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## CONDITIONAL PROBABILITY AND INDEPENDENCE

### 1. Conditional Probability

Conditional Probability simply defines the probability of an event  $A$ , given that another event  $B$  has already occurred. This is represented as  $A \mid B$ . We can formally define condition probability as follows.

**Definition 2.1.** Let  $A$  and  $B$  be two events. The conditional probability of  $A$  given  $B$  is defined by

$$\mathbb{P}[A \mid B] = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]}$$

**Note.** Conditional Probability  $A \mid B$  is defined only if  $\mathbb{P}[B] \neq 0$

**Remark.** For a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , let  $A \in \mathcal{F}$  be an event such that  $\mathbb{P}[A] \neq 0$ . Then,  $\mathbb{P}[\cdot \mid A]$  is also a probability measure for the sample space  $\Omega$ , as well as for  $A$ .

**Theorem 2.1** (Total Probability Theorem). Let  $\{E_n\}$  be a countable collection of mutually exclusive and exhaustive events. Then, for any other event  $E$

$$\begin{aligned} \mathbb{P}[E] &= \sum_n \mathbb{P}[E \cap E_n] \\ &= \sum_n \mathbb{P}[E \mid E_n] \mathbb{P}[E_n] \end{aligned}$$

**Proof.** Since the collection  $\{E_n\}$  is exhaustive, we can say that

$$\begin{aligned} \mathbb{P}[E] &= \mathbb{P}\left[E \cap \left(\bigcup_n E_n\right)\right] \\ &= \mathbb{P}\left[\bigcup_n (E \cap E_n)\right] \\ &= \sum_n \mathbb{P}[E \cap E_n] \\ &= \sum_n \mathbb{P}[E \mid E_n] \mathbb{P}[E_n] \end{aligned}$$

□

**Theorem 2.2** (Bayes' Theorem). Let  $S = \{E_n\}$  be a countable collection of mutually exclusive and exhaustive events such that  $\mathbb{P}[E] \forall E \in S$ . Then, for any  $j$  such that  $E_j \in S$

$$\mathbb{P}[E_j \mid E] = \frac{\mathbb{P}[E \mid E_j] \mathbb{P}[E_j]}{\sum_n \mathbb{P}[E \mid E_n] \mathbb{P}[E_n]}$$

**Exercise 2.1.** Give a proof of Bayes' Theorem

## 2. Independence of Events

**Definition 2.2.** Events  $E_1, E_2 \dots E_n$  are said to be

(i) pairwise independent if

$$\mathbb{P}[E_i \cap E_j] = \mathbb{P}[E_i] \mathbb{P}[E_j]$$

(ii) mutually independent if  $\forall k \in \{2, 3 \dots n\}$  and distinct  $d_1, d_2 \dots d_k \in \{1, 2 \dots, n\}$

$$\mathbb{P}[E_{d_1} \cap E_{d_2} \dots \cap E_{d_k}] = \mathbb{P}[E_{d_1}] \cdot \mathbb{P}[E_{d_2}] \dots \mathbb{P}[E_{d_k}]$$

**Note.** Pairwise Independence does not imply Mutual Independence