

- Lecture (11)
- Conditional expectations
 - Examples

Conditional expectation:-

Definition:-

Let (X, Y) be a two-dimensional r.v. and let $g(\cdot, \cdot)$ be a function of two variables. The conditional expectation of $g(X, Y)$ given $X=x$, denoted by:

$E[g(X, Y) | X=x]$, is defined to be

$$E[g(X, Y) | X=x] = \begin{cases} \sum_y g(x, y) f(y|x) & \text{if } Y \text{ is a discrete type} \\ \int_{-\infty}^{\infty} g(x, y) f(y|x) dy & \text{if } Y \text{ is continuous type} \end{cases}$$

Similarly

$$E[g(X, Y) | Y=y] = \begin{cases} \sum_x g(x, y) f(x|y) & \text{if } X \text{ is a discrete r.v.} \\ \int_{-\infty}^{\infty} g(x, y) f(x|y) dx & \text{if } X \text{ is a continuous r.v.} \end{cases}$$

EX:- Let $(X, Y) : (0, 0), (1, 2), (1, 3)$ with
 $f(x, y) : \frac{1}{3}, \frac{1}{3}, \frac{1}{3}$

Find $E(Y | X=1)$, $E(Y^2 | X=1)$ and $E[Y + \frac{1}{2} Y^2 | X=1]$

sol:-

$$E(Y|X=1) = \sum_{Y_K} Y_K f(Y|X=1)$$

$$f(Y|X) = \frac{f(x,y)}{f(x)}$$

$$X: 0, 1$$

$$f(x): \frac{1}{3}, \frac{2}{3}$$

$$Y|X=1: 2, 3$$

$$f(Y|X=1): \frac{1}{2}, \frac{1}{2}$$

$$f(Y_1|X=1) = \frac{f(1,2)}{f(X=1)} = \frac{\frac{1}{3}}{\frac{2}{3}} = \frac{1}{2}$$

$$f(Y_2|X=1) = \frac{f(1,3)}{f(X=1)} = \frac{\frac{1}{3}}{\frac{2}{3}} = \frac{1}{2}$$

Therefore

$$E(Y|X=1) = \sum_{Y_K} Y_K f(Y|X=1) = 2\left(\frac{1}{2}\right) + 3\left(\frac{1}{2}\right) = \frac{5}{2}$$

$$E(Y^2|X=1) = \sum_{Y_K} Y_K^2 f(Y|X=1) = (2)^2\left(\frac{1}{2}\right) + (3)^2\left(\frac{1}{2}\right) = \frac{13}{2}$$

$$\begin{aligned} E\left(Y + \frac{1}{2} Y^2 | X=1\right) &= E(Y|X=1) + \frac{1}{2} E(Y^2|X=1) \\ &= \frac{5}{2} + \frac{1}{2} \left(\frac{13}{2}\right) = \frac{5}{2} + \frac{13}{4} = \frac{10+13}{4} = \frac{23}{4} \end{aligned}$$

ex:-

$$\text{Let } (X, Y) \sim P(x, y) = \begin{cases} 1 & 0 \leq X, Y \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\text{Find } E(Y|X=x), E(Y^2|X=x), E(Y^2|X=\frac{1}{2})$$

sol:-

$$E(Y|X=x) = \frac{f(x,y)}{f(x)} = \int_0^1 f(x,y) dy = \int_0^1 1 dy = y \Big|_0^1 = 1$$

$$\Rightarrow f(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$f(y|X=x) = \begin{cases} 1 & 0 \leq y \leq 1, 0 \leq x \leq 1 \\ 0 & \text{o.w.} \end{cases}$$

$$E(Y|X=x) = \int_0^1 y dy = \frac{y^2}{2} \Big|_0^1 = \frac{1}{2}$$

$$E(Y^2|X=x) = \int_0^1 y^2 dy = \frac{y^3}{3} \Big|_0^1 = \frac{1}{3}$$

$$E(Y^2|X=\frac{1}{2}) = \int_0^1 y^2 dy = \frac{y^3}{3} \Big|_0^1 = \frac{1}{3}$$

Ex:- Let $(X, Y) \sim f(x, y) = \begin{cases} x+y & 0 < x, y < 1 \\ 0 & \text{o.w.} \end{cases}$

Find $E(Y|X=\frac{1}{2})$

Sol:-

$$E(Y|X=\frac{1}{2}) = \int_0^1 y f(Y|X=\frac{1}{2}) dy$$

$$f(Y|X) = \frac{f(X, Y)}{f(X)}$$

$$f(X) = \int_0^1 f(x, y) dy = \int_0^1 (x+y) dy = xy + \frac{y^2}{2} \Big|_0^1$$

$$= x + \frac{1}{2}$$

$$\therefore f(X) = \begin{cases} x + \frac{1}{2} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore f(Y|X) = \begin{cases} \frac{x+y}{x+\frac{1}{2}} & 0 < y < 1, 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore f(Y|X=\frac{1}{2}) = \begin{cases} \frac{y+\frac{1}{2}}{\frac{1}{2}+\frac{1}{2}} = \frac{y+\frac{1}{2}}{1} & 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore E(Y|X=\frac{1}{2}) = \int_0^1 y f(Y|X=\frac{1}{2}) dy$$

