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# Real and Functional Analysis

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These are unreviewed notes and may contain errors.**

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# Chapter 1

## Set Theory

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### 1.1 Basic notions

**Definition 1.1.1.** Let  $X, Y$  be sets. We say:

- $X, Y$  are **equipotent** if there exists a bijection  $f : X \rightarrow Y$ .
- $X$  has a **cardinality greater or equal** to  $Y$  if there exists an surjection  $f : X \rightarrow Y$ .
- $X$  is **finite** if it is equipotent to  $\{1, 2, \dots, n\}$  for some  $n \in \mathbb{N}$ .  $X$  is infinite otherwise.

**Remark:**  $X$  is infinite  $\iff$  it is equipotent to a proper subset of itself.

**E.g.:** The set of natural numbers  $\mathbb{N}$  is infinite. In fact, the set of even natural numbers  $E = \{2, 4, 6, \dots\} \subset \mathbb{N}$  is equipotent to  $\mathbb{N}$ , as we can define the bijection  $f : \mathbb{N} \rightarrow E$  as  $f(n) = 2n$ .

**Definition 1.1.2.** Let  $X$  be an infinite set. We say  $X$  is **countable** if it is equipotent to  $\mathbb{N}$ .  $X$  is **uncountable** otherwise, in which case it is **more than countable**.

**Definition 1.1.3.**  $X$  has the **cardinality of the continuum** if it is equipotent to  $[0, 1] \subset \mathbb{R}$ . Any such set is uncountable.

**E.g.:** We have that:

- $\mathbb{N}, \mathbb{Z}, \mathbb{Q}$  are countable.
- $\mathbb{R}, \mathbb{R}^n, (0, 1), [0, 1]$  are uncountable.
- Countable union of countable sets is countable.

## 1.2 Families of subsets

Let  $X$  be a set. The “Power set” of  $X$  is the set of all subsets of  $X$ , denoted by  $\mathcal{P}(X)$ .

$$\mathcal{P}(X) = \{E : E \subseteq X\}$$

Note that  $\mathcal{P}(X)$  has always a cardinality greater than  $X$ . For example, if  $X = \mathbb{N}$ , then  $\mathcal{P}(X)$  has the cardinality of the continuum.

**Definition 1.2.1.** Let  $X$  be a set. A **family of subsets** of  $X$  is a set  $E$  such that:

$$E \subseteq \mathcal{P}(X)$$

We usually denote  $E = \{E_i\}_{i \in I}$ , where  $I$  is an index set.

**Definition 1.2.2.** Let  $E = \{E_i\}_{i \in I}$  be a family of subsets of  $X$ . We define:

- The **union** of  $E$  as:

$$\bigcup_{i \in I} E_i = \{x \in X : x \in E_i \text{ for some } i \in I\}$$

- The **intersection** of  $E$  as:

$$\bigcap_{i \in I} E_i = \{x \in X : x \in E_i \text{ for all } i \in I\}$$

**Definition 1.2.3.** Let  $E = \{E_i\}_{i \in I}$  be a family of subsets of  $X$ . We say  $F$  is **pairwise disjoint** if:

$$E_i \cap E_j = \emptyset \quad \forall i, j \in I, i \neq j$$

**Definition 1.2.4.** We say that the family  $E = \{E_i\}_{i \in I}$  of subsets of  $X$  is a **covering** of  $X$  if:

$$X = \bigcup_{i \in I} E_i$$

Any subfamily of  $E$ ,  $E' = \{E_i\}_{i \in I'}$  is a **subcovering** of  $X$  if it is a covering of  $X$  itself.

**E.g.:** Let  $X = \mathbb{R}$ . We define:

$$\mathcal{T} = \{E \subset X : E \text{ is open}\}$$

We say that  $\mathcal{T}$  is the standard topology of  $X$ . More generally, this can be done in

“metric spaces”  $(X, d)$ .

**Properties of  $\mathcal{T}$  (open sets):**

- $\emptyset, X \in \mathcal{T}$ .
- Finite intersection of elements in  $\mathcal{T}$  is in  $\mathcal{T}$ .
- Arbitrary union of elements in  $\mathcal{T}$  is in  $\mathcal{T}$ .

We can also define **sequences of sets**. Let  $X$  be a set. A sequence of sets in  $X$  is a family of sets  $\{E_n\}_{n \in \mathbb{N}}$ .

**Definition 1.2.5.** Let  $X$  be a set. A sequence of sets  $\{E_n\}_{n \in \mathbb{N}}$  is said to be:

- **Increasing** if:

$$E_n \subseteq E_{n+1} \quad \forall n \in \mathbb{N}$$

It is denoted by  $\{E_n\} \uparrow$ .

- **Decreasing** if:

$$E_n \supseteq E_{n+1} \quad \forall n \in \mathbb{N}$$

It is denoted by  $\{E_n\} \downarrow$ .

Let now  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{P}(X)$  be a sequence of sets in  $X$ :

**Definition 1.2.6.** We define the following:

- The **limit superior** of  $\{E_n\}$  as:

$$\limsup_{n \rightarrow \infty} E_n = \bigcap_{n \in \mathbb{N}} \bigcup_{k \geq n} E_k$$

- The **limit inferior** of  $\{E_n\}$  as:

$$\liminf_{n \rightarrow \infty} E_n = \bigcup_{n \in \mathbb{N}} \bigcap_{k \geq n} E_k$$

- If the limit superior and limit inferior are equal, we say that

$$\lim_{n \rightarrow \infty} E_n = \limsup_{n \rightarrow \infty} E_n = \liminf_{n \rightarrow \infty} E_n$$

**Exercise:** Let  $X$  be a set and  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{P}(X)$  be a sequence of sets in  $X$ . Prove that:

$$(i) \quad \{E_n\} \uparrow \Rightarrow \lim_{n \rightarrow \infty} E_n = \bigcup_{n \in \mathbb{N}} E_n \quad (ii) \quad \{E_n\} \downarrow \Rightarrow \lim_{n \rightarrow \infty} E_n = \bigcap_{n \in \mathbb{N}} E_n$$

## 1.3 Characteristic functions

**Definition 1.3.1.** Let  $X$  be a set and  $E \subseteq X$ . The **characteristic function** of  $E$  is the function  $\mathbb{1}_E : X \rightarrow \{0, 1\}$  defined as:

$$\mathbb{1}_E(x) = \begin{cases} 1 & \text{if } x \in E \\ 0 & \text{if } x \notin E \end{cases}$$

This function is also called the **indicator function** of  $E$ .

**Remark:** Let  $E, F \subseteq X$ . We have that:

- $\mathbb{1}_{E \cap F} = \mathbb{1}_E \cdot \mathbb{1}_F$ .
- $\mathbb{1}_{E \cup F} = \mathbb{1}_E + \mathbb{1}_F - \mathbb{1}_{E \cap F}$ .
- $\mathbb{1}_{E^c} = 1 - \mathbb{1}_E$ .
- $\mathbb{1}_{\limsup_{n \rightarrow \infty} E_n} = \limsup_{n \rightarrow \infty} \mathbb{1}_{E_n}$ .
- $\mathbb{1}_{\liminf_{n \rightarrow \infty} E_n} = \liminf_{n \rightarrow \infty} \mathbb{1}_{E_n}$ .

## 1.4 Equivalence relations and Quotient sets

**Definition 1.4.1.** A relation  $R$  on a set  $X$  is a subset of  $X \times X$ . For any  $x, y \in X$ , we say that  $x$  is related to  $y$  if  $(x, y) \in R$ . We denote this as  $xRy$ .

**Definition 1.4.2.** A relation  $R$  on a set  $X$  is an **equivalence relation** if it satisfies:

- **Reflexivity:**

$$xRx \quad \forall x \in X$$

- **Symmetry:**

$$xRy \Rightarrow yRx \quad \forall x, y \in X$$

- **Transitivity:**

$$xRy, yRz \Rightarrow xRz \quad \forall x, y, z \in X$$

Every equivalence relation on  $X$  induces a partition of  $X$ . We define the **equivalence class** of  $x \in X$  as:

$$[x] = \{y \in X : xRy\}$$

The set of all equivalence classes is called the **quotient set** of  $X$  by  $R$ , denoted by  $X/R$ .

$$X/R = \{[x] : x \in X\}$$

**E.g.:** Let  $X = \mathbb{Z} \times \mathbb{Z}_0$  such that  $\mathbb{Z}_0 = \mathbb{Z} \setminus \{0\}$ . We define the relation  $R$  on  $X$  as:

$$(a, b)R(c, d) \iff ad = bc$$

We can prove that  $R$  is an equivalence relation. The equivalence classes are:

$$[(a, b)] = \{(c, d) \in X : ad = bc\}$$

Notice that:

$$[(a, b)] = \{(a, b), (2a, 2b), (3a, 3b), \dots\}$$

If we denote a class  $[(a, b)]$  as  $[a/b]$ , then we have that:

$$X/R = \{[a/b] : a, b \in \mathbb{Z}_0\} = \mathbb{Q}$$

## Chapter 2

# Measure Theory

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## 2.1 Measure spaces

**Definition 2.1.1.** Let  $X$  be a non-empty set. A family  $\mathcal{M} \subseteq \mathcal{P}(X)$  is a  **$\sigma$ -algebra** if:

- (i)  $\emptyset \in \mathcal{M}$ .
- (ii)  $E \in \mathcal{M} \implies E^c \in \mathcal{M}$ .
- (iii)  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M} \implies \bigcup_{n \in \mathbb{N}} E_n \in \mathcal{M}$ .

