# Measure Theory & Probability

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# **Basic Notions and Notation**

#### Example 1.1.

Simplest  $\sigma$ -algebra:

- $\{\emptyset, \Omega\}$ , contained in every  $\sigma$ -algebra on  $\Omega$ ,
- Family of all subsets of  $\Omega$ , containing every  $\sigma$ -algebraon  $\Omega$ .

#### Exercise 1.1.

Let  $\mathcal{F}$  be a  $\sigma$ -algebra. Then  $A_n \in \mathcal{F}$  for every integer  $n \geqslant 1 \Rightarrow \bigcap_{n=1}^{\infty} A_n \in \mathcal{F}$ .

## Proposition (Unknown).

Let  $A, B \in \mathcal{F}$ . Then  $\mu(A \cap B) \leq \mu(A)$ .  $Hint: \sigma$ -additivity and  $A = (A \cap B) \cup (A \setminus B)$ .

# **Expectation Integrals**

#### Proposition (Unknown).

Let  $A, B \subset \Omega$ . Then the following equalities hold:

- $\mathbf{1}_{A^C} = 1 \mathbf{1}_A$ ,
- $\mathbf{1}_{A}\mathbf{1}_{B} = \mathbf{1}_{A\cap B}$ .

#### Lemma 3.3.

Let X be a **non-negative** random variable. Then there exists a sequence of **non-negative**, **simple** random variables  $X_n$  converging to X for every  $\omega \in \Omega$ .

Hint:  $h_n(x) = \min\{\lfloor 2^n x \rfloor / 2^n, n\}$  is non-negative, simple and increasing, approaching x. Consider  $X_n := h(X) \to X$ .

**Lemma** (Simple Function Integral Properties). Let  $f,g:\Omega\to\overline{\mathbb{R}}$  be a **non-negative**, simple functions and  $a,b\geqslant 0$ . Then the following holds:

- $\int_{\Omega} f \, d\mu \geqslant 0$ ,
- $\int_{\Omega} (af + bg) d\mu = a \int_{\Omega} f + b \int_{\Omega} g d\mu$ .

#### Lemma 3.3 (General).

Let  $f: \Omega \to \overline{\mathbb{R}}$  be a **non-negative**, measurable function. The there exists a sequence  $f_n$  of **non-negative**, simple functions such that:

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

*Hint:* : Use  $h_n$  from Lemma 3.3's hint.

#### Exercise 3.5.

Let  $A \in \mathcal{F}$  s.t.  $\mu(A) = 0$ . Then for **any** measurable function  $f : \Omega \to \overline{\mathbb{R}}$ :

$$\int_A f \, d\mu = 0.$$

**Theorem 3.8** (Monotone Convergence). Let  $(f_n)_{n=1}^{\infty}$  be increasing sequence of non-negative, measurable functions  $f_n: \Omega \to \overline{\mathbb{R}}$ , converging to some f. Then:

$$\int_{\Omega} \lim_{n \to \infty} f_n \, d\mu = \lim_{n \to \infty} \int_{\Omega} f_n \, d\mu$$

**Proposition 3.18** (Markov-Chebyshev's Inequality).

Let X be a **non-negative** R.V., then

$$P(X \ge \lambda) \le \lambda^{-\alpha} E(X^{\alpha}) \quad \forall \lambda > 0, \alpha > 0.$$

#### Remark 3.3.

Let  $(\Omega, \mathcal{F}, \mu)$  be measure space,  $f: \Omega \to \overline{\mathbb{R}}$  non-negative  $\mathcal{F}$ -measurable, then

$$\mu(f \geqslant \lambda) \leqslant \lambda^{-\alpha} \int_{\Omega} f^{\alpha} d\mu \quad \forall \lambda > 0, \alpha > 0.$$

**Proposition** (Restricted Expectation). Let X be a random variable and  $A \in \mathcal{F}$ , then:

$$E(X\mathbf{1}_A) = \int_A X \, dP.$$

# **Definitions**

In the following,  $\Omega$  is a set,  $\mathcal{F}$  a  $\sigma$ -algebra on  $\Omega$ . If used, then  $\mu$  is a measure. Otherwise, the measure is the probability measure P.

# Definition 1.1.

Let  $\mathcal{F}$  be a family of subsets of set  $\Omega$ .  $\mathcal{F}$  is called a  $\sigma$ -algebra if:

- Closed Under Complement:  $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$ ,
- Closed Under Arbitrary Union:  $A_n \in \mathcal{F}$  for integer  $n \geqslant 1$  $\Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$ ,

• Contains Entire Set:  $\Omega \in \mathcal{F}$ 

**Definition 1.2.** Let  $\mathcal{C}$  be a family of subsets of  $\Omega$ . There exists a  $\sigma$ -algebra which contains  $\mathcal{C}$  and which is contained in every  $\sigma$ -algebra that contains  $\mathcal{C}$  (take intersection of all  $\sigma$ -algebras. Such  $\sigma$ -algebra is unique and called smallest  $\sigma$ -algebra containing  $\mathcal{C}$  or  $\sigma$ -algebra generated by  $\mathcal{C}$ , denoted by  $\sigma(\mathcal{C})$ . Simplest example, let  $A \subseteq \Omega$ :

$$\sigma(A) = \{\emptyset, A, A^c, \Omega\}.$$

#### Definition 2.1.1.

Let  $A \subseteq \Omega$  and  $\mathbf{1}_A$  be defined as follows:

$$\mathbf{1}_{A}(\omega) = \begin{cases} 1, & \omega \in A \\ 0, & \omega \notin A \end{cases}.$$

Then  $\mathbf{1}_A$  is a R.V. and called the *indicator* (function) of (events) A.

 ${\bf Definition~(Indicator~Integral)}.$ 

Let  $A \subset \Omega$ , then:

$$\int_{\Omega} \mathbf{1}_A \, d\mu = \mu(A).$$

# Definition (Simple Function).

Let  $f:\Omega\to\mathbb{R}$  be a *simple function*, then f takes finitely many values. Formally, if I is a finite index set,  $(A_i)_{i\in I}$  a famility of disjoint subsets of  $\Omega$  and  $(c_i)_{i\in I}$  a family of real numbers, then:

$$f(\omega) = \sum_{i \in I} c_i \mathbf{1}_{A_i}(\omega).$$

# Definition (Lebesgue Integral for Expectation).

Let X be a random variable. Then we write:

$$EX = \int_{\Omega} X \, dP.$$

# Definition (Unknown).

Let  $A \in \mathcal{F}$  and  $f : \Omega \to \overline{\mathbb{R}}$  is a measurable function, then we define:

$$\int_A f \, d\mu = \int_{\Omega} \mathbf{1}_A f \, d\mu,$$

when the integral of  $\mathbf{1}_A f$  w.r.t  $\mu$  exists.

## Definition (Unknown).

Let X be a random variable. Then X has finite second moment if  $EX^2 < \infty$ .