

# CSCI 2540: PROBABILISTIC METHODS IN COMPUTER SCIENCE (NOTES)

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## 1. EVENTS AND PROBABILITY

### 1.2. Axioms of Probability.

**Definition.** A probability space has three components:

- (1) a sample space  $\Omega$ , which is the set of all possible outcomes of the random process modeled by the probability space;
- (2) a family of sets  $\mathcal{F}$  representing the allowable events, where each set in  $\mathcal{F}$  is a subset of the sample space  $\Omega$ ; and
- (3) a probability function  $\Pr : \mathcal{F} \rightarrow \mathbb{R}$  satisfying the following definition.

**Definition.** A probability function is any function  $\Pr : \mathcal{F} \rightarrow \mathbb{R}$  that satisfies the following conditions:

- (1) for any event  $E$ ,  $0 \leq \Pr(E) \leq 1$ ;
- (2)  $\Pr(\Omega) = 1$ ; and
- (3) for any finite or countably infinite sequence of pairwise mutually disjoint events  $E_1, E_2, E_3, \dots$ ,

$$\Pr \left( \bigcup_{i \geq 1} E_i \right) = \sum_{i \geq 1} \Pr(E_i).$$

**Lemma.** For any two events  $E_1$  and  $E_2$ ,

$$\Pr(E_1 \cup E_2) = \Pr(E_1) + \Pr(E_2) - \Pr(E_1 \cap E_2).$$

**Lemma** (Union Bound). For any finite or countably infinite sequence of events  $E_1, E_2, \dots$ ,

$$\Pr \left( \bigcup_{i \geq 1} E_i \right) \leq \sum_{i \geq 1} \Pr(E_i).$$

**Lemma** (1.3). Let  $E_1, \dots, E_n$  be any  $n$  events. Then

$$\begin{aligned} \Pr \left( \bigcup_{i=1}^n E_i \right) &= \sum_{i=1}^n \Pr(E_i) - \sum_{i < j} \Pr(E_i \cap E_j) + \sum_{i < j < k} \Pr(E_i \cap E_j \cap E_k) \\ &\quad - \dots + (-1)^{l+1} \sum_{i_1 < i_2 < \dots < i_l} \Pr \left( \bigcap_{r=1}^l E_{i_r} \right) + \dots \end{aligned}$$

**Definition.** Two events  $E$  and  $F$  are independent if and only if

$$\Pr(E \cap F) = \Pr(E) \cdot \Pr(F).$$

More generally, events  $E_1, E_2, \dots, E_k$  are mutually independent if and only if, for any subset  $I \subseteq [1, k]$ ,

$$\Pr\left(\bigcap_{i \in I} E_i\right) = \prod_{i \in I} \Pr(E_i).$$

**Definition.** The conditional probability that event  $E$  occurs given that event  $F$  occurs is

$$\Pr(E \mid F) = \frac{\Pr(E \cap F)}{\Pr(F)}.$$

The conditional probability is well-defined only if  $\Pr(F) > 0$ .

**Theorem** (Law of Total Probability). Let  $E_1, E_2, \dots, E_n$  be mutually disjoint events in the sample space  $\Omega$ , and let  $\bigcup_{i=1}^n E_i = \Omega$ . Then

$$\Pr(B) = \sum_{i=1}^n \Pr(B \cap E_i) = \sum_{i=1}^n \Pr(B \mid E_i) \Pr(E_i).$$

**Theorem** (Bayes' Law). Assume that  $E_1, E_2, \dots, E_n$  are mutually disjoint events in the sample space  $\Omega$  such that  $\bigcup_{i=1}^n E_i = \Omega$ . Then

$$\Pr(E_j \mid B) = \frac{\Pr(E_j \cap B)}{\Pr(B)} = \frac{\Pr(B \mid E_j) \Pr(E_j)}{\sum_{i=1}^n \Pr(B \mid E_i) \Pr(E_i)}.$$