

# PROBABILITY AND STATISTICS

## (UCS401)

### Lecture-18

(Conditional mean and variance, Correlation and regression)

Two-dim. r.v.'s and Joint Distributions (Unit -V)



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# Independent Random Variables

Let  $(X, Y)$  be two dimensional random variable with joint probability function  $f_{(X,Y)}(x, y) = P[X = x, Y = y]$ . Then  $(X, Y)$  is said to be independent if

$$f_{(X,Y)}(x, y) = f_X(x) \cdot f_Y(y)$$

**Note:** If  $X, Y$  are independent, then

a)  $f_{X|Y}(x|y) = f_X(x),$

b)  $f_{Y|X}(y|x) = f_Y(y),$

**Example 1:** Let X,Y have following joint probability distribution,  $f_{(X,Y)}(x, y)$  .Show that X,Y are independent.

	X=2	X= 4
Y=1	0.10	0.15
Y=3	0.20	0.30
Y=5	0.10	0.15

**Solution:**

Y \ X	2	4	$P(Y = y)$
1	0.10	0.15	0.25
3	0.20	0.30	0.50
5	0.10	0.15	0.25
$P(X = x)$	0.40	0.60	1

From the above table we can see that

$$f_X(x) \cdot f_Y(y) = P(X=x) \cdot P(Y=y) = f_{(X,Y)}(x, y) \text{ for each pair of values } (x, y).$$

e.g for Pair (2,1), we have  $f_{(X,Y)}(2,1) = 0.10$  and  $P(X=2)=0.40$  and  $P(Y=1)=0.25$

$$f_X(x) \cdot f_Y(y) = P(X=x) \cdot P(Y=y) = 0.25 * 0.40 = 0.10 = f_{(X,Y)}(x, y)$$

**Example 2:** If X,Y have the joint PDF

$$f(x,y) = \begin{cases} x+y, & 0 < x < 1, 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

Check whether X and Y are independent or not.

**Solution:** The marginal density function of X is given by

$$f_X(x) = \int_{-\infty}^{\infty} f(x,y)dy = \int_0^1 (x+y)dy = x + \frac{1}{2}, \quad 0 < x < 1$$

The marginal density function of Y is given by

$$f_Y(y) = \int_{-\infty}^{\infty} f(x,y)dx = \int_0^1 (x+y)dx = y + \frac{1}{2}, \quad 0 < y < 1$$

Now,

$$f_X(x) \cdot f_Y(y) = (x + 1/2)(y + 1/2) \neq f(x,y)$$

Ans: X and Y are not independent.

# Conditional Mean and Variance

If  $(X,Y)$  is a two-dimensional random variable, then the mean or expectation of  $(X,Y)$  is defined as follows

- **Case 1:** when  $X,Y$  are discrete random variables, then

$$E(X) = \sum_{x_i} x_i \cdot P(X = x_i)$$

$$E(Y) = \sum_{y_j} y_j \cdot P(Y = y_j)$$

$$E(X/Y) = \sum_{x_i} x_i \cdot P(X = x_i / Y = y_j)$$

$$E(Y/X) = \sum_{y_j} y_j \cdot P(Y = y_j / X = x_i)$$

$$E(XY) = \sum_{x_i} \sum_{y_j} x_i y_j P(X = x_i, Y = y_j)$$

**Case 2:** When X,Y are continuous random variables, then

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

$$E(Y) = \int_{-\infty}^{\infty} y f(y) dy$$

**Conditional Expected Values**

$$E(X/Y) = \int_{-\infty}^{\infty} x f(x/y) dx$$

$$E(Y/X) = \int_{-\infty}^{\infty} y f(y/x) dy$$

$$E(XY) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xyf(x, y) dx dy$$

**Conditional Variance:** If  $(X, Y)$  is a two-dimensional random variable, then the conditional variance of  $(X, Y)$  is

$$\text{Var}(Y/X) = E(Y^2/X) - [E(Y/X)]^2$$

$$\text{Var}(X/Y) = E(X^2/Y) - [E(X/Y)]^2$$

**Notes:** If  $X$  and  $Y$  are independent random variables, then

$$E(X/Y) = E(X)$$

$$E(Y/X) = E(Y)$$

$$E[E(Y/X)] = E(Y)$$

$$E[E(X/Y)] = E(X)$$



**Example 1:** The joint distribution of X and Y is given by

$$f(x, y) = \frac{x+y}{21}, \quad x = 1, 2, 3, \quad y = 1, 2$$

Find the marginal distributions of X and Y. Find the mean of X and Y also.

