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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

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

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 **σ -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 σ -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 σ -algebra.
- $\{\emptyset, X\}$ is a σ -algebra.
- $\{\emptyset, E, E^c, X\}$ is a σ -algebra for any $E \subseteq X$.
- $X = \mathbb{R}$, $\mathcal{M} = \{E \subseteq \mathbb{R} : E \text{ is open}\}$ is NOT a σ -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 Ω 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 σ -algebra

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

- (i) $\sigma_0(\mathcal{A})$ is a σ -algebra.
- (ii) $\mathcal{A} \subseteq \sigma_0(\mathcal{A})$.
- (iii) If \mathcal{M} is a σ -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 σ -algebra as it satisfies the properties of a σ -algebra, denoted in definition 2.1.1. ■

Remark: This is relevant. Often, to check that a σ -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 σ -algebra generated by \mathcal{T} is called the **Borel σ -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_δ -sets)
- Countable unions of closed sets (F_σ -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) $\{E_n\}_{n \in \mathbb{N}} \subseteq \mathcal{M}, \text{ disjoint} \implies \mu\left(\bigcup_{n \in \mathbb{N}} E_n\right) = \sum_{n \in \mathbb{N}} \mu(E_n)$

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

Terminology: Let X, \mathcal{M}, μ defined as above:

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

Definition 2.4.2. A measure μ is:

1. **Finite** if $\mu(X) < \infty$
2. **σ -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 μ 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 δ_{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}, μ) be a measure space. Then:*

- (i) μ 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}, μ) 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 (σ -subadditivity). *Let (X, \mathcal{M}, μ) 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}, μ) 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}, μ) be a measure space. Then, we say that μ is a **complete measure** (alternatively, that (X, \mathcal{M}, μ) is a **complete measure space**) all negligible sets are measurable.

Remark (Completion of a measure space): A measure space (X, \mathcal{M}, μ) 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 σ -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 λ 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” μ^* on $\mathcal{P}(\mathbb{R})$.
- Restrict μ^* to a well-chosen σ -algebra $\mathcal{M} \subsetneq \mathcal{P}(\mathbb{R})$.

Definition 2.6.1. Let X be a set. An **outer measure** μ^* 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. (σ -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 μ 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, μ^* 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 μ^* is called the **Lebesgue outer measure**. We will denote it by λ^* (or λ_N^* in the second case). Note that in this case, $\lambda^*(E) = f(E)$ for any $E \in \mathcal{E}$.

Remark: Any μ 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 μ^* be an outer measure on $\mathcal{P}(X)$. A set $E \subseteq X$ is μ^* -**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 μ^* -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 μ^* -measurable. This concludes the proof. ■

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

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

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

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

Lemma 2.6.5. *Let (X, \mathcal{M}, μ) be the measure space as in Carathéodory’s theorem. Then, any $N \subseteq X$ s.t. $\mu^*(N) = 0$ is μ -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 μ^* -measurable. By Carathéodory’s theorem, we have that N is μ -measurable. Finally, we have that $\mu(N) = \mu^*(N) = 0$. ■

Corollary 2.6.5.1. *μ 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, λ^* is the **Lebesgue outer measure** on \mathbb{R} .

Theorem 2.7.1. Let λ^* 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 σ -algebra, called the **Lebesgue σ -algebra**, and λ^* 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 λ 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 λ^* :

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

As ε 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:

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

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_δ 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_σ 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 λ^* , 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_δ 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 λ 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 $(\forall E \in \rho, f^{-1}(E) \in \mathcal{M})$

- Σ is a σ -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 Ψ 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}, μ) be a complete measure space. We say that a property $P(x)$ holds μ -almost everywhere (a.e) if:

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

In other words, $P(x)$ holds μ -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}, μ) 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 χ_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. σ -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 μ is σ -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}, μ) 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, ν is a measure on (X, \mathcal{M}) . Moreover, for $f : X \rightarrow [0, \infty]$ measurable:

