ , f and g on F, we may assume that all three conditions hold everywhere since modifying on a null set does not change the integral.

We now consider the  $\overline{\mathbb{R}}$ -valued case only. Note that the second and the third conditions imply that f is integrable since  $|f| \leq g(x)$ . We then see that  $g + f_n \geq 0$  and  $g - f_n \geq 0$  because  $-g \leq f_n \leq g$ . From Theorem 2.3, we have

Check
C-valued
case

$$\int g + f \le \liminf_{n \to \infty} \int g + f_n, \quad \int g - f \le \liminf_{n \to \infty} \int g - f_n.$$

From the linearity of integral, we have

$$\int g + \int f \le \int g + \liminf_{n \to \infty} \int f_n, \quad \int g - \int f \le \int g - \liminf_{n \to \infty} \int f_n.$$

Now, since  $\int g < \infty$ , we can cancel it, which gives

$$\int f \le \liminf_{n \to \infty} \int f_n, \quad -\int f \le \liminf_{n \to \infty} \int -f_n = -\limsup_{n \to \infty} \int f_n,$$

which implies

$$\int f \le \liminf_{n \to \infty} f_n \le \limsup_{n \to \infty} \int f_n \le \int f.$$

This shows that the limit exists, and the desired result indeed holds.

Corollary 2.7 (Tonelli's theorem for series and integrals). Suppose  $f_n$  are integrable functions such that

$$\sum_{n=1}^{\infty} \int |f_n| < \infty,$$

then we have

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

*Proof.* Take G(x) to be

$$G(x) := \sum_{n=1}^{\infty} |f_n(x)|,$$

then we see

$$G(x) \ge |F_N(x)|$$

where

$$F_N(x) := \sum_{n=1}^N f_n(x).$$

By Corollary 2.4, we have

$$\int G(x) = \sum_{n=1}^{\infty} |f_n(x)| < \infty.$$

Lastly, from Theorem 2.4, the result follows.

Remark. Compare to Corollary 2.4, we see that we further generalize the result!

# Lecture 14: $L^1$ Space

7 Feb. 11:00

# 2.4 $L^1$ Space

We now introduce another space called  $L^p$  spaces, which are function spaces defined using a natural generalization of the p-norm for finite-dimensional vector spaces. We sometimes call it Lebesgue spaces also.

Before we start, we need to define a *norm*.

**Definition 2.11 (Seminorm).** Let V be a vector space over filed  $\mathbb{R}$  or  $\mathbb{C}$ . A *seminorm* on V is

$$\|\cdot\|:V\to[0,\infty)$$

such that

- ||cv|| = |c| ||v|| for every  $v \in V$  and every scalar c.
- $||v + w|| \le ||v|| + ||w||$  for every  $v, w \in V$ .

Definition 2.12 (Norm). A norm is a seminorm with

 $\bullet \|v\| = 0 \iff v = 0.$ 

Lemma 2.6. A normed vector space is a metric space with metric

$$\rho(v, w) = ||v - w||.$$

Proof.

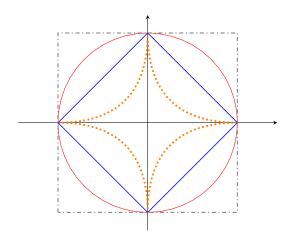
$$\left\|x\right\|_{p} = \begin{cases} \left(\sum_{i=1}^{d}\left|x_{i}\right|^{p}\right)^{1/p}, & \text{if } p \in [0, \infty); \\ \max_{1 \leq i \leq d}\left|x_{i}\right|, & \text{if } p = \infty \end{cases}$$

is a normed vector space. The unit ball

Example (p-norm).  $V = \mathbb{R}^d$  with

$$\{x \in \mathbb{R}^d \mid ||x||_n \le 1\}$$

for different p has the following figures.



**Remark.** All  $\|\cdot\|_p$  norms induce the same topology. i.e., if U is open in p-norm, it is open in p'-norm as well.

**Note.** Recall that we say f is integrable means

$$\int |f| < \infty,$$

$$\int f = \int g$$

and if f = g a.e., then

$$\int f = \int g$$

**Definition 2.13** ( $L^1$  Space). Given  $(X, \mathcal{A}, \mu)$ ,

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

means that f is an integrable function on X.

**Lemma 2.7.**  $L^1(X, \mathcal{A}, \mu)$  is a vector space with seminorm

$$\|f\|_1 = \int |f| \,.$$

Proof.

Check this is indeed a seminorm.

Definition 2.14 ( $L^1$  Space with equivalence class). Define  $f \sim g$  if f = g a.e., then

$$L^1(X, \mathcal{A}, \mu) /_{\sim} = L^1(X, \mathcal{A}, \mu),$$

i.e., we simply denote the collection of equivalence classes by itself.  $^a$ 

 $<sup>^{</sup>a}$ By some abusing of notation of  $L^{1}$ .

Remark. We have

- With Definition 2.14,  $L^1(X, \mathcal{A}, \mu)$  is a normed vector space.
- We say that the  $L^1$ -metric  $\rho(f,g)$  is simply

$$\rho(f,g) = \int |f - g|.$$

#### 2.4.1 Dense Subsets of $L^1$

**Note.** Recall the definition of a dense  $set^{23}$ .

**Definition 2.15 (Step function).** A step function on  $\mathbb{R}$  is

$$\psi = \sum_{i=1}^{N} c_i \mathbb{1}_{I_i},$$

where  $I_i$  is an <u>interval</u>.

**Theorem 2.5.** We have the following.

- 1. {integrable simple functions} is dense in  $L^1(X, \mathcal{A}, \mu)$  (with respect to  $L^1$ -metric).
- 2.  $(X, \mathcal{A}, \mu) = (\mathbb{R}, \mathcal{A}_{\mu}, \mu)$ , where  $\mu$  is a Lebesgue-Stieltjes-measure. Then {integrable simple functions} is dense in  $L^1(\mathbb{R}, \mathcal{A}_{\mu}, \mu)$ .
- 3.  $C_c(\mathbb{R})$  is dense in  $L^1(\mathbb{R}, \mathcal{L}, m)$ .

**Notation.** We denote the collection of continuous functions with compact support by  $C_c(\mathbb{R})$ .

*Proof.* We prove this one by one.

1. Since there exists simple functions  $0 \le |\phi_1| \le |\phi_2| \le \ldots \le |f|$ , where  $\phi_n \to f$  pointwise. Then by Theorem 2.4, we have

$$\lim_{n \to \infty} \int \underbrace{|f_n - f|}_{\le |\phi_n| + |f| \le 2|f|} = 0$$

where 2|f| is in  $L^1$ .

2. Let  $\mathbb{1}_E$  approximate by  $\sum_{i=1}^{\infty} c_i \mathbb{1}_{I_i}$ . From Theorem 1.6 for Lebesgue-Stieltjesmeasure,

$$\forall \epsilon' > 0 \; \exists I = \bigcup_{i=1}^{N} I_i \text{ such that } \mu(E \triangle I) \leq \epsilon'.$$

<sup>23</sup> https://en.wikipedia.org/wiki/Dense\_set

3. To approximate  $\mathbb{1}_{(a,b)}$ , we simply consider function  $g \in C_c(\mathbb{R})$  such that

$$\int \left| \mathbb{1}_{(a,b)} - g \right| \le \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon.$$

# Lecture 15: Riemann Integral

9 Feb. 11:00

# 2.5 Riemann Integrability

We are now working in  $(\mathbb{R}, \mathcal{L}, m)$ . Let's first revisit the definition of Riemann Integral. Let P be a partition of [a, b] as

$$P = \{a = t_0 < t_1 < \dots < t_k = b\}.$$

Then the lower Riemann sum of f using P is equal to  $L_P$ , which is defined as

$$L_P = \sum_{i=1}^{K} \left( \inf_{[t_{i-1}, t_i]} f \right) (t_i - t_{i-1}),$$

and the upper Riemann sum of f using P is equal to  $U_P$ , which is defined as

$$U_P = \sum_{i=1}^{K} \left( \sup_{[t_{i-1}, t_i]} f \right) (t_i - t_{i-1}).$$

Then we call

- Lower Riemann integral of  $f = \underline{I} = \sup_{P} L_{P}$
- Upper Riemann integral of  $f = \overline{I} = \inf_P U_P$

**Definition 2.16 (Riemann (Darboux) integrable).** A <u>bounded</u> function  $f: [a, b] \to \mathbb{R}$  is called *Riemann (Darboux) integrable* if

$$I = \overline{I}$$

If so, then  $\underline{I} = \overline{I} = \int_a^b f(x) dx$ .

