

Recall :- X - a discrete random variable. $E[X] < \infty$

• $X \geq 0, c > 0$
$$\underbrace{P(X \geq c)}_{\text{tail probabilities}} \leq \frac{E[X]}{c} \quad [\text{Markov inequality}]$$

• $\mu = E[X] < \infty, \sigma = SD[X] = \sqrt{E[(X - E[X])^2]} < \infty$

$k > 0, \underbrace{P(|X - \mu| \geq k\sigma)}_{X \notin (\mu - k\sigma, \mu + k\sigma)} \leq \frac{1}{k^2} \quad [\text{Tchebyshev inequality}]$

• A - Event $P(A) > 0$ $X|A := Y$ Conditional distribution of X given A

$P(Y=t) \text{ or } P(X=t|A) = \frac{P(X=t, A)}{P(A)}$

• Conditional Expectation of X given A

$$E[X|A] = \sum_{t \in \text{Range}(X)} t \cdot P(X=t|A)$$

• Conditional variance of X given A

$$\begin{aligned} \text{Var}[X|A] &= E[(X - E[X|A])^2 | A] \\ &= \sum_{t \in \text{Range}(X)} (t - E[X|A])^2 P(X=t) \end{aligned}$$

Example 4.4.3:-

Deck of cards — standard 52 cards.

— Cards are drawn, one at a time, randomly with replacement.

— $X \equiv \#$ of draws till the first King is drawn.
 $Y \equiv \#$ of draws till the first Ace is drawn.

Question:- $E[X | Y=3] = ?$

$$L: \sum_{t \in \text{Range}(X)} t P(X=t | Y=3)$$

$Y=3$ $\equiv$ Ace was drawn for the 1st time in 3rd draw.
[Ace was not drawn in 1 and 2]

$$P(X=t | Y=3) = P(\text{King was drawn for 1st time at } t^{\text{th}} \text{ draw} | \underline{Y=3}) = ?$$

$$P(\text{King is drawn in } t^{\text{th}} \text{ draw} | Y=3) = \begin{cases} \frac{4}{48} & t=1, 2 \\ 0 & t=3 \\ \frac{4}{52} & t \geq 4 \end{cases}$$

$$\Rightarrow P(X=t | Y=3) = \begin{cases} \left(\frac{44}{48}\right)^{t-1} \cdot \frac{4}{48} & t=1, 2 \\ 0 & t=3 \\ \left(\frac{44}{48}\right)^2 \left(\frac{48}{52}\right)^{t-4} \left(\frac{4}{52}\right) & t \geq 4 \end{cases}$$

$t \geq 4$, $X=t \equiv$ no king in 1 and 2 draws - $\frac{44}{48}$
 $\equiv$ at 3rd draw given Ace ✓
 $\equiv$ no king for rest of $t-4$ draws and draw a king in t draw

$$\begin{aligned}
 E[X | Y=3] &= \sum_{t=1}^{\infty} t \cdot P(X=t | Y=3) \\
 &= 1 \cdot \frac{4}{48} + 2 \left(\frac{44}{48} \right) \frac{4}{48} + \sum_{t=4}^{\infty} t \left(\frac{44}{48} \right)^2 \left(\frac{48}{52} \right)^{t-4} \frac{4}{52}
 \end{aligned}$$

$$\underline{E_x} = \frac{985}{72} \quad \left[\text{Geometric Series } \sum_{n=1}^{\infty} n r^{n-1} \equiv ? \right]$$

$$\approx 13.68$$

Given Ace appeared for the 1st time in 3rd draw, the
 Conditional expectation of $X \approx \underline{\underline{13.68}}$.

$X \stackrel{d}{=} \text{Geometric}(4/52)$; $E(X) = \frac{1}{52/4} = \frac{4}{52} = 13.$
 [last weeks] □

Example 4.4.7 $Y \sim \text{Uniform}\{1, 2, \dots, n\}$ for some $n \geq 1$
 $X = \#$ of heads on Y flips of a fair coin.

What is $E[X]$?

• $\text{Range}(Y) = \{1, \dots, n\}$

$E(X|Y=y) \equiv \checkmark$ know how to do.
 $X|Y=y \stackrel{d}{=} \text{Binomial}(y, \frac{1}{2})$.

$$\Rightarrow E[X|Y=y] = y \cdot \frac{1}{2} = \frac{y}{2}$$

• Define: $g: \{1, \dots, n\} \rightarrow \mathbb{R}$ by
 $g(y) := E[X|Y=y]$
 $= \frac{y}{2}$ [By above computation]

• Q: $E[g(Y)] = ?$

By def: $E[g(Y)] = \sum_{y \in \text{Range}(Y)} g(y) P(Y=y) = \sum_{y=1}^n \frac{y}{2} \cdot \frac{1}{n}$

$$= \frac{1}{2n} \sum_{y=1}^n y = \frac{1}{2n} \cdot \frac{n(n+1)}{2} = \frac{n+1}{4}$$

claim:- $E[X]$

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

$$E[g(Y)] = \sum_{y \in \text{Range}(Y)} g(y) P(Y=y)$$

$$= \sum_{y=1}^{\infty} \underbrace{E[X | Y=y]} P(Y=y)$$

$$= \sum_{y=1}^{\infty} \left[\sum_{t \in \text{Range}(X)} t P(X=t | Y=y) \right] \cdot P(Y=y)$$

$$= \sum_{y=1}^{\infty} \left[\sum_{t \in \text{Range}(X)} t \frac{P(X=t, Y=y)}{P(Y=y)} \right] P(Y=y)$$

$$= \sum_{y=1}^{\infty} \sum_{t \in \text{Range}(X)} t P(X=t, Y=y)$$

$$= \sum_{t \in \text{Range}(X)} t \left[\sum_{y=1}^{\infty} P(X=t, Y=y) \right]$$

$$= \sum_{t \in \text{Range}(X)} t P\left(\bigcup_{y=1}^{\infty} (X=t, Y=y)\right)$$

$$= \sum_{t \in \text{Range}(X)} t P(X=t) = E[X]$$

The previous computation proves the below
Theorem

Theorem 4.4.6 : let X and Y be two discrete random variables on the sample space S .