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

Proof. The proof goes as follows:

- $\nu : \mathcal{M} \rightarrow [0, \infty]$: Trivial
- $\nu(\emptyset) = 0$: Trivial
- σ -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:

$$\begin{aligned}\nu(E) &= \int_E \Phi \, d\mu = \int_X \Phi \cdot \chi_E \, d\mu = \sum_n \int_X \Phi \cdot \chi_{E_n} \, d\mu \\ &= \sum_n \int_{E_n} \Phi \, d\mu = \sum_n \nu(E_n)\end{aligned}$$

■

Lemma 2.9.7 (Fatou). *Let (X, \mathcal{M}, μ) be a complete measure space, and $\{f_n\}_{n \in \mathbb{N}}$ be a sequence of measurable functions. Then:*

$$\int_X \liminf_n f_n \, d\mu \leq \liminf_n \int_X f_n \, d\mu$$

Proof. Recall that:

$$\begin{aligned}\liminf_n f_n &= \lim_{n \rightarrow \infty} \left(\inf_{k \geq n} f_k \right) \\ &= \sup_n \left(\inf_{k \geq n} f_k \right)\end{aligned}$$

Then, we define:

$$g_n = \inf_{k \geq n} f_k$$

We have the following properties $\forall n$:

- g_n is measurable.
- $g_n \geq 0$
- $g_n \leq g_{n+1}$
- $g_n \leq f_n$

Then, by the MCT, we have that:

$$\begin{aligned}\int_X \liminf_n f_n d\mu &= \int_X \lim_n g_n d\mu = \lim_n \int_X g_n d\mu \\ &= \liminf_n \int_X g_n d\mu \leq \liminf_n \int_X f_n d\mu\end{aligned}$$

■

2.9.3 Integral of real-valued measurable functions

Let $f : X \rightarrow \mathbb{R}$ be a measurable function. Then, we can write $f = f^+ - f^-$, where:

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

Notice that $f^+, f^- \geq 0$ are measurable functions. Then, we define:

$$|f| = f^+ + f^-$$

We also notice that $|f| = f^+ + f^- \geq 0$ is measurable.

Definition 2.9.3. We say $f : X \rightarrow \mathbb{R}$ is **integrable** on X if it is measurable and:

$$\int_X |f| d\mu < \infty$$

We define the set of **integrable functions** as:

$$\mathcal{L}^1(X, \mathcal{M}, \mu) = \{f : X \rightarrow \mathbb{R} : f \text{ is integrable}\}$$

For $f \in \mathcal{L}^1(X, \mathcal{M}, \mu)$, and $E \in \mathcal{M}$, we define:

$$\int_E f d\mu = \int_E f^+ d\mu - \int_E f^- d\mu$$

Proposition 2.9.8. Let $f : X \rightarrow \mathbb{R}$ be a measurable function. Then:

$$(i) \quad f \in \mathcal{L}^1 \iff |f| \in \mathcal{L}^1 \iff (f^+ \in \mathcal{L}^1 \text{ and } f^- \in \mathcal{L}^1)$$

(ii) (Triangular inequality):

$$\left| \int_E f d\mu \right| \leq \int_E |f| d\mu$$

Proof. The proof goes as follows:

(i) Trivial (but see next remark)

(ii) We have that:

$$\begin{aligned}
\left| \int_E f \, d\mu \right| &= \left| \int_E f^+ \, d\mu - \int_E f^- \, d\mu \right| \\
&\leq \left| \int_E f^+ \, d\mu \right| + \left| \int_E f^- \, d\mu \right| = \int_E f^+ \, d\mu + \int_E f^- \, d\mu \\
&= \int_E f^+ + f^- \, d\mu = \int_E |f| \, d\mu
\end{aligned}$$

■

Remark: In general, it is not true that $|f|$ measurable $\implies f$ measurable. Take $F \subset X$, $F \notin \mathcal{M}$ and:

$$f(x) = \chi_F(x) - \chi_{X \setminus F}(x)$$

Then, $|f| = 1$ is measurable, but f is not.

