



Probability and Stochastic Processes

Discrete Random Variables, Continuous Random Variables, Multiple Random Variables

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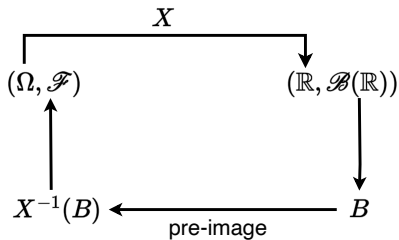
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Recap: Random Variable and CDF

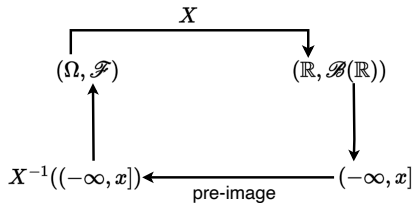
Random Variable



$$\forall B \in \mathcal{B}(\mathbb{R}), \quad X^{-1}(B) \in \mathcal{F}$$

Figure: A pictorial representation of the definition of a random variable X

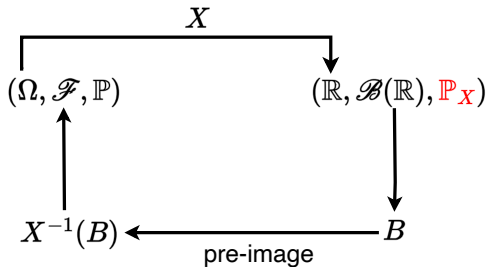
Equivalent Definition of Random Variable



$$\forall x \in \mathbb{R}, \quad X^{-1}((-\infty, x]) \in \mathcal{F}$$

Figure: Simplified, yet equivalent, definition of random variable

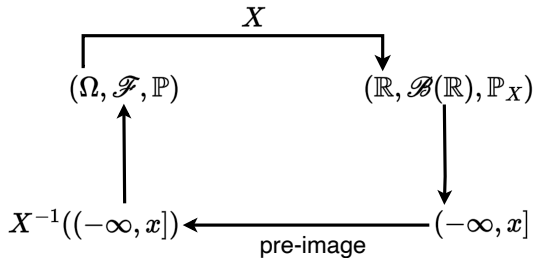
Probability Law



$$\mathbb{P}_X(B) = \mathbb{P} \circ X^{-1}(B) = \mathbb{P}(X^{-1}(B)) \quad \forall B \in \mathcal{B}(\mathbb{R})$$

Figure: Pictorial representation of probability law

Cumulative Distribution Function (CDF)



$$\textcolor{red}{F}_X(x) = \mathbb{P}_X((-\infty, x]) = \mathbb{P}(X^{-1}((-\infty, x])), \quad x \in \mathbb{R}$$

CDF \longleftrightarrow Probability Law

- If we know $\mathbb{P}_X = \{\mathbb{P}_X(B) : B \in \mathcal{B}(\mathbb{R})\}$, then we can extract the CDF $F_X : \mathbb{R} \rightarrow [0, 1]$ by using the formula

$$F_X(x) = \mathbb{P}_X((-\infty, x]), \quad x \in \mathbb{R}.$$

- Given the CDF $F_X : \mathbb{R} \rightarrow [0, 1]$, let

$$\mathbb{P}_X((-\infty, x]) = F_X(x), \quad x \in \mathbb{R}.$$

Then, by **Caratheodory's extension theorem**, \mathbb{P}_X can be extended to all $B \in \mathbb{R}$

Properties of CDF

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

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable with respect to \mathcal{F} with CDF F_X

- $\lim_{x \rightarrow -\infty} F_X(x) = 0, \quad \lim_{x \rightarrow +\infty} F_X(x) = 1$
- (**Monotonicity**) If $x \leq y$, then $F_X(x) \leq F_X(y)$
- (**Right-Continuity**) F_X is right-continuous, i.e., for all $x \in \mathbb{R}$,

$$\lim_{\varepsilon \downarrow 0} F_X(x + \varepsilon) = F_X(x).$$

Continuity of CDF \longleftrightarrow Zero Mass

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

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable with respect to \mathcal{F} with CDF F_X