If instead of (iii) we have that  $E_1, E_2 \in \mathcal{M} \implies E_1 \cup E_2 \in \mathcal{M}$ , then  $\mathcal{M}$  is called an **algebra**.

**Remark:** If  $\mathcal{M}$  is a  $\sigma$ -algebra, then we say that  $(X, \mathcal{M})$  is a **measurable space**. Any set  $E \in \mathcal{M}$  is called a **measurable set**.

**E.g.:** Let  $X \neq \emptyset$ . Then:

- $\mathcal{P}(X)$  is a  $\sigma$ -algebra.
- $\{\emptyset, X\}$  is a  $\sigma$ -algebra.
- $\{\emptyset, E, E^c, X\}$  is a  $\sigma$ -algebra for any  $E \subseteq X$ .
- $X = \mathbb{R}$ ,  $\mathcal{M} = \{E \subseteq \mathbb{R} : E \text{ is open}\}$  is NOT a  $\sigma$ -algebra.



**Properties 2.1.1.** Let  $(X, \mathcal{M})$  be a measurable space. Then:

- (i)  $X = \emptyset^c \in \mathcal{M}$
- (ii)  $\mathcal{M}$  is also an algebra. Indeed, if  $\{E_1, E_2\} \subseteq \mathcal{M}$ ,  $E_n = \emptyset \forall n \geq 3$ , then  $E_1 \cup E_2 = \bigcup_{n \in \mathbb{N}} E_n \in \mathcal{M}$ .
- (iii)  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M} \implies \bigcap_n E_n \in \mathcal{M}$ .
- (iv)  $E, F \in \mathcal{M} \implies E \setminus F \in \mathcal{M}$
- (v)  $\Omega \subseteq X$ . Then, the **restriction** of  $\mathcal{M}$  to  $\Omega$  is:

$$\mathcal{M}|_{\Omega} := \{F \subseteq \Omega : F = E \cap \Omega, E \in \mathcal{M}\}$$

Then,  $(\Omega, \mathcal{M}|_{\Omega})$  is a measurable space.

## 2.2 Generation of a $\sigma$ -algebra

**Theorem 2.2.1.** Take any family  $\mathcal{A} \subseteq \mathcal{P}(X)$ . Then, it is well-defined the  $\sigma$ -algebra generated by  $\mathcal{A}$ , denoted by  $\sigma_0(\mathcal{A})$ , as the smallest  $\sigma$ -algebra containing  $\mathcal{A}$ . It is characterized by:

- (i)  $\sigma_0(\mathcal{A})$  is a  $\sigma$ -algebra.
- (ii)  $\mathcal{A} \subseteq \sigma_0(\mathcal{A})$ .
- (iii) If  $\mathcal{M}$  is a  $\sigma$ -algebra and  $\mathcal{A} \subseteq \mathcal{M}$ , then  $\sigma_0(\mathcal{A}) \subseteq \mathcal{M}$ .

*Sketch of proof.* Define  $V = \{\mathcal{M} \subseteq \mathcal{P}(X) : \mathcal{M} \text{ is } \sigma\text{-algebra, } \mathcal{A} \subseteq \mathcal{M}\}$ . Notice that  $V \neq \emptyset$  because  $\mathcal{P}(X) \in V$ . Then, define:

$$\sigma_0(\mathcal{A}) = \bigcap_{\mathcal{M} \in V} \mathcal{M}$$

Then,  $\sigma_0(\mathcal{A})$  is a  $\sigma$ -algebra as it satisfies the properties of a  $\sigma$ -algebra, denoted in definition 2.1.1. ■

**Remark:** This is relevant. Often, to check that a  $\sigma$ -algebra has certain properties, it is enough to check the property on a set of generators.

## 2.3 Borel sets

Take  $(X, d)$  as a metric space, so that open sets are defined. Recall that:

$$\mathcal{T} = \{E \subseteq X : E \text{ is open}\}$$

**Definition 2.3.1.** The  $\sigma$ -algebra generated by  $\mathcal{T}$  is called the **Borel  $\sigma$ -algebra** of  $X$ , denoted by:

$$\mathcal{B}(X) := \sigma_0(\mathcal{T})$$

Any set  $E \in \mathcal{B}(X)$  is a **Borel set**.

**Remark:** The following are Borel sets:

- Open sets
- Closed sets
- Countable intersections of open sets ( $G_\delta$ -sets)
- Countable unions of closed sets ( $F_\sigma$ -sets)

We will deal with:

$$X = \mathbb{R} = (-\infty, \infty)$$

but also:

$$X = \overline{\mathbb{R}} = [-\infty, \infty] = \mathbb{R} \cup \{-\infty, \infty\}$$

Let us define the arithmetic operations on  $\overline{\mathbb{R}}$ . Let  $a \in \mathbb{R}$ :

- $a \pm \infty = \pm\infty$
- $a > 0 : a \cdot \pm\infty = \pm\infty$
- $a < 0 : a \cdot \pm\infty = \mp\infty$
- $a = 0 : 0 \cdot \pm\infty = 0$
- $\infty - \infty, \infty/\infty, 0/0$  are not defined.

Also, the open intervals in  $\overline{\mathbb{R}}$  are the following:

- $(a, b)$ , with  $a, b \in \mathbb{R}$
- $[-\infty, b)$
- $(a, \infty]$

**Remark:** We have that:

$$\begin{aligned}\mathcal{B}(\mathbb{R}) &:= \sigma_0(\{\text{open sets}\}) \\ &= \sigma_0(\{(a, b) : a < b\}) \\ &= \sigma_0(\{[a, b] : a < b\}) \\ &= \sigma_0(\{(a, \infty) : a \in \mathbb{R}\})\end{aligned}$$

$$\begin{aligned}\mathcal{B}(\overline{\mathbb{R}}) &:= \sigma_0(\{\text{open sets}\}) \\ &= \sigma_0(\{(a, \infty] : a \in \mathbb{R}\})\end{aligned}$$

$$\mathcal{B}(\mathbb{R}^N) = \sigma_0(\{\text{open rectangles}\})$$

## 2.4 Measures

Let  $(X, \mathcal{M})$  be a measurable space.

**Definition 2.4.1.** A function  $\mu : \mathcal{M} \rightarrow [0, \infty]$  is a (positive) **measure** on  $\mathcal{M}$  if:

- (i)  $\mu(\emptyset) = 0$
- (ii)  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M}, \text{ disjoint} \implies \mu(\bigcup_{n \in \mathbb{N}} E_n) = \sum_{n \in \mathbb{N}} \mu(E_n)$

**Note:** To avoid nonsenses, we always assume that  $\exists E \in \mathcal{M} \text{ s.t. } \mu(E) < \infty$

**Terminology:** Let  $X, \mathcal{M}, \mu$  defined as above:

- $(X, \mathcal{M}, \mu)$  is a **measure space**.
- If  $\mu(X) = 1$ , then  $(X, \mathcal{M}, \mu)$  is a **probability space** and  $\mu$  is a **probability measure**.

