



# Probability and Stochastic Processes

Jointly Continuous Random Variables, Joint PDF, Conditional PDF,  
Transformations of Random Variables

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# Jointly Continuous Random Variables

## Jointly Continuous Random Variables

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

Let  $X : \Omega \rightarrow \mathbb{R}$  and  $Y : \Omega \rightarrow \mathbb{R}$  be random variables defined with respect to  $\mathcal{F}$ .

### Definition (Jointly Continuous Random Variables)

$X$  and  $Y$  are said to be **jointly continuous** if  $(X, Y) : \Omega \rightarrow \mathbb{R}^2$  is a continuous random variable, i.e., there exists a function  $f_{X,Y} : \mathbb{R}^2 \rightarrow [0, +\infty)$  such that the joint CDF of  $X$  and  $Y$  may be expressed as

$$F_{X,Y}(x, y) = \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(u, v) dv du \quad \forall x, y \in \mathbb{R}.$$

The function  $f_{X,Y}$  is called the **joint PDF** of  $X$  and  $Y$ .

Remark:

$X$  continuous,  $Y$  continuous  $\not\Rightarrow X, Y$  jointly continuous

## Properties of Joint PDF

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- $$\int_{-\infty}^{+\infty} f_{X,Y}(u, v) du = f_Y(v) \text{ for all } v \in \mathbb{R}.$$

This says if  $X$  and  $Y$  are jointly continuous, then  $Y$  is a continuous RV

## Conditional CDF and Conditional PDF

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

Let  $X : \Omega \rightarrow \mathbb{R}$  and  $Y : \Omega \rightarrow \mathbb{R}$  be **jointly continuous** random variables defined with respect to  $\mathcal{F}$ .

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Remedy:

Fix  $y \in \mathbb{R}$  and  $\varepsilon > 0$  such that  $\mathbb{P}(\{Y \in (y - \varepsilon, y + \varepsilon)\}) > 0$ .

Define conditional probability with respect to the event  $\{Y \in (y - \varepsilon, y + \varepsilon)\}$ , and let  $\varepsilon \downarrow 0$ .

## Conditional CDF and Conditional PDF

$$\mathbb{P}(\{X \leq x\} | \{Y \in (y - \varepsilon, y + \varepsilon)\}) = \frac{\mathbb{P}(\{X \leq x\} \cap \{Y \in (y - \varepsilon, y + \varepsilon)\})}{\mathbb{P}(\{Y \in (y - \varepsilon, y + \varepsilon)\})}$$

## Conditional CDF and Conditional PDF

$$\begin{aligned}\mathbb{P}(\{X \leq x\} | \{Y \in (y - \varepsilon, y + \varepsilon)\}) &= \frac{\mathbb{P}(\{X \leq x\} \cap \{Y \in (y - \varepsilon, y + \varepsilon)\})}{\mathbb{P}(\{Y \in (y - \varepsilon, y + \varepsilon)\})} \\ &= \frac{\int_{-\infty}^x \int_{y-\varepsilon}^{y+\varepsilon} f_{X,Y}(u, v) \, dv \, du}{\int_{y-\varepsilon}^{y+\varepsilon} f_Y(v) \, dv}\end{aligned}$$

## Conditional CDF and Conditional PDF

$$\begin{aligned}\mathbb{P}(\{X \leq x\} | \{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\}) &= \frac{\mathbb{P}(\{X \leq x\} \cap \{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\})}{\mathbb{P}(\{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\})} \\&= \frac{\int_{-\infty}^x \int_{\gamma - \varepsilon}^{\gamma + \varepsilon} f_{X,Y}(u, v) \, dv \, du}{\int_{\gamma - \varepsilon}^{\gamma + \varepsilon} f_Y(v) \, dv} \\&\approx \frac{\int_{-\infty}^x f_{X,Y}(u, \gamma) \, du \cdot 2\varepsilon}{f_Y(\gamma) \cdot 2\varepsilon}\end{aligned}$$

