# 4COSCoo2W Mathematics for Computing

**Lecture 9** 

**Probability Theory** 

UNIVERSITY OF WESTMINSTER#

# What is Probability?

*Probability* measures the likelihood of an event occurring.

**Output range:** Between 0 (impossible) and 1 (certain) – [0,1]

## For example:

The chance that a human lives for 1000 years is **o**.

The chance we all die someday is 1.



## **Applications:**

Critical in computing for decision-making, algorithms, and risk assessment.

# **Theoretical Probability**

Probability of event *A* occurring:

$$P(A) = \frac{\text{Number of favourable outcomes}}{\text{Total number of outcomes}}$$

*Number of favourable outcomes:* The count of outcomes that are considered as a success or the desired outcome for the event A.

Total number of possible outcomes (probability space): The total count of all possible outcomes for the event, assuming each outcome is equally likely.

**NB!** Theoretical probability assumes that every outcome has an equal chance of occurring.

# **Common Examples of Probability**





# **Examples: Theoretical Probaility**

**Rolling a Die:** The probability of rolling a 4 on a six-sided die. There is one favorable outcome (rolling a 4), and six possible outcomes.

$$P(4) = \frac{1}{6}$$

**Flipping a Coin:** The probability of getting heads when flipping a fair coin. There is one favorable outcome (heads), and two possible outcomes (heads or tails).

$$P(\text{Heads}) = \frac{1}{2}$$

**Drawing a Card:** The probability of drawing an Ace from a standard deck of 52 cards. There are four favorable outcomes (the four Aces), and 52 possible outcomes.

$$P(Ace) = \frac{4}{52} = \frac{1}{13}$$

# **Complementary Events**

Events that together cover all possible outcomes are called *complementary*.

If A is an event, the complement of A, denoted as A' or sometimes  $\overline{A}$ , represents all outcomes that are not in A. The probability of the complementary event A is given by:

$$P(A') = 1 - P(A)$$

P(A) is the probability of event A occurring.

P(A') is the probability of event A not occurring.

## **Example:**

If the probability of raining is 0.3, the probability of not raining is 0.7.

# **Examples: Complementary Events**

## **Coin Toss:**

If P(Heads) = 0.5, then P(Tails) = 1 - P(Heads) = 1 - 0.5 = 0.5.

## **Drawing a Card:**

If the probability of drawing a king from a standard deck of cards is  $\frac{4}{52}$  or  $\frac{1}{13}$ , then the probability of not drawing a king is  $1 - \frac{1}{13} = \frac{12}{13}$ .

## **Dice Roll:**

If the probability of rolling a 6 on a standard six-sided die is  $\frac{1}{6}$ , then the probability of not rolling a 6 is  $1 - \frac{1}{6} = \frac{5}{6}$ .

# **Independent Events**

*Independent events* in probability theory are two or more events whose occurrence or non-occurrence does not affect the probability of the other events occurring.

If two events, A and B, are independent, the occurrence (or non-occurrence) of A does not change the probability of B occurring, and vice versa.

For two independent events, A and B, the probability of both events occurring is given by the product of their individual probabilities (*multiplication law*):

$$P(A \text{ and } B) = P(A) \times P(B)$$

# **Examples: Independent Events**

## **Coin Tosses:**

Tossing a coin twice. The outcome of the first toss (say, heads) does not affect the probability of the outcome of the second toss. If P(Heads) = 0.5 for each toss, then  $P(Heads) = 0.5 \times 0.5 = 0.25$ .

### **Dice and Cards:**

Rolling a die and drawing a card from a deck. The probability of rolling a 6 (1/6) and drawing an ace (4/52) are independent events. So, P(6 and Ace) =  $\frac{1}{6} \times \frac{4}{52}$ .

## **Traffic Lights:**

The probability of hitting a green light on two different traffic signals might be independent if the lights are not synchronised. If the probability of green at each light is 0.3, then  $P(Green at both) = 0.3 \times 0.3 = 0.09$ .

# The probability of rolling certain totals

<b>Total on Dice</b>	Pairs of Dice Probability	
2	1+1	1/36 = 3%
3	1+2, 2+1	2/36 = 6%
4	1+3, 2+2, 3+1	3/36 = 8%
5	1+4, 2+3, 3+2, 4+1	4/36 = 11%
6	1+5, 2+4, 3+3, 4+2, 5+1	5/36 = 14%
7	1+6, 2+5, 3+4, 4+3, 5+2, 6+1	6/36 = 17%
8	2+6, 3+5, 4+4, 5+3, 6+2	5/36 = 14%
9	3+6, 4+5, 5+4, 6+3	4/36 = 11%
10	4+6, 5+5, 6+4	3/36 = 8%
11	5+6, 6+5	2/36 = 6%
12	6+6	1/36 = 3%

