



Mathematical Foundations for Data Science (Probability)

Probability Measures on Discrete Sample Spaces, Conditional Probability, Bayes' Theorem, Independence of Events, Random Variables

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Probability Measures on Discrete Spaces

Discrete Sample Spaces

Let Ω be a non-empty, discrete (finite or countably infinite) sample space. Then, Ω may be represented as

$$\Omega = \{\omega_1, \omega_2, \dots\}$$

In this case, we simply take $\mathcal{F} = 2^\Omega$

Probability Assignment for Discrete Sample Spaces

Given (Ω, \mathcal{F}) , we define $\mathbb{P} : \mathcal{F} \rightarrow [0, 1]$ as

$$\mathbb{P}(A) = \sum_{\omega \in A} \mathbb{P}(\{\omega\}), \quad A \in \mathcal{F},$$

while making sure that the assignment \mathbb{P} satisfies $\sum_{\omega \in \Omega} \mathbb{P}(\{\omega\}) = 1$.

Examples

- $\Omega = \{H, T\}, \quad \mathcal{F} = 2^\Omega = \left\{ \emptyset, \Omega, \{H\}, \{T\} \right\}$

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$$\mathbb{P}(\{H\}) = p = 1 - \mathbb{P}(\{T\}), \quad p \in [0, 1]$$

- $\Omega = \mathbb{N}, \quad \mathcal{F} = 2^\Omega$

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- $\mathbb{P}(\{k\}) = p(1-p)^{k-1}, \quad k \in \Omega \quad (p \in [0, 1], \text{Geometric measure})$

- $\Omega = \mathbb{N} \cup \{0\}, \quad \mathcal{F} = 2^\Omega$

- $\mathbb{P}(\{k\}) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad k \in \Omega \quad (\lambda > 0, \text{Poisson measure})$

Conditional Probabilities

Conditional Probability Measure

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

Definition (Conditional Probability

Given $B \in \mathcal{F}$ such that $\mathbb{P}(B) > 0$, define

$$\mathbb{P}_B : \mathcal{F} \rightarrow [0, 1] \quad \text{via} \quad \mathbb{P}_B(A) := \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}, \quad A \in \mathcal{F}.$$

Then, \mathbb{P}_B is a valid probability measure on (Ω, \mathcal{F}) , and is called the **conditional probability measure conditioned on the event B** .

Notation: $\mathbb{P}_B(A)$ is denoted more commonly as $\mathbb{P}(A|B)$.



\mathbb{P}_B is a Valid Probability Measure

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

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\mathbb{P}_B is a Valid Probability Measure

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

- $\mathbb{P}_B(\emptyset) = 0$
- $\mathbb{P}_B(\Omega) = 1$
- For any mutually disjoint collection of sets $A_1, A_2, \dots \in \mathcal{F}$,

$$\mathbb{P}_B \left(\bigcup_{i=1}^{\infty} A_i \right) = \sum_{i=1}^{\infty} \mathbb{P}_B(A_i).$$

Conditional Probability – Properties

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

- Fix $B \in \mathcal{F}$ such that $0 < \mathbb{P}(B) < 1$. Then, for any $A \in \mathcal{F}$,

$$\mathbb{P}(A) = \mathbb{P}(A|B) \cdot \mathbb{P}(B) + \mathbb{P}(A|B^c) \cdot \mathbb{P}(B^c).$$

Conditional Probability – Properties

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

- (Law of Total Probability)

Let $B_1, B_2, \dots \in \mathcal{F}$ be a *partition* of Ω , i.e., $B_i \cap B_j = \emptyset$ for all $i \neq j$, and $\bigcup_{i=1}^{\infty} B_i = \Omega$. Then, for any $A \in \mathcal{F}$,

$$\mathbb{P}(A) = \sum_{i=1}^{\infty} \mathbb{P}(A|B_i) \cdot \mathbb{P}(B_i).$$

Conditional Probability – Properties

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

- (Bayes' Theorem)