## Conditional CDF and Conditional PDF

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 \mathbb{P}(\{X \leq x\} | \{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\}) &= \frac{\mathbb{P}(\{X \leq x\} \cap \{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\})}{\mathbb{P}(\{Y \in (\gamma - \varepsilon, \gamma + \varepsilon)\})} \\
 &= \frac{\int_{-\infty}^x \int_{\gamma - \varepsilon}^{\gamma + \varepsilon} f_{X,Y}(u, v) \, dv \, du}{\int_{\gamma - \varepsilon}^{\gamma + \varepsilon} f_Y(v) \, dv} \\
 &\approx \frac{\int_{-\infty}^x f_{X,Y}(u, \gamma) \, du \cdot 2\varepsilon}{f_Y(\gamma) \cdot 2\varepsilon} \\
 &= \int_{-\infty}^x \underbrace{\frac{f_{X,Y}(u, \gamma)}{f_Y(\gamma)}}_{\text{conditional PDF}} \, du
 \end{aligned}$$

## Conditional CDF for Jointly Continuous Random Variables

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

Let  $X : \Omega \rightarrow \mathbb{R}$  and  $Y : \Omega \rightarrow \mathbb{R}$  be **jointly continuous** random variables defined with respect to  $\mathcal{F}$ .

### Definition (Conditional CDF for Jointly Continuous Random Variables)

The **conditional CDF** of  $X$ , conditioned on the event  $\{Y = \gamma\}$ , is the function  $F_{X|Y=\gamma} : \mathbb{R} \rightarrow [0, 1]$  defined as

$$F_{X|Y=\gamma}(x) = \int_{-\infty}^x \frac{f_{X,Y}(u, \gamma)}{f_Y(\gamma)} du, \quad x \in \mathbb{R},$$

defined for all  $\gamma \in \mathbb{R}$  such that  $f_Y(\gamma) > 0$ .

## Conditional PDF for Jointly Continuous Random Variables

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

Let  $X : \Omega \rightarrow \mathbb{R}$  and  $Y : \Omega \rightarrow \mathbb{R}$  be **jointly continuous** random variables defined with respect to  $\mathcal{F}$ .

### Definition (Conditional PDF for Jointly Continuous Random Variables)

The **conditional PDF** of  $X$ , conditioned on the event  $\{Y = y\}$ , is the function  $f_{X|Y=y} : \mathbb{R} \rightarrow [0, +\infty)$  defined as

$$f_{X|Y=y}(x) = \frac{f_{X,Y}(x, y)}{f_Y(y)}, \quad x \in \mathbb{R},$$

defined for all  $y \in \mathbb{R}$  such that  $f_Y(y) > 0$ .

## Independence and Joint Continuity

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

Let  $X : \Omega \rightarrow \mathbb{R}$  and  $Y : \Omega \rightarrow \mathbb{R}$  be **jointly continuous** random variables defined with respect to  $\mathcal{F}$ .

### Definition (Joint Continuity and Independence)

$X$  and  $Y$  are **independent** if

$$f_{X,Y}(x, y) = f_X(x) \cdot f_Y(y) \quad \forall x, y \in \mathbb{R}.$$

Remark:

- $X \perp\!\!\!\perp Y \iff f_{X|Y=y} = f_X$  for all  $y$  such that  $f_Y(y) > 0$



## Conditional PMF Summary

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

Let  $X$  and  $Y$  be random variables defined with respect to  $\mathcal{F}$ .

- If  $X$  and  $Y$  are **jointly discrete**,

$$p_{X|Y=y}(x) = \frac{p_{X,Y}(x, y)}{p_Y(y)}, \quad x \in \mathbb{R}, p_Y(y) > 0.$$

- For any event  $A \in \mathcal{F}$ ,

$$\mathbb{P}(\{X \in A\} | \{Y = y\}) = \sum_{x \in A} p_{X|Y=y}(x).$$

- For any events  $A, B \in \mathcal{F}$ ,

$$\mathbb{P}(\{X \in A\} | \{Y \in B\}) = \frac{\mathbb{P}(\{X \in A\} \cap \{Y \in B\})}{\mathbb{P}(\{Y \in B\})} = \frac{\sum_{x \in A} \sum_{y \in B} p_{X,Y}(x, y)}{\sum_{y \in B} p_Y(y)}$$

## Conditional PDF Summary

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

Let  $X$  and  $Y$  be random variables defined with respect to  $\mathcal{F}$ .

