

TWO-DIMENSIONAL RANDOM VARIABLES

Defn:- let  $S$  be the sample space. let  $X=X(S)$  &  $Y=Y(S)$  be two functions each assigning a real numbers to each outcome  $s \in S$  then  $(X,Y)$  is a two dimensional random Variables.

Two types of two dimensional R.V are .

- i) Discrete R.V
- ii) Continuous R.V

T.D Discrete Random Variable:-

If the possible values of  $(X,Y)$  are finite or countably infinite, then  $(X,Y)$  is called a two dimensional discrete random variable.

TD continuous Random Variable:-

If  $(X,Y)$  can assumes all values in a specified Region  $R$  in  $xy$ -Plane then  $(X,Y)$  is called a Two dimensional continuous random Variable.

Joint Probability function (or) Joint Probability mass  $f_{ij}$ :-

If  $(X,Y)$  be a TDDR.V such that  $P(X=x_i, Y=y_j) = P(x_i, y_j) = p_{ij}$  is called the joint Probability  $f_{ij}$  (or) joint probability mass  $f_{ij}$ .

If (i)  $p_{ij} \geq 0 \quad \forall i \& j$

(ii)  $\sum_j \sum_i p_{ij} = 1$

Joint Probability density  $f_{ij}$ :-

If  $(X,Y)$  be a TD continuous Random Variable then  $f(x,y)$  is called the joint probability density  $f_{ij}$  of  $(X,Y)$ .

If i)  $f(x, y) \geq 0 \quad \forall (x, y) \in R$ ,  $R$  is the region.

ii)  $\iint_R f(x, y) dx dy = 1$  or  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$  (or)  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx = 1$

Joint cumulative distribution fn. :-

In discrete case :-

$$F(x, y) = P[X \leq x, Y \leq y] = \sum_{j=1}^m \sum_{i=1}^n P(x_i, y_j)$$

In continuous case :-

$$F(x, y) = P[X \leq x, Y \leq y] = \int_{-\infty}^x \int_{-\infty}^y f(x, y) dx dy$$

Marginal Probability function (or) Marginal distribution :-

i) In discrete case :-

\*  $P[X = x_i] = \sum_{j=1}^m P(x_i, y_j) = P_{i*} \rightarrow$  marginal probability fn. of  $x$ .

\*  $P[Y = y_j] = \sum_{i=1}^n P(x_i, y_j) = P_{*j} \rightarrow$  marginal probability fn. of  $y$ .

ii) In continuous case :-

\* Marginal <sup>prob.</sup> density fn. of  $x$  is

$$f_x(x) = f(x) = \int_{-\infty}^{\infty} f(x, y) dy$$

\* Marginal <sup>prob.</sup> density fn. of  $y$  is

$$f_y(y) = f(y) = \int_{-\infty}^{\infty} f(x, y) dx$$

Conditional distribution :-

In discrete case :-

\* Conditional probability fn. of  $x$  given  $y = y_j$  is

$$P[X = x_i / Y = y_j] = \frac{P[X = x_i, Y = y_j]}{P[Y = y_j]}$$

\* Conditional probability  $f_{h_2}$  of  $x$  given  $x = x_i$  is

$$P[Y = y_j / X = x_i] = \frac{P[X = x_i, Y = y_j]}{P[X = x_i]}$$

In continuous case:-

\* The conditional probability  $f_{h_2}$  of  $x$  given  $y$  is

$$f(x/y) = \frac{f(x,y)}{f(y)} ; f(y) > 0$$

\* The conditional probability  $f_{h_2}$  of  $y$  given  $x$  is

$$f(y/x) = \frac{f(x,y)}{f(x)} ; f(x) > 0$$

Independent condition:-

In discrete case:-

$$* P[X = x_i, Y = y_j] = P(x) \cdot P(y) \text{ (or) } P_{ij} = P_i \cdot P_j$$

In continuous case:-

$$* f(x,y) = f(x) \cdot f(y)$$

Note:-

$$1) P[X = x_i, Y = y_j] = P[X = x_i \cap Y = y_j]$$

$$2) P[a_1 \leq x \leq b_1 \cap a_2 \leq y \leq b_2] = \int_{a_2}^{b_2} \int_{a_1}^{b_1} f(x,y) dx dy$$

3)  $f(x,y)$  or  $f_{xy}(x,y)$  are both represents Joint probability  $f_{h_2}$ .

$$4) f(x,y) = \frac{\partial^2}{\partial x \partial y} F(x,y).$$

$$5) F(\infty, \infty) = 1 ; F(-\infty, y) = F(x, -\infty) = 0 ; 0 \leq F(x,y) \leq 1.$$