

LECTURE - 32

Agenda:

- ① Conditional expectation of random variables
- ② Properties involving conditional expectations

CONDITIONAL EXPECTATION OF RANDOM VARIABLES

Let us recollect that for two random variables X and Y , the conditional distribution of X given $Y = y$, describes the probability behaviour of the random variable X given the information that $Y = y$. In the same spirit, we defined the notion of the conditional expectation of X given $Y = y$.

Definition: If X and Y are two continuous random variables with joint probability density $f_{X,Y}$, then the conditional expectation of X given $Y = y$ is defined as

$$E(X|Y=y) = \int_{-\infty}^{\infty} x \underbrace{f_{X|Y=y}(x)}_{\text{conditional probability density of } X \text{ given } Y=y} dx.$$

conditional probability density of X given $Y = y$.

Definition: If X and Y are two discrete random variables with joint probability mass function $p_{X,Y}$, then the conditional expectation of X given $Y = y$ is defined as

$$E(X|Y=y) = \sum_{x \in \mathcal{X}} x \underbrace{p_{X|Y=y}(x)}$$

Conditional probability mass function of X given $Y=y$

Example: A soft-drink machine has a random supply Y at the beginning of a given day and dispenses a random amount X during the day (with measurements in gallons). It has been observed that X and Y have joint density

$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{2} & \text{if } 0 \leq x \leq y, 0 \leq y \leq 2, \\ 0 & \text{otherwise.} \end{cases}$$

~~Exercise~~ If it is known that on a particular day 1 gallon was supplied at the beginning of the day, ~~find~~ ^{find} the expected value of the amount of soft-drink consumed in that day.

Note that we need to find $E(X|Y=1)$.

Let us first find the marginal density of Y .

$$f_Y(y) = \begin{cases} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx & \text{if } 0 \leq y \leq 2, \\ 0 & \text{otherwise.} \end{cases}$$

$$= \begin{cases} \int_0^y \frac{1}{2} dx & \text{if } 0 \leq y \leq 2, \\ 0 & \text{otherwise.} \end{cases}$$

$$= \begin{cases} \frac{y}{2} & \text{if } 0 \leq y \leq 2, \\ 0 & \text{otherwise.} \end{cases}$$

Hence, the conditional density of X given $Y=y$ is given by

$$f_{X|Y=y}(x) = \begin{cases} \frac{f_{X,Y}(x,y)}{f_Y(y)} & \text{if } 0 \leq x \leq y, \\ 0 & \text{otherwise.} \end{cases}$$

$$= \begin{cases} \frac{1/2}{y/2} & \text{if } 0 \leq x \leq y, \\ 0 & \text{otherwise.} \end{cases}$$

$$= \begin{cases} \frac{1}{y} & \text{if } 0 \leq x \leq y. \\ 0 & \text{otherwise.} \end{cases}$$

Hence,

$$\begin{aligned} E(X|Y=1) &= \int_{-\infty}^{\infty} x f_{X|Y=1}(x) dx \\ &= \int_0^1 x \cdot 1 dx \\ &= \frac{1}{2}. \end{aligned}$$

Hence, the expected amount of the amount of soft drink consumed in that day is $\frac{1}{2}$ gallon.

PROPERTIES INVOLVING CONDITIONAL EXPECTATIONS

Result: Let X and Y ~~be~~ be two random variables. Then,

$$E(X) = E(E(X|Y)),$$

where on the right-hand side, the inside expectation is stated with respect to the conditional distribution of X given Y , and the outside expectation is stated with respect to the distribution of Y .

CONFUSED ??

Sometimes, the way the distribution of the random variable X is specified, it is not possible to directly evaluate its ~~expectation~~ expectation, and the previous identity comes in very handy.

HERE IS AN ALTERNATE WAY OF EVALUATING $E(X)$ BASED ON THE PREVIOUS IDENTITY.

- ① For every y , find $E(X|Y=y)$.
- ② Define the function h as $h(y) \triangleq E(X|Y=y)$.
- ③ ~~$E(X)$~~ $E(h(Y))$.

Hence, the way to interpret the previous relationship is as follows.

$$E(X) = E(h(Y)), \text{ where } h(y) = E(X|Y=y).$$

Example: A quality control plan for an assembly line involves sampling n finished items per day and counting X , the number of defective items. If p denotes the probability that an item is defective, then given p , X has a binomial distribution with parameters n and p . However, it is observed that p is a random quantity,

and has a uniform distribution on the interval $[0, \frac{1}{4}]$.

(a) Find $E(X)$.

Note that given $p = p_0$, X has a binomial distribution with parameters n and p_0 . Hence,

$$E(X|p=p_0) = np_0.$$

Hence, ~~$E(X) = E(np)$~~ $E(X) = E(E(X|p))$
 $= E(h(p)),$ where $h(p) = E(X|p=p)$.

Hence, $E(X) = E(np)$
 $= n E(p)$
 $= n \left(\frac{0 + \frac{1}{4}}{2} \right)$

($\because p$ is uniform on the interval $[0, \frac{1}{4}]$)
 $= \frac{n}{8}.$

(b) Find $SD(X)$.

Result: Let X and Y be two random variables. Then,
 $V(X) = E(V(X|Y)) + V(E(X|Y)).$

HERE IS AN ALGORITHM BASED ON THIS IDENTITY.
TO EVALUATE $V(X)$.

- (1) For every y , find $E(X|Y=y)$ and $V(X|Y=y)$.
- (2) Define the function h as $h(y) \triangleq E(X|Y=y)$.
- (3) Define the function \tilde{h} as $\tilde{h}(y) \triangleq V(X|Y=y)$.
- (4) $V(X) = \text{} V(h(Y)) + E(\tilde{h}(Y))$.

(Note that for any function g , $E(g(X)|Y=y)$ is $\int_{-\infty}^{\infty} g(x) f_{X|Y=y}(x) dx$ if X is continuous, and $\sum_{x \in \mathcal{X}} g(x) p_{X|Y=y}(x)$ if X is discrete.)

Hence, the way to interpret the previous relationship is as follows.

$$V(X) = V(h(Y)) + E(\tilde{h}(Y)), \text{ where } h(y) = E(X|Y=y) \\ \text{and } \tilde{h}(y) = V(X|Y=y).$$

Returning to the example,

$$V(X) = \text{} V(E(X|p)) + E(V(X|p)) \\ = V(np) + E[np(1-p)] \\ = n^2 V(p) + nE(p) - nE(p^2)$$

$$= (n^2 - n)V(p) + nE(p) - n(E(p))^2$$

$$(\because E(p^2) = V(p) + (E(p))^2)$$

$$= (n^2 - n) \cdot \frac{\left(\frac{1}{4}\right)^2}{\frac{1}{12}} + \frac{n}{8} - \frac{n}{64}$$

$$= \frac{n^2}{192} + \frac{5n}{48}$$

$$\text{Hence, } SD(x) = \sqrt{V(x)} = \sqrt{\frac{n^2}{192} + \frac{5n}{48}}$$