Note. We see that

• If  $P \subset P'$ , then

$$L_P \le L_{P'}, \quad U_{P'} \le U_P.$$

• Recall that continuous functions on [a, b] are Reimann integrable on [a, b].

**Theorem 2.6.** Let  $f: [a,b] \to \mathbb{R}$  be a bounded function. Then

1. If f is Reimann integrable, then f is Lebesgue measurable, thus Lebesgue integrable. Further,

$$\int_{a}^{b} f(x) \, \mathrm{d}x = \int_{[a,b]} f \, \mathrm{d}m.$$

2. If f is Reimann integrable  $\iff$  f is continuous Lebesgue a.e.<sup>a</sup>

 $^{a}$ Here, we mean that the set where f is discontinuous is a null set under Lebesgue measure.

*Proof.* There exists  $P_1 \subset P_2 \subset \dots$  such that  $L_{P_n} \nearrow \underline{I}$  and  $U_{P_n} \searrow \overline{I}$ .<sup>24</sup> Now, define simple (step) functions

• 
$$\phi_n = \sum_{i=1}^K \left( \inf_{[t_{i-1}, t_i]} f \right) \mathbb{1}_{(t_{i-1}, t_i]}$$

• 
$$\psi_n = \sum_{i=1}^K \left( \sup_{[t_{i-1}, t_i]} f \right) \mathbb{1}_{(t_{i-1}, t_i]}$$

if  $P_n = \{a = t_0 < t_1 < \ldots < t_K\}$ . Let  $\phi := \sup_n \phi_n$  and  $\psi := \inf_n \psi_n$ . We then see that  $\phi, \psi$  are Lebesgue (Borel) measurable function.

**Note.** Note that

- $\exists M > 0$  such that  $\bigvee_{n \in \mathbb{N}} |\phi_n|, |\psi_n| \leq M \mathbbm{1}_{[a,b]}$
- $\int \phi_n \, \mathrm{d}m = L_{P_n}, \int \psi_n \, \mathrm{d}m = U_{P_n}$

By Theorem 2.4 and the fact that  $M1_{[a,b]} \in L^1(\mathbb{R},\mathcal{L},m)$ , we have

$$\underline{I} = \lim_{n \to \infty} \int \phi_n \, \mathrm{d}m = \int \phi \, \mathrm{d}m, \quad \overline{I} = \int \psi \, \mathrm{d}m.$$

Thus,

$$f$$
 is Riemann integrable  $\iff \int \phi = \int \psi$   $\iff \int (\psi - \phi) = 0$   $\iff \psi = \phi$  Lebesgue a.e.

# 2.6 Modes of Convergence

As we should already see, there are different modes of convergence. Let's formalize them.

<sup>&</sup>lt;sup>24</sup>Here, we took refinements of  $P_n$  if needed.

Definition 2.17 (Pointwise, Uniformly convergence). Let

$$f_n, f: X \to \mathbb{C},$$

and  $S \subset X$ . Then we say

•  $f_n \to f$  pointwise on S if

$$\forall \exists \exists \forall \exists \forall |f_n(x) - f(x)| < \epsilon.$$

•  $f_n \to f$  uniformly on S if

$$\forall \exists \forall \forall \forall f \mid f_n(x) - f(x) \mid < \epsilon.$$

**Remark.** We see that we can replace  $\forall \epsilon > 0$  by  $\forall k \in \mathbb{N}$  with  $\epsilon$  changing to  $\frac{1}{k}$ .

**Lemma 2.8.** Let  $B_{n,k}$  be

$$B_{n,k} := \left\{ x \in X \mid |f_n(x) - f(x)| < \frac{1}{k} \right\}.$$

Then

1.  $f_n \to f$  pointwise on S if and only if

$$S \subset \bigcap_{k=1}^{\infty} \bigcup_{N=1}^{\infty} \bigcap_{n=N}^{\infty} B_{n,k}.$$

2.  $f_n \to f$  uniformly on S if and only if  $\exists N_1, N_2, \ldots \in \mathbb{N}$  such that

$$S \subset \bigcap_{k=1}^{\infty} \bigcap_{n=N_k}^{\infty} B_{n,k}.$$

*Proof.* This essentially follows from Definition 2.17.

Definition 2.18 (Converges a.e., Converges in  $L^1$ ). Let  $(X, \mathcal{A}, \mu)$  be a measure space. Assuming that  $f_n, f$  are measurable function, then

1.  $f_n \to f$  almost everywhere means

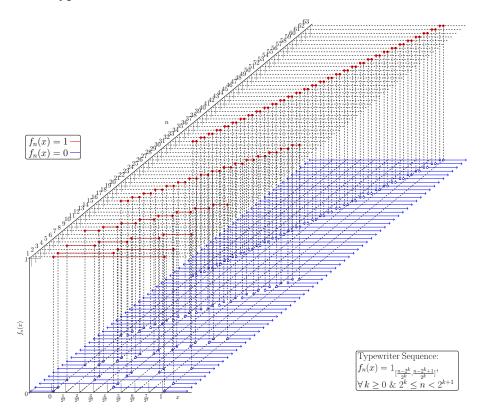
 $\exists$  null set E such that  $f_n \to f$  pointwise on  $E^c$ .

2.  $f_n \to f$  in  $L^1$  means

$$\lim_{n\to\infty} ||f_n - f|| = 0.$$

**Example.** Given  $(\mathbb{R}, \mathcal{L}, m)$  and let f = 0. We see the followings.

- 1.  $f_n = \mathbb{1}_{(n,n+1)}$
- 2.  $f_n = \frac{1}{n} \mathbb{1}_{(0,n)}$
- 3.  $f_n = n \mathbb{1}_{(0,\frac{1}{n})}$
- 4. Typewriter functions.



# Lecture 16: Modes of Convergence

11 Feb. 11:00

Let's start with a proposition.

Proposition 2.4 (Fast  $L^1$  convergence leads to a.e. convergence). Let  $(X, \mathcal{A}, \mu)$  be a measure space, and  $f_n, f$  are all measurable functions on X. Then

$$\sum_{n=1}^{\infty} \|f_n - f\|_1 < \infty \implies f_n \to f \text{ a.e.}$$

Proof. Let

$$E := \bigcup_{k=1}^{\infty} \bigcap_{N=1}^{\infty} \bigcup_{n=N}^{\infty} B_{n,k}^{c} = \{ x \in X \mid f_{n}(x) \nrightarrow f(x) \}.$$

By Lemma 2.3, we see that

$$\forall_{k} \forall_{N} \mu\left(B_{n,k}^{c}\right) \leq k \int |f_{n} - f| \implies \forall_{k} \mu\left(\bigcup_{n=N}^{\infty} B_{n,k}^{c}\right) \leq \sum_{n=N}^{\infty} k \|f_{n} - f\|_{1} \to 0$$

as  $N \to \infty$ . Now, by continuity of measure from above,

$$\forall \ \mu\left(\bigcap_{N=1}^{\infty}\bigcup_{n=N}^{\infty}B_{n,k}^{c}\right)=\lim_{N\to\infty}\mu\left(\bigcup_{n=N}^{\infty}B_{n,k}^{c}\right)=0 \implies \mu(E)=0$$

since  $f_n \to f$  pointwise on  $E^c$ .

