

# CSE400 – Fundamentals of Probability in Computing

## Lecture 4: Joint Probability and Conditional Probability

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## 1. Introduction

### 1.1 Course Motivation and Applications

#### Engineering Applications

The lecture presents engineering applications where probabilistic reasoning is used:

##### 1. Speech Recognition System

- The system uses vocabulary templates such as:
  - Hello
  - Yes
  - No
  - Bye

##### 2. Radar System

- Radar systems detect objects such as aircraft using transmitted and received signals.

##### 3. Communication Network

- Communication networks involve transmission of data from source nodes to destination nodes through interconnected systems.
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## 2. Lecture Outline

The lecture includes the following major components:

### 2.1 Introduction to Probability Theory

- Experiments, Sample Space, and Events
- Axioms of Probability
- Corollaries and Propositions from Probability Axioms
- How to Assign Probability: Classical and Relative Frequency Approaches

## 2.2 Joint Probability

- Motivation, Notation, and Concepts of Joint Probability
- Example 1: Card Deck Example
- Example 2: Costume Party Example

## 2.3 Conditional Probability

- Motivation, Notation, and Concepts of Conditional Probability
  - Example 3: Cards Without Replacement
  - Example 4: Game of Poker
  - Example 5: The Missing Key
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# 3. Introduction to Probability Theory

## 3.1 Experiments, Sample Space, and Events

### 3.1.1 Experiment

**Definition:**

An **Experiment (E)** is a procedure we perform that produces some result.

**Example:**

- Tossing a coin five times
- Denoted as  $E_5$

### 3.1.2 Outcome

**Definition:**

An **Outcome ( $\xi$ )** is a possible result of an experiment.

**Example:**

For experiment  $E_5$ :

- One possible outcome is

$$\xi_1 = HHTHT$$

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### 3.1.3 Event

#### Definition:

An **Event (any letter)** is a certain set of outcomes of an experiment.

#### Example:

Consider event  $C$  with experiment  $E_5$ :

$$C = \{all\ outcomes\ consisting\ of\ an\ even\ number\ of\ heads\}$$

## 3.2 Sample Space

### 3.2.1 Definition

The **Sample Space (S)** is the collection or set of all possible distinct outcomes of an experiment.

The outcomes in the sample space must satisfy:

#### 1. Mutually Exclusive

- Two outcomes cannot occur simultaneously.
- Example: In a coin flip, you can get heads or tails but not both.

#### 2. Collectively Exhaustive

- All possible outcomes must be included.
- Example: In a coin flip, outcomes are only heads or tails.

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### 3.2.2 Properties of Sample Space

- $S$  is the universal set of outcomes of an experiment.
- Sample space can be:
  - Discrete
  - Countably infinite
  - Continuous

### 3.2.3 Examples of Sample Spaces

1. Flipping a fair coin once
  2. Rolling a cubical die with numbered faces
  3. Rolling two dice
  4. Flipping a coin until a tail occurs
  5. Random number generator with interval  $[0, 1]$
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## 4. Axioms of Probability

### 4.1 Probability Definition

Probability is defined as:

- A measure of the likelihood of various events  
OR
- A function of an event that produces a numerical quantity measuring the likelihood of that event.

There are many ways to define such a function.

### 4.2 Probability Axioms

The axioms are statements that are taken to be self-evident and require no proof.

**Axiom 1:**

For any event  $A$ ,

$$0 \leq Pr(A) \leq 1$$

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**Axiom 2:**

If  $S$  is the sample space for a given experiment,

$$Pr(S) = 1$$

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**Axiom 3 (Finite Additivity):**

If  $A \cap B = \emptyset$ ,

$$Pr(A \cup B) = Pr(A) + Pr(B)$$

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**Axiom 4 (Countable Additivity):**

For an infinite number of mutually exclusive sets  $A_i$ ,  
where

$$A_i \cap A_j = \emptyset \quad \text{for all } i \neq j$$

Then:

$$Pr\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} Pr(A_i)$$

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## 5. Corollaries from Probability Axioms

### Definition

Let  $A$  be the null event for all values of  $i$  greater than  $n$ .

### Corollary 2.1

For  $M$  finite number of mutually exclusive sets  $A_i$ ,

$$A_i \cap A_j = \emptyset \quad \text{for all } i \neq j$$

Then:

$$Pr\left(\bigcup_{i=1}^M A_i\right) = \sum_{i=1}^M Pr(A_i)$$

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### Notes

- Axiom 3 is equivalent to Corollary 2.1 when the sample space is finite.
- The generality of Axiom 3 is necessary when the sample space contains an infinite number of points.

## 6. Propositions from Probability Axioms

### Proposition 2.1

For any event  $A$ ,

$$Pr(A^c) = 1 - Pr(A)$$

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### Proposition 2.2

If  $A \subset B$ , then:

$$Pr(A) \leq Pr(B)$$

## 7. Logical Flow of Concepts

The lecture establishes the following conceptual progression:

1. Define experiments and outcomes
2. Construct events as sets of outcomes
3. Define the sample space as the universal outcome set

4. Introduce probability as a measure over events
5. Establish probability axioms
6. Derive corollaries and propositions from axioms
7. Use these foundations to study joint probability and conditional probability (introduced later in lecture outline)