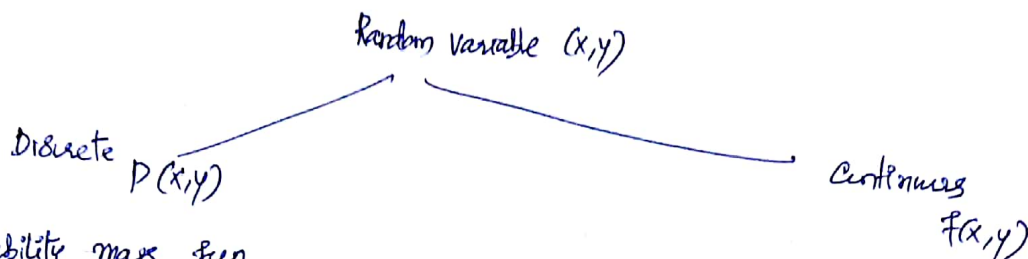


Two Dimensional Random Variables (x,y)

Let S be the sample space associated with a random experiment. Let $x=x(s)$ and $y=y(s)$ be two functions each of which assigns a real number to each outcome $s \in S$. Then, (x,y) is called a two dimensional random variable.



1. Joint probability mass fun

If (x,y) is the 2-dim random variable s.t $P(x=x_i, y=y_j) = P_{ij} = P(x,y)$, then $P(x,y)$ is called the joint probability mass fun, provided the following conditions.

- (i) $P(x,y) \geq 0$
- (ii) $\sum_y \sum_x P(x,y) = 1$

2. Cumulative distribution fun:

$$P(x \leq x, y \leq y) = \sum_{-\infty}^y \sum_{-\infty}^x P(x,y)$$

3. Marginal distribution fun (MDF):

In $P(x,y)$, finding only $P(x)$ or $P(y)$

is MDF

$$P(x) = \sum_y P(x,y), \quad P(y) = \sum_x P(x,y)$$

4. Conditional probability distribution:

$$P(x|y) = \frac{P(x,y)}{P(y)}, \quad P(y) \neq 0$$

$$P(y|x) = \frac{P(x,y)}{P(x)}, \quad P(x) \neq 0$$

5. Condition for independency:

$$P(x,y) = P(x) * P(y)$$

1. Joint probability density fun

$$(i) f(x,y) \geq 0$$

$$(ii) \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$$

2.

$$P(x \leq x, y \leq y) = \int_{-\infty}^y \int_{-\infty}^x f(x,y) dx dy$$

3. In $f(x,y)$, finding only $f(x)$ or $f(y)$

$$f(x) = \int_{-\infty}^{\infty} f(x,y) dy, \quad f(y) = \int_{-\infty}^{\infty} f(x,y) dx$$

4.

$$f(x|y) = \frac{f(x,y)}{f(y)}, \quad f(y) \neq 0$$

$$f(y|x) = \frac{f(x,y)}{f(x)}, \quad f(x) \neq 0$$

5.

$$f(x,y) = f(x) * f(y)$$

1. Let (X, Y) be two dimensional R.V with joint probability mass fun

$y \backslash x$	1	2	3
1	$\frac{1}{12}$	$\frac{1}{6}$	0
2	0	$\frac{1}{9}$	$\frac{1}{5}$
3	$\frac{1}{18}$	$\frac{1}{4}$	$\frac{2}{15}$

- Evaluate
- Marginal distribution of X and Y .
 - Conditional distribution of X given $Y=2$.
 - Conditional distribution of Y given $X=2$.
 - $P(X \leq 2, Y=3)$
 - $P(Y \geq 2)$
 - $P(X+Y < 4)$
 - Are X and Y independent?

(i) Marginal distribution of X and Y .