- If  $X$  and  $Y$  are **jointly continuous**,

$$f_{X|Y=y}(x) = \frac{f_{X,Y}(x, y)}{f_Y(y)}, \quad x \in \mathbb{R}, f_Y(y) > 0.$$

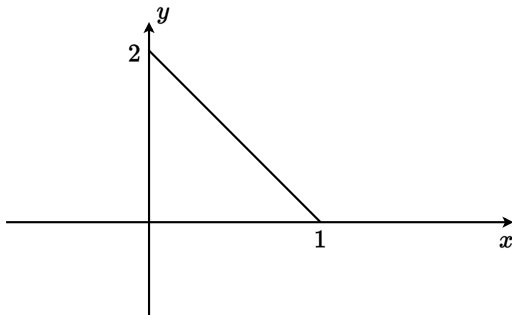
- For any event  $A \in \mathcal{F}$ ,

$$\mathbb{P}(\{X \in A\} | \{Y = y\}) = \int_A f_{X|Y=y}(u) du.$$

- For any events  $A, B \in \mathcal{F}$ ,

$$\mathbb{P}(\{X \in A\} | \{Y \in B\}) = \frac{\mathbb{P}(\{X \in A\} \cap \{Y \in B\})}{\mathbb{P}(\{Y \in B\})} = \frac{\int_{x \in A} \int_{y \in B} f_{X,Y}(x, y) dy dx}{\int_{y \in B} f_Y(y) dy}$$

## Example



Let  $f_{X,Y}(x, y) = 1$  inside the triangle, and 0 elsewhere.

Compute the marginal PDFs of  $X$  and  $Y$ , and the conditional PDF of  $X$  conditioned on  $\{Y = y\}$  for various values of  $y$ . Argue if  $X$  and  $Y$  are independent.

# Transformations of Random Variables

## Transformations of Random Variables

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

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

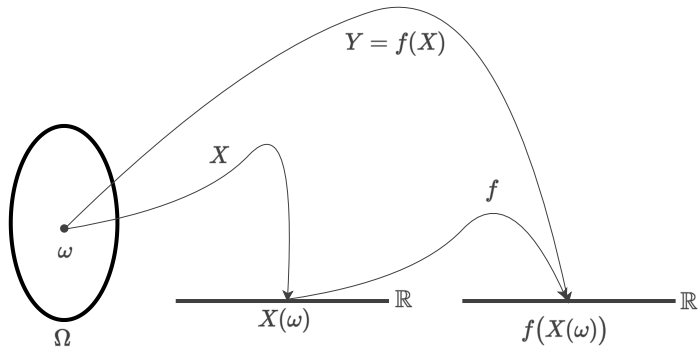
Consider  $f : \mathbb{R} \rightarrow \mathbb{R}$ .

Define  $Y : \Omega \rightarrow \mathbb{R}$  as  $Y = f(X)$ , i.e.,

$$Y(\omega) = f(X(\omega)), \quad \omega \in \Omega.$$

- For what functions  $f$  is  $Y = f(X)$  a random variable with respect to  $\mathcal{F}$ ?
- Given the CDF/PMF/PDF of  $X$ , what is the CDF/PMF/PDF of  $Y = f(X)$ ?