Proposition 2.9.9. *We propose two properties:*

(i) $\mathcal{L}^1(X, \mathcal{M}, \mu)$ is a (real) vector space.

(ii) The functional

$$I(\cdot) := \int_X \cdot \, d\mu : \mathcal{L}^1(X, \mathcal{M}, \mu) \rightarrow \mathbb{R}$$

is a linear functional.

Proof. The proof sketch goes as follows:

Let $u, v \in \mathcal{L}^1(X, \mathcal{M}, \mu)$, $\alpha, \beta \in \mathbb{R}$. We should show that:

$$\alpha u + \beta v \in \mathcal{L}^1(X, \mathcal{M}, \mu)$$

since:

$$|\alpha u + \beta v| \leq |\alpha u| + |\beta v|$$

Then:

$$\int_X (\alpha u + \beta v) d\mu \leq \int_X |\alpha u + \beta v| d\mu \leq \int_X |\alpha u| d\mu + \int_X |\beta v| d\mu < \infty$$

since $|\alpha u|, |\beta v| \in \mathcal{L}^1(X, \mathcal{M}, \mu)$. Then, we have that $\alpha u + \beta v \in \mathcal{L}^1(X, \mathcal{M}, \mu)$.

For the second property, we have that:

$$I(\alpha u + \beta v) = \int_X (\alpha u + \beta v) d\mu = \alpha \int_X u d\mu + \beta \int_X v d\mu = \alpha I(u) + \beta I(v)$$

■

Remark: All the other basic properties of the integral of non-negative functions can be extended to the integral of real-valued functions.

Theorem 2.9.10 (Vanishing lemma). *Let (X, \mathcal{M}, μ) be a complete measure space, and $f, g \in \mathcal{L}^1(X, \mathcal{M}, \mu)$. Then:*

$$f = g \text{ a.e.} \iff \int_X |f - g| d\mu = 0 \iff \int_E (f - g) d\mu = 0 \forall E \in \mathcal{M}$$

Proof. The “difficult” part of the proof is:

$$\int_E (f - g) d\mu = 0, \quad \forall E \in \mathcal{M} \implies f = g \text{ a.e.}$$

The proof goes as follows:

Let $E_1 = \{f \geq g\}$, and $E_2 = X \setminus E_1$. Then, we have that:

$$\begin{aligned} 0 &= \int_{E_1} (f - g) d\mu = \int_{E_1} (f - g)^+ d\mu \\ 0 &= \int_{E_2} (f - g) d\mu = - \int_{E_2} (f - g)^- d\mu \end{aligned}$$

Then, we have that:

$$(f - g)^+ = 0 \text{ and } (f - g)^- = 0 \text{ a.e. on } X$$

■

Remark: In particular, for $u \in \mathcal{L}^1$:

$$\int_E u d\mu = 0 \quad \forall E \in \mathcal{M} \implies u = 0 \text{ a.e.}$$

This is the same as:

$$\int_X u \varphi d\mu = 0 \quad \forall \varphi \text{ characteristic function} \implies u = 0 \text{ a.e.}$$

This can be true also replacing φ by “something else”. For instance, in the case of $u \in \mathcal{L}^1(\mathbb{R}, \mathcal{L}(\mathbb{R}), \lambda)$:

$$\int_{\mathbb{R}} u \varphi d\lambda = 0 \quad \forall \varphi \in V \implies u = 0 \text{ a.e.}$$

where $V = \{C_0^\infty(\mathbb{R})\}$, or $V = \{C_0^0(\mathbb{R})\}$.

This is the “fundamental lemma of calculus of variations”.

