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# Important Notions in Measure Theory

The modern approach to study Probability Theory is built on Measure Theory. In order to follow this lecture, it is essential to have a minimum understanding and familiarity with Measure Theory. For this reason, we will start by reviewing some notions in Measure Theory: we will recall the important definitions and theorems along with a few proofs. The proofs provided here are classical ones and the techniques involved will also be seen repeatedly in Probability Theory; those which are omitted require more technical details from Measure Theory. Interested readers are invited to take a careful look at Rudin's book "Real and Complex Analysis".

## 1.1 Measurable Spaces and Measures

Given a set, we desire to define a measure (測度) which is a function that attributes a mass to (potentially) all the subsets. Additionally, we want a measure to satisfy some specific properties, such as additivity (加法性) or even  $\sigma$ -additivity ( $\sigma$  加法性). Hence, we will first define measurable spaces and introduce the notion of  $\sigma$ -algebra, which is, vaguely speaking, a collection of subsets on which we can make sense of a measure.

### 1.1.1 Measurable Sets and $\sigma$ -algebras

First, let us start with the notion of  $\sigma$ -algebra.

**Definition 1.1.1:** Let E be a set. A subset  $A \subseteq \mathcal{P}(E)$  is called a  $\sigma$ -algebra ( $\sigma$  代數) if the following properties are satisfied.

- (a)  $E \in \mathcal{A}$ .
- (b) If  $A \in \mathcal{A}$ , then  $A^c \in \mathcal{A}$ .
- (c) If for all  $n \in \mathbb{N}$ ,  $A_n \in \mathcal{A}$ , then  $\bigcup_{n \in \mathbb{N}} A_n \in \mathcal{A}$ .

The elements of  $\mathcal{A}$  are called *measurable sets* (可測集合), or  $\mathcal{A}$ -measurable sets, if one wants to emphasize the underlying  $\sigma$ -algebra. The couple  $(E, \mathcal{A})$  is called a *measurable space* (可測空間).

**Remark 1.1.2**: In the above definition,  $\sigma$  means that in the condition (c), the union can be any *countable* union. If we replace this condition by any *finite* union, then the corresponding definition defines what we call an *algebra* (代數), in the sense of *algebra of sets*, or *set algebra*.

In Section 1.1.2 below, we will define the notion of measure, which is closely related to the  $\sigma$ -algebra of the considered space. Moreover, later in Probability Theory, we will also see the importance and applications of  $\sigma$ -algebra. For example, in Chapter 5, we will define the notion of *conditional probability* and *conditional expectation*.

**Example 1.1.3:** Given a set E. We have two extremal  $\sigma$ -algebras: the finest (最精緻的) is  $\mathcal{A} = \mathcal{P}(E)$  and the coarsest (最粗糙) is  $\mathcal{A} = \{\emptyset, E\}$ . The latter is also called the trivial (平凡)  $\sigma$ -algebra.

In order to construct other examples of  $\sigma$ -algebras, we introduce the notion of *generated*  $\sigma$ -algebra (生成 $\sigma$ 代數).

**Definition 1.1.4:** Let  $\mathcal{C}$  be a subset of  $\mathcal{P}(E)$ . Then the smallest  $\sigma$ -algebra containing  $\mathcal{C}$  exists, denoted by  $\sigma(\mathcal{C})$  and is given by,

$$\sigma(\mathcal{C}) = \bigcap_{\substack{\mathcal{A} \text{ is a } \sigma\text{-algebra} \\ \text{s.t. } \mathcal{C} \subseteq \mathcal{A}}} \mathcal{A}.$$

In the definition, the intersection is well defined because  $\mathcal{A} = \mathcal{P}(E)$  is a  $\sigma$ -algebra containing  $\mathcal{C}$ .

**Definition 1.1.5**: Let E be a set, and  $\mathcal{O}$  be a set of subsets of E satisfying

- (a)  $E \in \mathcal{O}$  and  $\emptyset \in \mathcal{O}$ .
- (b) For any finitely many open sets  $U_1, \ldots, U_n \in \mathcal{O}$ , we have  $U_1 \cap \cdots \cap U_n \in \mathcal{O}$ .
- (c) For any family of open sets  $(U_i)_{i\in I}$  with  $U_i\in\mathcal{O}$ , we have  $\bigcup_{i\in I}U_i\in\mathcal{O}$ .

