

## CBSE Class 12 Mathematics

### Chapter-13

### Probability

- **Sample Space:** The set of all possible outcomes of a random experiment. It is denoted by the symbol  $S$ .
- **Sample points:** Elements of the sample space.
- **Event:** A subset of the sample space.
- **Impossible Event:** The empty set.
- **Sure Event:** The whole sample space.
- **Complementary event or "not event":** The set " $S$ " or  $S - A$ .
- **The event A or B:** The set  $A \cup B$ .
- **The event A and B:** The set  $A \cap B$ .
- **The event A but not B:**  $A - B$ .
- **Mutually exclusive events:** A and B are mutually exclusive if  $A \cap B = \phi$ .
- **Exhaustive and Mutually exclusive events:** Events  $E_1, E_2, \dots, E_n$  are mutually exclusive and exhaustive if  $E_1 \cup E_2 \cup \dots \cup E_n = S$  and  $E_i \cap E_j = \phi$  for all  $i \neq j$ .
- **Exiomatic approach to probability:** To assign probabilities to various events, some axioms or rules have been described.

Let  $S$  be the sample space of a random experiment. The probability  $P$  is a real values function whose domain is the power set of  $S$  and range is the interval  $[0, 1]$  satisfying the following axioms:

(a) For any event  $E$ ,  $P(E) \geq 0$

(b)  $P(S) = 1$

(c) If  $E$  and  $F$  are mutually exclusive event, then  $P(E \cup F) = P(E) + P(F)$

If  $E_1, E_2, E_3, \dots$  are  $n$  mutually exclusive events, then  $P\left(\bigcup_{i=1}^n E_i\right) = \sum_{i=1}^n P(E_i)$

- **Probability of an event in terms of the probabilities of the same points (outcomes):** Let  $S$  be the sample space containing  $n$  exhaustive outcomes

$$W_1, W_2, W_3, \dots, W_n \text{ i.e., } S = (W_1, W_2, W_3, \dots, W_n)$$

Now from the axiomatic definition of the probability:

(a)  $0 \leq P(W_i) \leq 1$ , for each  $W_i \in S$ .

(b)  $P(W_1) + P(W_2) + \dots + P(W_n) = P(S) = 1$

(c) For any event A,  $P(A) = \sum P(W_i), W_i \in A$

- **Equally likely outcomes:** All outcomes with equal probability.
- **Classical definition of the probability of an event:** For a finite sample space with equally likely outcome, probability of an event A.

$$P(A) = \frac{n(A)}{n(S)}$$

where  $n(A)$  = Number of elements in the set A. and  $n(S)$  = Number of elements in set S.

- If A is any event, then  $P(\text{not } A) = 1 - P(A) \Rightarrow P(\bar{A}) = 1 - P(A) \Rightarrow P(A') = 1 - P(A)$
- The conditional probability of an event E, given the occurrence of the event F is given by  $P(E|F) = \frac{P(E \cap F)}{P(F)}, P(F) \neq 0$   
 $0 \leq P(E|F) \leq 1$ ,
- $P(E'|F) = 1 - P(E|F)$   
 $P((E \cup F)|G) = P(E|G) + P(F|G) - P((E \cap F)|G)$   
 $P(E \cap F) = P(E)P(E|F), P(E) \neq 0$
- $P(E \cap F) = P(F)P(E|F), P(F) \neq 0$   
 $P(E \cap F) = P(E)P(F)$
- $P(E|F) = P(E), P(F) \neq 0$   
 $P(F|E) = P(F), P(E) \neq 0$
- **Theorem of total probability:**



Let  $\{E_1, E_2, \dots, E_n\}$  be a partition of a sample space and suppose that each of  $E_1, E_2, \dots, E_n$  has non zero probability. Let  $A$  be any event associated with  $S$ , then

$$P(A) = P(E_1)P(A|E_1) + P(E_2)P(A|E_2) + \dots + P(E_n)P(A|E_n)$$

- **Bayes' theorem:** If  $E_1, E_2, \dots, E_n$  are events which constitute a partition of sample space  $S$ , i.e.  $E_1, E_2, \dots, E_n$  are pairwise disjoint and  $E_1 \cup E_2 \cup \dots \cup E_n = S$  and  $A$  be any event with non-zero probability, then,

$$P(E_i|A) = \frac{P(E_i)P(A|E_i)}{\sum_{j=1}^n P(E_j)P(A|E_j)}$$

- **Random variable:** A random variable is a real valued function whose domain is the sample space of a random experiment.
- **Probability distribution:** The probability distribution of a random variable  $X$  is the system of numbers

$$X : x_1 \quad x_2 \quad \dots \quad x_n$$

$$P(X): p_1 \quad p_2 \quad \dots \quad p_n$$

Where,  $p_i > 0, \sum_{i=1}^n p_i = 1, i = 1, 2, \dots, n$

- **Mean of a probability distribution:** Let  $X$  be a random variable whose possible values  $x_1, x_2, x_3, \dots, x_n$  occur with probabilities  $p_1, p_2, p_3, \dots, p_n$  respectively.

The mean of  $X$ , denoted by  $\mu$  is the number  $\sum_{i=1}^n x_i p_i$ . The mean of a random variable

$X$  is also called the expectation of  $X$ , denoted by  $E(X)$ .

- **Variance:** Let  $X$  be a random variable whose possible values  $x_1, x_2, x_3, \dots, x_n$  occur with probabilities  $p(x_1), p(x_2), \dots, p(x_n)$  respectively. Let  $\mu = E(X)$  be the mean of  $X$ . The variance of  $X$ , denoted by  $\text{Var}(X)$  or  $\sigma_x^2$  is defined as  $\sum x^2$

$Var(X) = \sum_{i=1}^n (x_i - \mu)^2 p(x_i)$  or equivalently  $\sigma_x^2 = E(X - \mu)^2$ . The non-

negative number,  $\sqrt{Var(X)} = \sqrt{\sum_{i=1}^n (x_i - \mu)^2 p(x_i)}$  is called the **standard deviation** of the random variable X.

$$Var(X) = E(X^2) - [E(X)]^2$$

- **Bernoulli Trials:** Trials of a random experiment are called Bernoulli trials, if they satisfy the following conditions:

- There should be a finite number of trials.
- The trials should be independent.
- Each trial has exactly two outcomes: success or failure.
- The probability of success remains the same in each trial.

For Binomial distribution  $B(n, p)$ ,  $P(X=x) = {}^n C_x q^{n-x} p^x$ ,  $x = 0, 1, \dots, n$  ( $q = 1 - p$ )