- For any $x \in \mathbb{R}$,

$$\lim_{\varepsilon \downarrow 0} F_X(x - \varepsilon) = \mathbb{P}(\{X < x\}).$$

- F_X is continuous at a point $x \in \mathbb{R}$ if and only if $\mathbb{P}(\{X = x\}) = 0$

Properties of CDF

Lemma

Suppose that $F : \mathbb{R} \rightarrow [0, 1]$ satisfies

- $\lim_{x \rightarrow -\infty} F(x) = 0, \quad \lim_{x \rightarrow +\infty} F(x) = 1.$
- $x \leq y \implies F(x) \leq F(y).$
- $\lim_{\varepsilon \downarrow 0} F(x + \varepsilon) = F(x)$ for all $x \in \mathbb{R}.$

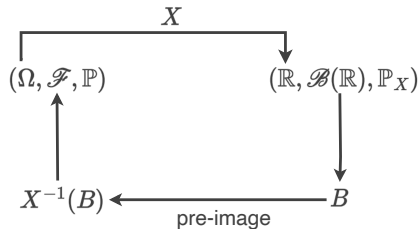
Then, there exists a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ and an \mathcal{F} -measurable random variable $X : \Omega \rightarrow \mathbb{R}$ such that $F = F_X$.

Set $\Omega = \mathbb{R}, \quad \mathcal{F} = \mathcal{B}(\mathbb{R}), \quad X(\omega) = \omega, \quad \mathbb{P}_X((-\infty, x]) = F(x), \quad \mathbb{P} = \mathbb{P}_X$

Starting with a function $F : \mathbb{R} \rightarrow [0, 1]$ satisfying the above properties, one can generate any probability measure on $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$

Discrete Random Variables

Discrete Random Variables



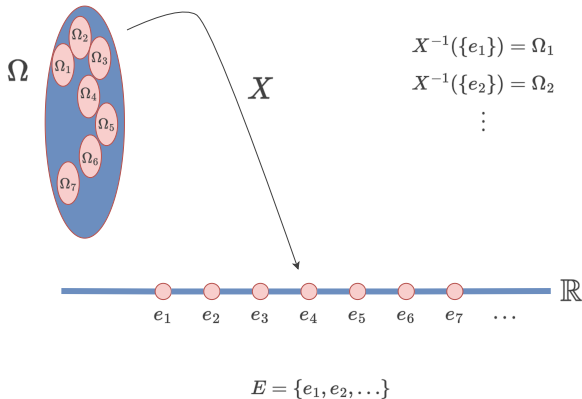
Definition (Discrete Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **discrete** if there exists a countable set $E \subset \mathbb{R}$, say $E = \{e_1, e_2, \dots\}$, such that $\mathbb{P}_X(E) = 1$.

$$1 = \mathbb{P}_X(E) = \sum_{i=1}^{\infty} \mathbb{P}_X(\{e_i\}) = \sum_{i=1}^{\infty} \mathbb{P}(\{X = e_i\});$$

$$\mathbb{P}_X(B) = \sum_{i: e_i \in B} \mathbb{P}_X(\{e_i\}).$$

Discrete Random Variable



$$\mathbb{P} \left(\bigcup_{i=1}^{\infty} \Omega_i \right) = \mathbb{P}_X(E) = 1$$

Probability Mass Function (PMF)

Definition (Probability Mass Function)

Given a random variable $X : \Omega \rightarrow \mathbb{R}$, the function $p_X : \mathbb{R} \rightarrow [0, 1]$ defined as

$$p_X(x) = \mathbb{P}_X(\{x\}) = \mathbb{P}(\{X = x\}), \quad x \in \mathbb{R},$$

is called the **probability mass function (PMF)** of X .

