

## APA lecture 2

### Probability measures

$(\Omega = \text{fundamental set}, \mathcal{F} = \sigma\text{-field}) \leftarrow \begin{matrix} \text{measurable} \\ \text{space} \end{matrix}$

Def: A probability measure on  $(\Omega, \mathcal{F})$  is a mapping

$\mathbb{P}: \mathcal{F} \rightarrow [0, 1]$  such that

(i)  $\mathbb{P}(\emptyset) = 0$  and  $\mathbb{P}(\Omega) = 1$

(ii) finite additivity: if  $A_1, \dots, A_n \in \mathcal{F}$  are such that  $A_j \cap A_k = \emptyset$  for  $j \neq k$

$$\text{then } \mathbb{P}\left(\bigcup_{j=1}^n A_j\right) = \sum_{j=1}^n \mathbb{P}(A_j)$$

(ii')  $\sigma$ -additivity: if  $(A_n, n \geq 1) \subset \mathcal{F}$  is s.t.  $A_j \cap A_k = \emptyset$  for  $j \neq k$

$$\text{then } \mathbb{P}\left(\bigcup_{n \geq 1} A_n\right) = \sum_{n \geq 1} \mathbb{P}(A_n)$$

Terminology:  $(\Omega, \mathcal{F}, \mathbb{P})$  is a probability space

Properties: If  $P$  is a probability measure on  $(\Omega, \mathcal{F})$ , then:

(ii')

(iii) If  $A_1, \dots, A_n \in \mathcal{F}$ , then  $P\left(\bigcup_{j=1}^n A_j\right) \leq \sum_{j=1}^n P(A_j)$



(iii') If  $(A_n, n \geq 1) \subset \mathcal{F}$ , then  $P\left(\bigcup_{n \geq 1} A_n\right) \leq \sum_{n \geq 1} P(A_n)$

(iv) If  $A, B \in \mathcal{F}$  &  $A \subset B$ , then  $P(A) \leq P(B)$

and  $P(B \setminus A) = P(B) - P(A)$ . Also  $P(A^c) = 1 - P(A)$

(v) If  $A, B \in \mathcal{F}$ , then  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

(vi) If  $(A_n, n \geq 1) \subset \mathcal{F}$  is st.  $A_n \subset A_{n+1} \forall n \geq 1$ , then

$$P\left(\bigcup_{n \geq 1} A_n\right) = \lim_{n \rightarrow \infty} P(A_n)$$

(vii) If  $(A_n, n \geq 1) \subset \mathcal{F}$  is st.  $A_n \supset A_{n+1} \forall n \geq 1$ , then

$$P\left(\bigcap_{n \geq 1} A_n\right) = \lim_{n \rightarrow \infty} P(A_n)$$

Example:  $\Omega = \{1..6\}$ ,  $\mathcal{F} = \mathcal{P}(\Omega)$

$$\bullet P_1(\{i\}) = \frac{1}{6}, \quad P_1(A) = \sum_{i \in A} P_1(\{i\}) = \frac{\#A}{6} \quad A \in \mathcal{F}$$

$$\text{ex: } P_1(\{1,3,5\}) = \sum_{i \in \{1,3,5\}} \frac{1}{6} = \frac{3}{6} = \frac{1}{2}$$

$$\bullet P_2(\{i\}) = 0 \quad \text{if } i \in \{1..5\} \quad \& \quad P_2(\{6\}) = 1$$

$$P_2(A) = \sum_{i \in A} P_2(\{i\}) = \begin{cases} 1 & \text{if } 6 \in A \\ 0 & \text{otherwise} \end{cases}$$

$$\Omega = \{1..6\}, \quad \mathcal{F} = \left\{ \emptyset, \{1\}, \{2..6\}, \Omega \right\}$$

$$\begin{array}{cccc} \uparrow & \uparrow & \uparrow & \uparrow \\ 0 & p & 1-p & 1 \end{array} \quad \leftarrow \text{possible } P$$

Example:  $\Omega = [0, 1]$

•  $\mathcal{F} = \sigma(I_1, \dots, I_n) \quad \bigcup_{j=1}^n I_j = [0, 1], I_j \cap I_k = \emptyset$

•  $\mathcal{F} = \mathcal{B}([0, 1]) = \sigma(\{]a, b[, 0 \leq a < b \leq 1\}, \{0\}, \{1\})$