**Definition 2.4.2.** A measure  $\mu$  is:

1. **Finite** if  $\mu(X) < \infty$
2.  **$\sigma$ -finite** if  $\exists \{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M} \text{ s.t.}$

$$\mu(E_n) < \infty \quad \forall n \in \mathbb{N} \quad \wedge \quad X = \bigcup_{n \in \mathbb{N}} E_n$$

**E.g.:** Some examples of measures are:

1. (Trivial measure): For any  $(X, \mathcal{M})$ , define  $\mu$  as  $\mu(E) = 0 \quad \forall E \in \mathcal{M}$
2. (Counting measure): For any  $(X, \mathcal{M})$ , typically  $\mathcal{M} = \mathcal{P}(X)$ , define  $\mu_{\#}$  as:

$$\mu_{\#}(E) = \begin{cases} \#\{\text{elements of } E\} & E \text{ finite} \\ \infty & \text{otherwise} \end{cases}$$

3. (Dirac measure): For any  $(X, \mathcal{M})$ , pick  $x_0 \in X$ . Then, define  $\delta_{x_0}$  as:

$$\delta_{x_0}(E) = \begin{cases} 1 & x_0 \in E \\ 0 & x_0 \notin E \end{cases}$$

### 2.4.1 Properties of measures

**Theorem 2.4.1** (Basic properties). *Let  $(X, \mathcal{M}, \mu)$  be a measure space. Then:*

- (i)  $\mu$  is finitely additive:  $E, F \in \mathcal{M}, E \cap F = \emptyset \implies \mu(E \cup F) = \mu(E) + \mu(F)$
- (ii) (Monotonicity):  $E, F \in \mathcal{M}, E \subseteq F \implies \mu(E) \leq \mu(F)$
- (iii) (Excision property):  $E, F \in \mathcal{M}, E \subseteq F, \mu(E) < \infty \implies \mu(F \setminus E) = \mu(F) - \mu(E)$

*Proof.* The proof is straightforward:

- (i) Let  $E, F \in \mathcal{M}, E \cap F = \emptyset$ . Then:

$$\mu(E \cup F) = \mu(E) + \mu(F)$$

*Proof.* Obvious, using  $E_n = \emptyset$  for  $n \geq 3$ . ■

- (ii) Let  $E, F \in \mathcal{M}, E \subseteq F$ . Then:

$$\mu(E) \leq \mu(F)$$

*Proof.* Let  $F = E \cup (F \setminus E)$ . Then:

$$\mu(F) = \mu(E) + \mu(F \setminus E) \geq \mu(E)$$
■

- (iii) Let  $E, F \in \mathcal{M}, E \subseteq F, \mu(E) < \infty$ . Then:

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

*Proof.* Let  $F = E \cup (F \setminus E)$ . Then:

$$\mu(F) = \mu(E) + \mu(F \setminus E)$$

This concludes the proof. ■

**Theorem 2.4.2** (Continuity among monotone sequences). *Let  $(X, \mathcal{M}, \mu)$  be a measure space. Let  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M}$  be a sequence of measurable sets. Then:*

(i) *If  $\{E_n\} \uparrow$ ,  $E := \lim_n E_n = \bigcup_n E_n$ , then:*

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

(ii) *If  $\{E_n\} \downarrow$ ,  $E := \lim_n E_n = \bigcap_n E_n$ , and  $\mu(E_1) < \infty$ , then:*

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

*Proof.* The proof goes as follows:

(i) If  $\mu(E_n) = \infty$  for some  $n$ , then the proof is trivial. Otherwise, let  $F_1 = E_1$  and  $F_n = E_n \setminus E_{n-1}$  for  $n \geq 2$ . Then, we can check that:

- $F_n \in \mathcal{M}, \forall n \in \mathbb{N}$
- $\{F_n\}$  is a disjoint sequence.
- $E_n = \bigcup_{k=1}^n F_k$
- Because of the disjointness, we have that:

$$\mu(E_n) = \sum_{k=1}^n \mu(F_k)$$

Then, we have that:

$$\begin{aligned} \mu(E) &= \mu\left(\lim_n E_n\right) = \mu\left(\bigcup_{n=1}^{\infty} E_n\right) = \\ &= \mu\left(\bigcup_{n=1}^{\infty} F_n\right) = \sum_{n=1}^{\infty} \mu(F_n) = \\ &= \sum_{n=1}^{\infty} (\mu(E_n) - \mu(E_{n-1})) = \lim_n \mu(E_n) \end{aligned}$$

(ii) Define  $G_n = E_1 \setminus E_n$ . Then, check that:

- $G_n \in \mathcal{M}, \forall n \in \mathbb{N}$
- $\{G_n\} \uparrow$

By the previous result, we have that:

$$\mu \left( \bigcup_{n=1}^{\infty} G_n \right) = \lim_n \mu(G_n)$$

Then, on the right-hand side:

$$\begin{aligned} \lim_n \mu(G_n) &= \lim_n \mu(E_1 \setminus E_n) = \\ &= \mu(E_1) - \lim_n \mu(E_n) \end{aligned}$$

On the left-hand side:

$$\begin{aligned} \mu \left( \bigcup_{n=1}^{\infty} G_n \right) &= \mu \left( \bigcup_{n=1}^{\infty} (E_1 \setminus E_n) \right) = \\ &= \mu \left( E_1 \setminus \bigcap_{n=1}^{\infty} E_n \right) = \\ &= \mu(E_1) - \mu(E) \end{aligned}$$

Finally, we have that:

$$\mu(E_1) - \mu(E) = \mu(E_1) - \lim_n \mu(E_n)$$

And because  $\mu(E_1) < \infty$ , we have that:

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

■

**Remark:** In (ii), the condition  $\mu(E_1) < \infty$  is essential. Consider  $(\mathbb{N}, \mathcal{P}(\mathbb{N}), \mu_{\#})$ . Then, define the following sequence:

$$E_n = \{n, n+1, n+2, \dots\}$$

Note that  $E_n \subseteq E_{n-1}$ . Also, note that for any  $n \in \mathbb{N}$ , we have that:

$$\mu_{\#}(E_n) = \infty$$

Then, we have that:

$$\mu_{\#} \left( \bigcap_n E_n \right) = \mu_{\#}(\emptyset) = 0$$

But:

$$\lim_n \mu_{\#}(E_n) = \infty$$

This shows that the condition  $\mu(E_1) < \infty$  is essential.

**Theorem 2.4.3** ( $\sigma$ -subadditivity). *Let  $(X, \mathcal{M}, \mu)$  be a measure space. Let  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M}$  be a sequence of measurable sets. Then:*

$$\mu \left( \bigcup_n E_n \right) \leq \sum_n \mu(E_n)$$

*Proof.* Let  $F_1 = E_1$  and  $F_n = E_n \setminus (\bigcup_{k=1}^{n-1} E_k)$  for  $n \geq 2$ . Then, we have that:

- $F_n \in \mathcal{M}, \forall n \in \mathbb{N}$
- $F_n \subseteq E_n, \forall n \in \mathbb{N}$
- $\{F_n\}$  is a disjoint sequence.
- $\bigcup_n E_n = \bigcup_n F_n$

Then, we have that:

$$\begin{aligned} \mu \left( \bigcup_n E_n \right) &= \mu \left( \bigcup_n F_n \right) = \\ &= \sum_n \mu(F_n) \leq \sum_n \mu(E_n) \end{aligned}$$

■

## 2.5 Sets of measure zero, negligible sets, complete measures

**Definition 2.5.1.** Let  $(X, \mathcal{M}, \mu)$  be a measure space. Then:

1. A set  $E \in \mathcal{M}$  is a **set of measure zero** if  $\mu(E) = 0$ .
2. A set  $F \in X$  (not necessarily measurable) is a **negligible set** if  $\exists E \in \mathcal{M}$  s.t.  $F \subseteq E$  and  $E$  is a set of measure zero.

**Definition 2.5.2.** Let  $(X, \mathcal{M}, \mu)$  be a measure space. Then, we say that  $\mu$  is a **complete measure** (alternatively, that  $(X, \mathcal{M}, \mu)$  is a **complete measure space**) all negligible sets are measurable.

**Remark** (Completion of a measure space): A measure space  $(X, \mathcal{M}, \mu)$  may not be complete. However, we can define the following:

$$\overline{\mathcal{M}} = \{E \subseteq X : \exists F_1, F_2 \in \mathcal{M} : F_1 \subseteq E \subseteq F_2, \mu(F_2 \setminus F_1) = 0\}$$

One can show that  $\overline{\mathcal{M}}$  is a  $\sigma$ -algebra, and that  $\mathcal{M} \subseteq \overline{\mathcal{M}}$ . Moreover, if  $E, F_1, F_2$  are as above, define:

$$\overline{\mu}(E) = \mu(F_1) = \mu(F_2)$$

One can check that  $(X, \overline{\mathcal{M}}, \overline{\mu})$  is a complete measure space.

## 2.6 Towards the Lebesgue measure

We would like to define a measure  $\lambda$  with  $X = \mathbb{R}$  (or  $X = \mathbb{R}^N$ ) s.t.  $\forall a < b$ :

- $\lambda((a, b)) = b - a$  (**length of the interval**)
- $\forall E, \lambda(E + x) = \lambda(E)$  (**translation invariance**)

In principle, we would like to define it in  $\mathcal{P}(\mathbb{R})$ . Such a measure should satisfy  $\lambda(\{a\}) = 0$ .

**Theorem 2.6.1** (Ulam). *The only measure on  $\mathcal{P}(\mathbb{R})$  that satisfies  $\lambda(\{a\}) = 0 \forall a \in \mathbb{R}$  is the trivial measure.*

Therefore, we need to choose an  $\mathcal{M} \subsetneq \mathcal{P}(\mathbb{R})$ . We can construct one as follows:

- Starting family with a “measure”, e.g.,  $\mathcal{T} = \{(a, b) : a < b\}$  and  $f((a, b)) = b - a$ .
- Construct an “outer measure”  $\mu^*$  on  $\mathcal{P}(\mathbb{R})$ .
- Restrict  $\mu^*$  to a well-chosen  $\sigma$ -algebra  $\mathcal{M} \subsetneq \mathcal{P}(\mathbb{R})$ .