**Solution:**

$X \backslash Y$	1	2	$P(X = x_i)$	
1	$\frac{2}{21}$	$\frac{3}{21}$	$\frac{5}{21}$	$P(X = 1)$
2	$\frac{3}{21}$	$\frac{4}{21}$	$\frac{7}{21}$	$P(X = 2)$
3	$\frac{4}{21}$	$\frac{5}{21}$	$\frac{9}{21}$	$P(X = 3)$
$P(Y = y_j)$	$\frac{9}{21}$	$\frac{12}{21}$	1	
	$P(Y = 1)$	$P(Y = 2)$		

The marginal distributions of  $X$  are

$$P(X = 1) = \frac{5}{21}, P(X = 2) = \frac{7}{21}, P(X = 3) = \frac{9}{21}$$

The marginal distributions of  $Y$  are

$$P(Y = 1) = \frac{9}{21}, P(Y = 2) = \frac{12}{21}$$

$$\begin{aligned} E(X) &= \sum_{x=1}^3 xP(X = x) = 1P(X = 1) + 2P(X = 2) + 3P(X = 3) \\ &= \frac{5}{21} + \frac{14}{21} + \frac{27}{21} = \frac{46}{21} \end{aligned}$$

$$\begin{aligned} E(Y) &= \sum_{y=1}^2 yP(Y = y) = 1P(Y = 1) + 2P(Y = 2) \\ &= \frac{9}{21} + \frac{24}{21} = \frac{33}{21} = \frac{11}{7} \end{aligned}$$

**Example 2:** The joint PDF of (X,Y) is given by

$$f(x, y) = \begin{cases} 24xy, & 0 < x, 0 < y, x + y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Find the conditional mean and variance of Y given X.

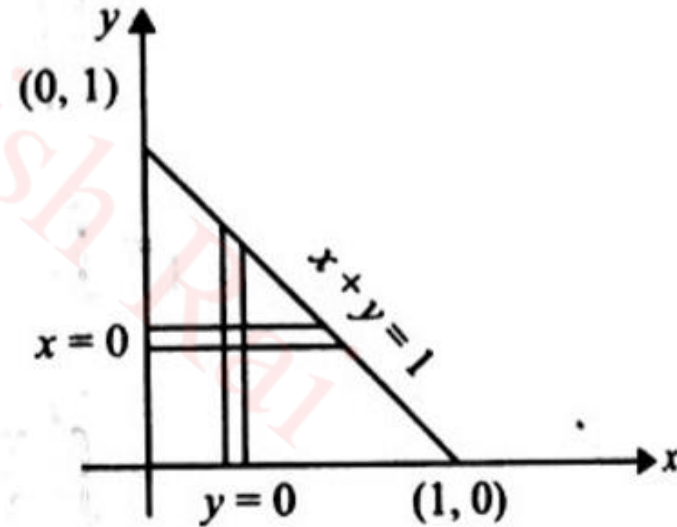
**Solution:**

Given:  $f(x, y) = 24xy, \quad x > 0, y > 0, x + y \leq 1$

$$\therefore f_X(x) = \int_0^{1-x} 24xy \, dy = 24x \int_0^{1-x} y \, dy,$$

$$\begin{aligned} &= 24x \left[ \frac{y^2}{2} \right]_0^{1-x} = 24x \frac{(1-x)^2}{2} \\ &= 12x (1-x)^2, \quad 0 < x < 1 \end{aligned}$$

$$f(y/x) = \frac{f(x, y)}{f_X(x)} = \frac{2y}{(1-x)^2}, \quad 0 < y < 1-x$$



$$\begin{aligned}
 E(Y/X) &= \int_0^{1-x} y f(y/x) dy \\
 &= \int_0^{1-x} \frac{2y^2}{(1-x)^2} dy = \frac{2}{(1-x)^2} \left[ \frac{y^3}{3} \right]_0^{1-x} = \frac{2}{3}(1-x), \quad x > 0
 \end{aligned}$$

$$\begin{aligned}
 E(Y^2/X = x) &= \int_0^{1-x} y^2 f(y/x) dy \\
 &= \int_0^{1-x} y^2 \frac{2y}{(1-x)^2} dy = \frac{2}{(1-x)^2} \left[ \frac{y^4}{4} \right]_0^{1-x} = \frac{1}{2}(1-x)^2, \quad x > 0
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(Y/X) &= E(Y^2/X) - [E(Y/X)]^2 \\
 &= \frac{1}{2}(1-x)^2 - \frac{4}{9}(1-x)^2 = \frac{1}{18}(1-x)^2, \quad x > 0
 \end{aligned}$$

# Correlation and regression

**Covariance:** Let  $X$  and  $Y$  be two random variables defined on the same probability space. The covariance of  $X$  and  $Y$ , denoted as  $cov(X, Y)$ , is defined as

$$cov(X, Y) = E[\{X - E[X]\}\{Y - E[Y]\}] \quad \text{OR}$$

$$cov(X, Y) = E[XY] - E[X] \cdot E[Y]$$

**Correlation coefficient:** The correlation coefficient of two random variables  $X$  and  $Y$ , denoted by  $\rho(X, Y)$ ,

is defined as  $\rho(X, Y) = \frac{cov(X, Y)}{\sqrt{var[X] \cdot var[Y]}} = \frac{cov(X, Y)}{\sqrt{\sigma_X \cdot \sigma_Y}}$ , where  $\sigma_X > 0$  and  $\sigma_Y > 0$ .