Corollary 2.8. Given  $\{f_n\}_n$  such that  $f_n \to f$  in  $L^1$ , there exists a subsequence  $\{f_{n_j}\}_{n_j}$  where  $f_{n_j} \to f$  a.e.

Proof. Since

$$\forall \forall \forall f_{n_j \in \mathbb{N}} ||f_{n_j} - f||_1 \le \frac{1}{j^2}.$$

Then,

$$\sum_{j=1}^{\infty} \left\| f_{n_j} - f \right\|_1 < \infty.$$

From Proposition 2.4, we have the desired result.

**Definition 2.19 (Converge in measure).** Let  $f_n, f$  be measurable functions on  $(X, \mathcal{A}, \mu)$ . Then  $f_n \to f$  in measure means

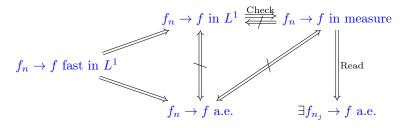
$$\bigvee_{\epsilon > 0} \lim_{n \to \infty} \mu\left(\left\{x \in X \mid |f_n(x) - f(x)| \ge \epsilon\right\}\right) = 0.$$

**Example.** Let  $f_n = n \mathbb{1}_{(0,\frac{1}{n})}$  and f = 0. We see that

$$\forall \epsilon > 0 \ \left\{ x \in X \mid |f_n(x) - f(x)| > \epsilon \right\} = \left(0, \frac{1}{n}\right),$$

 $f_n \to 0$  in measure. (Recall that  $f_n \nrightarrow 0$  in  $L^1$ )

Remark. We see that



**Definition 2.20 (Uniformly a.e., Almost uniformly).** Let  $f_n, f$  be measurable functions on  $(X, \mathcal{A}, \mu)$ .

- 1.  $f_n \to f$  uniformly almost everywhere means  $\exists \text{null set } F$  such that  $f_n \to f$  uniformly on  $F^c$ .
- 2.  $f_n \to f$  almost uniformly means  $\forall \epsilon > 0 \ \exists F \in \mathcal{A}$  such that  $\mu(F) < \epsilon$ ,  $f_n \to f$  uniformly on  $F^c$ .

#### Lemma 2.9. We have

$$f_n \to f$$
 uniformly on  $S \iff \exists N_1, N_2, \ldots \in \mathbb{N} \ S \subset \bigcap_{k=1}^{\infty} \bigcap_{n=N_k}^{\infty} B_{n,k}$ .

Theorem 2.7 (Egorov's Theorem). Let  $f_n, f$  be measurable functions on  $(X, \mathcal{A}, \mu)$ . Suppose  $\mu(X) < \infty$ , then

 $f_n \to f \ a.e. \iff f_n \to f \ \text{almost uniformly}.$ 

*Proof.* We prove two directions.

- ← DIY
- $\implies$  Fix  $\epsilon > 0$ . We see that

$$f_n \to f \text{ a.e.} \implies \mu \left( \bigcup_{k=1}^{\infty} \bigcap_{N=1}^{\infty} \bigcup_{n=N}^{\infty} B_{n,k}^c \right) = 0$$

$$\implies \forall \mu \left( \bigcap_{N=1}^{\infty} \bigcup_{n=N}^{\infty} B_{n,k}^c \right) = 0.$$

From continuity of measure from above and  $\mu(X) < \infty$ , we further have

$$\forall \lim_{k} \lim_{N \to \infty} \mu \left( \bigcup_{n=N}^{\infty} B_{n,k}^{c} \right) = 0 \implies \forall \lim_{k} \lim_{N_{k} \in \mathbb{N}} \mu \left( \bigcup_{n=N_{k}}^{\infty} B_{n,k}^{c} \right) < \frac{\epsilon}{2^{k}}.$$

Now, let

$$F\coloneqq \bigcup_{k=1}^\infty \bigcup_{n=N_k}^\infty B_{n,k}^c,$$

we see that  $\mu(F) < \epsilon$ , hence  $f_n \to f$  uniformly.

## 3 Product Measure

# 3.1 Product $\sigma$ -algebra

Before we start, we see the setup.

3 PRODUCT MEASURE

• Product space.

$$X = \prod_{\alpha \in I} X_{\alpha}$$

where  $x = (x_{\alpha})_{{\alpha} \in I} \in X$ .

• Coordinate map.

$$\pi_{\alpha} \colon X \to X_{\alpha}$$
.

Now we see the formal definition.

**Definition 3.1 (Product**  $\sigma$ -algebra). Let  $(X_{\alpha}, \mathcal{A}_{\alpha})$  be a measurable space for all  $\alpha \in I$ . Then a product  $\sigma$ -algebra on  $X = \prod_{\alpha \in I} X_{\alpha}$  is

$$\bigotimes_{\alpha \in I} \mathcal{A}_{\alpha} = \left\langle \bigcup_{\alpha \in I} \pi_{\alpha}^{-1} \left( \mathcal{A}_{\alpha} \right) \right\rangle,\,$$

where

$$\pi_{\alpha}^{-1}(\mathcal{A}_{\alpha}) = \left\{ \pi_{\alpha}^{-1}(E) \mid E \in \mathcal{A}_{\alpha} \right\}.$$

**Notation.** We denote  $I = \{1, \ldots, d\} \implies X = \prod_{i=1}^d X_i, x = (x_1, \ldots, x_d)$ . Also,

$$\bigotimes_{i=1}^d \mathcal{A}_i = \mathcal{A}_1 \otimes \ldots \otimes \mathcal{A}_d.$$

**Lemma 3.1.** If I is countable, then

$$\bigotimes_{i=1}^{\infty} \mathcal{A}_i = \left\langle \left\{ \prod_{i=1}^{\infty} E_i \mid \forall E_i \in \mathcal{A}_i \right\} \right\rangle.$$

Proof.

DIY

# Lecture 17: Product Measure

14 Feb. 11:00

We now see a lemma.

**Lemma 3.2.** Suppose  $\mathcal{A}_{\alpha} = \langle \mathcal{E}_{\alpha} \rangle$  for every  $\alpha \in I$ . Then

1. 
$$\pi_{\alpha}^{-1}(\mathcal{A}_{\alpha}) = \langle \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \rangle$$

2. 
$$\bigotimes_{\alpha} \mathcal{A}_{\alpha} = \left\langle \bigcup_{\alpha} \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \right\rangle$$

3. If I is countable, then

$$\bigotimes_{i}^{\infty} \mathcal{A}_{i} = \left\langle \left\{ \prod_{i=1}^{\infty} E_{i} \mid \forall E_{i} \in \mathcal{E}_{i} \right\} \right\rangle$$

*Proof.* We prove this one by one.

1. Note that for  $f\colon Y\to Z$ , and  $\mathcal B$  be a  $\sigma$ -algebra on Z, then  $f^{-1}(\mathcal B)$  is also a  $\sigma$ -algebra.<sup>25</sup> Hence,  $\pi_\alpha^{-1}$  is a  $\sigma$ -algebra on X, i.e.,

$$\pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \subset \pi_{\alpha}^{-1}(\mathcal{A}_{\alpha}) \stackrel{!}{\Longrightarrow} \langle \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \rangle \subset \pi_{\alpha}^{-1}(\mathcal{A}_{\alpha}).$$

To show the other direction, let  $\mathcal{M}$  being

$$\mathcal{M} = \left\{ B \subset X_{\alpha} \mid \pi_{\alpha}^{-1}(B) \in \left\langle \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \right\rangle \right\}.$$

We now check

•  $\mathcal{M}$  is a  $\sigma$ -algebra.