Let $B_1, B_2, \dots \in \mathcal{F}$ be as before. For any $A \in \mathcal{F}$ such that $\mathbb{P}(A) > 0$,

$$\mathbb{P}(B_i|A) = \frac{\mathbb{P}(A|B_i) \cdot \mathbb{P}(B_i)}{\sum_{j=1}^{\infty} \mathbb{P}(A|B_j) \cdot \mathbb{P}(B_j)}.$$

Conditional Probability – Properties

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

- (Decomposition Rule for Conditional Probabilities)

Let $A_1, A_2, \dots \in \mathcal{F}$. Then,

$$\begin{aligned}\mathbb{P}\left(\bigcap_{i=1}^{\infty} A_i\right) &= \mathbb{P}(A_1) \cdot \mathbb{P}(A_2|A_1) \cdot \mathbb{P}(A_3|A_1 \cap A_2) \cdots \\ &= \mathbb{P}(A_1) \cdot \prod_{i=2}^{\infty} \mathbb{P}\left(A_i \mid \bigcap_{j=1}^{i-1} A_j\right),\end{aligned}$$

provided each of the conditional probabilities on the right-hand side is defined.

Independence

Independence of Events

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

Definition (Independence of Events)

Events $A, B \in \mathcal{F}$ are said to be **independent** if

$$\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B).$$

We write $A \perp\!\!\!\perp B$ as a shorthand notation to denote that A and B are independent.

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Remarks:

- The definition of independence does not involve conditional probabilities
- If $\mathbb{P}(B) > 0$, then

$$A \perp\!\!\!\perp B \iff \mathbb{P}(A|B) = \mathbb{P}(A).$$

Independence of Events

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

Definition (Independence of Events)

- Events $A_1, A_2, \dots, A_n \in \mathcal{F}$ are said to be **independent** if for all $\mathcal{I}_0 \subseteq \{1, 2, \dots, n\}$,

$$\mathbb{P} \left(\bigcap_{i \in \mathcal{I}_0} A_i \right) = \prod_{i \in \mathcal{I}_0} \mathbb{P}(A_i).$$

- Let \mathcal{I} be an arbitrary index set. A collection of events $\{A_i : i \in \mathcal{I}\}$ is independent if for every **finite** subset $\mathcal{I}_0 \subseteq \mathcal{I}$, the collection of events $\{A_i : i \in \mathcal{I}_0\}$ is independent.

Random Variables

Random Variables

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

Definition (Random Variables)

A function $X : \Omega \rightarrow \mathbb{R}$ is called a **random variable** with respect to \mathcal{F} if

$$\{\omega \in \Omega : X(\omega) \leq x\} \in \mathcal{F} \quad \forall x \in \mathbb{R}.$$

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- If X is a random variable with respect to \mathcal{F} , it is called an **\mathcal{F} -measurable function**
- The definition of a random variable does not involve \mathbb{P}

Examples

- $\Omega = \{1, 2, \dots, 6\}, \quad \mathcal{F} = \left\{ \emptyset, \Omega \right\}, \quad X(\omega) = \omega$

Is X a random variable with respect to \mathcal{F} ?

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- $\Omega = \{1, 2, \dots, 6\}, \quad \mathcal{F} = \left\{ \emptyset, \Omega \right\}, \quad X(\omega) = \omega$

Is X a random variable with respect to \mathcal{F} ?

- What functions X are random variables with respect to \mathcal{F} ?

Examples

- $\Omega = \{1, 2, 3, 4, 5\}$, $\mathcal{F} = \left\{ \emptyset, \Omega, A, A^c \right\}$ for a fixed $A \subseteq \Omega$

What functions X are random variables with respect to \mathcal{F} ?

Examples

- $\Omega = \{1, 2, 3, 4, 5\}, \quad \mathcal{F} = \sigma \left(\left\{ \{1\}, \{2, 3\} \right\} \right)$
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Examples

- $\Omega = \mathbb{N}$, $\mathcal{F} = 2^\Omega$

What functions X are random variables with respect to \mathcal{F} ?