# It can be verify that  $-1 \leq \rho(X, Y) \leq 1$ .

## Two random variables  $X$  and  $Y$  are uncorrelated if

$$\rho(X, Y) = 0 \quad \text{or} \quad cov(X, Y) = 0$$

$\Rightarrow X$  and  $Y$  are independent random variables.

# Correlation and regression contd..

## In particular,  $E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy \cdot f_{X,Y}(x, y) dx dy$ ,  $E[X] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \cdot f_{X,Y}(x, y) dx dy$  and  $E[Y] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y \cdot f_{X,Y}(x, y) dx dy$ .

**Conditional Expectation:** Let  $(X, Y)$  be a two dimensional continuous random variable with joint continuous density function  $f_{X,Y}(x, y)$ , then the conditional expectation of  $g(X, Y)$  given  $X = x$ , denoted by  $E[g(X, Y)|X = x]$ , is defined as  $E[g(X, Y)|X = x] = \int_{-\infty}^{\infty} g(x, y) \cdot f_{Y|X}(y|x) dy$ .

In particular,  $E[Y|X = x] = \int_{-\infty}^{\infty} y \cdot f_{Y|X}(y|x) dy$ .

**Regression curve:**

1. The relation  $y = E[Y|X = x]$  is called the regression curve of  $Y$  on  $X = x$ .
2. The relation  $x = E[X|Y = y]$  is called the regression curve of  $X$  on  $Y = y$ .



**Example 2.** Let the joint p.d.f. of  $X$  and  $Y$  be  $f(x, y) = \begin{cases} (x + y), & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{otherwise.} \end{cases}$

Find (i)  $P\left[0 < X < \frac{1}{2}, 0 < Y < \frac{1}{4}\right]$ , (ii)  $E[X]$ ,  $E[Y]$ ,  $E[XY]$  and  $E[X + Y]$  (iii)  $\sigma(X, Y)$ .

**Solution:** We have  $P\left[0 < X < \frac{1}{2}, 0 < Y < \frac{1}{4}\right] = \int_0^{\frac{1}{2}} \left[ \int_0^{\frac{1}{4}} (x + y) dy \right] dx$

$$= \int_0^{1/2} \left[ xy + \frac{1}{2}y^2 \right]_0^{1/4} dx = \int_0^{1/2} \left( \frac{1}{4}x + \frac{1}{32} \right) dx = \frac{3}{64}.$$

Now  $E[X] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x \cdot f(x, y) dx dy = \int_0^1 \int_0^1 x \cdot (x + y) dx dy = \int_0^1 \left( x^2 + \frac{1}{2}x \right) dx = \frac{7}{12}.$

Similarly,  $E[Y] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y \cdot f(x, y) dx dy = \int_0^1 \int_0^1 y \cdot (x + y) dx dy = \frac{7}{12}.$

Now  $E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy \cdot f(x, y) dx dy = \int_0^1 \int_0^1 xy \cdot (x + y) dx dy = \frac{1}{3}.$

$$E[X + Y] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x + y) \cdot f(x, y) dx dy = \int_0^1 \int_0^1 (x + y) \cdot (x + y) dx dy = \frac{7}{6}.$$

We have  $cov[X, Y] = E[XY] - E[X] \cdot E[Y] = -\frac{1}{144}.$

$$E[X^2] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x^2 \cdot f(x, y) \, dx dy = \int_0^1 \int_0^1 x^2 \cdot (x + y) \, dx dy = \frac{5}{12}.$$

$$E[Y^2] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y^2 \cdot f(x, y) \, dx dy = \int_0^1 \int_0^1 y^2 \cdot (x + y) \, dx dy = \frac{5}{12}.$$

$$\text{Now } \text{var}[X] = E[X^2] - (E[X])^2 = \frac{5}{12} - \left(\frac{7}{12}\right)^2 = \frac{11}{144}.$$

$$\text{var}[Y] = E[Y^2] - (E[Y])^2 = \frac{11}{144}.$$

$$\text{Hence } \rho(X, Y) = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}[X] \cdot \text{var}[Y]}} = \frac{-\frac{1}{144}}{\sqrt{\frac{11}{144} \cdot \frac{11}{144}}} = -\frac{1}{11}.$$



THANK YOU

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