$$\# P(X) = \sum_y P(X, Y) \quad (\text{we need to find } P(X=1), P(X=2), P(X=3))$$

$$\bullet P(X=1) = \sum_y P(1, Y) = P(1, 1) + P(1, 2) + P(1, 3) = \frac{1}{12} + 0 + \frac{1}{18} = \frac{5}{36}$$

$$\bullet P(X=2) = \sum_y P(2, Y) = P(2, 1) + P(2, 2) + P(2, 3) = \frac{1}{6} + \frac{1}{9} + \frac{1}{4} = \frac{19}{36}$$

$$\text{III}^y \bullet P(X=3) = \sum_y P(3, Y) = 0 + \frac{1}{5} + \frac{2}{15} = \frac{1}{3}$$

$$\# P(Y) = \sum_x P(X, Y) \quad (\text{we need to find } P(Y=1), P(Y=2), P(Y=3))$$

$$P(Y=1) = \sum_x P(X, 1) = P(1, 1) + P(2, 1) + P(3, 1) = \frac{1}{12} + \frac{1}{6} + 0 = \frac{1}{4}$$

$$P(Y=2) = 0 + \frac{1}{9} + \frac{1}{5} = \frac{14}{45}$$

$$P(Y=3) = \frac{1}{18} + \frac{1}{4} + \frac{2}{15} = \frac{79}{180}$$

(ii) $P(X|Y=2)$ [we need to find $P(X=1|Y=2)$, $P(X=2|Y=2)$, $P(X=3|Y=2)$]

$$P(X=1|Y=2) = \frac{P(X=1, Y=2)}{P(Y=2)} = \frac{0}{14/45} = 0$$

$$P(X=2|Y=2) = \frac{P(X=2, Y=2)}{P(Y=2)} = \frac{1/9}{14/45} = \frac{5}{14}$$

$$P(X=3|Y=2) = \frac{P(X=3, Y=2)}{P(Y=2)} = \frac{1/5}{14/45} = \frac{9}{14}$$

(iii) $P(Y|X=2)$

$$\cdot P(Y=1|X=2) = \frac{P(Y=1, X=2)}{P(X=2)} = \frac{1/6}{19/36} = \frac{6}{19}$$

$$\cdot P(Y=2|X=2) = \frac{P(Y=2, X=2)}{P(X=2)} = \frac{1/9}{19/36} = \frac{4}{19}$$

$$\cdot P(Y=3|X=2) = \frac{P(Y=3, X=2)}{P(X=2)} = \frac{1/4}{19/36} = \frac{9}{19}$$

(iv) $P(X \leq 2, Y=3) = P(X=1, Y=3) + P(X=2, Y=3)$
 $= \frac{1}{18} + \frac{1}{4} = \frac{11}{36}$

(v) $P(Y \geq 2) = P(Y=2) + P(Y=3)$
 $= \frac{14}{45} + \frac{79}{180}$

(vi) $P(X+Y < 4) = P(1,1) + P(1,2) + P(2,1)$
 $= \frac{1}{12} + 0 + \frac{1}{6}$
 $= \frac{1}{4}$

(vii) Are X and Y independent

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

let us take $X=1, Y=1$

$$P(X=1, Y=1) = P(X=1) * P(Y=1)$$

$$\frac{1}{12} \neq \frac{5}{36} * \frac{1}{4}$$

$\therefore X$ and Y are not independent.

2. The joint probability mass function of (X, Y) is given by

$$P(X, Y) = k(2x+3y), \quad x=0,1,2; \quad y=1,2,3$$

find (i) Marginal distribution of X and Y .

(ii) Conditional distribution of X given $Y=1$.

(iii) probability distribution of $X+Y$.

3. The joint probability density function of 2-D R.V. (x, y) is given by

$$f(x, y) = \begin{cases} k(6-x-y), & 0 < x < 2, \text{ ~~2 < y < 4~~ } \\ 0, & \text{otherwise.} \end{cases}$$

Find (i) k (ii) $P(x < 1, y < 3)$ (iii) $P(x+y < 3)$ (iv) $P(x < 1 | y < 3)$
 (v) Are x and y independent?

Soln:

(i) Find k

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$$

$$\Rightarrow \int_2^4 \int_0^2 k(6-x-y) dx dy = 1$$

$$\int_2^4 k \left(6x - \frac{x^2}{2} - xy \right) dy = 1$$

$$k \int_2^4 (12 - 2 - 2y) dy = 1$$

$$k \left[10y - \frac{2y^2}{2} \right]_2^4 = 1$$

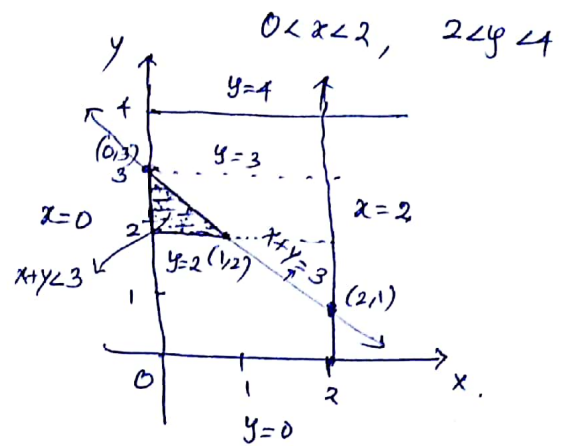
$$k \left([40 - 16] - [20 - 4] \right) = 1$$

