

Definition (28) Let X and Y be any two discrete r.v.'s. The real-valued function $F: \mathbb{R}^2 \rightarrow \mathbb{R}$ is called the joint cumulative probability distribution function of X and Y iff:

$$F(x, y) = P(X \leq x, Y \leq y) \quad \checkmark$$

From this definition it can be shown that for any real numbers a, b, c and d :

$$P(a \leq X \leq b, c \leq Y \leq d) = F(a, b) + F(a, c) - F(a, d) - F(c, a) \quad \checkmark$$

2. Bivariate Continuous Random Variables:

Definition (29) The joint probability density function of the r.v.'s X and Y is an integrable function $f(x, y)$ such that

(a) $f(x, y) \geq 0$ for all $(x, y) \in \mathbb{R}^2$; and

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

Example (22) Let the joint density of the continuous r.v.'s X and Y be:

$$f(x, y) = \begin{cases} \frac{6}{5}(x^2 + 2xy), & 0 \leq x \leq 1; 0 \leq y \leq 1 \\ 0, & \text{o.w.} \end{cases}$$

what is the probability of the event $(X \leq Y)$?

Solution: Let $A = (X \leq Y)$, we want to find:

$$\begin{aligned} P(A) &= \int \int_A f(x, y) dx dy \\ &= \int_0^1 \left[\int_0^y \frac{6}{5}(x^2 + 2xy) dx \right] dy \\ &= \frac{6}{5} \int_0^1 \left[\frac{x^3}{3} + x^2 y \right]_0^y dy = \frac{6}{5} \int_0^1 \frac{4}{3} y^3 dy \\ &= \frac{8}{5} \left[\frac{y^4}{4} \right]_0^1 = \frac{2}{5} \quad \checkmark \end{aligned}$$

Definition (30): Let (X, Y) be a continuous bivariate r.v.s.
 Let $f(x, y)$ be the j.p.d.f. of X and Y . The function

$$f_1(x) = \int_{\forall y} f(x, y) dy$$

is called the m.p.d.f. of X . Similarly, the function

$$f_2(y) = \int_{\forall x} f(x, y) dx$$

is called the m.p.d.f. of Y .

Example (23):

If the j.p.d.f. for X and Y is given by:

$$f(x, y) = \begin{cases} \frac{3}{4}, & 0 < y^2 < x < 1 \\ 0, & \text{o.w} \end{cases}$$

then what is the m.p.d.f. of X for $0 < x < 1$?

Solution:

$$\begin{aligned} f_1(x) &= \int_{\forall y} f(x, y) dy = \int_{-\sqrt{x}}^{\sqrt{x}} \frac{3}{4} dy \\ &= \left[\frac{3y}{4} \right]_{-\sqrt{x}}^{\sqrt{x}} = \frac{3}{4} [\sqrt{x} + \sqrt{x}] = \frac{3}{2} \sqrt{x} \end{aligned}$$

Example (24). Let (X, Y) be distributed uniformly on the circular disk centered at $(0, 0)$ with radius $\frac{2}{\sqrt{\pi}}$. what is the m.p.d.f. of X where nonzero?

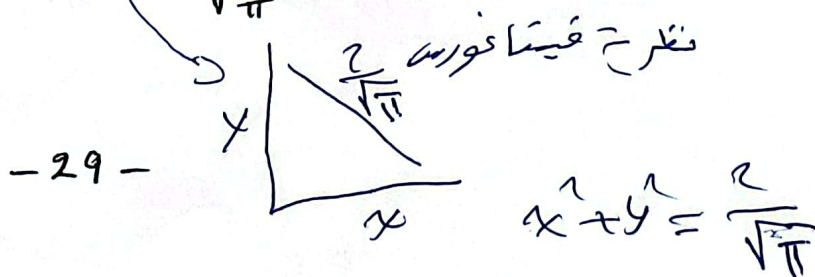
Solution:

The equation of a circular with radius $\frac{2}{\sqrt{\pi}}$ at Center at the origin is

$$x^2 + y^2 = \left(\frac{4}{\pi} \right)$$

$\Rightarrow y = \pm \sqrt{\frac{4}{\pi} - x^2}$; then the m.p.d.f. of X is

$$f_1(x) = \int_{-\sqrt{\frac{4}{\pi} - x^2}}^{\sqrt{\frac{4}{\pi} - x^2}} f(x, y) dy = \int_{-\sqrt{\frac{4}{\pi} - x^2}}^{\sqrt{\frac{4}{\pi} - x^2}} \frac{1}{\text{area of circle}} dy$$



$$x^2 + y^2 = \frac{4}{\pi}$$

$$\therefore f_1(x) = \int_{-\sqrt{\frac{4}{\pi}-x^2}}^{+\sqrt{\frac{4}{\pi}-x^2}} \left(\frac{1}{4}\right) dy$$

$$= \left[\frac{1}{4} y \right]_{-\sqrt{\frac{4}{\pi}-x^2}}^{+\sqrt{\frac{4}{\pi}-x^2}} = \frac{1}{2} \sqrt{\frac{4}{\pi}-x^2}$$