Then, we call  $(E, \mathcal{O})$  a topological space (拓撲空間) and the elements in  $\mathcal{O}$  open sets (開集).

**Definition 1.1.6**: Let  $(E, \mathcal{O})$  be a topological space.

- (1) We denote by  $\sigma(\mathcal{O})$  the  $\sigma$ -algebra generated by  $\mathcal{O}$ , the open sets of the topological space.
- (2) It is also called Borel  $\sigma$ -algebra (伯雷爾  $\sigma$  代數) of E, which can be denoted by  $\mathcal{B}(E)$ , if there is a canonical choice for  $\mathcal{O}$ .
- (3) We call Borel sets (伯雷爾集合) the elements of  $\mathcal{B}(E)$ .

In other words, Borel  $\sigma$ -algebra is the smallest  $\sigma$ -algebra containing all the open sets of E.

Next, whenever we talk about topological spaces, the  $\sigma$ -algebra we consider will always be its Borel  $\sigma$ -algebra. For example, when we want to equip  $\mathbb{R}$  or  $\mathbb{R}^d$  with a  $\sigma$ -algebra to turn them into measurable spaces, we take the open sets to be given by any equivalent norm on  $\mathbb{R}$  or  $\mathbb{R}^d$ , and the corresponding Borel  $\sigma$ -algebra. Actually, in the class of Probability Theory, the most of the spaces we will discuss fall into this setting.

**Question 1.1.7:** Prove that the  $\sigma$ -algebra generated by each of the three following sets of intervals is equal to  $\mathcal{B}(\mathbb{R})$ ,

- (1)  $\mathcal{I} = \{(a, b) : a < b, a, b \in \mathbb{R}\},\$
- (2)  $\mathcal{I} = \{(-\infty, a) : a \in \mathbb{R}\},\$
- (3)  $\mathcal{I} = \{(-\infty, a) : a \in \mathbb{Q}\}.$

The next important notion to introduce is *prodcut*  $\sigma$ -algebra.

**Definition 1.1.8**: Given two measurable spaces  $(E_1, A_1)$  and  $(E_2, A_2)$ , we can define the *product*  $\sigma$ -algebra (積  $\sigma$  代數) on their product space  $E_1 \times E_2$ ,

$$\mathcal{A}_1 \otimes \mathcal{A}_2 = \sigma(A_1 \times A_2 : A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2).$$

**Question 1.1.9:** Show that  $\mathcal{B}(\mathbb{R}^2) = \mathcal{B}(\mathbb{R}) \otimes \mathcal{B}(\mathbb{R})$ .

#### 1.1.2 Measures

Given a measurable space (E, A), we want to define a *measure* on it.

**Definition 1.1.10**: A function  $\mu: \mathcal{A} \longrightarrow [0, \infty]$  is called a *positive measure* (正測度) or simply *measure* (測度) if the following axioms (公理) are satisfied,

- $\mu(\varnothing) = 0$ ,
- ( $\sigma$ -additivity) For any disjoint sequence  $(A_n)_{n\geqslant 1}$  on  $\mathcal{A}$ , we have

$$\mu\Big(\bigcup_{n\geqslant 1} A_n\Big) = \sum_{n\geqslant 1} \mu(A_n). \tag{1.1}$$

Moreover, we call  $(E, A, \mu)$  a measure space (測度空間).

**Proposition 1.1.11:** Let  $(E, A, \mu)$  be a measure space. We have the following properties.

(1) If  $A, B \in \mathcal{A}$ ,  $A \subseteq B$ ,  $\mu(A) \leqslant \mu(B)$  and  $\mu(A) < \infty$ , then

$$\mu(B \backslash A) = \mu(B) - \mu(A).$$

(2) If  $A, B \in \mathcal{A}$ , then

$$\mu(A) + \mu(B) = \mu(A \cup B) + \mu(A \cap B).$$

(3) (Lower continuity) Let  $(A_n)_{n\geqslant 1}$  be a sequence of increasing elements in  $\mathcal{A}$  (i.e.,  $A_n\subseteq A_{n+1}$  for all n). Then, we have

$$\mu\Big(\bigcup_{n>1} A_n\Big) = \lim_{n\to\infty} \uparrow \mu(A_n).$$

(4) (Upper continuity) Let  $(B_n)_{n\geqslant 1}$  be a sequence of decreasing elements in  $\mathcal{A}$  (i.e.,  $B_{n+1}\subseteq B_n$  for all n). If there exists  $n_0$  such that  $\mu(B_{n_0})<\infty$ , then