$$\boxed{k = \frac{1}{8}}$$

$$\begin{aligned} \text{(ii) } P(x < 1, y < 3) &= \int_{-\infty}^3 \int_{-\infty}^1 f(x, y) dx dy \\ &= \int_2^3 \int_0^1 \frac{1}{8} (6-x-y) dx dy \\ &= \frac{1}{8} \int_2^3 \left(6x - \frac{x^2}{2} - xy \right) dy \\ &= \frac{1}{8} \int_2^3 \left(6 - \frac{1}{2} - y \right) dy \\ &= \frac{3}{8} \end{aligned}$$

$$(iii) P(x+y < 3) = \int_2^3 \int_0^{3-y} \frac{1}{8} (6-x-y) dx dy$$

$$= \frac{10}{24}$$



limits, for the region of $x+y < 3$
 $x = 0$ to $x = 3-y$
 $y = 2$ to 3

$$(iv) P(x < 1 | y < 3) = \frac{P(x < 1, y < 3)}{P(y < 3)}$$

we have found $P(x < 1, y < 3) = \frac{3}{8}$

Let us find $P(y < 3) = \int_2^3 f(y) dy$

here we need to find $f(y)$

$$f(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^2 \frac{1}{8} (6-x-y) dx$$

$$= \int_0^2 \frac{1}{8} \left(6x - \frac{x^2}{2} - xy \right) dx$$

$$= \frac{1}{8} [12 - 2 - 2y]$$

$$f(y) = \frac{5-y}{4}$$

$$\Rightarrow P(y < 3) = \int_2^3 f(y) dy$$

$$= \int_2^3 \left(\frac{5-y}{4} \right) dy = \frac{1}{4} \left[5y - \frac{y^2}{2} \right]_2^3$$

$$= \frac{1}{4} \left[\left(15 - \frac{9}{2} \right) - \left(10 - \frac{4}{2} \right) \right]$$

$$= \frac{5}{8}$$

$$\therefore P(x < 1 | y < 3) = \frac{P(x < 1, y < 3)}{P(y < 3)} = \frac{3/8}{5/8} = \frac{3}{5}$$

(V) To check x and y are independent

$$f(x, y) = f(x) * f(y)$$

$$\begin{aligned} f(x) &= \int_{-\infty}^{\infty} f(x, y) dy \\ &= \int_2^4 \frac{1}{8} (6-x-y) dy \end{aligned}$$

$$f(x) = \frac{6-2x}{8}$$

we have found

$$f(y) = \frac{5-y}{4}$$

$$f(x) * f(y) = f(x, y)$$

$$\frac{6-2x}{8} * \frac{5-y}{4} \neq \frac{1}{8} (6-x-y)$$

$\therefore x$ and y are dependent.

4. The joint pdf of (x, y) is given by

$$f(x, y) = \begin{cases} x^2 + \frac{xy}{3} & , 0 < x < 1, 0 < y < 2 \\ 0 & , \text{otherwise.} \end{cases}$$

Find (i) $p(x > \frac{1}{2})$ (ii) $p(y < x)$, (iii) $p(y < \frac{1}{2} | x < \frac{1}{2})$.

Covariance and Correlation Coefficient

Covariance :

As the Variance σ_x^2 (or $\text{Var}(X) = E(X^2) - [E(X)]^2$) is a measure of the Variation of the Random Variable (R.V) X from its mean Value $E(X)$, the quantity $E[(X-E(X))(Y-E(Y))]$ measures the simultaneous Variation of two R.V's X and Y from their respective means and hence it is called the Covariance of X, Y and it is denoted as $\text{Cov}(X, Y)$.