Check (Easy)!

•  $\mathcal{E}_{\alpha} \subset \mathcal{M}$ . This is true by definition of  $\mathcal{M}$ .

Thus,  $\langle \mathcal{E}_{\alpha} \rangle = \mathcal{A}_{\alpha} \subset \mathcal{M}$ . Hence, if  $E \in \mathcal{A}_{\alpha}$ ,  $E \in \mathcal{M}$ , implying

$$\pi_{\alpha}^{-1}(E) \in \langle \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \rangle$$
,

i.e.,  $\mathcal{A}_{\alpha} \subset \langle \pi_{\alpha}^{-1}(\mathcal{E}_{\alpha}) \rangle$ .

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3.

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**Theorem 3.1.** Suppose  $X_1, \ldots, X_d$  are metric spaces. Let  $X = \prod_{i=1}^d X_i$  with product metric defined as

$$\rho(x,y) = \sum_{i=1}^{d} \rho_i(x_i, y_i).$$

Then,

1. 
$$\bigotimes_{i=1}^{d} \mathcal{B}(X_i) \subset \mathcal{B}(X)$$

2. If in addition, each  $X_i$  has a countable dense subset,

$$\bigoplus_{i=1}^{d} \mathcal{B}(X_i) = \mathcal{B}(X).$$

Proof.

■ DIY

Remark. We see that

• 
$$\mathcal{B}(\mathbb{R}^d) = \mathcal{B}(\mathbb{R}) \otimes \ldots \otimes \mathcal{B}(\mathbb{R})$$

<sup>25</sup>Since 
$$f^{-1}(\emptyset) = \emptyset$$
,  $f^{-1}(B)^c = f^{-1}(B^c)$ , and  $\bigcup_n f^{-1}(B_n) = f^{-1}(\bigcup_n B_n)$ .

• let  $f = u + iv : X \to \mathbb{C}$ , and  $\mathcal{A}$  be a  $\sigma$ -algebra on X. Then

$$\bigvee_{E \in \mathcal{B}(\mathbb{R})} u^{-1}(E), v^{-1}(E) \in \mathcal{A} \iff f^{-1}(F) \in \mathcal{A}, \forall F \in \mathcal{B}(\mathbb{C})$$

with  $\mathcal{B}(\mathbb{C}) = \mathcal{B}(\mathbb{R}^2) = \mathcal{B}(\mathbb{R}) \otimes \mathcal{B}(\mathbb{R})$ .

**Definition 3.2** (x-section, y-section). Let X, Y be two sets. Then

1. For  $E \subset X \times Y$ ,

$$E_x = \{ y \in Y \mid (x, y) \in E \}, \quad E^y = \{ x \in X \mid (x, y) \in E \}.$$

2. For  $f: X \times Y \to \mathbb{C}$ , define

$$f_x \colon Y \to \mathbb{C}, \quad f^y \colon X \to \mathbb{C}$$

by

$$f_x(y) = f(x, y) = f^y(x).$$

**Example.** We see that

$$(1_E)_r = 1_{E_r}$$

and

$$(\mathbb{1}_E)^y = \mathbb{1}_{E^y}.$$

**Proposition 3.1.** Given two measurable spaces (X, A) and (Y, B), then

1. If  $E \in \mathcal{A} \otimes \mathcal{B}$ , then

$$\bigvee_{x \in X} \bigvee_{y \in Y} E_x \in \mathcal{B}, E^y \in \mathcal{A}.$$

2. If  $f: X \times Y \to \mathbb{C}$  is  $\mathcal{A} \otimes \mathcal{B}$ -measurable, then

 $\underset{x \in X}{\forall} \ \underset{y \in Y}{\forall} \ f_x \text{ is } \mathcal{B}\text{-measurable}, f^y \text{ is } \mathcal{A}\text{-measurable}.$ 

*Proof.* We prove this one by one.

1. Let 
$$\mathcal{F} := \left\{ E \subset X \times Y \mid \begin{subarray}{c} \forall \ x \in X \ y \in Y \end{subarray} \mid E_x \in \mathcal{B}, E^y \in \mathcal{A} \right\}$$
, then

- $\mathcal{F}$  is a  $\sigma$ -algebra (Easy to show since  $(E^c)_x = E_x^c, \dots$ )
- Let  $\mathcal{R}_0 := \{A \times B \mid A \in \mathcal{A}, B \in \mathcal{B}\} \subset \mathcal{F}$ , which is again easy to show from definition.

Hence, we see that  $\langle R_0 \rangle = \mathcal{A} \otimes \mathcal{B} \subset \mathcal{F}$ .

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3 PRODUCT MEASURE

#### 3.2 Product Measures

We start with the definition.

**Definition 3.3 (Rectangle).** Given two measurable spaces, a *(measurable) rectangle* is  $R = A \times B$  where  $A \in \mathcal{A}$  and  $B \in \mathcal{B}$ . Furthermore, we let

$$\mathcal{R}_0 := \{ R = A \times B \mid A \in \mathcal{A}, B \in \mathcal{B} \},\,$$

and

$$\mathcal{R} \coloneqq \left\{ \bigcup_{i=1}^{N} R_i \mid N \in \mathbb{N}, R_1, \dots, R_N \text{ disjoint rectangles} \right\}.$$

Note. Whenever we're talking about rectangle, they're always measurable.

**Lemma 3.3.**  $\mathcal{R}$  is an algebra, and

$$\langle \mathcal{R}_0 \rangle = \langle \mathcal{R} \rangle = \mathcal{A} \otimes \mathcal{B}.$$

*Proof.* Simply observe that

$$(A \times B)^c = (A^c \times Y) \cup (A \times B)$$

#### DIY

#### Lecture 18: Monotone Class

16 Feb. 11:00

Let's start with a theorem.

**Theorem 3.2.** Let  $(X, \mathcal{A}, \mu), (Y, \mathcal{B}, \nu)$  be measure spaces. Then

1. There is a measure  $\mu \times \nu$  on  $\mathcal{A} \otimes \mathcal{B}$  satisfying

$$(\mu \times \nu)(A \times B) = \mu(A)\nu(B)$$

for every  $A \in \mathcal{A}, B \in \mathcal{B}$ .

2. If  $\mu, \nu$  are  $\sigma$ -finite, then  $\mu \times \nu$  is unique.

*Proof.* We prove this one by one.

1. Define  $\mu \colon \mathcal{R} \to [0, \infty]$  by  $\mu(A \times B) = \mu(A)\nu(B)$ , and extending linearly, we have

$$\pi(A \times B) = \mu(A)\nu(B),$$

hence

$$\pi\left(\coprod_{i=1}^{N} A_i \times B_i\right) = \sum_{i=1}^{n} \pi(A_i \times B_i).$$

We claim that  $\pi$  is a pre-measure. To show this, it's enough to check that  $\pi(A \times B) = \sum_{n=1}^{\infty} \pi(A_n \times B_n)$  if  $A \times B = \coprod_n A_n \times B_n$ . Since  $A_n \times B_n$  are

disjoint, so

$$\mathbb{1}_{A\times B}(x,y) = \sum_{n=1}^{\infty} \mathbb{1}_{A_n\times B_n}(x,y).$$

Thus,

$$\mathbb{1}_{A}(x)\mathbb{1}_{B}(y) = \sum_{n=1}^{\infty} \mathbb{1}_{A_{n}}(x)\mathbb{1}_{B_{n}}(y).$$

Integrating with respect to x, and applying Proposition 1.3, we have

$$\int_X \mathbb{1}_{A}(x) \mathbb{1}_{B}(y) \, d\mu(x) = \sum_{n=1}^{\infty} \int_x \mathbb{1}_{A_n}(x) \mathbb{1}_{B_n}(y) \, d\mu(x),$$

which implies

$$\mu(A)\mathbb{1}_{B}(y) = \sum_{n=1}^{\infty} \mu(A_n)\mathbb{1}_{B_n}(y)$$

for every y. We can then integrate again with respect to y and apply Proposition 1.3, we have

$$\int_{Y} \mu(A) \mathbb{1}_{B}(y) \, d\nu(y) = \sum_{n=1}^{\infty} \int_{Y} \mu(A_{n}) \mathbb{1}_{B_{n}}(y) \, d\nu(y),$$

which gives us

$$\mu(A)\nu(B) = \sum_{n=1}^{\infty} \mu(A_n)\nu(B_n).$$

Hence, we see that  $\mu$  is indeed a pre-measure, so Theorem 1.3 gives  $\mu \times \nu$  on  $\langle \mathcal{R} \rangle = \mathcal{A} \otimes \mathcal{B}$  extending  $\pi$  on  $\mathcal{R}$ .