$$\mu\Big(\bigcap_{n>1} B_n\Big) = \lim_{n\to\infty} \downarrow \mu(B_n).$$

(5) If  $(A_n)_{n\geqslant 1}$  is a sequence of elements in A, then

$$\mu\Big(\bigcup_{n\geqslant 1}A_n\Big)\leqslant \sum_{n\geqslant 1}\mu(A_n).$$

**Proof**: See Exercise 1.5.

#### **Definition 1.1.12**: Given a measure $\mu$ .

- The quantity  $\mu(E)$  is called the *total mass* (總質量) of  $\mu$ .
- If  $\mu(E)=1$ , we say that  $\mu$  is a probability measure (機率測度).
- If  $\mu(E) < \infty$ , we say that  $\mu$  is a finite measure (有限測度).
- If there exists an increasing measurable subsets  $(E_n)$  such that  $E = \bigcup E_n$  and that for all n, we have  $\mu(E_n) < \infty$ , then we say that  $\mu$  is a  $\sigma$ -finite measure ( $\sigma$  有限測度).
- Let  $x \in E$  and suppose that  $\{x\} \in A$ . If  $\mu(\{x\}) > 0$ , then we say that x is an atom (原子) of  $\mu$  and that  $\mu$  is an atomic measure (原子測度). Otherwise, we say that  $\mu$  is an atomless measure (無原子測度).

#### **Example 1.1.13:**

- (1) If  $\mu$  is a finite measure on  $(E, \mathcal{A})$ , then  $\frac{\mu}{\mu(E)}$  is a probability measure on the same measurable space.
- (2) On the measurable space  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ , the Lebesgue measure  $\mu$  is the unique measure such that  $\mu([a,b]) = \mu((a,b)) = b-a$  for any a < b. Its uniqueness is a consequence of the monotone class lemma, see Corollary 1.1.19 and Remark 1.1.20 for details. Its existence can be derived from Theorem 1.2.29.
- (3) (Dirac measure) Given a measurable space (E, A). Let  $x \in E$  and suppose  $\{x\} \in A$ . The measure  $\mu = \delta_x$  is the *Dirac measure* (狄拉克測度) at x.

**Proposition 1.1.14** (Upper and lower continuity): Let (E, A) be a measurable space. Define  $\mu : A \longrightarrow [0, \infty]$  satisfying

- $\mu(\emptyset) = 0$ ,
- (Finite additivity) For any finite disjoint sequence of  $(A_n)_{1 \le n \le N}$  with values in A, we have

$$\mu\Big(\bigcup_{n=1}^{N} A_n\Big) = \sum_{n=1}^{N} \mu(A_n).$$

Then, the following properties are equivalent.

(i)  $(\sigma$ -additivity)  $\mu$  satisfies  $\sigma$ -additivity mentioned in Eq. (1.1), that is,  $(E, \mathcal{A}, \mu)$  is a measure space.

(ii) (Lower continuity) For any increasing sequence  $(A_n)_{n\geqslant 1}$  with values in A, we have

$$\mu\Big(\bigcup_{n=1}^{\infty} A_n\Big) = \lim_{n \to \infty} \uparrow \mu(A_n).$$

(iii) (Upper continuity) For any decreasing sequence  $(A_n)_{n\geqslant 1}$  with values in A, if there exists  $n_0$  such that  $\mu(A_{n_0})<\infty$ , then we have

$$\mu\Big(\bigcap_{n=1}^{\infty} A_n\Big) = \lim_{n \to \infty} \downarrow \mu(A_n).$$

**Proof**: Let us prove (i)  $\Longrightarrow$  (ii). Given an increasing sequence  $(A_n)_{n\geqslant 1}$ , define the sequence  $(B_n)_{n\geqslant 1}$  as follows. Set  $B_1:=A_1$  and  $B_n:=A_n\backslash A_{n-1}$  for all  $n\geqslant 2$ . Under this construction, the elements  $B_n$  are pairwise disjoint and for all  $n\geqslant 1$ , we have

$$\bigcup_{k=1}^{n} B_k = A_n = \bigcup_{k=1}^{n} A_k \quad \text{and} \quad \bigcup_{k=1}^{\infty} B_k = \bigcup_{k=1}^{\infty} A_k.$$

