

UNIT-3

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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

TD 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) $\sum_i \sum_j p_{ij} = 1$

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 ^{prob.} density fn. of x is

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

* Marginal ^{prob.} 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_i} 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_i} of x given y is

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

* The conditional probability f_{h_i} 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_i} .

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

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