

School of Engineering and Applied Science (SEAS), Ahmedabad University

**CSE 400: Fundamentals of Probability in Computing**

**Lecture 5: Bayes' Theorem, Random Variables, and PMF**

**Instructor:** Dhaval Patel, PhD

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**1. Bayes' Theorem: Weighted Average of Conditional Probabilities**

Let  $A$  and  $B$  be events. We may express  $A$  as

$$A = AB \cup AB^c$$

for, in order for an outcome to be in  $A$ , it must either be in both  $A$  and  $B$  or be in  $A$  but not in  $B$ .

As  $AB$  and  $AB^c$  are mutually exclusive, by Axiom 3,

$$\begin{aligned} \Pr(A) &= \Pr(AB) + \Pr(AB^c) \\ &= \Pr(A | B)\Pr(B) + \Pr(A | B^c)[1 - \Pr(B)] \end{aligned}$$

The probability of event  $A$  is a weighted average of the conditional probabilities with weights given as the probability of the event on which it is conditioned has of occurring.

**2. Bayes' Theorem: Learning by Example**

**Example 3.1 (Part 1)**

An insurance company believes that people can be divided into two classes: those who are accident prone and those who are not. The company's statistics show that an accident-prone person will have an accident within a fixed 1-year period with probability 0.4, whereas this probability decreases to 0.2 for a person who is not accident prone. If 30% of the population is accident prone, find the probability that a new policyholder will have an accident within a year.

**Solution:**

Let  $A_1$  denote the event that the policyholder has an accident within a year, and let  $A$  denote the event that the policyholder is accident prone.

$$\begin{aligned} \Pr(A_1) &= \Pr(A_1 | A)\Pr(A) + \Pr(A_1 | A^c)\Pr(A^c) \\ &= (0.4)(0.3) + (0.2)(0.7) = 0.26 \end{aligned}$$

**3. Bayes Formula: Learning by Example**

**Example 3.1 (Part 2)**

Suppose that a new policyholder has an accident within a year of purchasing a policy. What is the probability that the policyholder is accident prone?

**Solution:**

$$\begin{aligned}\Pr(A | A_1) &= \frac{\Pr(A \cap A_1)}{\Pr(A_1)} \\ &= \frac{\Pr(A) \Pr(A_1 | A)}{\Pr(A_1)} \\ &= \frac{(0.3)(0.4)}{0.26} = \frac{6}{13}\end{aligned}$$

#### 4. Law of Total Probability

If  $B_1, B_2, \dots, B_n$  are mutually exclusive and exhaustive events, then

$$\Pr(A) = \sum_{i=1}^n \Pr(A \cap B_i)$$

This is known as the **Law of Total Probability** (Formula 3.4).

#### 5. Bayes Formula (Proposition 3.1)

Using

$$\Pr(A \cap B_i) = \Pr(B_i | A) \Pr(A)$$

we obtain the Bayes Formula:

$$\Pr(B_i | A) = \frac{\Pr(A | B_i) \Pr(B_i)}{\sum_j \Pr(A | B_j) \Pr(B_j)}$$

Here,

- $\Pr(B_i)$  is the *a priori probability*
- $\Pr(B_i | A)$  is the *a posteriori probability*

#### 6. Bayes Formula: Card Example

Three cards are mixed: one red-red, one black-black, and one red-black. A card is selected randomly and placed face up. Given that the upper side is red, find the probability that the other side is black.

Let  $RR$ ,  $BB$ , and  $RB$  denote the three cards. Let  $R$  denote the event that the upper side is red.

$$\begin{aligned}\Pr(RB | R) &= \frac{\Pr(R | RB) \Pr(RB)}{\Pr(R | RR) \Pr(RR) + \Pr(R | RB) \Pr(RB) + \Pr(R | BB) \Pr(BB)} \\ &= \frac{(1/2)(1/3)}{(1)(1/3) + (1/2)(1/3) + (0)(1/3)} = \frac{1}{3}\end{aligned}$$

#### 7. Random Variables: Motivation and Concept

A random variable is a real-valued function defined on the sample space.

Values are determined by the outcomes of an experiment.

Probabilities are assigned to possible values of random variables.

### 8. Random Variable Example

Suppose an experiment consists of tossing 3 fair coins. Let  $Y$  denote the number of heads.

$$\Pr(Y = 0) = \frac{1}{8}, \quad \Pr(Y = 1) = \frac{3}{8}, \quad \Pr(Y = 2) = \frac{3}{8}, \quad \Pr(Y = 3) = \frac{1}{8}$$

Since  $Y$  must take one of the values 0, 1, 2, 3,

$$\sum_y \Pr(Y = y) = 1$$

### 9. Probability Mass Function

A random variable that can take on at most a countable number of possible values is said to be **discrete**.

Let  $X$  be a discrete random variable with range

$$R_X = \{x_1, x_2, x_3, \dots\}$$

The function

$$p(x_k) = \Pr(X = x_k)$$

is called the **Probability Mass Function (PMF)** of  $X$ .

Since  $X$  must take one of the values  $x_k$ ,

$$\sum_k p(x_k) = 1$$

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