Remark

For a discrete random variable X taking values in a countable set $E = \{e_1, e_2, \dots\}$, the PMF p_X gives the **full probabilistic description** of X (i.e., $p_X \longleftrightarrow F_X$), and

$$\sum_{i=1}^{\infty} p_X(e_i) = 1.$$

CDF in Terms of PMF

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Definition (Probability Mass Function)

Let $X : \Omega \rightarrow \mathbb{R}$ be a discrete random variable taking values in a countable set $E = \{e_1, e_2, \dots\} \subset \mathbb{R}$. Then,

$$F_X(x) = \sum_{i: e_i \leq x} \mathbb{P}(\{X = e_i\}) = \sum_{i: e_i \leq x} p_X(e_i), \quad x \in \mathbb{R}.$$

Examples of Discrete Random Variables

Definition (Discrete Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **discrete** if there exists a countable set $E \subset \mathbb{R}$, say $E = \{e_1, e_2, \dots\}$, such that $\mathbb{P}_X(E) = 1$.

- $X \sim \text{Bernoulli}(p)$, $p \in [0, 1]$

$$E = \{0, 1\}, \quad p_X(x) = \begin{cases} p, & x = 1, \\ 1 - p, & x = 0, \\ 0, & \text{otherwise.} \end{cases}$$

- $X \sim \text{unif}(\{1, \dots, n\})$ for some fixed $n \in \mathbb{N}$

$$E = \{1, \dots, n\}, \quad p_X(x) = \begin{cases} \frac{1}{n}, & x \in \{1, \dots, n\}, \\ 0, & \text{otherwise.} \end{cases}$$

Examples of Discrete Random Variables

Definition (Discrete Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **discrete** if there exists a countable set $E \subset \mathbb{R}$, say $E = \{e_1, e_2, \dots\}$, such that $\mathbb{P}_X(E) = 1$.

- $X \sim \text{Geometric}(p)$, $p \in (0, 1]$

$$E = \mathbb{N}, \quad p_X(x) = \begin{cases} p(1-p)^{x-1}, & x \in \mathbb{N}, \\ 0, & \text{otherwise.} \end{cases}$$

- $X \sim \text{Binomial}(n, p)$ for some fixed $n \in \mathbb{N} \cup \{0\}$ and $p \in [0, 1]$

$$E = \{0, \dots, n\}, \quad p_X(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x}, & x \in \{0, \dots, n\}, \\ 0, & \text{otherwise.} \end{cases}$$

Examples of Discrete Random Variables

Definition (Discrete Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **discrete** if there exists a countable set $E \subset \mathbb{R}$, say $E = \{e_1, e_2, \dots\}$, such that $\mathbb{P}_X(E) = 1$.

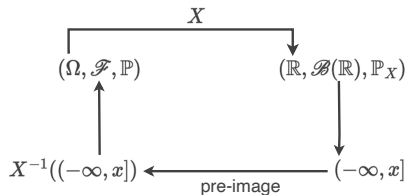
- $X \sim \text{Poisson}(\lambda)$, $\lambda > 0$

$$E = \{0, 1, 2, \dots\}, \quad p_X(x) = \begin{cases} e^{-\lambda} \frac{\lambda^x}{x!}, & x \in \{0, 1, 2, \dots\}, \\ 0, & \text{otherwise.} \end{cases}$$

- $E = \{1, 2, \dots\}, \quad p_X(x) = \begin{cases} \frac{6}{\pi^2} \frac{1}{x^2}, & x \in \{1, 2, \dots\}, \\ 0, & \text{otherwise.} \end{cases}$

Continuous Random Variables

Continuous Random Variables



Definition (Continuous Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **continuous** if there exists a function $f_X : \mathbb{R} \rightarrow [0, \infty)$ such that

$$F_X(x) = \mathbb{P}_X((-\infty, x]) = \int_{-\infty}^x f_X(t) dt, \quad \forall x \in \mathbb{R}.$$