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

Proof :

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

$$E(X) = \bar{x}, \quad E(Y) = \bar{y}$$

$$= E[(X - \bar{x})(Y - \bar{y})]$$

$$= E[XY - X\bar{y} - \bar{x}Y + \bar{x}\bar{y}]$$

$$= E[XY] - E[X\bar{y}] - E[\bar{x}Y] + E[\bar{x}\bar{y}]$$

$$= E[XY] - \bar{y}E[X] - \bar{x}E[Y] + E[\bar{x}\bar{y}]$$

$$= E[XY] - \bar{y}\bar{x} - \bar{x}\bar{y} + \bar{x}\bar{y}$$

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

Correlation Co-efficient :

$$\rho_{xy} = r(x, y) = \rho(x, y) = \frac{\text{Cov}(X, Y)}{\sigma_x \sigma_y}$$

* Correlation can be -ve but Covariance can't be -ve.

The Correlation Co-efficient is a statistical measure that calculates the strength of the relationship b/w the relative movements of two variables. The values $-1 \leq r(x, y) \leq 1$. A Correlation -1 shows that perfect negative Correlation, while a Correlation 1 shows a perfect positive Correlation. If the Correlation is 0 , then there is no relationship b/w variables however other functional relationship may exist.

Raw data:

$$X: x_1 \quad x_2 \quad \dots \quad x_n$$

$$Y: y_1 \quad y_2 \quad \dots \quad y_n$$

$$\rho(x, y) = \frac{E(xy) - E(x)E(y)}{\sigma_x \sigma_y}$$

where

$$E(x) = \frac{\sum x_i}{n}$$

$$E(y) = \frac{\sum y_i}{n}$$

$$E(xy) = \frac{\sum x_i y_i}{n}$$

$$\sigma_x = \sqrt{E(x^2) - [E(x)]^2}$$

$$\sigma_x = \sqrt{\frac{\sum (x - \bar{x})^2}{n}}$$

For discrete:

$$E(x) = \sum x p(x)$$

$$E(x^2) = \sum x^2 p(x)$$

$$E(xy) = \sum \sum xy p(xy)$$

$$E(y) = \sum y p(y)$$

$$E(y^2) = \sum y^2 p(y)$$

For Continuous:

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

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

$$E(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x, y) dx dy$$

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

$$E(y^2) = \int_{-\infty}^{\infty} y^2 f(y) dy$$

Note:

(i) $\text{Cov}(x, y) = 0$ if x and y are independent.

(ii) $\text{Var}(x \pm y) = \text{Var}(x) + \text{Var}(y) \pm 2\text{Cov}(x, y)$

(iii) $\text{Var}(x \pm y) = \text{Var}(x) + \text{Var}(y)$ if x and y are independent.

1. Find the Correlation Coefficient b/w X and Y.

X: 2 4 5 6 8 11

Y: 18 12 10 8 7 5

Soln:

$$r_{xy} = \rho(x, y) = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y} = \frac{E(xy) - E(x)E(y)}{\sigma_x \sigma_y}$$

$$E(xy) = \frac{\sum x_i y_i}{n} = \frac{(2 \times 18) + (4 \times 12) + (5 \times 10) + (6 \times 8) + (8 \times 7) + (11 \times 5)}{6} = \frac{293}{6}$$

$$E(xy) = 48.83$$

$$E(x) = \frac{\sum x_i}{n} = \frac{2+4+5+6+8+11}{6} = 6$$

$$E(y) = \frac{\sum y_i}{n} = \frac{60}{6} = 10$$

$$\text{Cov}(x, y) = E(xy) - E(x)E(y)$$

$$\text{Cov}(x, y) = -11.17$$

$$\sigma_x = \sqrt{E(x^2) - [E(x)]^2}$$

$$E(x^2) = \frac{\sum x_i^2}{n} = \frac{266}{6} = 44.33$$

$$\sigma_x = \sqrt{44.33 - 36} = 2.886$$

$$\sigma_y = \sqrt{E(y^2) - [E(y)]^2}$$

$$E(y^2) = \frac{\sum y_i^2}{n} = 117.67$$

$$\sigma_y = \sqrt{117.67 - 100} = 4.2$$

$$r_{xy} = \rho(x, y) = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y} = \frac{-11.17}{2.886 \times 4.2} = -0.9207$$

2. Find $\rho(x, y)$ for the following

X: 55 56 58 59 60 60 62

Y: 35 38 37 39 44 43 44

3. let (x, y) be a 2D-R.V.S. whose joint pdf is given by

$$f(x, y) = \begin{cases} x+y, & 0 < x < 1, 0 < y < 1 \\ 0, & \text{otherwise} \end{cases}$$

find $r(x, y)$.

