

# Measure Theory & Probability

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## Example 1.1.

Simplest  $\sigma$ -algebra:

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

## Exercise 1.1.

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

## Expectation Integrals

### Exercise 3.5.

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

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

### Theorem 3.8 (Monotone Convergence).

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

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

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

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

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

### Remark 3.3.

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

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

### 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  $n \geq 1$   
 $\Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$ ,
- **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)  $A$** .

### Definition 2.1.1.

Let  $A \subseteq \Omega$ , then:

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

### 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  $f : \Omega \rightarrow \overline{\mathbb{R}}$  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  $EX^2 < \infty$ .