Definition (31) Let X and Y be the continuous r.v.s with j.p.d.f. $f(x,y)$. The joint Cumulative distribution Function $F(x,y)$ of X and Y is defined as

$$F(x,y) = P(X \leq x, Y \leq y) = \int_{-\infty}^y \int_{-\infty}^x f(u,v) du dv$$

From the fundamental theorem of calculus, we get:

$$f(x,y) = \frac{\partial^2}{\partial x \partial y} F(x,y)$$

Example (25) If the j.c.d.f. of X and Y is given by,

$$F(x,y) = \begin{cases} \frac{1}{5} (2x^3y + 3x^2y^2) & , 0 < x, y < 1 \\ 0 & , \text{o.w} \end{cases}$$

then what is the j.p.d.f. of X and Y ?

Solution:

$$f(x,y) = \frac{1}{5} \frac{\partial}{\partial x} \cdot \frac{\partial}{\partial y} (2x^3y + 3x^2y^2)$$

$$= \frac{1}{5} \cdot \frac{\partial}{\partial x} (2x^3 + 6x^2y)$$

$$= \frac{1}{5} (6x^2 + 12xy) = \frac{6}{5} (x^2 + 2xy)$$

$$\text{then } f(x,y) = \begin{cases} \frac{6}{5} (x^2 + 2xy) & , 0 < x, y < 1 \\ 0 & , \text{o.w} \end{cases}$$

8/6

Example(26). Let X and Y have the joint density function

$$f(x,y) = \begin{cases} x+y & , 0 \leq x \leq 1; 0 \leq y \leq 1 \\ 0 & , \text{o.w} \end{cases}$$

what is the $P(2x \leq 1 / x+y \leq 1)$?

Solution:

$$P(2x \leq 1 / x+y \leq 1) = \frac{P[(X \leq \frac{1}{2}) \cap (x+y \leq 1)]}{P(x+y \leq 1)}$$

$$P(x+y \leq 1) = \int_0^1 \left[\int_0^{1-x} (x+y) dy \right] dx$$

$$= \int_0^1 \left[xy + \frac{y^2}{2} \right]_0^{1-x} dx$$

$$= \int_0^1 \left[x(1-x) + \frac{(1-x)^2}{2} \right] dx$$

$$= \int_0^1 \left[x - x^2 + \frac{(1-x)^2}{2} \right] dx$$

$$= \left[\frac{x^2}{2} - \frac{x^3}{3} - \frac{(1-x)^3}{6} \right]_0^1 = \frac{1}{2} - \frac{1}{3} = \frac{1}{6}$$

$$P[(X \leq \frac{1}{2}) \cap (x+y \leq 1)] = \int_0^{1/2} \int_0^{1-x} (x+y) dy dx$$

$$= \int_0^{1/2} \left[xy + \frac{y^2}{2} \right]_0^{1-x} dx$$

$$= \int_0^{1/2} \left[x - x^2 + \frac{(1-x)^2}{2} \right] dx$$

$$= \left[\frac{x^2}{2} - \frac{x^3}{3} - \frac{(1-x)^3}{6} \right]_0^{1/2}$$

$$= \frac{1}{8} - \frac{1}{24} - \frac{1}{48} = \frac{3}{48} = \frac{1}{16} = \frac{1}{48}$$

$$\therefore P(2x \leq 1 / x+y \leq 1) = \frac{\frac{1}{16}}{\frac{1}{6}} = \frac{1}{16} \cdot \frac{6}{1}$$

$$= \frac{3}{8}$$

Definition (32) Let X and Y be any two r.v's with joint density $f(x, y)$ and marginals $f_1(x)$ and $f_2(y)$. The conditional probability density function g of X , given $Y=y$ is defined as

$$\underline{g(x/y) = \frac{f(x, y)}{f_2(y)}, \quad f_2(y) > 0.}$$

Similarly, the Conditional probability density Function h of Y , given $X=x$, is defined as:

$$\underline{h(y/x) = \frac{f(x, y)}{f_1(x)}, \quad f_1(x) > 0}$$

Ex (27) Let X and Y be r.v's such that X has density function

$$\underline{f_1(x) = \begin{cases} 24x^2 & , 0 < x < 1/2 \\ 0 & , \text{o.w} \end{cases}}$$

and the conditional density of Y given $X=x$ is

$$\underline{h(y/x) = \begin{cases} \frac{y}{2x^2} & , 0 < y < 2x \\ 0 & , \text{o.w} \end{cases}}$$

what is the conditional density of X given $Y=y$ over the appropriate domain?