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$$= \int_0^1 y \cdot \frac{y + \frac{1}{2}}{\frac{1}{4}} dy = 4 \left[\frac{y^3}{3} + \frac{1}{2} \cdot \frac{y^2}{2} \right]_0^1$$
$$= 4 \left[\frac{y^3}{3} + \frac{y^2}{4} \right]_0^1 = \frac{7}{3}$$

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Theorem 1:

Let (X, Y) be a two-dimensional r.v. and $g_1(\cdot)$ and $g_2(\cdot)$ be functions of one variable. Then

- 1) $E_X[g_1(Y) + g_2(Y) | X=x] = E[g_1(Y) | X=x] + E[g_2(Y) | X=x]$
- 2) $E_Y[g_1(Y) g_2(X) | X=x] = g_2(x) E_Y[g_1(Y) | X=x]$

Theorem 2:

Let X, Y be two r.v.'s, having joint probability function $f(x, y)$ and marginal $f(x)$ and $f(y)$ respectively. Then X and Y are indep. iff:-

$f(x) = f(x|y)$ for all x and y .

proof:-

If X and Y are indep.

$f(x, y) = f(x) f(y)$. Therefore, for $f(y) > 0$, we have:-

$$f(x|y) = \frac{f(x, y)}{f(y)} = \frac{f(x) f(y)}{f(y)} = f(x).$$

$$\therefore f(x) = f(x|y)$$

Similarly, we can prove that if X and Y are independent then

$f(y) = f(y|x)$ for all x and y .

Example 1:

$$\text{Let } (X, Y) \sim f(x, y) = \begin{cases} 8xy & 0 < x < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

Find $E(Y|X=x)$, $E(XY|X=x)$ and $\text{Var}(Y|X=x)$

Sol:-

$$E(Y|X=x) = \int_{-\infty}^{\infty} y f(y|x) dy = \int_x^1 y f(y|x) dy$$

$$f(y|x) = \frac{f(x, y)}{f(x)}$$

$$f(x) = \int_x^1 f(x, y) dy = \int_x^1 8xy dy = 8x \left. \frac{y^2}{2} \right|_x^1$$

$$= \begin{cases} 4x(1-x^2) & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore f(y|x) = \frac{f(x,y)}{f(x)} = \frac{8xy}{4x(1-x^2)}$$

$$= \begin{cases} \frac{2y}{1-x^2} & x < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\therefore E(Y|X=x) = \int_x^1 y \cdot \frac{2y}{1-x^2} dy = \frac{2}{1-x^2} \int_x^1 y^2 dy$$

$$= \frac{2}{1-x^2} \left[\frac{y^3}{3} \right]_x^1 = \frac{2}{1-x^2} \left[\frac{1}{3} - \frac{x^3}{3} \right]$$

$$= \frac{2}{1-x^2} \cdot \frac{1-x^3}{3} = \begin{cases} \frac{2(1-x^3)}{3(1-x^2)} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$E(XY|X=x) = X E(Y|X=x)$$

$$= \begin{cases} \frac{2}{3} \cdot \frac{x(1-x^3)}{(1-x^2)} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$E(Y^2|X=x) = \int_x^1 y^2 f(y|x) dy = \int_x^1 y^2 \cdot \frac{2y}{1-x^2} dy$$

$$= \frac{2}{1-x^2} \int_x^1 y^3 dy = \frac{2}{1-x^2} \left(\frac{y^4}{4} \right) \Big|_x^1$$

$$= \begin{cases} \frac{1-x^4}{2(1-x^2)} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

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

$$= \begin{cases} \frac{1-x^4}{2(1-x^2)} - \frac{4}{9} \cdot \frac{(1-x^3)^2}{(1-x^2)^2} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

Ex:-
Let $(X, Y) \sim f(x, y) = \begin{cases} e^{-(x+y)}, & x > 0, y > 0 \\ 0, & \text{otherwise} \end{cases}$

- Find the marginal and conditional distributions.
- Find $P(X > 1)$, $P(0 < X < 1 | Y = 2)$, $E(X, Y)$ and $E(X, Y)$.
- Find the covariance and the correlation coefficient of X and Y , and then find the regression line of Y on X .

