

Probability and Random Variable

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1 The Law of total probability

Definition 1 (Partition) A collection of events $\{B_1, B_2, \dots, B_n\}$ forms a **partition** of the sample space Ω if:

- $B_i \cap B_j = \emptyset$ for all $i \neq j$.
- $\bigcup_{k=1}^k B_i = \Omega$.

Theorem 1 (Law of Total Probability) Let $\{B_1, B_2, \dots, B_n\}$ be a partition of Ω with $P(B_i) > 0$ for all i . Then for any event A :

$$P(A) = \sum_{k=1}^k P(AB_k) = \sum_{k=1}^k P(A | B_k)P(B_k). \quad (1)$$

proof: Using the countable additivity axiom and conditional probability definition:

$$\begin{aligned} A &= \bigcup_{k=1}^k AB_k \\ (AB_k) \cap (AB_l) &= \emptyset \\ P(A) &= P\left(\bigcup_{k=1}^k AB_k\right) \\ &= \sum_{i=1}^n P(A \cap B_i) \\ &= \sum_{i=1}^n P(A|B_i)P(B_i) \end{aligned}$$

2 Bayesian Theorem

Theorem 2 (Bayes' Theorem) For any events $B_k (k = 1, 2, \dots)$ and A with $P(A) > 0$:

$$P(B_k|A) = \frac{P(A|B_k)P(B_k)}{\sum_{k=1}^k P(A|B_k)P(B_k)}. \quad (2)$$

Example 1 (Naive Bayes Spam Filter) *Suppose:*

- 70% of emails are spam ($P(\text{Spam}) = 0.7$).
- 20% of emails are spam ($P(\text{Low Priority}) = 0.2$).
- 10% of emails are spam ($P(\text{High Priority}) = 0.1$).
- The word "essay writing" appears in:
 - 90% of spam emails ($P(\text{"essay writing"}|\text{Spam}) = 0.9$).
 - 1% of legitimate emails ($P(\text{"essay writing"}|\text{Low Priority}) = 0.01$).
 - 1% of legitimate emails ($P(\text{"essay writing"}|\text{High Priority}) = 0.01$).

Calculate the probability that an email containing "essay writing" is spam.

Solution:

$$\begin{aligned}P(\text{Spam}|\text{"essay writing"}) &= \frac{P(\text{"essay writing"}|\text{Spam})P(\text{Spam})}{P(\text{"essay writing"})}, \\P(\text{"essay writing"}) &= P(e.w.|\text{Spam})P(\text{Spam}) + P(e.w.|\text{Low Priority})P(\text{Low Priority}) \\&\quad + P(e.w.|\text{High Priority})P(\text{High Priority}) \\&= 0.9 \times 0.7 + 0.01 \times 0.2 + 0.01 \times 0.1 \\&= 0.633, \\P(\text{Spam}|\text{"essay writing"}) &= \frac{0.9 \times 0.7}{0.633} = \frac{0.63}{0.633} \approx 0.9953.\end{aligned}$$

3 Continuity of Probability

Definition 2 (Monotonic Sequences) *A sequence of events $\{A_n\}$ is:*

- **Increasing** if $A_1 \subseteq A_2 \subseteq \dots$ (denoted $A_n \uparrow A$).
- **Decreasing** if $A_1 \supseteq A_2 \supseteq \dots$ (denoted $A_n \downarrow A$),

where $A = \lim_{n \rightarrow \infty} A_n$.

Theorem 3 (Continuity of Probability) *For any probability measure P :*

1. If $A_n \uparrow A$, then $\lim_{n \rightarrow \infty} P(A_n) = P(A)$.
2. If $A_n \downarrow A$ and $P(A_1) < \infty$, then $\lim_{n \rightarrow \infty} P(A_n) = P(A)$.

Proof for Increasing Sequence: Let $B_1 = A_1$, $B_k = A_k \setminus \bigcup_{i=1}^{k-1} A_i$ for $n \geq 2$.

Then:

$$\begin{aligned} \lim_{n \rightarrow \infty} P(A_n) &= \lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n B_i\right) = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(B_i) \\ &= P\left(\bigcup_{i=1}^{\infty} B_i\right) = P\left(\bigcup_{i=1}^{\infty} A_i\right) \\ &= P(A). \end{aligned}$$

4 Random Variable

Definition 3 (Random Variable) A *random variable* X is a measurable function from a probability space (Ω, \mathcal{F}, P) to $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$:

$$X : \Omega \rightarrow \mathbb{R}$$

- X : is a mapping, measurable function.
- Ω : outcomes.
- $\omega \in \Omega, X(\omega) \in \mathbb{R}$.