$\mathbb{P}(]a, b[) = b - a$  length of the interval  $]a, b[$

Carathéodory's extension theorem: There exists a unique extension of  $\mathbb{P}$  on  $\mathcal{B}([0, 1])$ , that is:

•  $\mathbb{P}(]a, b[) = b - a \quad \forall 0 \leq a < b \leq 1$

•  $\mathbb{P}$  is a probability measure on  $\mathcal{B}([0, 1])$

$\mathbb{P} =$  Lebesgue measure on  $[0, 1]$  (= uniform measure on  $[0, 1]$ )

## Lebesgue measure on $\mathbb{R}$ and $\mathbb{R}^2$

• on  $\mathbb{R}$ :  $\mathcal{B}(\mathbb{R}) = \sigma(\mathcal{I}_{a,b[}, a < b)$

$\mu(\mathcal{I}_{a,b[}) = b - a$  Lebesgue measure of  $\mathcal{I}_{a,b[}$

Carathéodory's extension theorem:  $\exists$  a unique extension of  $\mu$  on  $\mathcal{B}(\mathbb{R})$ .

Notation:  $\mu(B) = |B| \in [0, +\infty]$

• on  $\mathbb{R}^2$ :  $\mathcal{B}(\mathbb{R}^2) = \sigma(\mathcal{I}_{a,b[} \times \mathcal{I}_{c,d[} : a < b, c < d)$

$\mu(\mathcal{I}_{a,b[} \times \mathcal{I}_{c,d[}) = (b-a) \cdot (d-c)$

Carathéodory  $\Rightarrow \exists$  a unique extension of  $\mu$  on  $\mathcal{B}(\mathbb{R}^2)$



Notation:  $\mu(B) = |B|$

## Terminology:

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space

An event  $A \in \mathcal{F}$  is said to be negligible if  $\mathbb{P}(A) = 0$ ,  
resp. almost sure if  $\mathbb{P}(A) = 1$ .

Ex.:  $\Omega = [0, 1]$ ,  $\mathcal{F} = \mathcal{B}([0, 1])$ ,  $\mathbb{P} =$  Lebesgue measure

$\Rightarrow$  every singleton  $\{x\}$  is negligible:  $([0, 1] \setminus \{x\})$  is almost sure

$$\{x\} = \bigcap_{n \geq 1} ]x - \frac{1}{n}, x + \frac{1}{n}[ , \text{ so } \mathbb{P}(\{x\}) = \lim_{n \rightarrow \infty} \frac{1}{2n} = 0$$

•  $\Omega = \{1..5\}$ ,  $\mathcal{F} = \mathcal{P}(\Omega)$ ,  $\mathbb{P}_2(A) = \begin{cases} 1 & \text{if } 6 \in A \\ 0 & \text{otherwise} \end{cases}$

$\Rightarrow A = \{1..5\}$  is negligible:  $\mathbb{P}_2(A) = 0$ .

&  $A = \{6\}$  is almost-sure:  $\mathbb{P}_2(A) = 1$

## Proposition

• Let  $(A_n, n \geq 1)$  be a collection of negligible sets

Then  $\bigcup_{n \geq 1} A_n$  is also negligible.

• Let  $(B_n, n \geq 1)$  be a collection of almost-sure sets

Then  $\bigcap_{n \geq 1} B_n$  is also almost sure.

Proof: •  $P\left(\bigcup_{n \geq 1} A_n\right) \leq \sum_{n \geq 1} \underbrace{P(A_n)}_{=0} = 0$

• Take  $A_n = B_n^c$  #

Consequence:  $A = \mathbb{Q} \cap [0, 1]$  is negligible w.r.t.  
the Lebesgue measure.

## Distribution of a random variable

Def: Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space

$X: \Omega \rightarrow \mathbb{R}$  be an  $\mathcal{F}$ -measurable random variable

The distribution of  $X$  is the mapping  $\mu_X: \mathcal{B}(\mathbb{R}) \rightarrow [0, 1]$  defined as:

$$\begin{aligned} \mu_X(B) &= \mathbb{P}(\underbrace{\{\omega \in \Omega : X(\omega) \in B\}}_{= \{x \in B\} = X^{-1}(B)} \in \mathcal{F}) \quad B \in \mathcal{B}(\mathbb{R}) \end{aligned}$$

Note:  $(\mathbb{R}, \mathcal{B}(\mathbb{R}), \mu_X)$  is another probability space.