Continuous Random Variables

Definition (Continuous Random Variable)

A random variable $X : \Omega \rightarrow \mathbb{R}$ is said to be **continuous** if there exists a function $f_X : \mathbb{R} \rightarrow [0, \infty)$ such that

$$F_X(x) = \mathbb{P}_X((-\infty, x]) = \int_{-\infty}^x f_X(t) dt, \quad \forall x \in \mathbb{R}.$$

Remarks:

- If $X : \Omega \rightarrow \mathbb{R}$ is a continuous random variable, its CDF F_X is **absolutely continuous** (hence continuous)
- The function f_X in the definition is called the **probability density function** (PDF) of the random variable X
- For a continuous random variable X , its PDF f_X provides the **full probabilistic description** of X (i.e., $f_X \longleftrightarrow F_X$)

Examples

- $X \sim \text{Uniform}([a, b])$,
$$f_X(x) = \begin{cases} \frac{1}{b-a}, & x \in [a, b], \\ 0, & \text{otherwise.} \end{cases}$$
- $X \sim \text{Exponential}(\lambda)$ for some fixed $\lambda > 0$,
$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0, \\ 0, & \text{otherwise.} \end{cases}$$
- $X \sim \text{Gaussian}(\mu, \sigma^2)$ for some fixed $\mu \in \mathbb{R}, \sigma > 0$,

$$f_X(x) = \frac{1}{\sigma \sqrt{2\pi}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right), \quad x \in \mathbb{R}.$$

- $X \sim \text{Normal} = \text{Gaussian}(0, 1)$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), \quad x \in \mathbb{R}.$$

Some Remarks

- PDF \neq probabilities; PDF can take values greater than 1
integrals of PDF = probability; total area under PDF = 1
- If X is a continuous random variable, then $\mathbb{P}_X(\{x\}) = 0$ for all $x \in \mathbb{R}$
Furthermore, if $E \subset \mathbb{R}$ is any countable set, then $\mathbb{P}_X(E) = 0$
Contrast this with discrete random variable!

Multiple Random Variables

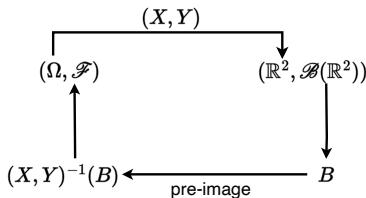
Two Random Variables

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Definition (Two Random Variables)

Given two \mathcal{F} -measurable random variables $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$, we say $(X, Y) : \Omega \rightarrow \mathbb{R}^2$ is a **random variable** with respect to \mathcal{F} if

$$(X, Y)^{-1}(B) = \{\omega \in \Omega : (X(\omega), Y(\omega)) \in B\} \in \mathcal{F} \quad \forall B \in \mathcal{B}(\mathbb{R}^2).$$



$$\forall B \in \mathcal{B}(\mathbb{R}^2), \quad (X, Y)^{-1}(B) \in \mathcal{F}$$

Joint Probability Law of Two Random Variables

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Definition (Joint Probability Law of Two Random Variables)

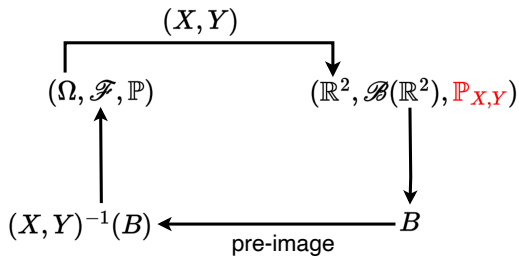
Given two random variables $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$ defined with respect to \mathcal{F} , their **joint probability law** $\mathbb{P}_{X,Y} : \mathcal{B}(\mathbb{R}^2) \rightarrow [0, 1]$, is the probability measure defined as

$$\mathbb{P}_{X,Y}(B) = \mathbb{P}(\{\omega \in \Omega : (X(\omega), Y(\omega)) \in B\}), \quad B \in \mathcal{B}(\mathbb{R}^2).$$