2. If  $\mu, \nu$  are  $\sigma$ -finite, then  $\pi$  is  $\sigma$ -finite on  $\mathcal{R}$ , then Theorem 1.4 applies. Moreover, we have that

$$(\mu \times \nu)(E) = \inf \left\{ \sum_{i=1}^{\infty} \mu(A_i) \nu(B_i) \mid E \subset \bigcup_{i=1}^{\infty} A_i \times B_i, A_i \in \mathcal{A}, B_i \in \mathcal{B} \right\}.$$

#### 3.3 Monotone Class Lemma

Let's start with a definition.

**Definition 3.4 (Monotone Class).** If X is a set, and  $C \subset \mathcal{P}(X)$ , we say that C is a monotone class on X if

- C is closed under countable increasing unions.
- C is closed under countable decreasing intersections.

Example. We see that

1. Every  $\sigma$ -algebra is a monotone class.

2. If  $C_{\alpha}$  are (arbitrarily many) monotone classes on a set X, then  $\bigcap_{\alpha} C_{\alpha}$  is a monotone class. Furthermore, if  $\mathcal{E} \subset \mathcal{P}(X)$ , there is a unique smallest monotone class containing  $\mathcal{E}$ .

Theorem 3.3 (Monotone Class Lemma). Suppose  $A_0$  is an algebra on X. Then  $\langle A_0 \rangle^a$  is the monotone class generated by  $A_0$ .

 ${}^{a}\langle \mathcal{A}_{0}\rangle$  is the  $\sigma$ -algebra generated by  $\mathcal{A}_{0}$  by Definition 1.3.

*Proof.* Let  $\mathcal{A} = \langle \mathcal{A}_0 \rangle$  and let  $\mathcal{C}$  be the monotone class generated by  $\mathcal{A}_0$ . Since  $\mathcal{A}$  is a  $\sigma$ -algebra, it's a monotone class. Note that it contains  $\mathcal{A}_0$ , hence  $\mathcal{A} \supset \mathcal{C}$ .

To show  $\mathcal{C} \supset \mathcal{A}$ , it's enough to show that  $\mathcal{C}$  is a  $\sigma$ -algebra. We check that

- 1.  $\varnothing \in \mathcal{A}_0 \subseteq \mathcal{C}$ .
- 2. Let  $\mathcal{C}' = \{ E \subset X \mid E^c \in \mathcal{C} \}.$ 
  - C' is a monotone class.
  - $\mathcal{A}_0 \subset \mathcal{C}'$  because if  $E \in \mathcal{A}_0$ , then  $E^c \in \mathcal{A}_0$ , so  $E^c \in \mathcal{C}$ , thus  $E \in \mathcal{C}'$ .

We see that  $\mathcal{C}' \subset \mathcal{C}'$ , so  $\mathcal{C}$  is closed under complements.

- 3. For  $E \subset X$ , let  $\mathcal{D}(E) = \{ F \in \mathcal{C} \mid E \cup F \in \mathcal{C} \}$ .
  - $\mathcal{D}(E) \subset \mathcal{C}$ .
  - $\mathcal{D}(E)$  is a monotone class.
  - If  $E \in \mathcal{A}_0$ , then  $\mathcal{A}_0 \subset \mathcal{D}(E)$ . We see this by picking  $F \in \mathcal{A}_0$ , then  $E \cup F \in \mathcal{A}_0 \supset \mathcal{C}$ .

Hence,  $C = \mathcal{D}(E)$  if  $E \in \mathcal{A}_0$ .

- 4. Let  $\mathcal{D} = \{E \in \mathcal{C} \mid \mathcal{D}(E) = \mathcal{C}\}$ . That is  $\mathcal{D} = \{E \in \mathcal{C} \mid E \cup F, \forall F \in \mathcal{C}\}$ . Then we have
  - $A_0 \subset \mathcal{D}$  by 3.
  - $\mathcal{D}$  is a monotone class.
  - $\mathcal{D} \subset \mathcal{C}$  by definition.

Thus,  $\mathcal{D} = \mathcal{C}$ , so if  $E, F \in \mathcal{C}$ , then  $E \cup F \in \mathcal{C}$ . This implies that  $\mathcal{C}$  is closed under finite unions.

5. Now to show that C is closed under countable unions, let  $E_1, E_2, \ldots \in C$ . We may then define

$$F_N = \bigcup_{n=1}^N E_n \in \mathcal{C}.$$

Then we see that  $F_1 \subset F_2 \subset ...$ , hence  $\bigcup_N F_N \in \mathcal{C}$ . But this simply implies

$$\bigcup_{N} F_N = \bigcup_{n} E_n,$$

so we're done.

#### Lecture 19: Fubini-Tonelli's Theorem

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As previously seen. If  $E \in A \otimes B \implies E_x \in \mathcal{B}, E^y \in \mathcal{A} \ \forall x \in X, \forall y \in Y$ . Note that the reverse is not true.

### 3.4 Fubini-Tonelli Theorem

We start with a theorem.

Theorem 3.4 (Tonelli for charactistic functions). Given  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be  $\sigma$ -finite measure space. Suppose  $E \in \mathcal{A} \otimes \mathcal{B}$ , then

- 1.  $\alpha(x) := \nu(E_x) : X \to [0, \infty]$  is a  $\mathcal{A}$ -measurable function.
- 2.  $\beta(x) := \mu(E^y) : Y \to [0, \infty]$  is a  $\mathcal{B}$ -measurable function.
- 3.  $(\mu \times \nu)(E) = \int_X \nu(E_x) \, d\mu(x) = \int_Y \mu(E^y) \, d\nu(y).$

*Proof.* We prove this one by one.

1. Assume that  $\mu, \nu$  are finite measure. Let

$$C := \{ E \in \mathcal{A} \otimes \mathcal{B} \mid \text{ Conditions 1., 2., 3., hold} \}.$$

It's enough to prove that  $\langle \mathcal{R} \rangle = \mathcal{A} \otimes \mathcal{B} \subset C$ . We further observe that from the monotone class lemma and the fact that  $\mathcal{R}$  is an algebra, it's also citation enough to show that

- $\mathcal{R} \subset C$ .
- $\bullet$  C is a monotone class.

From condition 1.,

$$\alpha(x) = \nu\left((A \times B)_x\right) = \begin{cases} \nu(B), & \text{if } x \in A; \\ 0, & \text{if } x \notin A \end{cases} = \nu(B) \mathbb{1}_A.$$

And from condition 2.,

$$(\mu \times \nu)(A \times B) = \mu(A)\nu(B)$$

and

$$\int_X \nu((A \times B)_x) \, \mathrm{d}\mu(x) = \nu(B)\mu(A).$$

Let  $E_n \in C$ ,  $E_1 \subset E_2 \subset \ldots$  We need to show  $E = \bigcup_{n=1}^{\infty} E_n \in C$ . We now see that

$$E_x = \bigcup_{n=1}^{\infty} (E_n)_x, (E_1)_x \subset (E_2)_x \subset \dots$$
  
$$\Longrightarrow \alpha(x) = \nu(E_n)_x \stackrel{!}{=} \lim_{n \to \infty} \nu((E_n)_x) \ \forall x \in X.$$

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This implies that 1. is proved.