Solution:

$$\begin{aligned} \underline{f(x, y)} &= h(y/x) f_1(x) \\ &= \frac{y}{2x^2} \cdot 24x^2 \\ &= \begin{cases} 12y & , 0 < y < 2x < 1 \\ 0 & , \text{o.w} \end{cases} \end{aligned}$$

$$\begin{aligned} 0 < x < \frac{1}{2} & \times 2 \\ 0 < 2x < 1 \\ 0 < y < 2x \end{aligned}$$

The m.p.d.f. of X is given by

$$\begin{aligned} f_2(y) &= \int_{-\infty}^{\infty} f(x, y) dx = \int_{y/2}^{1/2} 12y dx \\ &= \begin{cases} 6y(1-y) & , 0 < y < 1 \\ 0 & , \text{o.w} \end{cases} \end{aligned}$$

$y < x < \frac{1}{2}$
 $\Rightarrow \frac{y}{2} < x < \frac{1}{2}$
 $\Rightarrow \int_{y/2}^{1/2}$

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

$$g(x/y) = \frac{f(x,y)}{f_2(y)} = \frac{12y}{6y(1-y)}$$

$$= \begin{cases} \frac{2}{1-y} & , \quad 0 < y < 2x < 1 \\ 0 & , \quad \text{o.w} \end{cases}$$

Definition (33): Let X and Y be any two r.v's with joint density $f(x,y)$ and marginals $f_1(x)$ and $f_2(y)$. The random variables X and Y are stochastically independent iff

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

for all $(x,y) \in R_X \times R_Y$.

Example (28): Let X and Y be two independent r.v's with identical p.d.f. given by

$$f(x) = \begin{cases} e^{-x} & , x > 0 \\ 0 & , \text{o.w} \end{cases}$$

what is the probability density function of $W = \min(X,Y)$?

Solution: Let $G(w)$ be the c.d.f. of W . Then

$$G(w) = P(W \leq w)$$

$$= 1 - P(W > w)$$

$$= 1 - P(\min\{X,Y\} > w)$$

$$= 1 - P(X > w \text{ and } Y > w)$$

$$= 1 - P(X > w)P(Y > w) \quad (\text{since } X \text{ and } Y \text{ are independent}).$$

$$= 1 - \left(\int_w^\infty e^{-x} dx \right) \left(\int_w^\infty e^{-y} dy \right)$$

$$= 1 - e^{-2w}$$

Thus, the p.d.f. of W is

$$g(w) = \frac{d}{dw} G(w) = \frac{d}{dw} (1 - e^{-2w}) = 2e^{-2w}$$

$$= g(w) = \begin{cases} 2e^{-2w} & , w > 0 \\ 0 & , \text{o.w} \end{cases}$$

Definition (34): Let X and Y be any two r.v's with j.p.d.f $f(x,y)$. The product moment of X and Y , denoted by $E(XY)$, is defined as:

$$E(XY) = \begin{cases} \sum_{x \in R_X} \sum_{y \in R_Y} xy p(x,y) & \text{if } X \text{ and } Y \text{ are discrete r.v's} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x,y) dx dy & \text{if } X \text{ and } Y \text{ are continuous r.v's} \end{cases}$$

Definition (35) Let X and Y be any two r.v's with j.p.d.f $f(x,y)$. The covariance between X and Y , denoted by $\text{cov}(X,Y)$ or (σ_{xy}) , is defined as

$$\text{cov}(X,Y) = E[(X - \mu_X)(Y - \mu_Y)]$$

where μ_X and μ_Y are mean of X and Y , respectively. the mean of X is given by:

$$\mu_X = E(X) = \int_{-\infty}^{\infty} x f_1(x) dx = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x,y) dx dy$$

and similarly, the mean of Y is given by:

$$\mu_Y = E(Y) = \int_{-\infty}^{\infty} y f_2(y) dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x,y) dy dx$$

Theorem (15) Let X and Y be any two r.v's. Then:

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

Example (29): Let X and Y be continuous r.v's with j.p.d.f:

$$f(x,y) = \begin{cases} 2 & , 0 < y < 1-x, 0 < x < 1 \\ 0 & \text{o.w} \end{cases}$$

What is the covariance between X and Y ?