4. If (x, y) be a 2D-R.V with joint pdf is given by

$y \backslash x$	1	2	3
1	$\frac{2}{21}$	$\frac{3}{21}$	$\frac{4}{21}$
2	$\frac{3}{21}$	$\frac{4}{21}$	$\frac{5}{21}$

find $\sigma(x, y)$.

5. If the joint pdf of 2-D RV is given by

$$f(x, y) = \begin{cases} 3(x+y), & x > 0, y > 0, x+y \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$

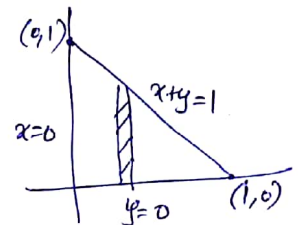
Find $\text{Cov}(x, y)$, $\text{Var}(x)$, $\text{Var}(y)$.

Solu:

$$\text{Cov}(x, y) = E(xy) - E(x)E(y)$$

The marginal density function of x is given by

$$\begin{aligned} f(x) &= \int_{-\infty}^{\infty} f(x, y) dy \\ &= \int_0^{1-x} 3(x+y) dy \\ &= 3 \left[xy + \frac{y^2}{2} \right]_0^{1-x} \\ &= 3 \left[(x(1-x) + \frac{1}{2}(1-x)^2) - (0) \right] \\ &= \frac{3}{2}(1-x^2), \quad 0 \leq x \leq 1 \end{aligned}$$



The marginal density fun of y

$$\begin{aligned} f(y) &= \int_{-\infty}^{\infty} f(x, y) dx \\ &= \int_0^{1-y} 3(x+y) dx = \end{aligned}$$

$$f(y) = \frac{3}{2}(1-y^2), \quad 0 \leq y \leq 1.$$

$$\begin{aligned} \bullet E(X) &= \int_{-\infty}^{\infty} x f(x) dx \\ &= \int_0^1 x \frac{3}{2}(1-x^2) dx = \frac{3}{2} \int_0^1 (x-x^3) dx \\ &= \frac{3}{2} \left(\frac{x^2}{2} - \frac{x^4}{4} \right) \Big|_0^1 = \frac{3}{8} \end{aligned}$$

$$\begin{aligned} \bullet E(X^2) &= \int_{-\infty}^{\infty} x^2 f(x) dx \\ &= \int_0^1 x^2 \left(\frac{3}{2}(1-x^2) \right) dx = \frac{3}{2} \int_0^1 (x^2-x^4) dx \\ &= \frac{3}{2} \left[\frac{x^3}{3} - \frac{x^5}{5} \right] \Big|_0^1 = \frac{1}{5} \end{aligned}$$

$$\begin{aligned} \bullet E(Y) &= \int_{-\infty}^{\infty} y f(y) dy \\ &= \int_0^1 y \left(\frac{3}{2}(1-y^2) \right) dy = \frac{3}{2} \int_0^1 (y-y^3) dy = \frac{3}{8} \end{aligned}$$

$$\begin{aligned} \bullet E(Y^2) &= \int_{-\infty}^{\infty} y^2 f(y) dy \\ &= \int_0^1 y^2 \left(\frac{3}{2}(1-y^2) \right) dy = \frac{3}{2} \int_0^1 (y^2-y^4) dy = \frac{1}{5} \end{aligned}$$

$$\neq \text{Var}(X) = E(X^2) - [E(X)]^2 = \frac{1}{5} - \left(\frac{3}{8}\right)^2 = \frac{19}{320}$$

$$\neq \text{Var}(Y) = \frac{19}{320}$$

$$\begin{aligned} \bullet E(XY) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x,y) dx dy \\ &= \int_0^1 \int_0^{1-y} xy \cdot 3(x+y) dx dy = \frac{1}{10} \\ &= \frac{1}{10} \end{aligned}$$

$$\begin{aligned} \therefore \text{Cov}(X,Y) &= E(XY) - E(X)E(Y) \\ &= \frac{1}{10} - \left(\frac{3}{8}\right)\left(\frac{3}{8}\right) = -\frac{13}{320} \\ \text{Cov}(X,Y) &= -\frac{13}{320} \end{aligned}$$

characteristic function

$$\phi_x(w) = E[e^{iwx}] = \begin{cases} \sum e^{iwx} p(x) & \text{if } x \text{ is discrete} \\ \int_{-\infty}^{\infty} e^{iwx} f(x) dx & \text{if } x \text{ is continuous} \end{cases}$$