**Definition 2.6.1.** Let  $X$  be a set. An **outer measure**  $\mu^*$  on  $X$  is a function

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

such that:

1.  $\mu^*(\emptyset) = 0$
2. (Monotonicity)  $E \subseteq F \subseteq X \implies \mu^*(E) \leq \mu^*(F)$
3. ( $\sigma$ -subadditivity)  $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{P}(X) \implies \mu^*\left(\bigcup_{n \in \mathbb{N}} E_n\right) \leq \sum_{n \in \mathbb{N}} \mu^*(E_n)$

**Remark:** Any measure  $\mu$  is an outer measure. However, the converse is not true.

**Proposition 2.6.2.** Let  $\mathcal{E} \subseteq \mathcal{P}(X)$ ,  $f : \mathcal{E} \rightarrow [0, \infty]$ . Assume that  $\emptyset, X \in \mathcal{E}$ ,  $f(\emptyset) = 0$ . Then,  $\forall E \subseteq X$  define:

$$\mu^*(E) = \inf \left\{ \sum_{n=1}^{\infty} f(A_n) : A_n \in \mathcal{E}, E \subseteq \bigcup_n A_n \right\}$$

Then,  $\mu^*$  is an outer measure.

*Proof.* The proof is omitted. ■

**Remark:** In this generality, if  $E \in \mathcal{E}$ , then  $f(E)$  and  $\mu^*(E)$  may not be equal. We can only guarantee that  $\mu^*(E) \leq f(E)$ .

**E.g.:** There are some important examples:

- $X = \mathbb{R}$ ,  $\mathcal{E} = \{(a, b) : a, b \in \overline{\mathbb{R}}, a \leq b\}$

$$f((a, b)) = \text{length}((a, b)) = b - a$$

- $X = \mathbb{R}^N$ ,  $\mathcal{E} = \{(a_1, b_1) \times \dots \times (a_N, b_N) : a_i, b_i \in \overline{\mathbb{R}}, a_i \leq b_i\}$

$$f(\underline{a}, \underline{b}) = \text{volume}(\underline{a}, \underline{b}) = \prod_{i=1}^N (b_i - a_i)$$

In both cases, the outer measure  $\mu^*$  is called the **Lebesgue outer measure**. We will denote it by  $\lambda^*$  (or  $\lambda_N^*$  in the second case). Note that in this case,  $\lambda^*(E) = f(E)$  for any  $E \in \mathcal{E}$ .

**Remark:** Any  $\mu$  measure on  $\mathcal{P}(X)$  is an outer measure. However, the converse is not true. In particular,  $\exists A, B \subseteq \mathbb{R}$  s.t.  $A \cap B = \emptyset$  and  $\lambda^*(A \cup B) < \lambda^*(A) + \lambda^*(B)$ .

### 2.6.1 Carathéodory's criterion

**Definition 2.6.2** (Carathéodory's condition). Let  $\mu^*$  be an outer measure on  $\mathcal{P}(X)$ . A set  $E \subseteq X$  is  $\mu^*$ -**measurable** if  $\forall A \subseteq X$ :

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

**Lemma 2.6.3** (Equivalence of Carathéodory's condition).  $E$  is  $\mu^*$ -measurable  $\iff \forall A \subseteq X, \mu^*(A) < \infty$ :

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

*Proof.* The proof is as follows:

$(\Rightarrow)$  : Trivial

$(\Leftarrow)$  : Let  $A \subseteq X$ , such that  $\mu^*(A) < \infty$  and:

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

Then, notice that  $\{A \cap E, A \cap E^c\}$  is a covering of  $A$ . By subadditivity:

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

Therefore, we have that:

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

Meaning that  $E$  is  $\mu^*$ -measurable. This concludes the proof. ■

**Theorem 2.6.4** (Carathéodory). *Let  $\mu^*$  be an outer measure on  $\mathcal{P}(X)$ . The family:*

$$\mathcal{M} = \{E \subseteq X : E \text{ is } \mu^*\text{-measurable}\}$$

*is a  $\sigma$ -algebra, and  $\mu^*$  restricted to  $\mathcal{M}$  (denoted  $\mu = \mu^*|_{\mathcal{M}}$ ) is a complete measure.*

**Remark:**  $(X, \mathcal{M}, \mu)$  as in the above theorem is sometimes called the “abstract Lebesgue measure space”. We will only prove the completeness of  $\mu$ .

**Lemma 2.6.5.** *Let  $(X, \mathcal{M}, \mu)$  be the measure space as in Carathéodory’s theorem. Then, any  $N \subseteq X$  s.t.  $\mu^*(N) = 0$  is  $\mu$ -measurable, i.e.,  $N \in \mathcal{M}$ , and  $\mu(N) = 0$ .*

*Proof.* We have to show that  $N$  satisfies Carathéodory’s condition, or equivalently, that it satisfies the lemma 2.6.3. Let  $A \subseteq X$  be arbitrary. Then, notice that:

$$\mu^*(A \cap N) \leq \mu^*(N) = 0$$

Also, notice that:

$$\mu^*(A \cap N^c) \leq \mu^*(A)$$

Therefore, we have that:

$$\mu^*(A \cap N) + \mu^*(A \cap N^c) \leq 0 + \mu^*(A) = \mu^*(A)$$

By lemma 2.6.3, we have that  $N$  is  $\mu^*$ -measurable. By Carathéodory’s theorem, we have that  $N$  is  $\mu$ -measurable. Finally, we have that  $\mu(N) = \mu^*(N) = 0$ . ■

**Corollary 2.6.5.1.**  *$\mu$  as in Carathéodory’s theorem is a complete measure.*

*Proof.* Let  $N \subseteq E$ , and  $\mu(E) = 0$  ( $E \in \mathcal{M}$ ). Then, we have that:

$$\mu(E) = 0 \implies \mu^*(E) = 0$$

By monotonicity, we have that:

$$\mu^*(N) \leq \mu^*(E) = 0$$

Then,  $\mu(N) = \mu^*(N) = 0$ , thus  $N \in \mathcal{M}$ . This concludes the proof. ■

## 2.7 Lebesgue measure

**Definition 2.7.1.** Let  $\mathcal{E} = \{(a, b) : a, b \in \overline{\mathbb{R}}, a \leq b\}$ . Define:

$$\lambda^*((a, b)) = b - a$$

Then,  $\lambda^*$  is the **Lebesgue outer measure** on  $\mathbb{R}$ .

**Theorem 2.7.1.** Let  $\lambda^*$  be the Lebesgue outer measure on  $\mathcal{E} = \{(a, b) : a, b \in \overline{\mathbb{R}}, a \leq b\}$ . Then, by Carathéodory's theorem, the family:

$$\mathcal{L}(\mathbb{R}) = \{E \subseteq \mathbb{R} : E \text{ is } \lambda^*\text{-measurable}\}$$

is a  $\sigma$ -algebra, called the **Lebesgue  $\sigma$ -algebra**, and  $\lambda^*$  restricted to  $\mathcal{L}(\mathbb{R})$  (denoted  $\lambda = \lambda^*|_{\mathcal{L}(\mathbb{R})}$ ) is a complete measure, called the **Lebesgue measure**.

*Proof.* The proof is omitted. ■

**Remark:** The measure space  $(\mathbb{R}, \mathcal{L}(\mathbb{R}), \lambda)$  is called the **Lebesgue measure space**.

**Proposition 2.7.2.** Let  $\lambda$  be the Lebesgue measure on  $\mathbb{R}$ . Then:

- (i)  $a \in \mathbb{R} \implies \{a\} \in \mathcal{L}(\mathbb{R})$  and  $\lambda(\{a\}) = 0$
- (ii)  $E \subset \mathbb{R}$  at most countable  $\implies E \in \mathcal{L}(\mathbb{R})$  and  $\lambda(E) = 0$

*Proof.* The proof is as follows:

- (i) Let  $a \in \mathbb{R}$ . Then, we have that, for any  $\varepsilon > 0$ :

$$E_1 = (a - \varepsilon, a + \varepsilon), \quad E_2 = E_3 = \dots = \emptyset$$

is a covering of  $\{a\}$ . Then, by definition of  $\lambda^*$ :

$$0 \leq \lambda^*(\{a\}) \leq \sum_{n=1}^{\infty} \lambda(E_n) = 2\varepsilon$$

As  $\varepsilon$  is arbitrary, we have that  $\lambda^*(\{a\}) = 0$ . By Lemma 2.6.5, we then have that  $\{a\} \in \mathcal{L}(\mathbb{R})$  and  $\lambda(\{a\}) = 0$ .

