

Probability and Statistics: Lecture-30

Monsoon-2020

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on October 21, 2020

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 $g(y) = ay + b, \quad E[X|Y] = aY + b$
- * Since $E[X | Y]$ is a RV, we can find its PMF, CDF, Variance, etc

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Solved Example

Let X, Y be RV with joint PMF given as follows

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	$Y = 0$	$Y = 1$
$X = 0$	$\frac{1}{5}$	$\frac{2}{5}$
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1. Find the **Marginal PMFs** of X and Y .
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need $P_{X|Y}(x|y)$
 $\sum_x P_{X|Y}(x|y)$

$$P(0,0) = \frac{1}{5}$$

$$= P_X(0) P_Y(0) ?$$

$$\frac{1}{5} - \frac{2}{5} = \frac{9}{25}$$

$$\Rightarrow P(0,0) \neq P_X(0) P_Y(0)$$

$\Rightarrow X, Y$ not independent.

Let $Z = E[X | Y]$.

1. Find the Marginal PMFs of X and Y .
2. Find the conditional PMF of X given $Y=0$ and $Y=1$, i.e., find $P_{X|Y}(x|0)$ and $P_{X|Y}(x|1)$
3. Find the PMF of Z
4. Find $E[Z]$, and check that $E[Z] = E[X]$
5. Find $\text{Var}(Z)$

» Answer to previous problem...

$$a) P_x(0) = \frac{1}{5} + \frac{2}{5} = \frac{3}{5}$$

$$P_x(1) = \frac{2}{5} + 0 = \frac{2}{5}$$

$$P_y(0) = \frac{1}{5} + \frac{2}{5} = \frac{3}{5}, \quad P_y(1) = \frac{2}{5} + 0 = \frac{2}{5}$$

$$P_x(x) = \begin{cases} 3/5, & x=0 \\ 2/5, & x=1 \end{cases} \quad \left. \vphantom{\begin{matrix} 3/5 \\ 2/5 \end{matrix}} \right\} \text{Bernoulli}$$

$$P_y(y) = \begin{cases} 3/5, & y=0 \\ 2/5, & y=1 \end{cases} \quad \left. \vphantom{\begin{matrix} 3/5 \\ 2/5 \end{matrix}} \right\} \text{Bernoulli}$$

Q Are x, y independent? Ans No

Recall: $P_{x,y}(x,y) = P_x(x)P_y(y)$
 \uparrow
 found in table $\forall x, y$

b) We need to find $P_{x|y}(x|y)$

$$P_{x|y}(0|0) = \frac{P_{xy}(0,0)}{P_y(0)} = \frac{1/5}{3/5} = \frac{1}{3}$$

$$P_{x|y}(1|0) = \frac{P_{xy}(1,0)}{P_y(0)} = \frac{2/5}{3/5} = 2/3$$

Similarly, find

$$P_{x|y}(0|1) = \frac{P_{xy}(0,1)}{P_y(1)} = \frac{2/5}{2/5} = 1$$

$$P_{x|y}(1|1) = 0$$

In words: Given $y=1$, we always have $x=0$.

» Answer to previous problem...

③ $z = E[X|Y]$, Noting Y takes 0, 1
value.

$$= \begin{cases} E[X|Y=0] & \text{if } y=0 \\ E[X|Y=1] & \text{if } y=1 \end{cases}$$

$$E[X|Y=0] = \sum_{x_i} x_i P_{X|Y}(x|y=0)$$

$$= 0 \cdot P_{X|Y}(0|0) + 1 \cdot P_{X|Y}(1|0)$$

$$= 0 \cdot \frac{1}{3} + 1 \cdot \frac{2}{3} = \frac{2}{3}$$

$$E[X|Y=1] = \sum_{x_i} x_i P_{X|Y}(x|y=1)$$

$$= 0 \cdot P_{X|Y}(0|1) + 1 \cdot P_{X|Y}(1|1)$$

$$= 0 \cdot 1 + 1 \cdot 0 = 0$$

Since, $P(Y=0) = \frac{2}{5}$, $P(Y=1) = \frac{2}{5}$

$$z = \begin{cases} 2/3 & \text{if } y=0 \\ 0 & \text{if } y=1 \end{cases} \quad \textcircled{*}$$

We need $P_z(z)$, that what
is the probab. of $z=z$
i.e. $P(z=2/3)$ & $P(z=0)$?

④ i.e.
i.e.) $\begin{cases} P(z=2/3) = P(Y=0) \\ P(z=0) = P(Y=1) \end{cases}$

» Answer to previous problem...

$$P_Z(z) = \begin{cases} 3/5 & \text{if } z = 2/3 \\ 2/5 & \text{if } z = 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\textcircled{c} E[Z] = \frac{2}{3} \cdot \frac{3}{5} + 0 \cdot \frac{2}{5} = \frac{2}{5}$$

$$E[X] = 0 \cdot \frac{3}{5} + 1 \cdot \frac{2}{5} = \frac{2}{5}$$

$$\Rightarrow E[Z] = E[X]$$

$$\Rightarrow \underbrace{E[E[X|Y]]}_Z = E[X]$$

true for $\forall X, Y$?
even for X, Y ind?