For 3., we see that

$$(\mu \times \nu)(E) \stackrel{!}{=} \lim_{n \to \infty} (\mu \times \nu)(E_n)$$
$$= \lim_{n \to \infty} \int_X \nu((E_n)_x) \, \mathrm{d}\mu(x)$$
$$\stackrel{!}{=} \int_X \nu(E_x) \, \mathrm{d}\mu(x).$$

Now let  $F_n \in C$ ,  $F_1 \supset F_2 \supset \ldots$  We need to show that  $F = \bigcap_{n=1}^{\infty} F_n \in C$ . Instead of using Theorem 2.2, we now want to use Theorem 2.4, which is applicable since  $\mu(X), \nu(Y) < \infty$  by assumption. Then assume that  $\mu, \nu$  are  $\sigma$ -finite, then

$$X \times Y = \bigcup_{n=1}^{\infty} (X_n \times Y_n), \begin{cases} X_1 \subset X_2 \subset \dots, & \mu(X_k) < \infty \\ Y_1 \subset Y_2 \subset \dots, & \nu(Y_k) < \infty. \end{cases}$$

DIY

**Theorem 3.5 (Fubini-Tonelli's Theorem).** We have the following two versions.

**Tonelli's ver.** If  $f: X \times Y \to [0, \infty]$  is  $\mathcal{A} \otimes \mathcal{B}$ -measurable, then

- 1.  $g(x) := \int_{V} f(x,y) d\nu(y), X \to [0,\infty]$  is a  $\mathcal{A}$ -measurable function.
- 2.  $h(x) := \int_X f(x,y) d\mu(x), Y \to [0,\infty]$  is a  $\mathcal{B}$ -measurable function.
- 3. We have

$$\int\limits_{X\times Y} f \,\mathrm{d}(\mu\times\nu) = \int\limits_X \left(\int\limits_Y f(x,y) \,\mathrm{d}\nu(y)\right) \mathrm{d}\mu(x) = \int\limits_Y \left(\int\limits_X f(x,y) \,\mathrm{d}\mu(x)\right) \mathrm{d}\nu(y).$$

**Fubini's ver.** If  $f \in L^1(X \times Y, \mu \times \nu)$ , then

- 1.  $f_x \in L^1(Y, \nu)$  for  $\mu$ -a.e. x, and  $g(x) \in L^1(X, \mu)$  defined  $\mu$ -a.e.
- 2.  $f^y \in L^1(X,\mu)$  for  $\nu$ -a.e. y, and  $h(x) \in L^1(Y,\nu)$  defined  $\mu$ -a.e.
- 3. The iterated integral formulas hold.

Proof. Read [FF99].

Lecture 20: Lebesgue Measure on  $\mathbb{R}^d$ 

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# 3.5 Lebesgue Measure on $\mathbb{R}^d$

**Example.** We first see some examples.

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- 1.  $(\mathbb{R}^2, \mathcal{L} \otimes \mathcal{L}, m \times m)$  is not complete.
  - Let  $A \in \mathcal{L}$ ,  $A \neq \emptyset$ , m(A) = 0.
  - Let  $B \subset [0,1]$ ,  $B \notin \mathcal{L}$  (Vital set for example).
  - Let  $E = A \times B$ ,  $F = A \times [0, 1]$ .

We see that  $E \subset F$ ,  $F \in \mathcal{L} \otimes \mathcal{L}$ ,  $(m \times m)(F) = m(A)m([0,1]) = 0$ , i.e., F is a null set. But E is **not**  $\mathcal{L} \otimes \mathcal{L}$ -measurable-function since otherwise, its sections are all measurable.

**Definition 3.5.** Let  $(\mathbb{R}^d, \mathcal{L}^d, m^d)$  be the *completion* of

$$(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d), m \times \ldots \times m),$$

which is same as the completion of

$$(\mathbb{R}^d, \mathcal{L} \otimes \ldots \otimes \mathcal{L}, m \times \ldots m).$$

Remark. We see that

$$\mathcal{L}^d \supsetneq \mathcal{L} \otimes \ldots \otimes \mathcal{L} = \left\langle \left\{ \prod_{i=1}^d E_i \mid E_i \in \mathcal{L} \right\} \right\rangle.$$

**Definition 3.6 (General rectangle).** A rectangle in  $\mathbb{R}^d$  is  $R = \prod_{i=1}^d E_i$  where  $E_i \in \mathcal{B}(\mathbb{R})$ .

**Definition 3.7.** We let

$$m^d(E) := \inf \left\{ \sum_{k=1}^{\infty} m^d(R_k) \mid E \subset \bigcup_{k=1}^{\infty} R_k, R_k \text{ is rectangles} \right\}.$$

**Theorem 3.6.** Let  $E \subset \mathcal{L}^d$ . Then

- 1.  $m^d(E) = \inf \{ m^d(0) \mid \text{open } O \supset E \} = \sup \{ m^d(K) \mid \text{compact } K \subset E \}.$
- 2.  $E = A_1 \cup N_1 = A_2 \setminus N_2$ , where  $A_1$  is  $F_{\sigma}$ ,  $A_2$  is  $G_{\delta}$ , and  $N_i$  are null.
- 3. If  $m^d(E) < \infty$ ,  $\forall \epsilon > 0$ ,  $\exists R_1, \dots, R_m$  rectangles whose sides are intervals such that

$$m^d \left( E \triangle \left( \bigcup_{i=1}^m R_i \right) \right) < \epsilon.$$

*Proof.* Similar to d = 1 case.

**Theorem 3.7.** Integrable step functions and  $C_c(\mathbb{R}^d)$ , the collection of continuous functions, are dense in  $L^1(\mathbb{R}^d, \mathcal{L}^d, m^d)$ 

**Theorem 3.8.** Lebesgue measure in  $\mathbb{R}^d$  is translation-invariant.

Theorem 3.9. Effect of linear transformation on Lebesgue measure.

# 4 Differentiation on Euclidean Space

As previously seen. Given  $f:[a,b]\to\mathbb{R}$ , there are two versions of fundamental theorem of calculus:

1.

$$\int_a^b f'(x) \, \mathrm{d}x = f(b) - f(a).$$

2.

$$\frac{\mathrm{d}}{\mathrm{d}x} \int_{a}^{x} f(t) \, \mathrm{d}t = f(x),$$

which follows from

$$\lim_{r \to 0^+} \frac{1}{r} \int_x^{x+r} f(t) \, \mathrm{d}t = f(x) = \lim_{r \to 0^+} \frac{1}{r} \int_{x-r}^x f(t) \, \mathrm{d}t.$$

Remark. We see that

$$\lim_{r \to 0^+} \frac{1}{r} \int_x^{x+r} (f(t) - f(x)) dt = 0 = \lim_{r \to 0^+} \frac{1}{r} \int_{x-r}^x (f(t) - f(x)) dt,$$

where we have

$$f(x) = \frac{1}{r} \int_{x}^{x+r} f(t) \, \mathrm{d}t.$$

This generalized to  $f: \mathbb{R}^d \to \mathbb{R}$ , namely

$$\lim_{r \to 0^+} \frac{1}{\operatorname{vol}\left(B(x,r)\right)} \int_{B(x,r)} \left(f(t) - f(x)\right) \underbrace{\operatorname{d}\! t}_{\operatorname{\mathbb{R}}^d} \stackrel{?}{=} 0.$$

# 4.1 Hardy-Littlewood Maximal Function

We first see our notation.