(ii) Let  $E \subseteq \mathbb{R}$  be at most countable. Then, we have that:

$$E = \bigcup_{a \in E} \{a\}$$

Because  $\{a\} \in \mathcal{L}(\mathbb{R})$  and  $\lambda(\{a\}) = 0$ , we have that  $E \in \mathcal{L}(\mathbb{R})$  and:

$$\lambda(E) = \lambda\left(\bigcup_{a \in E} \{a\}\right) = \sum_{a \in E} \lambda(\{a\}) = 0$$

■

**Remark:** We can point out 2 important consequences of the above proposition:

1. Arguing as in (i), we can show that the Lebesgue measure is translation invariant. That is,  $\forall E \in \mathcal{L}(\mathbb{R}), \forall x \in \mathbb{R}$ :

$$\lambda(E + x) = \lambda(E)$$

2. In particular, since  $\mathbb{Q}$  is countable, we have that  $\mathbb{Q} \in \mathcal{L}(\mathbb{R})$  and  $\lambda(\mathbb{Q}) = 0$ . In the measure sense,  $\mathbb{Q}$  has very few elements with respect to  $\mathbb{R}$ . On the other hand,  $\mathbb{Q}$  is dense in  $\mathbb{R}$ . In the topology sense,  $\mathbb{Q}$  has a lot of points.

**Proposition 2.7.3.** *We have that:  $\mathcal{B}(\mathbb{R}) \subset \mathcal{L}(\mathbb{R})$*

*Proof.* Since  $\mathcal{B}(\mathbb{R}) = \sigma_0(\{(a, \infty) : a \in \mathbb{R}\})$ , if we show that  $(a, \infty) \in \mathcal{L}(\mathbb{R}), \forall a \in \mathbb{R}$ , then the prop. follows.

Take  $A \subset \mathbb{R}$ , s.t.  $\lambda^*(A) < \infty$ . Then, we must show that:

$$\lambda^*(A) \geq \lambda^*(A \cap (a, \infty)) + \lambda^*(A \cap (-\infty, a])$$

Moreover, by a previous remark, one can assume that  $a \notin A$ . Then, take any countable covering of  $A$  by open intervals:

$$A \subseteq \bigcup_n I_n$$

Then, let us define  $A_{left} = A \cap (-\infty, a]$  and  $I_{n,left} = I_n \cap (-\infty, a]$ . Then, we notice that  $\{I_{n,left}\}$  is a covering of  $A_{left}$ .

In the same way, we define  $A_{right} = A \cap (a, \infty)$  and  $I_{n,right} = I_n \cap (a, \infty)$ . Then, we notice that  $\{I_{n,right}\}$  is a covering of  $A_{right}$ .

Then, we have that:

$$\lambda^*(A_{left}) \leq \sum_n \lambda^*(I_{n,left})$$

$$\lambda^*(A_{right}) \leq \sum_n \lambda^*(I_{n,right})$$

Summing both inequalities, we have that:

$$\begin{aligned} \lambda^*(A_{left}) + \lambda^*(A_{right}) &\leq \sum_n \lambda^*(I_{n,left}) + \sum_n \lambda^*(I_{n,right}) \\ &= \sum_n \lambda^*(I_n) \end{aligned}$$

Taking the infimum over all countable coverings of  $A$ , we have that:

$$\lambda^*(A_{left}) + \lambda^*(A_{right}) \leq \lambda^*(A)$$

■

**Remark:** In particular, we have that  $\forall (a, b) \subset \mathbb{R}$ :

$$(a, b) \in \mathcal{L}(\mathbb{R}), \quad \lambda((a, b)) = b - a$$

We know that:

$$\mathcal{B}(\mathbb{R}) \subset \mathcal{L}(\mathbb{R}) \subset \mathcal{P}(\mathbb{R})$$

Are these inclusions strict? We already know that  $\mathcal{L}(\mathbb{R}) \neq \mathcal{P}(\mathbb{R})$ , by Ulam's theorem. In particular,  $\exists E \subset \mathbb{R}$  not Lebesgue measurable (**Vitali sets**). More precisely:

$$\forall E \in \mathcal{L}(\mathbb{R}), \text{ s.t } \lambda(E) > 0, \exists V \subset E, \text{ s.t } V \notin \mathcal{L}(\mathbb{R})$$

The relation between  $\mathcal{B}(\mathbb{R})$  and  $\mathcal{L}(\mathbb{R})$  is more subtle. It is clarified by the following proposition:

**Proposition 2.7.4** (Regularity of the Lebesgue measure). *Let  $E \in \mathbb{R}$ . Then, the following are equivalent:*

(i)  $E \in \mathcal{B}(\mathbb{R})$

(ii)  $\forall \varepsilon > 0, \exists A \subset \mathbb{R}$  open set s.t.

$$E \subset A \quad \text{and} \quad \lambda^*(A \setminus E) < \varepsilon$$

(iii)  $\forall \varepsilon > 0, \exists G \subset \mathbb{R}$  of class  $G_\delta$  s.t.

$$E \subset G \quad \text{and} \quad \lambda^*(G \setminus E) = 0$$

(iv)  $\forall \varepsilon > 0, \exists C \subset \mathbb{R}$  closed set s.t.

$$C \subset E \quad \text{and} \quad \lambda^*(E \setminus C) < \varepsilon$$

(v)  $\forall \varepsilon > 0, \exists F \subset \mathbb{R}$  of class  $F_\sigma$  s.t.

$$F \subset E \quad \text{and} \quad \lambda^*(E \setminus F) = 0$$

We get as a consequence the following:

**Corollary 2.7.4.1.**  $\forall E \in \mathcal{L}(\mathbb{R}), \exists F, G \in \mathcal{B}(\mathbb{R})$  s.t.  $F \subset E \subset G$  and

$$\lambda(G \setminus F) = \lambda(G \setminus E) + \lambda(E \setminus F) = 0$$

In other words:

$$\mathcal{L}(\mathbb{R}) = \overline{\mathcal{B}(\mathbb{R})}$$

(But  $\mathcal{L}(\mathbb{R}) \neq \mathcal{B}(\mathbb{R})$ ).

*Proof. (Regularity of the Lebesgue measure).* The proof goes as follows:

(i)  $\Rightarrow$  (ii) :

Let  $E \in \mathcal{B}(\mathbb{R})$ . Note that, since  $A \in \mathcal{L}(\mathbb{R})$  for all  $A$  open, we have that:

$$\lambda^*(A \setminus E) = \lambda(A \setminus E)$$

By definition of  $\lambda^*$ , we have that  $\forall \varepsilon > 0, \exists \{I_n\}_{n \in \mathbb{N}}$  s.t.

$$E \subset \bigcup_n I_n \quad \text{and} \quad \sum_n \lambda(I_n) < \lambda^*(E) + \varepsilon$$

Then, set  $A = \bigcup_n I_n$ . We have that  $A$  is open,  $E \subset A$  and:

$$\begin{aligned} \lambda(A) &\leq \sum_n \lambda(I_n) < \lambda(E) + \varepsilon \\ \implies \lambda(A \setminus E) &= \lambda(A) - \lambda(E) < \varepsilon \end{aligned}$$

(ii)  $\Rightarrow$  (iii) :

Assume  $\forall \varepsilon > 0$ ,  $\exists A_\varepsilon$  open s.t.  $E \subset A_\varepsilon$  and  $\lambda(A_\varepsilon \setminus E) < \varepsilon$ . Then, set  $\varepsilon = 1/n$ ,  $n \geq 1$  (for ease of notation,  $A_n = A_{1/n}$ ) and define:

$$G = \bigcap_n A_n$$

Then,  $G$  is a  $G_\delta$  set,  $E \subset G$  and:

$$0 \leq \lambda^*(G \setminus E) \leq \lambda^*(A_n \setminus E) < 1/n$$

Taking the limit, we have that  $\lambda(G \setminus E) = 0$ .