As a consequence of  $\sigma$ -additivity, we find

$$\mu\left(\bigcup_{k=1}^{\infty} A_k\right) = \mu\left(\bigcup_{k=1}^{\infty} B_k\right) = \sum_{k=1}^{\infty} \mu(B_k) = \lim_{n \to \infty} \sum_{k=1}^{n} \mu(B_k)$$
$$= \lim_{n \to \infty} \mu\left(\bigcup_{k=1}^{n} B_k\right) = \lim_{n \to \infty} \mu(A_n).$$

Let us prove (ii)  $\Longrightarrow$  (i). Consider a sequence  $(A_n)_{n\geqslant 1}$  of disjoint elements and define the sequence  $(B_n)_{n\geqslant 1}$  as follows,

$$\forall n \geqslant 1, \qquad B_n := \bigcup_{k=1}^n A_k.$$

We note that  $(B_n)_{n\geqslant 1}$  is an increasing sequence and using the property (ii), we find

$$\mu\left(\bigcup_{k=1}^{\infty} A_k\right) = \mu\left(\bigcup_{k=1}^{\infty} B_k\right) = \lim_{n \to \infty} \mu(B_n) = \lim_{n \to \infty} \sum_{k=1}^{n} \mu(A_k) = \sum_{k=1}^{\infty} \mu(A_k).$$

Let us prove (ii)  $\Longrightarrow$  (iii). Consider a decreasing sequence  $(A_n)_{n\geqslant 1}$  with  $n_0$  such that  $\mu(A_{n_0})<\infty$ . Set  $B_n:=A_n^c\cap A_{n_0}$  for all  $n\geqslant 1$ . Then, the sequence  $(B_n)_{n\geqslant 1}$  is increasing. Using the property (ii), we find

$$\mu\left(\bigcap_{n=1}^{\infty} A_n\right) = \mu\left(\bigcap_{n=n_0}^{\infty} (A_n \cap A_{n_0})\right) = \mu\left(A_{n_0} \setminus \left(\bigcup_{n=n_0}^{\infty} B_n\right)\right)$$
$$= \mu(A_{n_0}) - \mu\left(\bigcup_{n=n_0}^{\infty} B_n\right)$$
$$= \mu(A_{n_0}) - \lim_{n \to \infty} \mu(B_n) = \lim_{n \to \infty} \mu(A_n).$$

To finish the proof, we use a similar method to prove (iii)  $\Longrightarrow$  (ii).

#### 1.1.3 Monotone Classes

In the previous subsections, we defined the notion of  $\sigma$ -algebras and measures. However,  $\sigma$ -algebras are not easy to manipulate directly, since in general, one does not have a way to write down a generic element from them. Apart from this, in the definition of a measure, we talk about countable unions of *disjoint* sequences of subsets; but in the definition of a  $\sigma$ -algebra, this can be countable unions of *any* sequence of subsets. As a consequence, when we need to construct a measure on a measurable space, or to prove some specific conditions such as the uniqueness of the measure, it can be a bit tricky. In this subsection, we will introduce the notion of *monotone classes*, state the monotone class lemma (Theorem 1.1.18) and discuss about the uniqueness of the measure (Corollary 1.1.19).

**Definition 1.1.15:** Let  $\mathcal{M}$  be a subset of  $\mathcal{P}(E)$ . We call  $\mathcal{M}$  a monotone class (單調類) if the following conditions are satisfied.

- (a)  $E \in \mathcal{M}$ .
- (b) If  $A, B \in \mathcal{M}$  and  $A \subseteq B$ , then  $B \setminus A \in \mathcal{M}$ .
- (c) If  $(A_n)_{n\in\mathbb{N}}$  is an increasing sequence of elements in  $\mathcal{M}$ , then  $\bigcup_{n\in\mathbb{N}} A_n \in \mathcal{M}$ .

**Remark** 1.1.16 : Any  $\sigma$ -algebra is also a monotone class.

As in the case of  $\sigma$ -algebras, we can also define the notion of a generated monoton class (生成單調類).

**Definition 1.1.17:** For any  $C \subseteq \mathcal{P}(E)$ , there exists a smallest monotone class such that it contains C, denoted

$$\mathcal{M}(\mathcal{C}) = \bigcap_{\substack{\mathcal{M} \text{ is a monotone class} \\ \text{s.t. } \mathcal{C} \subset \mathcal{M}}} \mathcal{M}.$$

The following lemma tells us the conditions under which a generated monotone class and a generated  $\sigma$ -algebra are equal.

