



Mathematical Foundations for Data Science (Probability)

Independence of Random Variables, Jointly Discrete Random Variables, Joint PMF, Conditional PMF, Joint PDF, Conditional PDF, Transformations of Random Variables

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Independence of Two Random Variables

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

Definition (Independence of Two Random Variables)

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

$$\begin{aligned} X \perp\!\!\!\perp Y &\iff F_{X,Y}(x,y) = F_X(x) \cdot F_Y(y) \quad \forall x,y \in \mathbb{R} \\ &\iff \mathbb{P}(\{X \leq x\} \cap \{Y \leq y\}) = \mathbb{P}(\{X \leq x\}) \cdot \mathbb{P}(\{Y \leq y\}) \quad \forall x,y \in \mathbb{R} \\ &\iff \{X \leq x\} \perp\!\!\!\perp \{Y \leq y\} \quad \forall x,y \in \mathbb{R}. \end{aligned}$$

Implications:

- $\{X \leq x\} \perp\!\!\!\perp \{Y > y\}$ for all $x,y \in \mathbb{R}$, i.e.,

$$\mathbb{P}(\{X \leq x, Y > y\}) = \mathbb{P}(\{X \leq x\}) \cdot \mathbb{P}(\{Y > y\}) \quad \forall x,y \in \mathbb{R}$$

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Implications:

- $\{X > x\} \perp\!\!\!\perp \{Y > y\}$ for all $x, y \in \mathbb{R}$, i.e.,

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Implications:

- $\{X = x\} \perp\!\!\!\perp \{Y = y\}$ for all $x, y \in \mathbb{R}$, i.e.,

$$\mathbb{P}(\{X = x, Y = y\}) = \mathbb{P}(\{X = x\}) \cdot \mathbb{P}(\{Y = y\}) \quad \forall x, y \in \mathbb{R}$$

Example

Let X_1 and X_2 be distributed exponentially with parameters $\lambda_1 > 0$ and $\lambda_2 > 0$ respectively. Determine the distribution of $Z = \min\{X_1, X_2\}$.

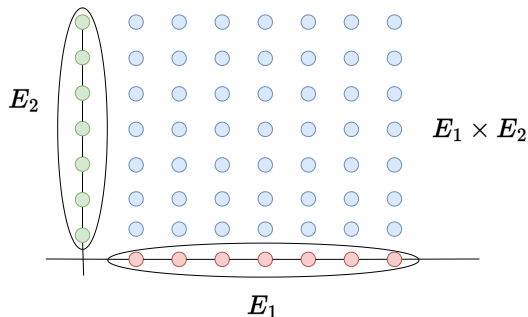
Jointly Discrete Random Variables

Jointly Discrete Random Variables

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

Definition (Jointly Discrete Random Variables)

Random variables $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$ defined with respect to \mathcal{F} are said to be **jointly discrete** if X and Y are individually discrete random variables.



Joint PMF

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

Definition (Joint PMF)

The joint PMF of jointly discrete random variables $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$ defined on \mathcal{F} is a function $p_{X,Y} : \mathbb{R}^2 \rightarrow [0, 1]$ defined as

$$p_{X,Y}(\mathbf{x}, \mathbf{y}) = \mathbb{P}(\{X = \mathbf{x}\} \cap \{Y = \mathbf{y}\}), \quad \mathbf{x}, \mathbf{y} \in \mathbb{R}.$$

Note:

$$\mathbb{P}(\{(X, Y) \in E_1 \times E_2\}) = \sum_{\mathbf{x} \in E_1} \sum_{\mathbf{y} \in E_2} p_{X,Y}(\mathbf{x}, \mathbf{y}) = 1,$$

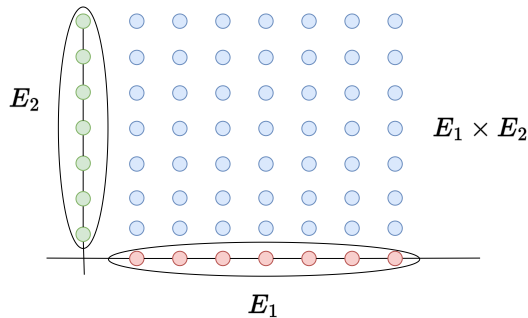
$$\mathbb{P}(\{(X, Y) \in B\}) = \sum_{(\mathbf{x}, \mathbf{y}) \in B \cap (E_1 \times E_2)} p_{X,Y}(\mathbf{x}, \mathbf{y}), \quad B \subseteq \mathbb{R}^2.$$