Sol:-

$$f_X(x) = \int_0^{\infty} f(x, y) dy = \int_0^{\infty} e^{-(x+y)} dy = e^{-x} \int_0^{\infty} e^{-y} dy$$

$$= \int_0^{\infty} e^{-x-y} dy = e^{-x} \left[-e^{-y} \right]_0^{\infty} = e^{-x} \cdot 1 = e^{-x} \quad x > 0$$

$$f_Y(y) = \int_0^{\infty} f(x, y) dx = \int_0^{\infty} e^{-(x+y)} dx = e^{-y} \int_0^{\infty} e^{-x} dx = e^{-y} \cdot 1 = e^{-y} \quad y > 0$$

Therefore

$$f(X|Y) = \frac{f(X, Y)}{f(Y)} = \frac{e^{-(x+y)}}{e^{-y}} = e^{-x} \quad x > 0$$

$$f(Y|X) = \frac{f(X, Y)}{f(X)} = \frac{e^{-(x+y)}}{e^{-x}} = e^{-y} \quad y > 0$$

Since the conditional distributions are the same of the marginal distributions, this means X and Y are independent r.v's.

$$b) P(X > 1) = \int_1^{\infty} f_X(x) dx = \int_1^{\infty} e^{-x} dx = e^{-1}$$

$$P(0 < X < 1 | Y = 2) = \int_0^1 f(X|Y=2) dx = \int_0^1 e^{-x} dx = 1 - e^{-1}$$

$$E(X+Y) = E(X) + E(Y) = \int_0^{\infty} x e^{-x} dx + \int_0^{\infty} y e^{-y} dy$$

$$\text{let } u = x \Rightarrow du = dx$$

$$dy = e^{-x} dx \Rightarrow y = \int e^{-x} dx = -e^{-x}$$

$$\Rightarrow \int_0^{\infty} x e^{-x} = -x e^{-x} \Big|_0^{\infty} + \int_0^{\infty} e^{-x} dx$$

$$= 0 + -e^{-x} \Big|_0^{\infty} = 1$$

$$\therefore \int_0^{\infty} y e^{-y} dy = 1 \Rightarrow E(X+Y) = 1+1=2$$

$$E(XY) = \int_0^{\infty} \int_0^{\infty} xy e^{-(x+y)} dx dy$$

$$= \int_0^{\infty} y e^{-y} \left[\int_0^{\infty} x e^{-x} dx \right] dy = \int_0^{\infty} y e^{-y} dy = 1$$

$$c) \text{Cov}(X, Y) = E(XY) - E(X) E(Y) = 1 - (1)(1) = 0$$

Note, this result can be obtained directly since X and Y are indep. as shown in part a)

To find the regression curve of Y on X , we should note that it is equal to $E(Y|X=x)$. Therefore

$$E(Y|X=x) = \int_0^{\infty} y f(Y|X=x) dy = \int_0^{\infty} y e^{-y} dy = 1$$

That is the regression curve of Y on X is constant, it does not depend on the value of X (why?).

Ex:-

Suppose that X and Y are two jointly distributed discrete r.v.s. Let X has a Bernoulli distⁿ with parameter p . That is, $p(X=1) = p = 1 - p(X=0)$. Suppose that

$E(Y|X=0) = 1$ and $E(Y|X=1) = 2$. What is $E(Y)$?

Sol:-

$$E(Y) = E[E(Y|X=x)]$$

$$= \sum_{x=0}^1 E(Y|X=x) f(x)$$

$$= (1)(1-p) + 2(p) = 1 - p + 2p = 1 + p$$

Ex:-

The discrete density of X is given by:-

$$f(x) = \begin{cases} \frac{x}{3} & x=1, 2 \\ 0 & \text{o.w.} \end{cases}$$

Let $f(Y|X)$ be a Binomial density with parameters X and $\frac{1}{2}$, that is

$$f(Y|X) = p(Y=y|X=x) = \begin{cases} \binom{x}{y} \left(\frac{1}{2}\right)^y \left(\frac{1}{2}\right)^{x-y} & y=0, 1, \dots, x \\ 0 & \text{o.w.} \end{cases}$$

$$= \begin{cases} \binom{x}{y} \left(\frac{1}{2}\right)^x & y=0, 1, 2, \dots, x \\ 0 & \text{o.w.} \end{cases}$$

Finda) $E(Y)$ b) The joint distⁿ of X and Y Sol:-

$$a) E(Y) = E[E(Y|X=x)] = \sum_Y Y f(Y|X=x) = E(Y|X)$$

$$= \sum_{y=0}^x Y \binom{x}{y} \left(\frac{1}{2}\right)^x = \frac{x}{2} \quad \text{ans.}$$

The expected value of Y having $\text{Bin}(X, \frac{1}{2})$

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$$\begin{aligned} \therefore E(Y)E\left[\frac{1}{2}X\right] &= \frac{1}{2}E(X) = \frac{1}{2}\sum X P(X) \\ &= \frac{1}{2}\sum_{X=1}^2 X \cdot \frac{X}{3} = \frac{1}{6}(1^2 + 2^2) = \frac{5}{6} \end{aligned}$$

6. To find the joint distⁿ. of X and Y , we have;

$$f(x, y) = f(y|x)f(x) = \begin{cases} \binom{x}{y} \left(\frac{1}{2}\right)^x \left(\frac{x}{3}\right) & , x=1, 2, \\ & Y=1, 2, \dots, x \\ 0 & \text{o.w.} \end{cases}$$