## Cumulative distribution function (cdf)

Def: The cdf of a random variable  $X$  is the mapping  $F_X: \mathbb{R} \rightarrow [0,1]$  defined as

$$\begin{aligned} F_X(t) &= \mathbb{P}(\{X \leq t\}) \quad t \in \mathbb{R} \\ &= \mu_X([-\infty, t]) \end{aligned}$$

Fact: The knowledge of  $\mu_X$  is equivalent to the knowledge of  $F_X$ .

$$\text{e.g.: } \mu_X([a, b]) = F_X(b) - F_X(a)$$

## Properties:

$$(i) \lim_{t \rightarrow -\infty} F_X(t) = 0, \quad \lim_{t \rightarrow +\infty} F_X(t) = 1$$

(ii)  $F_X$  is non-decreasing: if  $a \leq b$ , then  $F_X(a) \leq F_X(b)$

(iii)  $F_X$  is right-continuous:  $\lim_{\varepsilon \downarrow 0} F_X(t+\varepsilon) = F_X(t) \quad \forall t \in \mathbb{R}$

## Fact:



Any function  $F: \mathbb{R} \rightarrow [0, 1]$  satisfying properties (i), (ii) and (iii) is the cdf of a random variable  $X$ .

## Rmks:

- $F_X$  is not necessarily left-continuous
- $F_X$  has at most a countable number of jumps

# 1. Discrete random variables

Def:  $X$  is a discrete random variable if there exists a finite or countable set  $D$  such that  $X(\omega) \in D \quad \forall \omega \in \Omega$ .

Def: for  $x \in D$ ,  $p_x = P(\{X=x\})$

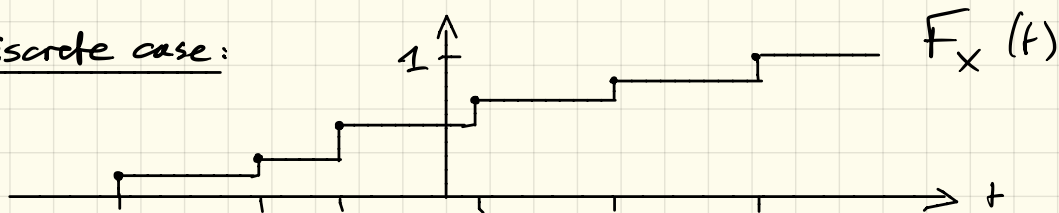
The sequence of numbers  $(p_x, x \in D)$  is called the probability mass function (pmf) of  $X$

Properties:  $0 \leq p_x \leq 1 \quad \forall x \in D$ ,  $\sum_{x \in D} p_x = 1$

$$P_X(B) = \sum_{x \in D \cap B} p_x \quad \forall B \in \mathcal{B}(\mathbb{R})$$

$$F_X(t) = \sum_{x \in D, x \leq t} p_x \quad \forall t \in \mathbb{R}$$

Discrete case:



## 2. Continuous random variables

Def:  $\left\{ \begin{array}{l} X \text{ is continuous if } P(\{X \in B\}) = 0 \\ (\forall B \in \mathcal{B}(\mathbb{R}) \text{ such that } \underbrace{|B|}_{= \text{Lebesgue measure of } B}) = 0 \end{array} \right.$

In particular,  $P(\{X = x\}) = 0 \quad \forall x \in \mathbb{R}$

Theorem: (Radon-Nikodym) Borel-measurable

In this case, there exists a function  $p_X: \mathbb{R} \rightarrow \mathbb{R}_+$  such that  $M_X(B) = \int_B p_X(x) dx \quad \forall B \in \mathcal{B}(\mathbb{R})$ .