Theorem 2.9.11 (Dominated convergence theorem (DCT)). *Let (X, \mathcal{M}, μ) be a complete measure space and $\{f_n\}_{n \in \mathbb{N}}$ be a sequence of measurable functions $f_n : X \rightarrow \mathbb{R}$, and $f : X \rightarrow \mathbb{R}$. Assume that:*

$$(i) \quad |f_n| \leq g \text{ a.e. on } X, \quad \forall n \in \mathbb{N}, \text{ where } g \in \mathcal{L}^1(X, \mathcal{M}, \mu)$$

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

Then, $f \in \mathcal{L}^1(X, \mathcal{M}, \mu)$, and:

$$\lim_{n \rightarrow \infty} \int_X |f_n - f| d\mu = 0$$

In particular:

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

Proof. First, we have 2 basic facts:

1. $|f_n| \leq g \text{ a.e. on } X, \quad \forall n \in \mathbb{N} \implies f_n \in \mathcal{L}^1(X, \mathcal{M}, \mu)$
2. $|f| \leq g \text{ a.e. on } X \implies f \in \mathcal{L}^1(X, \mathcal{M}, \mu)$

Then, consider the sequence $h_n = 2g - |f_n - f|$. We have that:

- h_n is measurable.
- $h_n \leq 2g$

- $h_n \geq 0$. Indeed:

$$|f_n - f| \leq |f_n| + |f| \leq 2g \implies 2g - |f_n - f| \geq 0$$

We now apply the Fatou's lemma to the sequence h_n :

$$\begin{aligned} \int_X (\liminf_n h_n) d\mu &\leq \liminf_n \int_X h_n d\mu \\ &= \int_X 2g d\mu - \limsup_n \int_X |f_n - f| d\mu \end{aligned}$$

Also, notice that:

$$\liminf_n h_n = 2g$$

Then, we have that:

$$\begin{aligned} \int_X 2g d\mu &\leq \int_X 2g d\mu - \limsup_n \int_X |f_n - f| d\mu \\ \implies \limsup_n \int_X |f_n - f| d\mu &\leq 0 \end{aligned}$$

Then, we have that:

$$\limsup_n \int_X |f_n - f| d\mu \geq \liminf_n \int_X |f_n - f| d\mu \geq 0$$

In the end:

$$\lim_n \int_X |f_n - f| d\mu = 0$$

■

Remark: If $\mu(X) < \infty$, then the constants are integrable. Then, if $|f_n(x)| \leq M$ a.e, for some $M \in \mathbb{R}$, then:

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

(We are using the DCT with $g = M$)

Corollary 2.9.11.1 (Dominated Convergence for series). *Let $\{f_n\}_{n \in \mathbb{N}}$ be a sequence of measurable functions $f_n : X \rightarrow \mathbb{R}$, s.t $f_n \in \mathcal{L}^1(X, \mathcal{M}, \mu)$. If $\sum_n \int_X |f_n| d\mu < \infty$, then:*

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

2.9.4 Comparison between Riemann and Lebesgue integrals

Theorem 2.9.12. *Let $I = [a, b] \subset \mathbb{R}$ be a closed interval, and $f : I \rightarrow \mathbb{R}$. If f is **Riemann integrable** on I , then f is **Lebesgue integrable** on I , i.e., $f \in \mathcal{L}^1(I, \mathcal{L}(I), \lambda)$, and the two integrals coincide:*

$$\int_I f d\lambda = \int_a^b f(x) dx$$

Theorem 2.9.13. *Let $I = (\alpha, \beta)$, such that $-\infty \leq \alpha < \beta \leq \infty$. If $|f|$ is **Riemann integrable** on I (in the generalized sense), then f is **Lebesgue integrable** on I :*

$$\int_I f d\lambda = \int_\alpha^\beta f(x) dx$$

Remark: If the generalized Riemann integral of $|f|$ diverges, then:

$$\int_I |f| d\lambda = \infty$$

but $\int_I f d\lambda$ is not defined (unless $f = \pm|f|$) and:

$$\int_\alpha^\beta f(x) dx \text{ and } \int_I f d\lambda$$

are not related.