## Picture



## Borel-Measurable Functions – 1

### Definition (Borel-Measurable Function)

A function  $f : \mathbb{R} \rightarrow \mathbb{R}$  is said to be **Borel-measurable** if

$$f^{-1}(B) = \{x \in \mathbb{R} : f(x) \in B\} \in \mathcal{B}(\mathbb{R}) \quad \forall B \in \mathcal{B}(\mathbb{R}).$$

Remarks:

- Every continuous function is Borel-measurable. Thus,

$$f(x) = |x|, \quad f(x) = x^2, \quad f(x) = e^x, \quad f(x) = \log x,$$

are Borel-measurable

- $X : \Omega \rightarrow \mathbb{R}$  random variable wrt  $\mathcal{F}$   $f : \mathbb{R} \rightarrow \mathbb{R}$  Borel-measurable  
 $\implies f(X) : \Omega \rightarrow \mathbb{R}$  random variable with respect to  $\mathcal{F}$

## Borel-Measurable Functions – 2

### Definition (Borel-Measurable Function)

A function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is said to be **Borel-measurable** if

$$f^{-1}(B) = \{x \in \mathbb{R}^n : f(x) \in B\} \in \mathcal{B}(\mathbb{R}^n) \quad \forall B \in \mathcal{B}(\mathbb{R}^m).$$

Implication for  $m = 1$ :

- $X_1, \dots, X_n$  random variables wrt  $\mathcal{F}$ ,  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  Borel-measurable  
 $\implies f(X_1, \dots, X_n) : \Omega \rightarrow \mathbb{R}$  random variable with respect to  $\mathcal{F}$
- Every continuous function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is Borel-measurable. Thus, for instance,

$$f(x_1, \dots, x_n) = \sum_{i=1}^n x_i$$

is Borel-measurable



## Maximum of Random Variables

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

Let  $X_1, \dots, X_n$  be random variables defined with respect to  $\mathcal{F}$ , with joint CDF  $F_{X_1, \dots, X_n}$ .

- Show that  $Y_n = \max\{X_1, \dots, X_n\}$  is a random variable with respect to  $\mathcal{F}$ .
- Derive the CDF of  $Y_n$ .
- Simplify the CDF of  $Y_n$  when  $X_1, \dots, X_n$  are i.i.d..

## Minimum of Random Variables

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

Let  $X_1, \dots, X_n$  be random variables defined with respect to  $\mathcal{F}$ , with joint CDF  $F_{X_1, \dots, X_n}$ .

- Show that  $Z_n = \min\{X_1, \dots, X_n\}$  is a random variable with respect to  $\mathcal{F}$ .
- Derive the CDF of  $Z_n$ .
- Simplify the CDF of  $Z_n$  when  $X_1, \dots, X_n$  are i.i.d..

## Minimum of i.i.d. Exponential Random Variables

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

Let  $X_1, \dots, X_n$  be independent, with  $X_i \sim \text{Exponential}(\lambda_i)$  for each  $i \in \{1, \dots, n\}$ .

Find the distribution of  $Z = \min\{X_1, \dots, X_n\}$ .

## Sums of Random Variables

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

Let  $X$  and  $Y$  be random variables with respect to  $\mathcal{F}$ .

- Show that  $X + Y$  is a random variable with respect to  $\mathcal{F}$ .
- In the cases when  $X$  and  $Y$  are jointly discrete/continuous, derive the PMF/PDF of  $X + Y$ .
- Simplify the PMF/PDF when  $X$  and  $Y$  are independent.

## Sum of Two Independent Exponentials

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

Let  $X \sim \text{Exponential}(\mu_1)$  and  $Y \sim \text{Exponential}(\mu_2)$ . Assume  $X \perp\!\!\!\perp Y$ .

Determine the distribution of  $Z = X + Y$ .

## Sum of Random Number of Random Variables

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

Let  $\{X_i : i \in \mathbb{N}\}$  be a collection of **i.i.d.** random variables defined with respect to  $\mathcal{F}$  and having a common CDF  $F$ .

Let  $N$  be a positive integer-valued random variable defined with respect to  $\mathcal{F}$  and having the PMF  $p_N$ .

Let  $N$  be independent of  $\{X_i : i \in \mathbb{N}\}$ .

Consider the sum

$$S_N := \sum_{i=1}^N X_i;$$

$$S_N(\omega) = \sum_{i=1}^{N(\omega)} X_i(\omega), \quad \omega \in \Omega.$$

- Show that  $S_N : \Omega \rightarrow \mathbb{R}$  is a random variable with respect to  $\mathcal{F}$ .
- Determine the CDF of  $S_N$ .