(iii)  $\Rightarrow$  (i) :

We know that  $E \subset G$ ,  $G \in \mathcal{L}(\mathbb{R})$  with  $\lambda(G \setminus E) = 0$ . Then, we have that:

$$E = G \setminus (G \setminus E) \in \mathcal{L}(\mathbb{R})$$

since  $G \in \mathcal{L}(\mathbb{R})$  and  $G \setminus E \in \mathcal{L}(\mathbb{R})$ . The last is because it is a negligible set and  $\lambda$  is complete. ■

**E.g.** (Cantor set): Let  $T_0 = [0, 1]$ . Then, construct  $T_{n+1}$  from  $T_n$  (recursively) by removing the inner third part of every interval in  $T_n$ :

$$\begin{aligned} T_0 &= [0, 1], \\ T_1 &= [0, 1/3] \cup [2/3, 1], \\ T_2 &= [0, 1/9] \cup [2/9, 1/3] \cup [2/3, 7/9] \cup [8/9, 1], \dots \end{aligned}$$



Then, define the **Cantor set** as:

$$C = \bigcap_n T_n$$

It can be proven that:

- $C$  has the cardinality of  $\mathbb{R}$
- $\lambda(C) = 0$
- $C$  is compact
- $C$  is nowhere dense (has no interior points), i.e.,  $\text{int}(C) = \emptyset$
- $\exists E \subset C$  s.t.  $E \in \mathcal{L}(\mathbb{R})$  but  $E \notin \mathcal{B}(\mathbb{R})$

## 2.8 Measurable functions

**Definition 2.8.1.** Given  $f : X \rightarrow Y$ , it is well-defined the **preimage** (or counterimage) of  $f$  as:

$$f^{-1} : \mathcal{P}(Y) \rightarrow \mathcal{P}(X), \quad f^{-1}(E) = \{x \in X : f(x) \in E\}$$

**Remark:** Preimages work well with set operations:

- $f^{-1}(E \cup F) = f^{-1}(E) \cup f^{-1}(F)$
- $f^{-1}(E \cap F) = f^{-1}(E) \cap f^{-1}(F)$
- $f^{-1}(E \setminus F) = f^{-1}(E) \setminus f^{-1}(F)$

Note this is not true for images.

**Definition 2.8.2.** Let  $(X, \mathcal{M})$  and  $(Y, \mathcal{N})$  be measurable spaces. A function  $f : X \rightarrow Y$  is **measurable** if  $\forall E \in \mathcal{N}$ , we have that  $f^{-1}(E) \in \mathcal{M}$ . We also say that  $f$  is  **$(\mathcal{M}, \mathcal{N})$ -measurable**.

**Proposition 2.8.1.** Let  $(X, \mathcal{M})$  and  $(Y, \mathcal{N})$  be measurable spaces, and  $\rho \subset \mathcal{N}$  s.t.  $\mathcal{N} = \sigma_0(\rho)$ . Then,  $f : X \rightarrow Y$  is measurable  $\iff \forall E \in \rho$ , we have that  $f^{-1}(E) \in \mathcal{M}$ .

*Proof.* The proofs goes as follows:

$(\Rightarrow)$  : Trivial

$(\Leftarrow)$  : Define  $\Sigma := \{E \subset Y : f^{-1}(E) \in \mathcal{M}\}$ . We have:

- $\rho \subset \Sigma$  as a consequence of  $(\forall E \in \rho, f^{-1}(E) \in \mathcal{M})$

- $\Sigma$  is a  $\sigma$ -algebra (check as an exercise)

Then, we have that  $\mathcal{N} = \sigma_0(\rho) \subset \Sigma$ . Therefore,  $f$  is measurable. ■

**Definition 2.8.3.** Suppose that  $\mathcal{M} \supseteq \mathcal{B}(X)$  and  $\mathcal{N} = \mathcal{B}(Y)$ . We say that  $f : X \rightarrow Y$  is:

- **Borel measurable** if  $f$  is  $(\mathcal{B}(X), \mathcal{B}(Y))$ -measurable.
- **Lebesgue measurable** if it is  $(\mathcal{M}, \mathcal{B}(Y))$ -measurable.

**Remark:** If  $f : X \rightarrow Y$  is Borel measurable, then it is Lebesgue measurable. The converse is not true.

Also, we never deal with  $\mathcal{L}(Y)$ .

**Corollary 2.8.1.1.**  $f$  is Borel measurable  $\iff f^{-1}(E) \in \mathcal{B}(X), \forall E \in Y$  open.  
Also,  $f$  is Lebesgue measurable  $\iff f^{-1}(E) \in \mathcal{M}, \forall E \in Y$  open.

*Proof.* It follows from the previous proposition, since  $\mathcal{B}(Y) = \sigma_0(\{E \subset Y : E \text{ open}\})$ . ■

**Definition 2.8.4.** We say that  $f$  is **continuous**  $\iff f^{-1}(E) \subset X$  is open  $\forall E \subset Y$  open.

**Proposition 2.8.2.** If  $f : X \rightarrow Y$  is continuous, then  $f$  is Borel measurable (and thus Lebesgue measurable).

*Proof.* Let  $E \subset Y$  be open. By continuity of  $f$ , we have that  $f^{-1}(E)$  is open. Then  $f^{-1}(E) \in \mathcal{B}(X)$ , and thus  $f$  is Borel measurable.

Note that the proposition is false when  $\mathcal{N} \supsetneq \mathcal{B}(Y)$ . ■

### 2.8.1 Operations on measurable functions

**Proposition 2.8.3.** *Let  $f : X \rightarrow Y$  be Lebesgue measurable, and  $g : Y \rightarrow Z$  be continuous. Then:*

$$g \circ f : X \rightarrow Z \text{ is Lebesgue measurable}$$

**Corollary 2.8.3.1.** *Let  $f : X \rightarrow Y$  be Lebesgue measurable. Then:*

- $f^+(x) = \max\{f(x), 0\}$  is Lebesgue measurable
- $f^-(x) = \max\{-f(x), 0\}$  is Lebesgue measurable
- $|f(x)| = f^+(x) + f^-(x)$  is Lebesgue measurable

*Proof.* Let  $f$  be Lebesgue measurable, and  $g : \mathbb{R} \rightarrow \mathbb{R}$  be continuous. Then, take  $E \subset Z$  open. We have that:

$$(g \circ f)^{-1}(E) = f^{-1}(g^{-1}(E))$$

Since  $g$  is continuous,  $g^{-1}(E)$  is open. Then,  $f^{-1}(g^{-1}(E)) \in \mathcal{M}$  ■

**Proposition 2.8.4.** *Let  $f, g : X \rightarrow \mathbb{R}$  be Lebesgue measurable, and  $\Phi : \mathbb{R}^2 \rightarrow \mathbb{R}$  be continuous. Then,  $h(x) = \Phi(f(x), g(x))$  is Lebesgue measurable.*

*Proof.* Let  $h(x) = \Phi(f(x), g(x)) = (\Phi \circ \Psi)(x)$ , where  $\Psi : X \rightarrow \mathbb{R}^2$  is defined as

$$\Psi(x) = (f(x), g(x))$$

Then, we should prove that  $\Psi$  is Lebesgue measurable for applying the previous proposition. For this, we have to show that  $\forall (a, b) \times (c, d) \subset \mathbb{R}^2$ , we have that:

$$\Psi^{-1}((a, b) \times (c, d)) = \{x \in X : f(x) \in (a, b), g(x) \in (c, d)\} \in \mathcal{M}$$

This can be done using the fact that  $f$  and  $g$  are Lebesgue measurable. ■

**Corollary 2.8.4.1.** *Let  $f, g : X \rightarrow \mathbb{R}$  be Lebesgue measurable. Then:*

- $f + g$  is Lebesgue measurable
- $f \cdot g$  is Lebesgue measurable

**Proposition 2.8.5.** *Let  $(X, \mathcal{M})$  be a measurable space (with  $\mathcal{M} \supseteq \mathcal{B}(X)$ ), and  $\{f_n\}_{n \in \mathbb{N}}$  be a sequence of Lebesgue measurable functions  $f_n : X \rightarrow \mathbb{R}$ . Then, the following functions are Lebesgue measurable:*

1.  $\sup_n f_n$
2.  $\inf_n f_n$
3.  $\limsup_n f_n$
4.  $\liminf_n f_n$

*In particular, if  $\lim_n f_n$  exists, then it is Lebesgue measurable.*

*Proof.* The proof goes as follows:

1. Since  $\mathcal{B}(\mathbb{R}) = \sigma_0(\{(a, \infty) : a \in \mathbb{R}\})$ , it is enough to show that  $\forall a \in \mathbb{R}$ , we have that:

$$(\sup_n f_n)^{-1}((a, \infty)) = \{x \in X : \sup_n f_n(x) > a\} \in \mathcal{M}$$

This can be done by using the fact that  $f_n$  is Lebesgue measurable. Indeed, we have that:

$$\begin{aligned} \{x \in X : \sup_n f_n(x) > a\} &= \bigcup_n \{x \in X : f_n(x) > a\} \\ &= \bigcup_n f_n^{-1}((a, \infty)) \in \mathcal{M} \end{aligned}$$

because  $f_n^{-1}((a, \infty)) \in \mathcal{M}$  for all  $n$ .

2. The proof is analogous to the previous case, taking that:

$$\inf_n f_n = -\sup_n (-f_n)$$

3. We have that:

$$\limsup_n f_n = \inf_n \sup_{k \geq n} f_k$$

4. We have that:

$$\liminf_n f_n = \sup_n \inf_{k \geq n} f_k$$

■

## 2.8.2 Properties holding almost everywhere

**Definition 2.8.5.** Let  $(X, \mathcal{M}, \mu)$  be a complete measure space. We say that a property  $P(x)$  holds  $\mu$ -almost everywhere (a.e) if:

$$\mu(\{x \in X : P(x) \text{ is false}\}) = 0$$

In other words,  $P(x)$  holds  $\mu$ -almost everywhere if it holds everywhere except for a set of measure zero.

