

ASSIGNMENT-2

1. The table can be shown as :

$X \backslash Y$	1	2	3	total
1	0.1	0.2	0.1	0.4
2	0.2	0.3	0.0	0.5
3	0.0	0.1	0.0	0.1
total	0.3	0.6	0.1	

a) The marginal probability distribution of X is given by

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

Hence, using formula,

Marginal distribution of X :

X	1	2	3
$P(X)$	0.4	0.5	0.1

marginal distribution of Y :

Y	1	2	3
$P(Y)$	0.3	0.6	0.1

b) Conditional probability of X given Y is

$$\begin{aligned} P(X|Y) &= \frac{P(X=X, Y=Y)}{P(Y)} \\ &= \frac{P(X, Y)}{P(Y)} \end{aligned}$$

So, we need $P(X=2 | Y=1)$

$$\begin{aligned} &= \frac{P(X=2, Y=1)}{P(Y=1)} \\ &= \frac{0.2}{0.3} = \frac{2}{3} = 0.66 \end{aligned}$$

c) Two RVs, X & Y are said to be independent if,

$$P(X, Y) = P(X) P(Y)$$

where, $P(X) \rightarrow$ marginal of X
 $P(Y) \rightarrow$ marginal of Y
+ pairs of (X, Y)

So,

say $(0, 0)$

$$P(0, 0) = 0.1$$

$$P(X=1) = 0.4$$

$$P(Y=1) = 0.3$$

Clearly,

$$P(1, 1) \neq P(X=1) \cdot P(Y=1)$$

$$0.1 \neq (0.4) \times (0.3)$$

Hence, the two random variables are not independent.

2. given PDF is

$$f(x) = 3x^2, \quad 0 \leq x \leq 1$$

$$f(x) = 0, \quad \text{otherwise}$$

a) The CDF is given by

$$F(x) = \int_{-\infty}^{x=x} f(y) dy$$

Hence, breaking it down.

when $x < 0$,

$$\cancel{F(x=0)} \quad F(x) = \int_{-\infty}^x f(y) dy$$

$$= 0$$

when $0 \leq x \leq 1$,

$$F(x) = \int_{-\infty}^x f(y) dy$$

$$= \int_{-\infty}^0 f(y) dy + \int_0^x f(y) dy$$

$$= 0 + \int_0^x 3y^2 dy$$

$$= \left[y^3 \right]_0^x$$

$$= x^3$$

For

$$x > 1$$

$$F(x) = \int_{-\infty}^x f(y) dy$$

$$= \int_{-\infty}^1 f(x) dx + \int_1^x f(x) dx$$

$$= (1)^3 + 0$$

$$= 1$$

Hence,

$$F(x) = \begin{cases} 0 & , x < 0 \\ x^3 & , 0 \leq x \leq 1 \\ 1 & , x > 1 \end{cases}$$

b) Mean of random variable X is given by:

$$\text{Mean} = \int_{-\infty}^{\infty} x f(x) dx$$

where $f(x)$ is PDF

$$\text{So, mean} = \int_{-\infty}^0 x \cdot f(x) dx + \int_0^1 x \cdot f(x) dx + \int_1^{\infty} x \cdot f(x) dx$$

$$= 0 + \int_0^1 x \cdot (3x^2) dx + 0$$

$$= 3 \int_0^1 x^3 dx$$

$$= 3 \cdot \left(\frac{x^4}{4} \right) \Big|_0^1$$

$$= \frac{3}{4}$$

So,

$$\text{mean}(x) = \frac{3}{4}$$

c)

$$\text{Variance}(x) = E(x^2) - (E(x))^2$$

where $E(x)$ is
the mean of x
 $E(x^2)$ is the
mean of x^2 .

Now,

$$E(x^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$$

$$= \int_0^1 x^2 (3x^2) dx$$

$$= 3 \int_0^1 x^4 dx$$

$$= \frac{3}{5} (x^5) \Big|_0^1$$

$$= \frac{3}{5}$$

and, $(E(x))^2 = \left(\frac{3}{4} \right)^2 = \frac{9}{16}$

So, $\text{variance}(x) = E(x^2) - (E(x))^2$

$$\begin{aligned}
 &= \frac{3}{5} - \frac{9}{16} \\
 &= \frac{16 \times 3 - 9 \times 5}{5 \times 16} \\
 &= \frac{48 - 45}{5 \times 16} \\
 &= \frac{3}{80}
 \end{aligned}$$

So,

$$\text{Var}(X) = \frac{3}{80}$$

3. Given X is uniformly distributed over $[0, 1]$, then

PDF of X is

$$f(x) = \begin{cases} 1 & ; 0 \leq x \leq 1 \\ 0 & ; \text{otherwise} \end{cases}$$

a) The new random variable is Y & it is related as

$$Y = X^2$$

Since $y = x^2$ is continuously differentiable & non decreasing for $0 \leq x \leq 1$, we can say,

$$x = \sqrt{y}, \quad y > 0$$

$$dx/dy = \frac{1}{2\sqrt{y}}$$

and,

$$0 < x \leq 1$$

$$0 \leq \sqrt{y} \leq 1$$

$$0 \leq y \leq 1$$

and,

Say

$$g(y) = F(x(y)) \left| \frac{dx}{dy} \right|$$

$$= \frac{1}{(1)} \left| \frac{1}{2\sqrt{y}} \right|$$

$$= \frac{1}{2\sqrt{y}}$$

So, the pdf of Y is

$$g(y) = \begin{cases} \frac{1}{2\sqrt{y}}, & 0 \leq y \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

b) Mean of Y is given by :