Note:

$$* \phi_x(w) = \sum_{r=0}^{\infty} \frac{i^r w^r}{r!} \mu_r'$$

$$* \mu_r' = \frac{1}{i^r} \left\{ \frac{d^r}{dw^r} \phi_x(w) \right\}_{w=0}$$

$$\text{mean} = \mu_1' \quad \text{variance} = \mu_2' - (\mu_1')^2$$

1. find the characteristic fun for the R.V x whose pdf is given by

$$f(x) = \frac{\alpha}{2} e^{-\alpha|x|}, \quad -\infty < x < \infty, \quad \alpha > 0.$$

and hence find $E(x)$ and $\text{Var}(x)$.

Soln:

$$\begin{aligned} \phi_x(w) &= \int_{-\infty}^{\infty} e^{iwx} f(x) dx \\ &= \int_{-\infty}^{\infty} e^{iwx} \frac{\alpha}{2} e^{-\alpha|x|} dx \\ &= \frac{\alpha}{2} \left[\int_{-\infty}^0 e^{iwx} e^{-\alpha(-x)} dx + \int_0^{\infty} e^{iwx} e^{-\alpha x} dx \right] \\ &= \frac{\alpha}{2} \left[\int_{-\infty}^0 e^{(\alpha+iw)x} dx + \int_0^{\infty} e^{-(\alpha-iw)x} dx \right] \\ &= \frac{\alpha}{2} \left[\left(\frac{e^{(\alpha+iw)x}}{(\alpha+iw)} \right)_{-\infty}^0 + \left(\frac{e^{-(\alpha-iw)x}}{-(\alpha-iw)} \right)_0^{\infty} \right] \\ &= \frac{\alpha}{2} \left[\left(\frac{1}{\alpha+iw} - 0 \right) + \left(0 + \frac{1}{\alpha-iw} \right) \right] \end{aligned}$$

$$= \frac{\alpha}{2} \left[\frac{1}{\alpha + iw} + \frac{1}{\alpha - iw} \right]$$

$$= \frac{\alpha}{2} \left[\frac{\alpha - iw + \alpha + iw}{\alpha^2 - i^2 w^2} \right]$$

$$= \frac{\alpha}{2} \left[\frac{2\alpha}{\alpha^2 + w^2} \right]$$

$$\because i^2 = -1$$

$$\phi_x(w) = \frac{\alpha^2}{\alpha^2 + w^2}$$

$$\text{mean} = \mu_1' = \frac{1}{i} \left\{ \frac{d}{dw} \left(\frac{\alpha^2}{\alpha^2 + w^2} \right) \right\}_{w=0}$$

$$= \frac{1}{i} \left\{ \frac{(\alpha^2 + w^2)(0) - \alpha^2(2w)}{(\alpha^2 + w^2)^2} \right\}_{w=0}$$

$$\mu_1' = 0.$$

$$\mu_2' = \frac{1}{i^2} \left\{ \frac{d^2}{dw^2} \left(\frac{\alpha^2}{\alpha^2 + w^2} \right) \right\}_{w=0}$$

$$= (-1) \left\{ \frac{d}{dw} \left(\frac{-2\alpha^2 w}{(\alpha^2 + w^2)^2} \right) \right\}_{w=0}$$

$$= (-1) \left\{ \frac{(\alpha^2 + w^2)^2 (-2\alpha^2) + 2\alpha^2 w (2(\alpha^2 + w^2) \cdot 2w)}{(\alpha^2 + w^2)^4} \right\}_{w=0}$$

$$= (-1) \left\{ \frac{-2\alpha^4}{\alpha^6} \right\}$$

$$\mu_2' = \frac{2}{\alpha^2}$$

$$\text{Var}(X) = \mu_2' - (\mu_1')^2 = \frac{2}{\alpha^2}$$

2. Find the characteristic fun of the R.V whose pmf is given by

$$p(x=r) = \frac{e^{-\lambda} \lambda^r}{r!}, \quad r=0, 1, 2, \dots \quad \lambda > 0$$