**Theorem 1.1.18** (Monotone class lemma) : If  $C \subseteq \mathcal{P}(E)$  is closed under finite intersection, then  $\mathcal{M}(C) = \sigma(C)$ .

This is one of the most important statements in Measure Theory. We only provide the necessary steps for the proof, the complete proof being an exercise.

**Proof**: From the above remark, we know that  $\mathcal{M}(\mathcal{C}) \subseteq \sigma(\mathcal{C})$ , so we only need to prove the other inclusion. We need to prove the two following points.

- (1) If  $\mathcal{M}$  is a monotone class and is closed under finite intersections, then  $\mathcal{M}$  is a  $\sigma$ -algebra.
- (2) Prove that  $\mathcal{M}(\mathcal{C})$  is closed under finite intersections in two steps:

- (a) Given  $A \in \mathcal{C}$ , let  $\mathcal{M}_1 = \{B \in \mathcal{M}(\mathcal{C}) : A \cap B \in \mathcal{M}(\mathcal{C})\}$  and prove  $\mathcal{M}_1 = \mathcal{M}(\mathcal{C})$ .
- (b) Given  $B \in \mathcal{M}(\mathcal{C})$ , let  $\mathcal{M}_2 = \{A \in \mathcal{M}(\mathcal{C}) : A \cap B \in \mathcal{M}(\mathcal{C})\}$  and prove  $\mathcal{M}_2 = \mathcal{M}(\mathcal{C})$ .

The monotone class lemma allows us to show the uniqueness of the measure.

**Corollary 1.1.19**: Let  $\mu$  and  $\nu$  be two measures on the measurable space (E, A). Assume that there exists a subset  $C \subseteq A$  which is closed under finite intersections satisfying  $\sigma(C) = A$  and such that for all  $A \in C$ , we have  $\mu(A) = \nu(A)$ .

- (1) (Finite measure) If  $\mu(E) = \nu(E) < \infty$ , then  $\mu = \nu$ .
- (2) ( $\sigma$ -finite measure) If there exists a increasing sequence  $(E_n)$  in  $\mathcal{C}$  such that  $E = \bigcup E_n$  and that for all n, we have  $\mu(E_n) = \nu(E_n)$ , then  $\mu = \nu$ .

#### **Proof:**

- (1) Let  $\mathcal{G} = \{A \in \mathcal{A} : \mu(A) = \nu(A)\}$ . From the hypothesis, we have  $\mathcal{C} \subseteq \mathcal{G}$ . Using the properties of a measure, we can show that  $\mathcal{G}$  is a monotone class. Moreover, since  $\mathcal{C}$  is closed under finite intersections, according to the monotone class lemma, we have  $\mathcal{M}(\mathcal{C}) = \sigma(\mathcal{C}) = \mathcal{A}$ . This proves that  $\mathcal{G} = \mathcal{A}$ , which is  $\mu = \nu$ .
- (2) For all n, we can define the restricted measures of  $\mu$  and  $\nu$  on  $E_n$ ,

$$\forall A \in \mathcal{A}, \quad \mu_n(A) = \mu(A \cap E_n), \quad \nu_n(A) = \nu(A \cap E_n).$$

From the first part of the proof, we obtain  $\mu_n = \nu_n$ , then from the property (3) of a measure, we get

$$\forall A \in \mathcal{A}, \quad \mu(A) = \lim \uparrow \mu_n(A \cap E_n), \quad \nu(A) = \lim \uparrow \nu_n(A \cap E_n).$$

This proves the desired statement.

**Remark 1.1.20**: The above corollary states that if there exists two measures  $\lambda$  and  $\lambda'$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  such that for all intervals (a,b), we have  $\lambda((a,b)) = \lambda'((a,b)) = b-a$ , then they must be the same measure  $\lambda = \lambda'$ .

**Question 1.1.21:** Please explain why it is important to assume that  $\mu$  and  $\nu$  are finite measures or  $\sigma$ -finite measures in Corollary 1.1.19. In other words, is it possible to find infinite measures  $\mu$  and  $\nu$  such that  $\mu(A) = \nu(A)$  for all  $A \in \mathcal{C}$ , but  $\mu \neq \nu$ ?