

# 1 Probability Measure

## Definition 1

- Given space  $\Omega$  ("abstract space")
- Need a  $\sigma$ -algebra  $A_R$  on  $\Omega$ . ("measurable events")
  - $A \in A_R \implies A^C \in A_R$
  - $(A_i)_{i \in \mathbb{N}} \subset A_R \implies \bigcup_{i=1}^{\infty} A_i \in A_R$  ("countable unions")
  - $\emptyset, \Omega \in A_R$
  - countable intersections
- A measure  $\mu$  on  $(\Omega, A_R)$  is a function  $\mu : A_R \rightarrow [0, \infty]$  that is countably additive: If  $(A_i)_{i \in \mathbb{N}}$  is a sequence of pairwise disjoint sets, then  $\mu(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \mu(A_i)$

A measure  $P$  on a measurable space  $(\Omega, A_R)$  is called a *probability measure* if  $P(\Omega) = 1$ . The elements of  $A_R$  are called events. Then  $(\Omega, A_R, P)$  is called a *probability space*.

### Example (1):

Throw a die

$\Omega = \{1, 2, \dots, 6\}$ ,  $A_R = P(\Omega)$  ( $\sigma$ -algebra generated by the "elementary events"  $\{1\}, \{2\}, \dots, \{6\}$ ).

$P$  can be defined uniquely by assigning  $P(\{1\}) = P(\{2\}) = \dots = P(\{6\}) = \frac{1}{6}$

For example  $P(\{1, 5\}) = P(\{1\}) + P(\{5\}) = \frac{1}{3}$

Throw two dice:

$\Omega = \{1, 2, \dots, 6\} \times \{1, 2, \dots, 6\} = \{ \underset{\text{first die, second die}}{(1, 1)}, (1, 2), \dots \}$  all of which are elementary events

$A_R = P(\Omega)$

$P(\{(i, j)\}) = \frac{1}{36}$

### Example (2): Normal distribution

$\Omega = \mathbb{R}$

$A_R = \text{Borel-}\sigma\text{-algebra}$

$f_{\mu, r} : \mathbb{R} \implies \mathbb{R}$

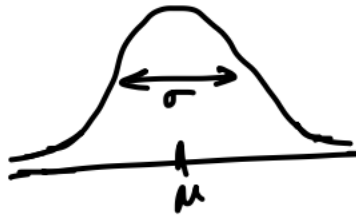


Figure 1:  $\mu$

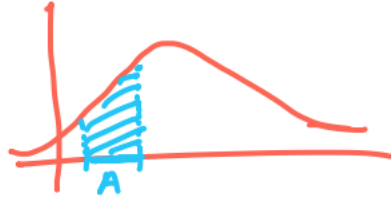


Figure 2:  $A$

$$x \mapsto \frac{1}{\sqrt{2\pi r^2}} \exp\left(\frac{-(x-\mu)^2}{2r^2}\right)$$

$$P : A_r \rightarrow [0, 1], P(A) := \int_A f_{\mu,r}(x) dx$$

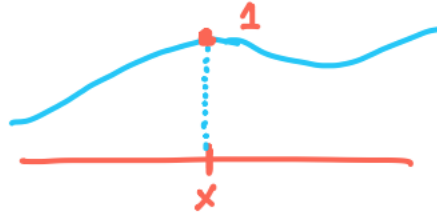


Figure 3: Dirac measure

## 2 Different Types of Probability Measures

**Definition 1** Discrete measure:

$\Omega = \{x_1, x_2, \dots\}$  finite and countable

$$A_r = P(\Omega)$$

We define a probability measure  $P : A_r \rightarrow [0, 1]$  by assigning probabilities to the "elementary events":

$$P(\{x_i\}) =: P_i$$

with  $0 \leq P_i \leq 1, \sum_i P_i = 1$

For  $A \in A_r$  we assign

$$P(A) = \sum_{\{i|x_i \in A\}} P_i.$$

Examples: a coin toss, distribution on  $Q$

**Definition 2** Dirac measure:

For  $x \in \mathbb{R}$ , we define the Dirac measure  $\delta_x$  on  $(\mathbb{R}, B(\mathbb{R}))$  by setting  $\delta_x(A) = \begin{cases} 1 & x \in A \\ 0 & \text{otherwise} \end{cases}$  sometimes this is called a point mass at a point  $x$ . A discrete measure on  $\mathbb{R}$  can be written as a sum of Dirac measures. For example, throwing a die can be considered as