**Notation.** Given a(n) (open) ball in  $\mathbb{R}^d$ , B = B(a, r), denote cB = B(a, cr) for c > 0.

**Lemma 4.1 (Vitali-type covering lemma).** Let  $B_1, \ldots, B_k$  be a finite collection of open balls in  $\mathbb{R}^d$ . Then there exists a sub-collection  $B'_1, \ldots, B'_m$  of disjoint open balls such that

$$\bigcup_{i=1}^{m} \left( 3B_j' \right) \supset \bigcup_{i=1}^{k} B_i.$$

*Proof.* Greedy Algorithm.

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Notation. We let

$$\int_{E} f \, \mathrm{d}m = \int_{E} f(x) \, \mathrm{d}x.$$

The problem we're working on is

$$\frac{1}{m(B(w,r))} \int_{B(x,r)} f(y) \, \mathrm{d} y \overset{r \to 0}{\xrightarrow{?}} f(x).$$

**Definition 4.1 (Locally integrable).** Given  $f: \mathbb{R}^d \to \mathbb{C}$  be Lebesgue measurable function. Then we say f is locally integrable if

$$\int_K |f| \ \mathrm{d} m < \infty, \forall \text{compact } K \subset \mathbb{R}^d.$$

We write  $f \in L^1_{loc}(\mathbb{R}^d)$ .

**Definition 4.2 (Hardy-Littlewood maximal function).** Given  $f \in L^1_{loc}(\mathbb{R}^d)$ , the *Hardy-Littlewood maximal function* for f is defined as

$$\mathrm{Hf}(x) \coloneqq \sup \left\{ \mathrm{A}_r(x) \mid r > 0 \right\},\,$$

where

$$A_r(x) := \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(y)| \, dy.$$

**Note.** We note that  $A_r(\cdot)$  means averaging function.

**Lemma 4.2.** Let  $f \in L^1_{loc}(\mathbb{R}^d)$ , then

- 1.  $A_r(x)$  is jointly continuous for  $(x,r) \in \mathbb{R}^d \times (0,\infty)$ .
- 2. Hf(x) is Borel measurable.

*Proof.* We outline the proof.

1. Let  $(x,r) \to (x^*,r^*) \Longrightarrow A_r(x) \to A_{r^*}(x^*)$ . Let  $(x_n,r_n)$  be any sequence which converges to  $x^*,r^*$ , then we consider  $\lim_{n\to\infty} A_{r_n}(x_n)$  and we can calculate

$$\int \underbrace{|f(y)| \, \mathbb{1}_{B(x_n, r_n)}(y)}_{:=h_n(y)},$$

then we apply Theorem 2.4 to  $h_n$ 

2. Observe that

$$(\mathrm{Hf})^{-1}(\underbrace{(a,\infty)}_{\mathrm{open}}) = \bigcup_{r>0} \mathrm{A}_r^{-1}((a,\infty))$$

is open, since  $A_r^{-1}((a,\infty))$  is open from the 1. Note that the equality comes from the fact that  $Hf = \sup_r A_r$ .

Theorem 4.1 (Hardy-Littlewood maximal inequality). There exists  $C_d > 0$  such that for every  $f \in L^1(\mathbb{R}^d)$ ,

$$\underset{\alpha>0}{\forall} m\left(\left\{x\in\mathbb{R}^d\mid \mathrm{Hf}(x)>\alpha\right\}\right)\leq \frac{C_d}{\alpha}\int |f(x)|\;\mathrm{d}x.$$

*Proof.* We first fix  $f \in L^1$  and  $\alpha > 0$ . We define

$$E \coloneqq \{x \mid \mathrm{Hf}(x) > \alpha\},\,$$

which is a Borel measurable set by Lemma 4.2.then

$$x \in E \implies \exists_{r_x > 0} A_{r_x}(x) > \alpha \implies m(B(x, r_x)) < \frac{1}{\alpha} \int_{B(x, r_x)} |f(y)| dy.$$

From inner regularity, we have

$$m(E) = \sup \{ m(K) \mid \text{compact } K \subset E \}.$$

Let  $K \subset E$  be compact, then

$$K \subset \bigcup_{x \in K} B(x, r_x) \overset{K \text{ compact}}{\Longrightarrow} K \subset \bigcup_{i=1}^N B_i \overset{!}{\Longrightarrow} K \subset \bigcup_{i=1}^m \left\{ 3B_j' \right\}.$$

From here, we further have

$$m(K) \le \sum_{i=1}^{m} m(3B'_j) = 3^d \sum_{j=1}^{m} m(B'_j) \le \frac{3^d}{\alpha} \sum_{j=1}^{m} \int_{B'_j} |f(y)| \, dy.$$

Now, since  $B'_i, \ldots, B'_m$  are disjoint, hence we finally have

$$m(K) \le \frac{3^d}{\alpha} \int_{\mathbb{R}^d} |f(y)| \, \mathrm{d}y.$$

# 4.2 Lebesgue Differentiation Theorem

We start with a theorem!

Theorem 4.2 (Lebesgue Differentiation Theorem). Let  $f \in L^1$ , then

$$\lim_{r \to 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(y) - f(x)| \, dy = 0$$

for a.e. x.

Corollary 4.1. Theorem 4.2 also holds for  $f \in L^1_{loc}(\mathbb{R}^d)$ .

*Proof.* Using the fact that  $m^d$  is  $\sigma$ -finite, and apply Theorem 4.2.

Corollary 4.2. For  $f \in L^1_{loc}$ , we have

$$\lim_{r\to 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} f(y) \,\mathrm{d}y = f(x)$$

for a.e. x.

Proof. DIY

**Definition 4.3 (Shrink nicely).** We say that  $\{E_r\}_{r>0}$  shrinks nicely to x as  $r \to 0$  if  $E_r \subset B(x,r)$  and

$$\underset{c>0}{\exists} c \cdot m(B(x,r)) \le m(E_r).$$

# **Appendix**

## A Additional Proofs

#### A.1 Measure

This section gives all additional proofs in Section 1.

**Theorem A.1 (Theorem 1.2 3.).** Under the setup of Theorem 1.2,  $(X, \mathcal{A}, \mu)$  is a complete measure space.

*Proof.* We see this in two parts.

1. Claim: If  $A \subset X$  satisfies  $\mu^*(A) = 0$ , then A is Carathéodory measurable with respect to  $\mu^*$ .

*Proof.* If  $A \subset X$  and  $\mu^*(A) = 0$ , where  $\mu^*$  is an outer measure on X, we'll show that A is Carathéodory measurable with respect to  $\mu^*$ .

Equivalently, we want to show that for any  $E \subset X$ ,

$$\mu^*(E) = \mu^*(E \cap A) + \mu^*(E \setminus A).$$

Firstly, noting that  $(E \cap A) \subset A$ , and by monotonicity of  $\mu^*$ , we see that

$$\mu^*(E \cap A) \le \mu^*(A) = 0,$$

and since  $\mu^* \geq 0$ , hence  $\mu^*(E \cap A) = 0$ . Now, we only need to show that

$$\mu^*(E) = \mu^*(E \setminus A).$$

Since  $E \setminus A = E \cap A^c$ , and hence we have  $E \cap A^c \subset E$ , so

$$\mu^*(E) \ge \mu^*(E \setminus A).$$

To show another direction, we note that

$$\mu^*(E) \le \mu^*(E \cup A) = \mu^*((E \setminus A) \cup A) \le \mu^*(E \setminus A),$$

hence we conclude that A is Carathéodory measurable with respect to  $\mu^*$  if  $\mu^*(A) = 0$ .

2. Claim: If A is  $\mu$ -subnull, then  $A \in \mathcal{A}$ .

*Proof.* Let  $\mathcal{A}$  denotes the Carathéodory  $\sigma$ -algebra, and  $\mu := \mu^*|_{\mathcal{A}}$ . We want to show if A is  $\mu$ -subnull, then  $A \in \mathcal{A}$ .

