

Lecture #19: Conditional Probability and Independence

Key Concepts

Almost everything here should be a part of a **review** of material from a prerequisite course — which might have presented the material somewhat differently.

Conditional Probability

Let Ω be a sample space and let $A, B \subseteq \Omega$ be events such that $P(B) > 0$. The **conditional probability** of A given B , denoted $P(A|B)$, is

$$P(A|B) = \frac{P(A \cap B)}{P(B)}.$$

$P(A|B)$ is not defined if $P(B) = 0$.

Several results, that can be useful when you need to compute probabilities, were introduced:

Claim (Law of Total Probability). *Let Ω be a sample space and let $P : \Omega \rightarrow \mathbb{R}$ be a probability distribution. Then, for any events A and B .*

$$P(A) = P(A|B) \cdot P(B) + P(A|\overline{B}) \cdot P(\overline{B}).$$

Claim (Extended Partition Rule). *Let Ω be a sample space, let $P : \Omega \rightarrow \mathbb{R}$ be a probability distribution, let k be a positive integer, and let A and $B_1, B_2, \dots, B_k$ be events satisfying the following properties.*

(a) $B_1, B_2, \dots, B_k$ are pairwise disjoint. That is, $B_i \cap B_j = \emptyset$ for all integers i and j such that $1 \leq i, j \leq k$ and $i \neq j$.

(b) $A \subseteq B_1 \cup B_2 \cup \dots \cup B_k$.

Then

$$P(A) = P(A|B_1) \cdot P(B_1) + P(A|B_2) \cdot P(B_2) + \dots + P(A|B_k) \cdot P(B_k).$$

Claim (Baye's Theorem). *Let Ω be a sample space, let $P : \Omega \rightarrow \mathbb{R}$, and let A and B be events such that $P(A) > 0$ and $P(B) > 0$. Then*

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}.$$

Independence

Events A and B are said to be **independent** if

$$P(A \cap B) = P(A) \times P(B).$$

This can be generalized in two — different — ways when three or more random variables are being considered:

Let Ω be a sample space with probability distribution $P : \Omega \rightarrow \mathbb{R}$. Let $A_1, A_2, \dots, A_k \subseteq \Omega$ be events, for some integer $k \geq 2$. The events $A_1, A_2, \dots, A_k$ are **mutually independent** if

$$P\left(\bigcap_{i \in S} A_i\right) = \prod_{i \in S} P(A_i) \quad (1)$$

for every S of $\{1, 2, \dots, k\}$.

Note: The condition at line (1) is guaranteed to hold whenever $|S| \leq 1$, so this condition only needs to be considered when $|S| \geq 2$.

The events $A_1, A_2, \dots, A_k$ are **pairwise independent** if

$$P(A_i \cap A_j) = P(A_i) \times P(A_j) \quad (2)$$

for every pair of integers i and j such that $1 \leq i, j \leq k$ and $i \neq j$.

If events $A_1, A_2, \dots, A_k$ are **mutually independent** then they are also **pairwise independent**. The converse does not always hold.