$$\frac{1}{6}(\delta_1 + \delta_2 + \dots + \delta_6)$$

Measures with a density

Consider  $(\mathbb{R}^n, B(\mathbb{R}^n))$  and the Lebesgue measure  $\lambda$ . Consider a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$  that is measurable and satisfies  $\int f d\lambda = 1 \implies \int f(x) dx = 1$ .

Then we define a measure  $\gamma$  on  $\mathbb{R}^n$  by setting, for all  $A \in A_r$ ,

$$\gamma(A) := \int_A f(x) dx$$

$\gamma$  is the probability measure on  $(\mathbb{R}^n, B(\mathbb{R}^n))$  with density  $f$ .



Figure 4:  $\mu(A) = 0 \implies \int_A f d\mu \equiv \gamma(A) = 0$



Figure 5:  $\gamma_A = \int_A f d\lambda$

Notation:  $\gamma = f * \lambda$

Question: Can we describe every probability measure on  $(\mathbb{R}^n, B(\mathbb{R}^n))$  in terms of density?

Answer: no!

Counterexample:  $\delta_0$  Dirac measure.

On the same measure space  $(\mathbb{R}^n, B(\mathbb{R}^n))$ , if we have two measures  $\lambda, \gamma$ .

Question:  $\gamma(A) = \int_A \varphi d\lambda$

Does  $\varphi$  exist?

Answer: No!

**Definition 1.** A probability measure on  $\gamma$  on  $(\mathbb{R}^n, B(\mathbb{R}^n))$  is called absolutely continuous with respect to another measure  $\mu$  on  $(\mathbb{R}^n, B(\mathbb{R}^n))$  if every  $\mu$ -null set is also a  $\gamma$ -null set

$$\forall B \in B(\mathbb{R}^n) : \mu(B) = 0 \implies \gamma(B) = 0.$$

Notation:  $\gamma \ll \mu$

$$\mu(A) = 0 \implies \int_A f d\mu \equiv \gamma(A) = 0$$

**Example:**  $N(0, 1) \ll \lambda$

$$\gamma_A = \int_A f d\lambda$$

**Example:**  $\delta_0 \not\ll \lambda$  because

$$\lambda(\{0\}) = 0 \text{ but } \delta_0(\{0\}) = 1$$

**Theorem 2.** (Radon-Nikodym): Consider two probability measures  $\gamma, \mu$  on  $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$ . Then the following two statements are equivalent:

(1)  $\gamma$  has a density w.r.t  $\mu$ .

(2)  $\gamma$  is absolutely continuous w.r.t  $\mu$ .

If  $\gamma \ll \mu$ , then  $\exists \phi$  such that  $\delta(A) = \int_A \phi d\mu$ ,  $\phi$  exists and is unique.

**Proof idea:**

(1)  $\implies$  (2) easy

(2)  $\implies$  (1) We need to construct a density!

Consider the set  $G$  of all functions  $g$  with the following properties:

$$\textcircled{\star} \begin{cases} \bullet g \text{ is measurable, } g \geq 0 \\ \bullet g * \mu \leq \gamma, \text{ that is } \forall A \in \mathcal{B}(\mathbb{R}^n) : \int_A g d\mu \leq \gamma(A). \end{cases}$$

- Observe:  $g = 0$  satisfies  $\textcircled{\star}$ , so  $G$  is not empty.
- If  $g, h$  both satisfy  $\textcircled{\star}$ , then  $\sup(g, h)$  satisfies  $\textcircled{\star}$ .
- Define  $\xi := \sup_{g \in G} \int g d\mu$  and construct a sequence  $(g_n)_{n \in \mathbb{N}}$  such that  $\lim \int g_n d\mu = \xi$ .
- Define "density"  $f := \sup g_n$ .
- Now prove:  $f$  is the density that we are looking for.  $\square$