

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

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

Lecture 5 Scribe

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## Objective

This scribe serves as an academic reference for exam preparation for **Lecture 5**. All definitions, derivations, and examples are strictly based on the lecture slides and are presented in a step-by-step manner suitable for a closed-notes examination.

## 1 Bayes' Theorem as a Weighted Average of Conditional Probabilities

### 1.1 Decomposition of an Event

Let  $A$  and  $B$  be two events. Any outcome that belongs to  $A$  must satisfy exactly one of the following:

- It lies in  $A \cap B$
- It lies in  $A \cap B^c$

Thus,

$$A = (A \cap B) \cup (A \cap B^c)$$

The events  $A \cap B$  and  $A \cap B^c$  are mutually exclusive.

### 1.2 Application of Axiom 3

By Axiom 3 (Additivity),

$$\Pr(A) = \Pr(A \cap B) + \Pr(A \cap B^c)$$

### 1.3 Use of Conditional Probability

Using the definition of conditional probability:

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

$$\Pr(A \cap B^c) = \Pr(A | B^c) \Pr(B^c)$$

Hence,

$$\Pr(A) = \Pr(A | B) \Pr(B) + \Pr(A | B^c) \Pr(B^c)$$

## 1.4 Interpretation

The probability of  $A$  is a weighted average of:

$$\Pr(A | B) \quad \text{and} \quad \Pr(A | B^c)$$

with weights  $\Pr(B)$  and  $\Pr(B^c)$  respectively.

## 2 Example: Insurance Policyholder Problem

### 2.1 Problem Statement

An insurance company classifies policyholders as:

- Accident prone
- Not accident prone

Given:

$$\Pr(A | B) = 0.4, \quad \Pr(A | B^c) = 0.2, \quad \Pr(B) = 0.3$$

where:

- $A$ : policyholder has an accident within one year
- $B$ : policyholder is accident prone

### 2.2 Step-by-Step Solution

Using the law of total probability:

$$\Pr(A) = \Pr(A | B) \Pr(B) + \Pr(A | B^c) \Pr(B^c)$$

$$\Pr(A) = (0.4)(0.3) + (0.2)(0.7)$$

$$\Pr(A) = 0.12 + 0.14 = 0.26$$

### 2.3 Bayes' Theorem Application

Find  $\Pr(B | A)$ :

$$\Pr(B | A) = \frac{\Pr(A | B) \Pr(B)}{\Pr(A)}$$

$$\Pr(B | A) = \frac{(0.4)(0.3)}{0.26} = \frac{6}{13}$$

## 3 Law of Total Probability

### 3.1 Partition of the Sample Space

Let  $B_1, B_2, \dots, B_n$  be mutually exclusive and exhaustive events such that:

$$\Pr(B_i) > 0 \quad \forall i$$

### 3.2 Derivation

Since:

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

and all intersections are mutually exclusive,

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

Using conditional probability:

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

## 4 Bayes' Theorem

$$\Pr(B_i | A) = \frac{\Pr(A | B_i) \Pr(B_i)}{\sum_{j=1}^n \Pr(A | B_j) \Pr(B_j)}$$

- $\Pr(B_i)$ : a priori probability
- $\Pr(B_i | A)$ : a posteriori probability

## 5 Three-Card Problem

### 5.1 Card Definitions

- RR: card with two red sides
- BB: card with two black sides
- RB: card with one red and one black side

Let  $R$  be the event that the upper side is red.

### 5.2 Required Probability

Find  $\Pr(RB | R)$ .

### 5.3 Solution

$$\begin{aligned} \Pr(RB | R) &= \frac{\Pr(R | RB) \Pr(RB)}{\Pr(R | RR) \Pr(RR) + \Pr(R | RB) \Pr(RB) + \Pr(R | BB) \Pr(BB)} \\ &= \frac{(1/2)(1/3)}{(1)(1/3) + (1/2)(1/3) + (0)(1/3)} = \frac{1/6}{1/2} = \frac{1}{3} \end{aligned}$$

## 6 Random Variables

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

## 6.1 Example

Let  $Y$  be the number of heads in three coin tosses.

$$Y \in \{0, 1, 2, 3\}$$

# 7 Probability Mass Function (PMF)

## 7.1 Definition

Let  $X$  be a discrete random variable with range:

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

The PMF is defined as:

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

## 7.2 PMF Properties

1.  $p(x_i) \geq 0$
  2.  $\sum_{i=1}^{\infty} p(x_i) = 1$
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*End of Lecture 5 Scribe*