Properties of Joint PMF

- $\sum_{x \in E_1} \sum_{y \in E_2} p_{X,Y}(x, y) = 1.$

- $p_X(x) = \sum_{y \in E_2} p_{X,Y}(x, y), \quad x \in \mathbb{R}$

- $p_Y(y) = \sum_{x \in E_1} p_{X,Y}(x, y), \quad y \in \mathbb{R}$



Conditional PMF

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

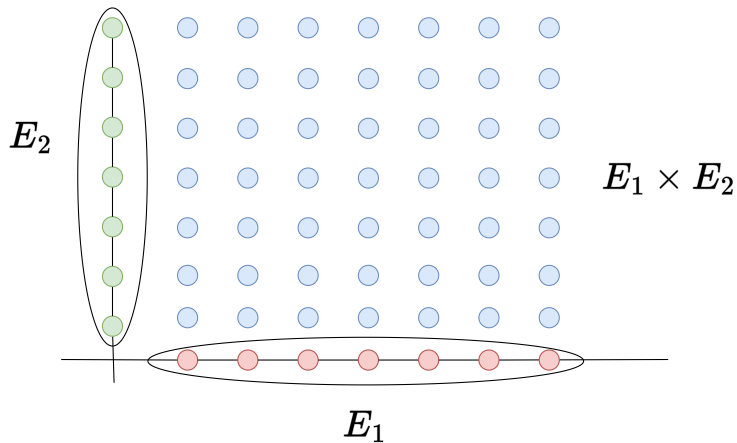
Definition (Conditional PMF)

Let X, Y be jointly discrete random variables defined with respect to \mathcal{F} . Fix $y \in \mathbb{R}$ such that $p_Y(y) = \mathbb{P}(\{Y = y\}) > 0$. The **conditional PMF** of X , conditioned on the event $\{Y = y\}$, is a function $p_{X|Y=y} : \mathbb{R} \rightarrow [0, 1]$ defined as

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

defined for all $y \in \mathbb{R}$ such that $p_Y(y) = \mathbb{P}(\{Y = y\}) > 0$.

Conditional PMF



Independence of Two Discrete Random Variables

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

Theorem

Let $X : \Omega \rightarrow \mathbb{R}$ and $Y : \Omega \rightarrow \mathbb{R}$ be discrete random variables with respect to \mathcal{F} . The following statements are equivalent.

1. $X \perp\!\!\!\perp Y$.
2. $\{X = x\} \perp\!\!\!\perp \{Y = y\}$ for all $x, y \in \mathbb{R}$.
3. $p_{X,Y}(x, y) = p_X(x) \cdot p_Y(y)$ for all $x, y \in \mathbb{R}$.
4. For all $y \in \mathbb{R}$ such that $p_Y(y) > 0$,

$$p_{X|Y=y}(x) = p_X(x) \quad \forall x \in \mathbb{R}.$$

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 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:

If X and Y are individually continuous, then they need not be jointly continuous.

Properties of Joint PDF

- $$\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_{X,Y}(u, v) dv du = 1.$$

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

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

Properties of Joint PDF

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

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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} .

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} .

Conditional CDF of X conditioned on $\{Y = y\}$: $\mathbb{P}(\{X \leq x\} | \{Y = y\})$.

However, this conditional probability is not defined because $\mathbb{P}(\{Y = y\}) = 0$.

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} .

Conditional CDF of X conditioned on $\{Y = y\}$: $\mathbb{P}(\{X \leq x\} | \{Y = y\})$.

However, this conditional probability is not defined because $\mathbb{P}(\{Y = y\}) = 0$.

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

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

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

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

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 and Conditional PDF – 1

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**,

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

Furthermore, for any event $A \in \mathcal{F}$,

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

Conditional PMF and Conditional PDF – 2

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

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

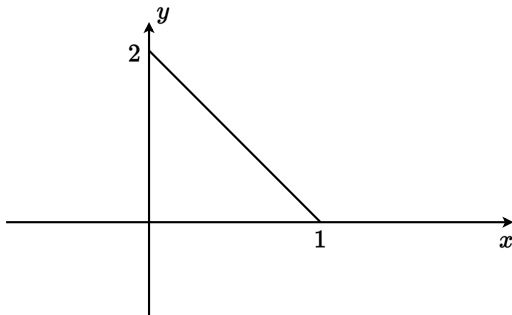
- If X and Y are **jointly continuous**,

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

Furthermore, for any event $A \in \mathcal{F}$,

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

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} .

Given a function $f : \mathbb{R} \rightarrow \mathbb{R}$, our interest is to characterise the CDF/PMF/PDF of the random variable $Y = f(X)$.

For ease of analysis, we shall consider functions f which are continuous and/or differentiable.

Examples

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

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable defined with respect to \mathcal{F} , with CDF F_X . Determine the CDF of $Y = aX + b$ for some $a, b \in \mathbb{R}$.

Examples

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

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable defined with respect to \mathcal{F} , with CDF F_X . Determine the CDF of $Y = X^2$.