## **Probabilities of Outcomes in Three-Coin Toss**

Specific Outcome	Probability of Specific Outcome	Total Heads in Outcome	Cumulative Probability of Total Heads (%)	<b>Probability Calculation</b>
TTT	1/8	O	12.5%	1/8 (1 outcome out of 8)
TTH	1/8	1	37.5%	3/8 (3 outcomes out of 8)
НТТ	1/8	1	37.5%	3/8 (3 outcomes out of 8)
THT	1/8	1	37.5%	3/8 (3 outcomes out of 8)
ТНН	1/8	2	37.5%	3/8 (3 outcomes out of 8)
ННТ	1/8	2	37.5%	3/8 (3 outcomes out of 8)
НТН	1/8	2	37.5%	3/8 (3 outcomes out of 8)
ннн	1/8	3	12.5%	1/8 (1 outcome out of 8)

# **Conditional Probability**

Conditional probability measures the probability of an event occurring, given that another event has already occurred.

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$
, where  $P(B)$  is not zero.

**Interpretation:** It quantifies how the probability of one event changes when another related event is known to occur.

**Example:** Probability of drawing an ace after drawing a king is  $\frac{4}{51}$ .

A fair die is rolled. Let  $\mathbf{A}$  be the event that the outcome is an *odd* number, i.e.,  $A=\{1,3,5\}$ . Also let  $\mathbf{B}$  be the event that the outcome is less than or equal to 3, i.e.,  $B=\{1,2,3\}$ .

- What is the Probability Space S?
- What is P(A)?
- What is P(B)?
- What is  $P(A \cap B)$ ?
- What is the conditional probability P(A|B)?

## **Solution:**

**Probability Space:** 

$$S = \{1,2,3,4,5,6\}$$

Probability of A:

$$P(A) = \frac{|A|}{|S|} = \frac{|\{1,3,5\}|}{6} = \frac{3}{6} = \frac{1}{2}$$

Given sets:  $A=\{1,3,5\}$   $B=\{1,2,3\}$ 

Let's find  $P(A \cap B)$ :

First, we need to define  $A \cap B = \{1,3\}$ .

$$P(A \cap B) = \frac{|A \cap B|}{|S|} = \frac{2}{6} = \frac{1}{3}.$$

Now we can find the conditional probability of A given that B occurred as we know:

$$P(A \cap B) = \frac{1}{3} \text{ and } P(B) = \frac{1}{2}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\frac{1}{3}}{\frac{1}{2}} = \frac{1}{3} \times \frac{2}{1} = \frac{2}{3}.$$

# **Bayes' Theorem**

Bayes' Theorem is an extension of conditional probability:

$$P(A|B) = \frac{P(B|A) \times P(A)}{P(B)}$$

## where:

P(A|B): The probability of event A occurring given that B is true.

P(B|A): The probability of event B occurring given that A is true.

P(A): The prior or initial probability of event A.

P(B): The prior or initial probability of event B.

Imagine a medical test for a particular disease. Let's denote:

A as the event of actually having the disease.

**B** as the event of testing positive for the disease.

Bayes' Theorem can be used to calculate P(A|B), the probability of having the disease given a positive test result.

### In this context:

P(A) would be the overall prevalence of the disease in the population.

P(B|A) is the probability of testing positive if you have the disease (true positive rate).

P(B) would be the overall probability of testing positive, which includes true positives and false positives.

## **Given Data:**

- 1. Prevalence of the Disease (P(A)): Let's say 1% of the population has a certain disease. So, P(A) = 0.01.
- **2. Sensitivity of the Test (P(B|A)):** The probability of the test correctly identifying the disease (true positive rate) is 95%. Hence, P(B|A) = 0.95.
- **3. False Positive Rate:** Assume the test has a 5% false positive rate. This means that the test incorrectly identifies the disease in healthy people 5% of the time.

## Calculating P(B):

P(B) is the overall probability of testing positive. This includes both true positives (sick people correctly diagnosed) and false positives (healthy people incorrectly diagnosed).

Let's calculate it:

**True Positives:**  $P(A) \times P(B|A) = 0.01 \times 0.95 = 0.0095$ .

**False Positives:**  $P(A') \times P(B|A')$  where P(A') = 0.99 (probability of not having the disease) and P(B|A') = 0.05 (false positive rate).

So, False Positives =  $0.99 \times 0.05 = 0.0495$ .

Therefore, P(B) = True Positives + False Positives = 0.0095 + 0.0495 = 0.059.

## **Applying Bayes' Theorem:**

We want to find P(A|B): the probability of having the disease given a positive test result.

Using the formula  $P(A|B) = \frac{P(B|A) \times P(A)}{P(B)}$ , we plug in the values:

$$P(A|B) = \frac{0.95 \times 0.01}{0.059}.$$

Let's calculate this.

The calculation yields that  $P(A|B) \approx 0.161$ , or approximately 16.1%.