Example 2 (Coin Toss) Consider the probability space (Ω, \mathcal{F}, P) where:

- Sample space $\Omega = \{H, T\}$ represents the elementary outcomes.
- Define the *indicator random variable* $X : \Omega \rightarrow \{0, 1\}$ by:

$$X(\omega) = \begin{cases} 1, & \text{if } \omega = H \text{ (Head).} \\ 0, & \text{if } \omega = T \text{ (Tail).} \end{cases}$$

Example 3 (Stock Price of Today is Higher than Yesterday) Consider a financial asset's daily closing prices over time. Define:

- **Sample space:** $\Omega = \{\omega\}_{t \in \mathbb{N}}$ where ω represents whether the stock price of today is higher than yesterday.
- **Event space:** $\mathcal{F} = \sigma(\{H, L\})$ where:
 - H , price increase.
 - L , price decrease.

- **Random variable:** $X : \Omega \rightarrow \{0, 1\}$ defined as:

$$X(\omega) = \begin{cases} 1, & \text{if } \omega \in H. \\ 0, & \text{if } \omega \in L. \end{cases}$$

Example 4 (Five Coin Tosses) Consider the canonical experiment of five independent coin tosses:

- **Sample space:** $\Omega = \{\omega = (\omega_1, \dots, \omega_5) \mid \omega_i \in \{H, T\}\}$ where H denotes Heads and T denotes Tails.
- **Random variable:** $X : \Omega \rightarrow \{0, 1, 2, 3, 4, 5\}$ defined as:

$$X(\omega) = \sum_{i=1}^5 \mathbf{1}_{\{\omega_i=H\}}.$$

counting the total number of Heads in the sequence

- **Realizations:**

$$X(H, H, H, H, H) = 5.$$

5 the Relationship Between Random Variable and Probability

- $X : \Omega \rightarrow \mathbb{R}$.

An outcome in the sample space is defined as a real number.

- Define a set $A = [a, b] \subseteq \mathbb{R}$, then $P(X \in A) = P(\omega \mid \omega \in \Omega, X(\omega) \in A)$.

Example 5 (Binomial Probability of Coin Tosses) Consider a sequence of n independent trials representing fair coin tosses. Let X be the random variable counting the number of heads observed.

where:

- $\Omega = \{HH\dots H, HH\dots T, \dots, TT\dots T\}$ (total $n + 1$ element).

- if $\omega = \overbrace{HH\dots H}^k TT\dots T$, then $X(\omega) = k$.

$$P(X(\omega) = k) = P(\omega = \overbrace{HH\dots H}^k TT\dots T) = \binom{n}{k} \left(\frac{1}{2}\right)^k \left(1 - \frac{1}{2}\right)^{n-k}.$$

Key Observations:

- The distribution is symmetric when $p = \frac{1}{2}$.

- *Sum of probabilities:*

$$P(\Omega) = 1 \Rightarrow \sum_{k=0}^n P(X(\omega) = k) = 1 \Rightarrow \sum_{k=0}^n \binom{n}{k} \left(\frac{1}{2}\right)^n = 2^n \left(\frac{1}{2}\right)^n = 1.$$

6 Cumulative Distribution Function(CDF)

Definition 4 (Cumulative Distribution Function) For a random variable X , its **Cumulative Distribution Function (CDF)** is the function $F_X : \mathbb{R} \rightarrow [0, 1]$ defined by:

$$F_X(x) = P(X \leq x) = P(\{\omega \in \Omega \mid X(\omega) \leq x\}).$$

Theorem 4 (CDF Characterization) Any valid CDF must satisfy:

1. **boundedness:** $0 \leq F_X(x) \leq 1$.
2. **Monotonic increasing:** $x_1 < x_2 \Rightarrow F_X(x_1) \leq F_X(x_2)$.
3. **Normalized:**

$$\lim_{x \rightarrow -\infty} F_X(x) = 0, \quad \lim_{x \rightarrow +\infty} F_X(x) = 1.$$

4. **Right-Continuity:** $\lim_{y \rightarrow x^+} F_X(y) = F_X(x^+) = F_X(x)$.

Theorem 5 (the Property of CDF) 1. $P(X = x) = F_X(x) - F_X(x^-)$.

2. if $F_X(x)$ is continuous, then $P(X = x) = 0$, $P(a \leq x \leq b) = P(a < x \leq b) = P(a < x < b) = P(a \leq x < b)$.
3. $P(x < X \leq y) = F_X(y) - F_X(x)$.
4. $P(X > x) = 1 - F_X(x)$.