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

$g(y) \rightarrow \text{pdf}$

$$= \int_0^1 y \cdot \left(\frac{1}{2\sqrt{y}} \right) dy$$

$$= \frac{1}{2} \int_0^1 \sqrt{y} dy$$

$$= \frac{1}{2} \left(\frac{y^{3/2}}{(3/2)} \right)_0^1$$

$$= \frac{1}{3} (y^{3/2})_0^1$$

$$= \frac{1}{3}$$

So, Mean $(Y) = \frac{1}{3}$

c) Variance of Y is $\rightarrow E(Y^2) - (E(Y))^2$

Now, $E(Y^2) = \int_{-\infty}^{\infty} y^2 g(y) dy$

$$= \int_0^1 y^2 \left(\frac{1}{2\sqrt{y}} \right) dy$$

$$= \frac{1}{2} \int_0^1 y^{3/2} dy$$

$$= \frac{1}{2} \frac{(y^{5/2})_0^1}{(5/2)}$$

$$= \frac{1}{5}$$

Also, $(E(Y))^2 = \left(\frac{1}{3} \right)^2 = \frac{1}{9}$

$$\text{So, } \text{Var}(Y) = \frac{1}{5} - \frac{1}{9}$$

$$= \frac{4}{45}$$

5. Stock Price Random Variable

Given X & $P(x)$:

$x(x_i)$	-2	-1	0	1	2
$P(x)$	0.1	0.2	0.3	0.25	0.15

$$a) \quad E(x) = \sum_{i \in \text{all}} x_i \cdot P(x_i)$$

$$= (-2)(0.1) + (-1)(0.2) + 0(0.3) + 1(0.25) + 2(0.15)$$

$$= -0.2 - 0.2 + 0.25 + 0.3$$

$$= 0.55 - 0.40$$

$$= 0.15$$

$$\text{So, } E(x) = 0.15$$

$$b) \quad V(x) = E(x^2) - (E(x))^2$$

Now,

$$E(x^2) = \sum_{i \in \text{all}} (x_i)^2 P(x_i)$$

$$= 4(0.1) + 1(0.2) + 0(0.3)$$

$$+ 1(0.25) + 4(0.15)$$

$$= 0.4 + 0.2 + 0.25 + 0.6$$

$$= 1.45$$

$$\Delta (E(x))^2 = 0.0225$$

$$\text{So, } V(x) = 1.4275 //$$

c) Standard deviation = $\sqrt{V(x)}$

$$= \sqrt{1.4275}$$

$$= 1.194 //$$

d) P(increase by atleast 1%) =

$$\sum_{x_i \geq 1} P(x_i)$$

$$= 0.25 + 0.15$$

$$= 0.4 //$$

e) P(increase or decrease by less than 2%)

$$= P(-1) + P(0) + P(1)$$

$$= 0.2 + 0.3 + 0.25$$

$$= 0.75 //$$

4. Given the random variables X and Y ,

X/Y	1	2	3
1	0.2	0.1	0.1
2	0.1	0.2	0.1
3	0.1	0.1	0

a) Finding marginal distribution of X :

X	1	2	3
$P(X)$	0.4	0.4	0.2

Marginal distribution of Y :

Y	1	2	3
$P(Y)$	0.4	0.4	0.2

Now, $\text{cov}(X, Y) = E(XY) - E(X) \cdot E(Y)$

$$\begin{aligned}\text{So, } E(X) &= \sum_i x_i p(x_i) \\ &= 1(0.4) + 2(0.4) + 3(0.2) \\ &= 0.4 + 0.8 + 0.6 \\ &= 1.8\end{aligned}$$

$$\begin{aligned}E(Y) &= \sum_i y_i p(y_i) \\ &= 1(0.4) + 2(0.4) + 3(0.2) \\ &= 0.4 + 0.8 + 0.6 \\ &= 1.8\end{aligned}$$

$$\text{Now, } E(XY) = \sum_{i,j} x_i y_j T_{ij}$$

where T_{ij} is the joint probability distribution.

$$\begin{aligned} \text{So, } E(XY) &= 0.2 + 2(0.1) + 3(0.1) + 2(0.1) + 4(0.2) + 6(0.1) + 3(0.1) + 6(0.1) + 9(0) \\ &= 0.2 + 0.2 + 0.3 + 0.2 + 0.8 + 0.6 + 0.3 + 0.6 \\ &= 3.2 \end{aligned}$$

$$\begin{aligned} \therefore \text{cov}(X, Y) &= 3.2 - (1.8)(1.8) \\ &= -0.040 \end{aligned}$$

b) $\because \text{cov}(X, Y) < 0$, we can conclude X and Y are negatively correlated.

c) Now,

$$\rho(X, Y) = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y}$$

$$\sigma_X = \sqrt{\text{VAR}(X)}$$

$$\begin{aligned} \therefore \text{VAR}(X) &= E(X^2) - [E(X)]^2 \\ &= 1(0.4) + 4(0.4) + 9(0.2) - (1.8)^2 \\ &= 0.4 + 1.6 + 1.8 - 3.24 \end{aligned}$$

$$\text{VAR}(X) = 0.56$$

$$\sigma_X = \sqrt{0.56} = 0.748$$

Similarly,

$$\sigma_Y = \sqrt{\text{VAR}(Y)}$$

$$\text{VAR}(Y) = E(Y^2) - [E(Y)]^2$$

$$= 0.4 + 4(0.4) + 9(0.2) - (1.8)^2$$

$$= 3.8 - 3.24$$

$$= 0.56$$

$$\sigma_Y = \sqrt{0.56} = 0.748$$

Hence,

$$\rho(X, Y) = \frac{-0.040}{(0.748)(0.748)}$$

$$= -0.071$$