**E.g.:** Let  $f(x) = x^2$ . Is it true that  $f(x) > 0$  a.e.?

We have that  $\{x : x^2 \leq 0\} = \{0\}$

- In  $(\mathbb{R}, \mathcal{L}(\mathbb{R}), \lambda)$ , the property is true a.e., since  $\lambda(\{0\}) = 0$
- In  $(\mathbb{R}, \mathcal{P}(\mathbb{R}), \mu_{\#})$  (counting measure), the property is false a.e., since  $\mu_{\#}(\{0\}) = 1$

**Proposition 2.8.6.** Let  $(X, \mathcal{M}, \mu)$  be a measure space:

1.  $f : X \rightarrow \overline{\mathbb{R}}$  s.t.  $f = g$  a.e, with  $g$  measurable  $\implies f$  is measurable
2.  $\{f_n\}_{n \in \mathbb{N}}$  a sequence of measurable functions s.t.  $f_n \rightarrow f$  a.e., then  $f$  is measurable.

### 2.8.3 Simple functions

**Definition 2.8.6.** Let  $(X, \mathcal{M})$  be a measurable space. A function  $s : X \rightarrow \overline{\mathbb{R}}$  is measurable and **simple** if  $s$  is measurable and  $s(X)$  is a finite set:

$$s(X) = \{a_1, a_2, \dots, a_k\}$$

where  $a_i \in \overline{\mathbb{R}} \forall i$ , with  $a_i \neq a_j$  for  $i \neq j$ . Then,  $s$  can be written as:

$$s(x) = \sum_{i=1}^k a_i \cdot \chi_{A_i}(x)$$

where  $A_i = s^{-1}(\{a_i\})$ ,  $A_i \cap A_j = \emptyset$  for  $i \neq j$ ,  $\bigcup_{i=1}^k A_i = X$  and  $A_i \in \mathcal{M}$ ,  $\forall i$ .

#### Particular case:

If  $X = \mathbb{R}$  (or  $(a, b) \subset \mathbb{R}$ ) and  $A_i$  is an interval  $\forall i$ , then  $s$  is called a **step function**.

On the other hand,  $\chi_{\mathbb{Q}}$  is a simple function, but not a step function:

$$\chi_{\mathbb{Q}}(x) = \begin{cases} 1 & x \in \mathbb{Q} \\ 0 & x \notin \mathbb{Q} \end{cases}$$

**Remark:** One may define simple functions without measurability requirements.

#### Goal:

Approximate any measurable function  $f : X \rightarrow \overline{\mathbb{R}}$  with (measurable and) simple functions.

**Theorem 2.8.7** (Simple approximation theorem (SAT)). *Take  $(X, \mathcal{M})$  measurable space and  $f : X \rightarrow [0, \infty]$ , measurable. Then  $\exists \{s_n\}_{n \in \mathbb{N}}$  a sequence of measurable, simple functions s.t.  $s_1 \leq s_2 \leq \dots \leq f$  pointwise (i.e.,  $\forall x \in X$ ) and:*

$$\lim_{n \rightarrow \infty} s_n(x) = f(x) \quad \forall x \in X$$

*Moreover, if  $f$  is bounded, the convergence is uniform:*

$$\lim_{n \rightarrow \infty} \sup_{x \in X} |s_n(x) - f(x)| = 0$$

*Proof.* In case  $f$  is bounded, say  $0 \leq f < 1$ .

For any  $n \geq 1$ , divide  $[0, 1)$  into  $2^n$  intervals of length  $2^{-n}$ , and define:

$$A_n^{(i)} = \{x \in X : \frac{i}{2^n} \leq f(x) < \frac{i+1}{2^n}\}$$

and:

$$s_n(x) = \sum_{i=0}^{2^n-1} \frac{i}{2^n} \cdot \chi_{A_n^{(i)}}(x)$$

One can show that such a sequence has the required properties ■

## 2.9 Lebesgue integral

- First, we will define the integral for non-negative simple (measurable) functions.
- Second, we will extend the definition to non-negative measurable functions.
- Finally, we will define the integral for general measurable functions.

### 2.9.1 Integral of non-negative simple functions

**Definition 2.9.1.** Let  $s : X \rightarrow [0, \infty]$  be a measurable and simple function:

$$s = \sum_{i=1}^k a_i \cdot \chi_{A_i}$$

where  $a_i \geq 0$  and  $A_i \in \mathcal{M}$ . Let  $E \in \mathcal{M}$ . Then, we define the **(Lebesgue) integral** of  $s$  over  $E$  as:

$$\int_E s d\mu = \sum_{i=1}^k a_i \cdot \mu(A_i \cap E)$$

**Remark:** There are some remarks:

1.  $s : [a, b] \rightarrow [0, \infty)$ ,  $\mu, \mu = \lambda$  (Lebesgue measure)  
Then,  $\int_{[a,b]} s d\mu = \text{area under the graph of } s \text{ in } [a, b]$
2. We are already using  $0 \cdot \infty = 0$  in the definition. In particular,

$$a_i \cdot \mu(A_i \cap E) = \begin{cases} 0 & a_i = 0 \\ \infty & a_i > 0 \end{cases}$$

if  $\mu(A_i \cap E) = \infty$ .

3.  $D \in \mathcal{M}$ , then  $\chi_D$  is a simple function, and:

$$\int_E \chi_D d\mu = \mu(D \cap E)$$

4. More generally,  $s$  simple and measurable,  $E \in \mathcal{M}$ , then:

$$\int_E s d\mu = \int_X s \cdot \chi_E d\mu$$

**Properties 2.9.1** (Basic properties). Let  $N, E, F \in \mathcal{M}$ ,  $s_1, s_2 : X \rightarrow [0, \infty)$  simple and measurable functions. Then:

(i) If  $\mu(N) = 0$ , then:

$$\int_N s_1 d\mu = 0$$

(ii) If  $0 \leq c \leq \infty$ , then:

$$\int_E c \cdot s_1 d\mu = c \cdot \int_E s_1 d\mu$$

(iii)  $\int_E (s_1 + s_2) d\mu = \int_E s_1 d\mu + \int_E s_2 d\mu$

(iv) If  $s_1 \leq s_2$ , then:

$$\int_E s_1 d\mu \leq \int_E s_2 d\mu$$

(v) if  $E \subset F$ , then:

$$\int_E s_1 d\mu \leq \int_F s_1 d\mu$$

The properties (ii) and (iii) are called **linearity** of the integral. The properties (iv) and (v) are called **monotonicity** of the integral.



**Proposition 2.9.1.** *Let  $s : X \rightarrow [0, \infty)$  be a simple measurable function. Then, the function:*

$$\phi(E) := \int_E s \, d\mu : \mathcal{M} \rightarrow [0, \infty]$$

*is a measure on  $(X, \mathcal{M})$ .*

*Proof.* Let  $s = \sum_{i=1}^k a_i \cdot \chi_{A_i}$ ,  $0 \leq a_i \leq \infty$ . We have to show that:

1.  $\phi : \mathcal{M} \rightarrow [0, \infty]$ ?: Yes, since  $s \geq 0$ ,  $\phi(E) \geq 0$ ,  $\forall E \in \mathcal{M}$ .
2.  $\phi(\emptyset) = 0$ ?: Yes, since  $\int_{\emptyset} s \, d\mu = 0$ , as  $\mu(\emptyset) = 0$ .
3.  $\sigma$ -additivity?: Let  $\{E_n\}_{n \in \mathbb{N}}$  be a sequence of pairwise disjoint sets in  $\mathcal{M}$ , and  $E = \bigcup_n E_n$ . Then, we have that:

$$\begin{aligned} \phi(E) &= \int_E s \, d\mu = \int_X s \cdot \chi_E \, d\mu = \sum_{i=1}^k a_i \cdot \mu(A_i \cap E) \\ &= \sum_{i=1}^k a_i \cdot \mu\left(\bigcup_n A_i \cap E_n\right) \end{aligned}$$

Since  $\mu$  is  $\sigma$ -additive, we have that:

$$\begin{aligned} &= \sum_{i=1}^k a_i \sum_n \mu(A_i \cap E_n) \\ &= \sum_n \sum_{i=1}^k a_i \cdot \mu(A_i \cap E_n) \\ &= \sum_n \int_{E_n} s \, d\mu = \sum_n \phi(E_n) \end{aligned}$$

■

## 2.9.2 Integral of non-negative measurable functions

**Definition 2.9.2.** Let  $f : X \rightarrow [0, \infty]$  be a measurable function,  $E \in \mathcal{M}$ . Then, we define the **(Lebesgue) integral** of  $f$  over  $E$  as:

$$\int_E f \, d\mu = \sup \left\{ \int_E s \, d\mu : s \text{ simple, measurable and } 0 \leq s \leq f \right\}$$

**Remark:** There are some remarks:

1. If  $f$  is simple, then the definition coincides with the previous one.
2.  $(\mathbb{N}, \mathcal{P}(\mathbb{N}), \mu_{\#})$ . Then  $f : \mathbb{N} \rightarrow [0, \infty]$  is a sequence. Indeed, if we name  $f_n = f(n)$ , then:

$$\int_{\mathbb{N}} f d\mu_{\#} = \sum_n f_n$$

3. All the basic properties of the integral for simple functions above hold for this new definition.

**Note:** The following propositions assume that  $(X, \mathcal{M}, \mu)$  is a complete measure space (needed for a.e. properties).