Terminology:  $p_x$  is the probability density function (pdf) of  $X$

Properties:  $\cdot p_x(x) \geq 0 \quad \forall x \in \mathbb{R}$  (but may be  $> 1$ )

$$\cdot \int_{\mathbb{R}} p_x(x) dx = p_x(\mathbb{R}) = \mathbb{P}(\{X \in \mathbb{R}\}) = 1$$

$$F_x(t) = \int_{-\infty}^t p_x(x) dx \quad t \in \mathbb{R}$$

So  $F_x$  is continuous & "differentiable":  $F_x'(t) = p_x(t)$



⚠  $p_x(x) \neq \mathbb{P}(\{X=x\})$  (which is equal to 0  $\forall x \in \mathbb{R}$ )

## Change of variable formula

Let  $X$  be a random variable with cdf  $F_X$

$f: \mathbb{R} \rightarrow \mathbb{R}$  increasing (i.e. if  $x_1 < x_2$ , then  $f(x_1) < f(x_2)$ )

$Y = f(X)$ . What is  $F_Y$ ?

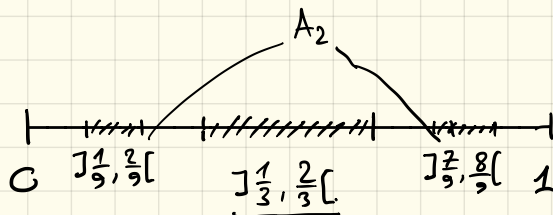
$$\begin{aligned} F_Y(t) &= \mathbb{P}(\{Y \leq t\}) = \mathbb{P}(\{f(X) \leq t\}) = \mathbb{P}(\{X \leq f^{-1}(t)\}) \\ &= F_X(f^{-1}(t)) \quad t \in \mathbb{R} \end{aligned}$$

If in addition  $X$  is a continuous r.v. and  $f$  is differentiable:

$$p_Y(t) = \frac{d}{dt} F_Y(t) = \frac{d}{dt} F_X(f^{-1}(t)) = p_X(f^{-1}(t)) \cdot \underbrace{\frac{1}{f'(f^{-1}(t))}}_{t \in \mathbb{R}}$$

# A strange cdf

Intro: the Cantor set



$$\rightarrow C = [0, 1] \setminus \left( \bigcup_{n \geq 1} A_n \right) \in \mathcal{B}([0, 1])$$

$$\text{What is } |C|? \quad |C| = 1 - \sum_{n \geq 1} |A_n|$$

$$|A_1| = \frac{1}{3}, \quad |A_2| = \frac{2}{9}, \quad |A_3| = \frac{4}{27} \quad \dots \quad |A_n| = \frac{2^{n-1}}{3^n}$$

$$|C| = 1 - \frac{1}{2} \sum_{n \geq 1} \left( \frac{2}{3} \right)^n = 1 - \frac{1}{2} \left( \frac{1}{1 - \frac{2}{3}} - 1 \right) = 1 - \frac{1}{2} \cdot 2 = 0$$

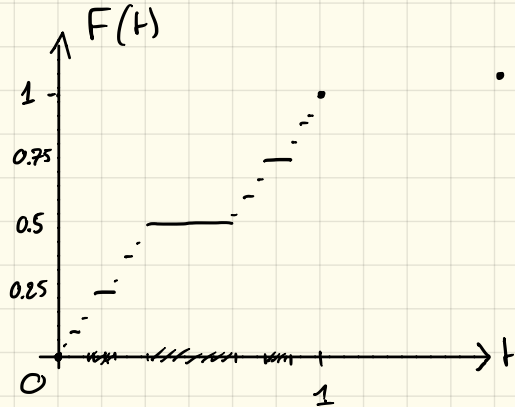
$$\begin{cases} x \in \underline{C} & \text{iff } x = \sum_{n \geq 1} \frac{a_n}{3^n} \text{ where } a_n \in \{0, 2\} \\ x \in [0, 1] & \text{iff } x = \sum_{n \geq 1} \frac{b_n}{2^n} \text{ where } b_n \in \{0, 1\} \end{cases} \Rightarrow C \text{ is } \underline{\text{uncountable}}$$

# The devil's stair case

Question: Does there exist a function  $F: [0,1] \rightarrow [0,1]$   
such that:  $F(0)=0$  &  $F(1)=1$  ✓

•  $F$  is continuous ✓

•  $|\{t \in [0,1] : F'(t)=0\}| = 1$  ? Yes! ✓



$$F'(t) = 0 \quad \forall t \in \bigcup_{n \geq 1} A_n$$

$F(t)$  is a cdf!

•  $X$  is discrete? no ( $F$  continuous) •  $X$  is continuous? no ( $p=F'=0$  almost everywhere)