Probability

 ${\bf Martin~Azpillaga}$ 

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## Contents

Ι	Definitions	3
1	Random Variables	5
	Probability Space	7
	Conditioned Probability	7
	Independency	7
	Random Variable	8
	Law	8
	Distribution Function	8
	Discrete Random Variable	9
	Probability Mass function	9
	Density function	10
	Absolutely Continuous random variable	10
	Expected Value	11
	Varaible change	12
2	Random Vectors	12
	Aleatory vector	15
	Law	15
	Distribution Function	15
	ith Marginal distribution function	16
	Discrete Random Variable	16
	Probability Mass function	17
	Density function	18
	Absolutely Continuous random variable	18
	Expected Value	19
	Varaible change	20

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II Propositions	20
III Examples	21
IV Exercises	23

### Block I

## **Definitions**

### 1. Random Variables

The study of probabilities was started by Blaise Pascal.

#### **Probability Space**

Let:

- $\cdot \Omega$  set
- $\cdot \mathcal{A}\sigma$ -algebra

$$\cdot \mathcal{P} : \mathcal{A} \to [0,1]$$

Then,  $(\Omega, \mathcal{A}, \mathcal{P})$  is a probability space if:

$$\cdot \mathcal{P}(\Omega) = 1$$

·  $\forall \{A_i\}_{i\geq 0} \subset \mathcal{A}$  ,  $\{A_i\}_{i\geq 0}$  mutually disjoint :

$$\mathcal{P}(\bigcup_{i\geq 0} A_i) = \sum_{i\geq 0} \mathcal{P}(A_i)$$

#### Conditioned Probability

Let:

$$\cdot (\Omega, \mathcal{A}, \mathcal{P})$$
 probability space  $B \in \mathcal{A}$ 

We call probability conditioned by B over  $(\Omega, \mathcal{A}, \mathcal{P})$  to:

$$\begin{array}{ccc} \mathcal{P}|_B : \mathcal{A} & \longrightarrow & [0,1] \\ A & \longmapsto & \frac{\mathcal{P}(A \cap B)}{\mathcal{P}(B)} \end{array}$$

#### Independency

Let:

$$\cdot (\Omega, \mathcal{A}, \mathcal{P})$$
 probability space  $\{A_i\}_{i=1}^n \subset \mathcal{A}$ 

Then,  $\{A_i\}_{i=1}^n$  is an independent family if:

$$\forall \{k_i\}_{i=1}^r \subset [0, n]_{\mathbb{N}}:$$

$$\mathcal{P}(\bigcap_{i=1}^r A_{k_i}) = \prod_{i=1}^n A_{k_i}$$

#### Random Variable

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $X: \Omega \to \mathbb{R}$

Then, X is a random variable if:

 $\cdot \ \forall \ A \in \mathbb{B}(\mathbb{R})$ :

$$X^{-1}(A) \in \mathcal{A}$$

#### Law

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $\cdot X : \Omega \to \mathbb{R}$  random variable

We call law of X to:

$$\begin{array}{ccc}
\mathcal{P}_X : \mathbb{B}(\mathbb{R}) & \longrightarrow & [0,1] \\
A & \longmapsto & \mathcal{P}(X^{-1}(A))
\end{array}$$

#### **Distribution Function**

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $\cdot X : \Omega \to \mathbb{R}$  random variable

We call  $\$ cumulative distribution function of X to:

$$\begin{array}{ccc}
\cdot & F_x : \mathbb{R} & \longrightarrow & [0,1] \\
& x & \longmapsto & \mathcal{P}(X \le x)
\end{array}$$

#### Discrete Random Variable

Let:

 $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space

 $X: \Omega \to \mathbb{R}$  random variable

Then, X is a discrete random variable if:

$$\cdot X(\Omega) \stackrel{\scriptscriptstyle \subset}{\scriptstyle \sim} \mathbb{N}$$

#### **Probability Mass function**

Let:

 $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space

 $\cdot X : \Omega \to \mathbb{R}$  discrete random variable

We call probability mass function of X to:

$$\begin{array}{ccc} p: X(\Omega) & \longrightarrow & [0,1] \\ x & \longmapsto & P(X=x) \end{array}$$

#### **Density function**

Let:

$$f: \mathbb{R} \to \mathbb{R}$$

Then, f is a density function if:

- $\cdot f \ge 0$
- $\cdot f\mathcal{R}(\mathbb{R})$
- $\int_{\mathbb{R}} f(x) dx = 1$

#### Absolutely Continuous random variable

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $\cdot X : \Omega \to \mathbb{R}$  random variable
- $\cdot F_X$  distribution function of X

Then, X is an absolutely continuous random variable if:

$$\cdot \exists f \in \mathcal{F}(\mathbb{R}, \mathbb{R}) \quad \forall x \in \mathbb{R}$$
:

$$F_x(x) = \int_{-\infty}^x f(x) dx$$

We denote:

 $\cdot\,X$  an absolutely continuous random variable : X abs cont

#### **Expected Value**

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $\cdot \, X \, : \, \Omega \to \mathbb{R}$  discrete random variable
- $\cdot\, p$  probability mass function of X
- $X': \Omega \to \mathbb{R} \text{ abs cont}$
- $\cdot f_X$  density function of X'

Then, X is expectable if:

$$\cdot \sum_{x \in X(\Omega)} |x| p(x) \in \mathbb{R}$$

We call expected value of X to:

$$\cdot \sum_{x \in X(\Omega)} x p(x)$$

Then, X' is expectable if:

$$\cdot \int_{\mathbb{R}} |x| f(x) dx \in \mathbb{R}$$

We call expected value of X' to:

$$\cdot \int_{\mathbb{R}} x f(x) dx$$

We denote:

· expected value of X : E(X)

#### Varaible change

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $X: \Omega \to \mathbb{R} \text{ abs cont}$

$$I = \mathring{I} \subset \mathbb{R} \mid_{\mathsf{I}} \mathcal{P}(X \in I) = 1$$

$$\cdot J = \mathring{J} \subset \mathbb{R}$$

$$g: I \to J$$

Then, g is a variable change over X if:

- $\cdot g$  biyective
- $g, g^{-1} \in \mathcal{C}^1$

### 2. Random Vectors

Generalization of random variables to n dimension

#### Aleatory vector

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $X: \Omega \to \mathbb{R}^m$

Then, X is a random vector if:

- $\cdot \ \forall \ i \in [1, m]_{\mathbb{N}}$ :
  - $X_i$  random variable

#### Law

Let:

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- $X: \Omega \to \mathbb{R}^m$  random vector

We call  $\$ cumulative distribution function of X to:

$$\begin{array}{ccc} F_x : \mathbb{R}^m & \longrightarrow & [0,1] \\ x & \longmapsto & \mathcal{P}(X \le x) \end{array}$$

#### ith Marginal distribution function

Let:

- $\cdot (\Omega, \mathcal{A}, \mathcal{P})$  probability space
- $X: \Omega \to \mathbb{R}^m$  random vector
- $i \in [1, m]_{\mathbb{N}}$

We call ith marginal distribution function of X to:

$$f: \mathbb{R} \longrightarrow [0,1]$$

$$x \longmapsto \mathcal{P}(X_i \le x)$$

We denote:

$$\cdot f : F_i$$

#### Discrete Random Variable

Let:

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Then, X is a discrete random variable if:

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III Examples 21

### Block II

# Propositions

IV Exercises 23

### **Block III**

Examples

IV Exercises 25

### Block IV

### **Exercises**