**Proposition 2.9.2** (Chebychev's inequality). *Let  $f : X \rightarrow [0, \infty]$  be a measurable function, and  $0 < c < \infty$ . Then:*

$$\mu(\{f \geq c\}) \leq \frac{1}{c} \int_{\{f \geq c\}} f d\mu \leq \frac{1}{c} \int_X f d\mu$$

where  $\{f \geq c\} = \{x \in X : f(x) \geq c\}$ .

*Proof.*

$$\int_X f d\mu \geq \int_{\{f < c\}} f d\mu \geq \int_{\{f < c\}} c d\mu = c \cdot \mu(\{f < c\})$$

Now we just divide by  $c$ . ■

**Note:** We have as a consequence the following lemmas:

**Lemma 2.9.3** (Vanishing lemma). *Let  $f : X \rightarrow [0, \infty]$  be a measurable function,  $E \in \mathcal{M}$ :*

$$\int_E f d\mu = 0 \iff f = 0 \text{ a.e. on } E$$

*Proof.* The proof goes as follows:

$(\Leftarrow)$  : Trivial

( $\Rightarrow$ ) : We have to show that:

$$\mu(\{x \in E : f(x) > 0\}) = 0$$

Let us define  $F = \{x : f(x) > 0\} = \bigcup_n F_n$ , where  $F_n = \{x : f(x) \geq 1/n\}$ . Then, we have that:

$$F_n \subset F_{n+1} \quad \forall n$$

so  $F_n \uparrow F$ . Then, we have that:

$$\mu(F_n) \rightarrow \mu(F)$$

and:

$$0 \leq \mu(F_n) = \mu(\{f \geq \frac{1}{n}\}) \leq \frac{1}{1/n} \int_E f \, d\mu = 0$$

Then,  $\mu(F) = 0$ .

■

**Remark:** The vanishing lemma applies to **every**  $f$  once  $\mu(E) = 0$ , indeed, every property is true a.e. on negligible sets. “The Lebesgue integral does not see negligible sets”.

**Lemma 2.9.4.** *Let  $f : X \rightarrow [0, \infty]$  be a measurable function. Then:*

$$\int_X f \, d\mu < \infty \implies \mu(\{f = \infty\}) = 0$$

*Proof.* Exercise. (Hint:  $\{f = \infty\} = \bigcap_n \{f \geq n\}$ )

■

**Theorem 2.9.5** (Monotone Convergence Theorem (MCT)). *Let  $\{f_n\}_{n \in \mathbb{N}}$  be a sequence of measurable functions  $f_n : X \rightarrow [0, \infty]$ . Assume that:*

$$(i) \quad f_n \leq f_{n+1} \quad \forall n$$

$$(ii) \quad \lim_{n \rightarrow \infty} f_n(x) = f(x) \quad \text{for a.e. } x \in X$$

*Then, we have that:*

$$\lim_{n \rightarrow \infty} \int_X f_n d\mu = \int_X f d\mu$$

**Remark:** All assumptions are essential

*Proof.* The proof goes as follows:

### Part 1:

Assume that assumptions (i) and (ii) hold  $\forall x \in X$ . We have some basic facts:

- $f(x) = \lim_{n \rightarrow \infty} f_n(x) \implies f(x) \geq 0$  and measurable.
- $\int_X f_n d\mu \leq \int_X f_{n+1} d\mu$ . Then, if we define:

$$\alpha_n = \int_X f_n d\mu, \quad \alpha = \lim_{n \rightarrow \infty} \alpha_n$$

we have that  $\alpha_n \leq \alpha_{n+1}$ , so  $\alpha_n \uparrow \alpha$ . Moreover, we have that:

$$\begin{aligned} f_n(x) \leq f(x) &\implies \int_X f_n d\mu \leq \int_X f d\mu \\ &\implies \alpha \leq \int_X f d\mu \end{aligned}$$

So, to complete part 1, we have to show that  $\alpha \geq \int_X f d\mu$ .

We use the definition of  $\int_X f d\mu$ :

Take any  $s : X \rightarrow [0, \infty)$  simple, measurable and  $0 \leq s \leq f$ . Take also  $0 \leq c < 1$ . Then, we have that:

$$0 < c \cdot s \leq f$$

Take  $f_n(x) \uparrow f(x) \forall x \in X$ . Consider  $E_n = \{x \in X : f_n(x) \geq c \cdot s(x)\} \in \mathcal{M}$ . Then, we have that:

- (a)  $E_n \subset E_{n+1}$ : indeed,  $x \in E_n \iff f_n(x) \geq c \cdot s(x) \implies f_{n+1}(x) \geq c \cdot s(x) \iff x \in E_{n+1}$
- (b)  $\bigcup_n E_n = X$ : indeed, either  $f(x) = 0 \implies x \in E_n \forall n$  or  $f(x) > 0$  and  $c \cdot s(x) < f(x)$ . Since  $f_n(x) \uparrow f(x)$ , we have that  $\exists N_0$  s.t.  $f_{N_0}(x) \geq c \cdot s(x)$ . Then  $x \in E_{N_0}$ .

Then, we have that:

$$\begin{aligned} \alpha \geq \alpha_n &= \int_X f_n d\mu \geq \int_{E_n} c \cdot s d\mu = c \cdot \int_{E_n} s d\mu \\ &= c \cdot \phi(E_n) \end{aligned}$$

(where  $\phi(E) = \int_E s d\mu$  is a measure). Then, notice that  $E_n \uparrow X$ , so  $\phi(E_n) \rightarrow \phi(X)$ .

Then, we have that:

$$\alpha \geq c \cdot \phi(X) = c \cdot \int_X s d\mu$$

Then,  $\forall c < 1, \forall s$ :

$$\alpha \geq c \int_X s d\mu$$

If we take the limit  $c \rightarrow 1$ , we have that  $\alpha \geq \int_X s d\mu$ . And if we take the supremum over all  $s$ , we have that:

$$\alpha \geq \int_X f d\mu$$

## **Part 2:**

Now, we have to show that the result holds for *a.e.*  $x \in X$ . Define

$$F = \{x \in X : \text{either (i) or (ii) fails}\}$$

Then we have that  $\mu(F) = 0$ , and  $E = X \setminus F$ . For any  $g$  (non-negative, measurable), we have that:

$$g - \chi_E \cdot g = 0 \quad \text{a.e. on } X$$

Then, we use the vanishing lemma to show that:

$$\begin{aligned} \int_X (g - \chi_E \cdot g) d\mu &= 0 \\ \iff \int_X g d\mu &= \int_E g d\mu \end{aligned}$$

Finally:

$$\int_X f d\mu = \int_E f d\mu = \lim_{n \rightarrow \infty} \int_E f_n d\mu = \lim_{n \rightarrow \infty} \int_X f_n d\mu$$

■

**Remark:** Note that we now have 2 ways to compute the integral of a non-negative measurable function:

- $\int_X f d\mu = \sup \left\{ \int_X s d\mu : s \text{ simple, measurable and } 0 \leq s \leq f \right\}$
- $\int_X f d\mu = \lim_{n \rightarrow \infty} \int_X f_n d\mu$  where  $f_n \uparrow f$  simple and measurable functions.

**Corollary 2.9.5.1** (Monotone convergence for series). *Let  $\{f_n\}_{n \in \mathbb{N}}$  be a sequence of measurable functions  $f_n : X \rightarrow [0, \infty]$ . Then, we have that:*

$$\int_X \sum_n f_n d\mu = \sum_n \int_X f_n d\mu$$

**Proposition 2.9.6.** *Take  $\Phi : X \rightarrow [0, \infty]$  measurable,  $E \in \mathcal{M}$ . Define:*

$$\nu(E) = \int_E \Phi d\mu$$

*Then,  $\nu$  is a measure on  $(X, \mathcal{M})$ . Moreover, for  $f : X \rightarrow [0, \infty]$  measurable:*

$$\int_X f d\nu = \int_X f \cdot \Phi d\mu$$