

## 1. What is Probability?

**Probability** is a branch of mathematics that deals with **uncertainty**. It measures how likely an event is to occur.

In simple words:

It tells us the **chance of something happening**.

### Formula:

$P(E) = \text{Total number of possible outcomes} / \text{Number of favorable outcomes}$

### Example:

If you toss a fair coin:

Possible outcomes = {Head, Tail}

Probability of Head:

$P(H) = 1/2$

### Range of Probability:

Probability always lies between:

$0 \leq P(E) \leq 1$

- **0** → Impossible event
- **1** → Certain event

## 2. Difference Between Mutually Exclusive and Independent Events

This is a very important concept in probability.

### Mutually Exclusive Events

Two events are **mutually exclusive** if they **cannot occur at the same time**.

If one happens, the other cannot happen.

### Example:

When rolling a die:

Event A = Getting 2

Event B = Getting 5

You cannot get both at once.

$P(A \cap B) = 0$

### Independent Events

Two events are **independent** if the occurrence of one **does not affect** the other.

**Example:**

Tossing a coin and rolling a die.

Coin result does not affect die result.

$$P(A \cap B) = P(A) \times P(B)$$

**Key Difference Table:**

Feature	Mutually Exclusive	Independent
Occur together?	No	Yes
Relation	One prevents other	No effect
Formula	$P(A \cap B) = 0$	$P(A \cap B) = P(A)P(B)$

**3. Addition Rule of Probability**

Used when we want probability of **A OR B**.

**General Formula:**

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

**For Mutually Exclusive Events:**

Since they cannot occur together:

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

**Example:**

Die roll:

$$P(\text{even}) = 3/6$$

$$P(\text{odd}) = 3/6$$

$$P(\text{even OR odd}) = 1$$

**4. Multiplication Rule of Probability**

Used for **A AND B** events.

**For Independent Events:**

$$P(A \cap B) = P(A) \times P(B)$$

**For Dependent Events:**

$$P(A \cap B) = P(A) \times P(B|A)$$

**Example:**

Two coin tosses:

$$P(HH) = 1/2 \times 1/2 = 1/4$$

**5. Conditional Probability**

Conditional probability means:

Probability of an event occurring **given another event has already happened**.

**Formula:**

$$P(A|B) = P(A \cap B) / P(B)$$

**Example:**

From a deck:

Event A = Drawing a King

Event B = Drawing a face card

$$P(A|B) = 4/12 = 1/3$$

**6. Difference Between Discrete and Continuous Data Distribution****Discrete Distribution**

Deals with **countable values**.

Examples:

- Number of students
- Number of cars

- Dice outcomes

Values are **finite or countable**.

### Continuous Distribution

Deals with **measurable values**.

Examples:

- Height
- Weight
- Time
- Temperature

Values are **infinite within a range**.

### Difference Table:

Feature	Discrete	Continuous
Values	Countable	Infinite
Examples	Dice roll	Height
Graph	Bar graph	Curve

## 7. Exponential Distribution

Exponential distribution is a **continuous probability distribution** that measures:

Time between events.

### Example:

- Time between customer arrivals
- Time until machine failure
- Waiting time in queue

### Formula:

$$f(x) = \lambda e^{-\lambda x}$$

Where:

$\lambda$  = rate of occurrence

**Key Property:**

It has **memoryless property**:

Future probability does not depend on past.

**8. Variance and Standard Deviation**

These measure **data spread**.

**Variance**

Measures how far values are from mean.

$$\sigma^2 = \sum (x - \mu)^2 / N$$

Large variance = Data widely spread.

**Standard Deviation**

Square root of variance.

$\sigma$  = square root of variance

Tells how much data deviates from mean.

**Example:**

Low SD → data close to mean

High SD → data spread out

**9. Bayes Theorem**

Bayes theorem calculates probability based on **prior knowledge**.

**Formula:**

$$P(A|B) = P(B|A) \times P(A) / P(B)$$

**Terms:**

- $P(A)$  = Prior probability
- $P(B|A)$  = Likelihood
- $P(A|B)$  = Posterior probability

**Simple Meaning:**

It updates probability when new information comes.

**10. Applications of Bayes Theorem**

Bayes theorem is widely used in real world.

**1. Medical Diagnosis**

Used to find disease probability based on test results.

Example:

COVID testing probability.

**2. Spam Email Detection**

Gmail uses Bayes theorem to detect spam.

**3. Machine Learning**

Used in:

- Naive Bayes classifier
- AI predictions
- Recommendation systems

**4. Weather Forecasting**

Predicts rain based on past data.

**5. Risk Analysis**

Used in finance and insurance.

**Final Summary**

Probability helps us:

- Measure uncertainty
- Predict outcomes

- Analyze data
- Build ML models
- Make decisions under risk