

Real valued function whose domain is the sample space of a random experiment

The probability distribution of a random variable  $x$  is the system of numbers  $x: x_1, x_2, \dots, x_n$   
 $P(x): p_1, p_2, \dots, p_n$  where,  $p_i > 0$ ,  
 $\sum_{i=1}^n p_i = 1, i=1, 2, \dots, n$ .

Let  $x$  be a R.V. whose possible values  $x_1, x_2, \dots, x_n$  occurs 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  $x$ ,  $\text{var}(x)$  or  $\sigma_x^2 = \sum_{i=1}^n (x_i - \mu)^2 p(x_i)$  or  $E(x - \mu)^2$ .

The non-negative number

$\sigma_x = \sqrt{\text{var}(x)} = \sqrt{\sum_{i=1}^n (x_i - \mu)^2 p(x_i)}$  is called the standard deviation of the R.V. 'X'. Also,  $\text{var}(x) = E(x^2) - [E(x)]^2$  For eg:  $E(x) = 3$  and  $E(x^2) = 10$ , then  $\text{var } x = 10 - 9 = 1$  and  $SD = \sqrt{1} = 1$ .

Variance and standard deviation

Bernoulli's Trials

Trials of a random experiment are called Bernoulli trials, if they satisfy the following conditions :  
(i) There should be a finite no. of trials.  
(ii) The trials should be independent.  
(iii) Each trial has exactly two outcomes: success or failure.  
(iv) The probability of success remains the same in each trial.

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

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 pair wise disjoint and  $E_1 \cup E_2 \cup \dots \cup E_n = S$  and  $A$  be any event with non-zero probability, then  $P(E_i | A) = \frac{P(E_i)P(A|E_i)}{\sum_{j=1}^n P(E_j)P(A|E_j)}$

The probability of the event  $E$  is called the conditional probability of  $E$  given that  $F$  has already occurred, and is denoted by  $P(E/F)$ . Also,

$$P(E/F) = \frac{P(E \cap F)}{P(F)}, P(F) \neq 0.$$

Random Variable

(i)  $0 \leq P(E/F) \leq 1, P(E'/F) = 1 - P(E/F)$   
(ii)  $P((E \cup F)/G) = P(E/G) + P(F/G) - P((E \cap F)/G)$

(iii)  $P(E \cap F) = P(E)P(F/E), P(E) \neq 0$

(iv)  $P(E \cap F) = P(F)P(E/F), P(F) \neq 0$

For eg: if  $P(A) = \frac{7}{13}, P(B) = \frac{9}{13}$  and  $P(A \cap B) = \frac{4}{13}$ , then

$$P(A/B) = \frac{P(A \cap B)}{P(B)} = \frac{\frac{4}{13}}{\frac{9}{13}} = \frac{4}{9}$$

Conditional probability

Properties

If  $E$  and  $F$  are independent, then  $P(E \cap F) = P(E)P(F), P(E|F) = P(E), P(F|E) = P(F)$  and  $P(F|E) = P(F), P(E) \neq 0$ .

Theorem of total Probability

Let,  $\{E_1, E_2, \dots, E_n\}$  be a partition of a sample space 'S' 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  $P(A) = P(E_1)P(A|E_1) + P(E_2)P(A|E_2) + \dots + P(E_n)P(A|E_n)$ .