let $g: \text{Range}(Y) \rightarrow \mathbb{R}$ given by

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

$$E[g(Y)] = E[X].$$

Remarks:-

$$g(y) := E[X | Y=y] \quad | \quad g(Y) = E[X | Y]$$

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

Expectation
of Y

conditional Expectation of
 X given $Y=y$

$$\text{Var}(X) = E[\text{var}[X | Y]] + \text{var}[E[X | Y]] \quad - \text{true}$$

$$h(y) = \text{var}[X | Y=y] \quad = E[h(Y)] + \text{var}(g(Y))$$

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

Definition 4.5.1. (Covariance of X and Y)

X and Y be two discrete random variables on a sample space S . Then the "covariance of X and Y " is defined as

$$\text{Cov}[X, Y] = E[(X - E[X])(Y - E[Y])]$$

↑
joint distribution

over X over Y

$$\left. \begin{array}{l} X > E[X] \text{ at the same time } Y > E[Y] \\ X < E[X] \text{ at the same time } Y < E[Y] \end{array} \right\} \Rightarrow \text{Cov}[X, Y] > 0$$

[Positively correlated]

$$\left. \begin{array}{l} X < E[X] \text{ at the time } Y > E[Y] \\ Y < E[Y] \text{ at the time } X > E[X] \end{array} \right\} \Rightarrow \text{Cov}[X, Y] < 0$$

[negatively correlated]

$$E[(X - E[X])(Y - E[Y])] = 0 \Leftrightarrow \text{Cov}[X, Y] = 0 \Rightarrow X \text{ and } Y \text{ uncorrelated}$$

$$\begin{array}{c} \Uparrow \\ E[XY - XE[Y] - E[X]Y + E[X]E[Y]] = 0 \end{array}$$

$$\begin{array}{c} \Downarrow \\ E[XY] - E[X]E[Y] - E[X]E[Y] + E[X]E[Y] = 0 \quad (=) \quad E[XY] = E[X]E[Y] \end{array}$$

Remark: X and Y are independent $\Rightarrow E[XY] = E[X]E[Y]$
(shown earlier)

Theorem :- let X and Y be discrete random variables with finite mean for which $E[XY]$ is also finite. Then

$$\text{Cov}[X, Y] = E[XY] - E[X]E[Y].$$

Ex: Z is another discrete random variable. $\alpha, \beta \in \mathbb{R}$

(a) $\text{Cov}[X, Y] = \text{Cov}[Y, X]$

(b) $\text{Cov}[X, \alpha Y + \beta Z] = \alpha \text{Cov}[X, Y] + \beta \text{Cov}[X, Z]$

(c) X and Y are independent with a finite covariance then $\text{Cov}[X, Y] = 0$

Example 4.5.6: X, Y with $\text{Range} = \{-1, 1\}$
be two random variables

$$P(X=x, Y=y) \equiv P(x, y)$$

$P(X=x, Y=y)$	$X=1$	$X=-1$	
$Y=1$	0.3	0.2	$P(Y=1) = 0.5$
$Y=-1$	0.3	0.2	$P(Y=-1) = 0.5$
	$P(X=1) = 0.6$	$P(X=-1) = 0.4$	

$\text{Cov}[X, Y] = ?$

$$= E[XY] - E[X]E[Y]$$

$$E[XY] = \sum_{\substack{x \in \text{Range}(X) \\ y \in \text{Range}(Y)}} xy P(X=x, Y=y) = 1 \cdot 1 (0.3) + (-1) \cdot 1 (0.2) + 1 \cdot (-1) (0.3) + (-1) \cdot (-1) (0.2) = 0$$

$$\begin{aligned}
 E[X] &= \sum_{x \in \text{Range}(X)} x P(X=x) = 1 P(X=1) + (-1) P(X=-1) \\
 &= 1 (0.6) + (-1) (0.4) \\
 &= 0.2
 \end{aligned}$$

$$E[Y] = \sum_{y \in \text{Range}(Y)} y P(Y=y) = 1 (0.5) + (-1) (0.5) = 0$$

$$\therefore \text{Cov}[X, Y] = E[XY] - E[X] E[Y] = 0.$$

Question :
[BP]

$$E(X|A) := \sum_{t \in T} t \mathbb{P}(X=t | A)$$

$$[MT] \quad \left\{ \begin{array}{l} E(X|A) := E[X | \sigma(A)], \quad \sigma(A) = \{\emptyset, A, A^c, \Omega\} \\ Y \equiv \sigma(A) \text{ m.b.k.} \quad \Leftarrow \\ \int_B Y dP = \int_B X dP \quad \forall B \in \sigma(A) \end{array} \right.$$

Is [BP] $\equiv$ [MT] definition?