Remarks:

- $\mathbb{P}_{X,Y}$ is called the **pushforward** of \mathbb{P} under the random variable (X, Y)
- $\mathbb{P}_{X,Y}$ is the **probability law** of the random variable (X, Y)
- $\mathbb{P}_{X,Y}$ gives the **full probabilistic description** of (X, Y)

The Picture to Have in Mind



$$\mathbb{P}_{X,Y}(B) = \mathbb{P}((X, Y)^{-1}(B)) \quad \forall B \in \mathcal{B}(\mathbb{R}^2)$$

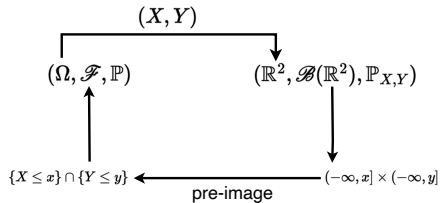
Remarks

- A special class of sets in $\mathcal{B}(\mathbb{R}^2)$ are semi-infinite rectangles of the form

$$(-\infty, x] \times (-\infty, y], \quad x, y \in \mathbb{R}.$$

- $\mathcal{B}(\mathbb{R}^2) = \sigma(\{(-\infty, x] \times (-\infty, y] : x, y \in \mathbb{R}\})$

Joint CDF of Two Random Variables



$$F_{X,Y}(x, y) = \mathbb{P}_{X,Y}((-\infty, x] \times (-\infty, y]) = \mathbb{P}(\{X \leq x\} \cap \{Y \leq y\}), \quad x, y \in \mathbb{R}$$

Definition (Joint CDF)

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Given random variables $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$ with respect to \mathcal{F} , their **joint CDF** $F_{X,Y} : \mathbb{R}^2 \rightarrow [0, 1]$ is defined as

$$F_{X,Y}(\mathbf{x}, \mathbf{y}) = \mathbb{P}_{X,Y}((-\infty, \mathbf{x}] \times (-\infty, \mathbf{y}]) = \mathbb{P}(\{X \leq \mathbf{x}\} \cap \{Y \leq \mathbf{y}\}), \quad \mathbf{x}, \mathbf{y} \in \mathbb{R}.$$

Notation

- $\{X \leq x\} \cap \{Y \leq y\} = \{X \leq x, Y \leq y\}$
- $\mathbb{P}(\{X \leq x\} \cap \{Y \leq y\}) = \mathbb{P}(X \leq x, Y \leq y)$

Joint CDF \longleftrightarrow Joint Probability Law

- If we know $\mathbb{P}_{X,Y} = \{\mathbb{P}_{X,Y}(B) : B \in \mathcal{B}(\mathbb{R}^2)\}$, then we can extract the CDF $F_{X,Y} : \mathbb{R}^2 \rightarrow [0, 1]$ by using the formula

$$F_{X,Y}(\mathbf{x}, \mathbf{y}) = \mathbb{P}_{X,Y}((-\infty, \mathbf{x}] \times (-\infty, \mathbf{y}]), \quad \mathbf{x}, \mathbf{y} \in \mathbb{R}.$$

- Given the joint CDF $F_{X,Y} : \mathbb{R}^2 \rightarrow [0, 1]$, let

$$\mathbb{P}_{X,Y}((-\infty, \mathbf{x}] \times (-\infty, \mathbf{y}]) = F_{X,Y}(\mathbf{x}, \mathbf{y}), \quad \mathbf{x}, \mathbf{y} \in \mathbb{R}.$$

Then, by **Caratheodory's extension theorem**, there exists a unique extension of $\mathbb{P}_{X,Y}$ to all Borel subsets of \mathbb{R}^2