Firstly, if A is  $\mu$ -subnull, then there exists a  $B \in \mathcal{A}$  such that  $A \subset B$  and  $\mu(B) = 0$ . But since from the monotonicity of  $\mu^*$ , we further have

$$0 = \mu(B) = \mu^*(B) \ge \mu^*(A),$$

hence  $\mu^*(A) = 0$ .

From the first claim, we immediately see that A is Carathéodory measurable with respect to  $\mu^*$ , which implies A is in Carathéodory  $\sigma$ -algebra, hence  $A \in \mathcal{A}$ .

We see that the second claim directly proves that  $(X, \mathcal{A}, \mu)$  is a complete measure space.

**Lemma A.1.** The function F defined in this example is a distribution function

Proof. We define

$$F_n(x) = \begin{cases} 1, & \text{if } x \ge r_n; \\ 0, & \text{if } x < r_n \end{cases}$$

where  $\{r_1, r_2, \ldots\} = \mathbb{Q}$ , and

$$F(x) = \sum_{n=1}^{\infty} \frac{F_n(x)}{2^n} = \sum_{n: r_n \le x} \frac{1}{2^n}$$

is both increasing and right-continuous.

• Increasing. Consider x < y. We see that

$$F(x) = \sum_{n; r_n \le x} \frac{1}{2^n} \le \sum_{n; r_n \le y} \frac{1}{2^n} = F(y)$$

clearly.26

• Right-continuous. We want to show  $F(x^+) = F(x)$ . Let  $x^+(\epsilon) := x + \epsilon$  with  $\epsilon > 0$ , we'll show that

$$\lim_{\epsilon \to 0} F(x^+(\epsilon)) = \lim_{\epsilon \to 0} F(x + \epsilon) = F(x).$$

Firstly, we have

$$F(x^{+}(\epsilon)) - F(x) = \sum_{n; r_n \le x + \epsilon} \frac{1}{2^n} - \sum_{n; r_n \le x} \frac{1}{2^n} = \sum_{n: x < r_n \le x + \epsilon^{27}} \frac{1}{2^n},$$

and we want to show

$$\lim_{\epsilon \to 0} F(x^+(\epsilon)) - F(x) = \lim_{\epsilon \to 0} \sum_{n; x < r_n \le x + \epsilon} \frac{1}{2^n} = 0.$$

Before we show how we choose  $\epsilon$ , <sup>28</sup> we see that

$$\sum_{n=k}^{\infty} \frac{1}{2^n} = 2^{1-k}.$$

<sup>&</sup>lt;sup>26</sup>This is trivial since we're always going to sum more strictly positive terms in F(y) than in F(x).

<sup>&</sup>lt;sup>27</sup>The strict is crucial to show the result, since if  $x = r_k$  for some fixed k, then we can't make the summation arbitrarily small.

<sup>&</sup>lt;sup>28</sup>To be precise, how  $\epsilon$  depends on  $r_n$ .

Now, we observe that

$$\sum_{n; x < r_n \le x + \epsilon} \frac{1}{2^n} \le \sum_{n = \arg\min_{k} x < r_k \le x + \epsilon}^{\infty} \frac{1}{2^n} = 2^{1-k}.$$

With this observation, it should be fairly easy to see that we can choose  $\epsilon$  based on how small we want to make  $2^{1-k}$  be,  $^{29}$  and we indeed see that

$$\lim_{k \to \infty} 2^{1-k} = 0,$$

which implies that F is right-continuous by squeeze theorem.

**Lemma A.2.** The function F defined in this example satisfies

- $\mu_F(\lbrace r_i \rbrace) > 0$  for all  $r_i \in \mathbb{Q}$ .
- $\mu_F(\mathbb{R} \setminus \mathbb{Q}) = 0$

given in this example.

*Proof.* We prove them one by one. And notice that F is indeed a distribution function as we proved in Lemma A.1.

1. To show  $\mu_F(\{r\}) > 0$  for every  $r \in \mathbb{Q}$ , we first note that  $\{r\} = \bigcap_{a-1 \le x < r} (x, r]$ . Then, we see that

$$\mu_F(\lbrace r \rbrace) = \mu_F \left( \bigcap_{a-1 \le x < a} (x, r] \right),$$

where each  $(x,r] \in \mathcal{A}$  and  $(x,r] \supset (y,r]$  whenever  $r-1 \le x \le y < r$ . Notice that we implicitly assign the order of the index by the order of x. Then, we see that  $\mu_F(r-1,r] < \infty$ .<sup>30</sup> Then, from continuity from above, we see that

$$\mu_F(\lbrace r \rbrace) = \lim_{i \to \infty} \mu_F((x_i, r]),$$

where we again implicitly assign an order to x as the usual order on  $\mathbb{R}$  by given index i. It's then clear that as  $i \to \infty$ ,  $x_i \to r$ . From the definition of F, we see that

$$F((x_i, r]) = F(r) - F(x_i) = \sum_{n; r_n \le r} \frac{1}{2^n} - \sum_{n; r_n \le x_i} \frac{1}{2^n} = \sum_{n; x_i < r_n \le r} \frac{1}{2^n}.$$

It's then clear that since  $r \in \mathbb{Q}$ , there exists an i' such that  $r_{i'} = r$ . Then, we immediately see that no matter how close  $x_i \to r$ , this sum is at least

$$\frac{1}{2^{i'}}$$

for a fixed i'. Hence, we conclude that  $\mu_F(\lbrace r \rbrace) > 0$  for every  $r \in \mathbb{Q}$ .

 $<sup>^{29} \</sup>mbox{We're}$  referring to  $\epsilon - \delta$  proof approach.

<sup>&</sup>lt;sup>30</sup>This will be  $\mu(A_1)$  in the condition of continuity from above. Furthermore, since  $\mathbb{Q}$  is countable, hence  $F(x) < \infty$  is promised.

2. Now, we show  $\mu_F(\mathbb{R} \setminus \mathbb{Q}) = 0$ . Firstly, we claim that

$$\mu_F(\mathbb{Q}) = 1$$

and

$$\mu_F(\mathbb{R}) = 1$$

as well. Since  $\mu_F(\mathbb{Q}) = 1$  is clear, we note that the second one essentially follows from the fact that we can write

$$\mathbb{R} = \lim_{N \to \infty} \bigcup_{i=1}^{N} (a - i, a + i]$$

for any  $a \in \mathbb{R}$ , say 0. From continuity from below, we have

$$\mu_F\left(\bigcup_{i=1}^{\infty} (-i, +i]\right) = \lim_{n \to \infty} \mu_F((-n, n]) = \sum_{n; r_n \in \mathbb{Q}} \frac{1}{2^n} = 1.$$

Given the above, from countable additivity of  $\mu_F$ , we have

$$\mu_F(\mathbb{R}\setminus\mathbb{Q}) + \underbrace{\mu_F(\mathbb{Q})}_{1} = \underbrace{\mu_F(\mathbb{R})}_{1} \implies \mu_F(\mathbb{R}\setminus\mathbb{Q}) = 0$$

as we desired.

Lemma A.3 (Cantor set has measure 0). Let C denotes the middle thirds Cantor set, then the Lebesgue measure of C is 0. i.e.,

$$m(C) = 0.$$

*Proof.* Since we're removing  $\frac{1}{3}$  of the whole interval at each n, we see that the measure of those removing parts, denoted by r, is

$$m(r) = \sum_{n=1}^{\infty} \frac{2^{n-1}}{3^n} = \frac{1}{2} \sum_{n=1}^{\infty} \left(\frac{2}{3}\right)^n = 1.$$

Then, by countable additivity of m, we see that

$$m(C) = m([0,1]) - m(r) = 1 - 1 = 0.$$

# A.2 Integration

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