$$E[E[X|Y]] = E[X] \quad (\text{Called Law of Iterated Expectation})$$

$$\text{Var}(Z) = E[Z^2] - (E[Z])^2$$

$$E[Z^2] = \sum z_i^2 P_Z(z)$$

$$= \left(\frac{2}{3}\right)^2 \cdot \frac{3}{5} + 0^2 \cdot \frac{2}{5} + 0^2 \cdot \frac{1}{5}$$

$$= \frac{4}{9} \cdot \frac{3}{5} = \frac{4}{15}$$

$$\text{Var}(Z) = \frac{4}{15} - \left(\frac{2}{5}\right)^2$$

» Fact...

$$E[c g(x)] = c E[g(x)]$$

Fact

Let X, Y be two RVs and g, h be two functions of X and Y respectively. Show that

$$E[g(X)h(Y) | X] = g(X)E[h(Y) | X]$$

Solution

$$E[g(x)h(Y) | X=x] \stackrel{\text{Lin. of Exp.}}{=} g(x) E[h(Y) | X=x].$$

↑ ↑
become constant fixed instantiated

Above line $\forall X=x$.

$$\Rightarrow E[g(x)h(Y) | X] = g(x) E[h(Y) | X].$$

» Iterated Expectations...

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Law of Iterated Expectations

Let X, Y be two RVs, then we have

$$E[X] = E[E[X | Y]]$$

Proof

Let $g(Y) = E[X | Y]$. Applying Law of Total Probability for Expectation.

$$\begin{aligned} E[X] &= \sum_{y_j \in R_Y} E[X | Y = y_j] P_Y(y_j) \\ &= \sum_{y_j \in R_Y} g(y_j) P_Y(y_j) \stackrel{\text{def}}{=} E[g(Y)] = E[\underbrace{E[X | Y]}_{g(Y)}] \end{aligned}$$

Partim: $\{Y = y_i\} = B_i$
 $UB_i = S$
 Φ

» Solved Example on Application of Iterated Expectation...

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Solved Example 2

Number of customers N visiting a fast food restaurant follows Poisson distribution $N \sim \text{Poisson}(\lambda)$.

~~$N \sim \text{Poisson}(\lambda)$~~ $X | N=n \sim \text{Binomial}(n, p)$

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Number of customers N visiting a fast food restaurant follows Poisson distribution $N \sim \text{Poisson}(\lambda)$. Each customer arriving in this restaurant purchases a drink with probability p ,

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Solved Example 2

Number of customers N visiting a fast food restaurant follows Poisson distribution $N \sim \text{Poisson}(\lambda)$. Each customer arriving in this restaurant purchases a drink with probability p , which is independent from other customers. What is the average number of customers who purchase drinks?

Solution

X : # customers who purchase drinks with prob. p
 N : # customer visiting fast food rest.

Recall: $X|N=n \sim \text{Binomial}(n, p)$

$$\text{To find } E[X] \stackrel{\text{L.O.I.E}}{=} E[E[X|N]] \stackrel{X|N \sim \text{Bin}(n,p)}{=} E[Np] = p E[N]$$

Since $N \sim \text{Poisson}(\lambda)$.

$$\Rightarrow E[X] = p\lambda$$

» Expectation for Independent RV...

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Recall X, Y ind $\Rightarrow \underline{P_{X|Y}(x|y)} = P_X(x)$.

Expectation for Independent RVs

Let X, Y be two independent RVs. Then we have the following

- ✓ 1. $E[X | Y] = E[X]$
- ✓ 2. $E[g(X) | Y] = E[g(X)]$
- ✓ 3. $E[XY] = E[X]E[Y]$
- ✓ 4. $E[g(X)h(Y)] = E[g(X)]E[h(Y)]$

» Answer to previous problem...

$$\textcircled{1} \quad E[X|Y=y] = \sum_{x \in \mathcal{R}_X} x P_{X|Y}(x|y) \rightarrow \textcircled{*}$$

$$\text{Since } P_{X|Y}(x|y) = P_X(x)$$

$$\Rightarrow \textcircled{*} \quad \sum_{x \in \mathcal{R}_X} x P_X(x) = E[X].$$

True for $\forall y \in \mathcal{R}_Y$.

$$\Rightarrow E[X|Y] = E[X], \text{ when } X, Y \text{ ind.}$$

$$\textcircled{2} \quad \text{Similarly.} \quad g(x, y) = xy. \quad E[g(X)|Y] = E[\underline{g(X)}]$$

$$\textcircled{3} \quad E[XY] = \sum_x \sum_y xy P_{X,Y}(x, y)$$

$$= \sum_x \sum_y xy P_X(x) P_Y(y)$$

$$= \sum_x x P_X(x) \sum_y y P_Y(y)$$

$$= E[X] E[Y]$$

» Answer to previous problem...

④ Easy. Try!

» Conditional Variance...

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Let X, Y be two RVs. By $\text{Var}(X \mid Y = y)$ the **conditional variance** of X given $Y = y$. Let

$\mu_{X|Y}(y) = E[X \mid Y = y]$. Then

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Definition of Conditional Variance

Let X, Y be two RVs. By $\text{Var}(X \mid Y = y)$ the conditional variance of X given $Y = y$. Let $\mu_{X|Y}(y) = E[X \mid Y = y]$. Then

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

Proof

Assignment