Solution: The marginal density function of X is given by

$$f_1(x) = \int_0^{1-x} 2 dy = 2(1-x)$$

Hence, the expected value of X is

$$\mu_X = E(X) = \int_0^1 x f_1(x) dx = \int_0^1 2(1-x) \cdot x dx = \frac{1}{3}$$

Similarly, the m.p.d.f. of Y is:

$$f_2(y) = \int_0^{1-y} 2 dx = 2(1-y)$$

Hence, the expected value of Y is:

$$\mu_Y = E(Y) = \int_0^1 y f_2(y) dy = \int_0^1 2y(1-y) dy = \frac{1}{3}$$

The product moment of X and Y is given by:

$$\begin{aligned} E(XY) &= \int_0^1 \int_0^{1-x} xy f(x,y) dy dx \\ &= \int_0^1 \int_0^{1-x} 2xy dy dx \\ &= 2 \int_0^1 x \left[\frac{y^2}{2} \right]_0^{1-x} dx = 2 \cdot \frac{1}{2} \int_0^1 x(1-x)^2 dx \\ &= \int_0^1 (x - 2x^2 + x^3) dx \\ &= \left[\frac{x^2}{2} - \frac{2x^3}{3} + \frac{x^4}{4} \right]_0^1 = \frac{1}{12} \end{aligned}$$

$$\begin{aligned} \text{Then; } \text{cov}(X,Y) &= E(XY) - E(X)E(Y) \\ &= \frac{1}{12} - \left(\frac{1}{3}\right)\left(\frac{1}{3}\right) = -\frac{1}{36} \end{aligned}$$

Theorem (16)

If X and Y are any two random variables and a, b, c and d are real constants, then

$$\text{cov}(aX + b, cY + d) = ac \text{cov}(X, Y).$$

Ex 5. $\int x f(x) dx$

$$E(X) = \int x f(x) dx$$

Ex

17

$E(XY)$

Example (30) If the product moment of X and Y is (3)

and the mean of X and Y are both equal to (2), then what is the covariance of the random variables

$(2X + 10)$ and $(-\frac{5}{2}Y + 3)$?

Solution: Since $E(XY) = 3$ and $E(X) = 2 = E(Y)$, the covariance of X and Y is given by:

$$\text{cov}(X, Y) = E(XY) - E(X)E(Y) = 3 - 4 = -1$$

Then the covariance of $(2X + 10)$ and $(-\frac{5}{2}Y + 3)$ is given by:

$$\begin{aligned} \text{cov}(2X + 10, -\frac{5}{2}Y + 3) &= (2)(-\frac{5}{2}) \text{cov}(X, Y) \\ &= (-5)(-1) \\ &= 5 \end{aligned}$$

Theorem (17) If X and Y are independent r.v's, then $E(XY) = E(X)E(Y)$.

Ex

Example (31) Let X and Y be two independent r.v's with respective density functions:

$$f(x) = \begin{cases} 3x^2, & 0 < x < 1 \\ 0, & \text{o.w} \end{cases}$$

and;

$$f(y) = \begin{cases} 4y^3, & 0 < y < 1 \\ 0, & \text{o.w} \end{cases}$$

what is $E(\frac{X}{Y})$?

Solution: Since X and Y are independent r.v's, then the j.p.d.f. of X and Y is given by:

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

$$\begin{aligned} \text{Therefore, } E\left(\frac{X}{Y}\right) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{x}{y} h(x, y) dx dy \\ &= \int_0^1 \int_0^1 \frac{x}{y} \cdot f(x) f(y) dx dy \end{aligned}$$

$$\begin{aligned}
 E\left(\frac{X}{Y}\right) &= \int_0^1 \int_0^1 \frac{x}{y} \cdot 3x^2 \cdot 4y^3 dx dy \\
 &= \left(\int_0^1 3x^3 dx \right) \left(\int_0^1 4y^2 dy \right) \\
 &= \left(\frac{3}{4} \right) \left(\frac{4}{3} \right) = 1
 \end{aligned}$$

Remark:

(1) $E\left(\frac{X}{Y}\right) \neq \frac{E(X)}{E(Y)}$, but only implies $E\left(\frac{X}{Y}\right) = E(X)E(Y^{-1})$.

(2) $E(Y^{-1}) \neq \frac{1}{E(Y)}$.

Theorem (18): If X and Y are independent r.v.'s, then the covariance between X and Y is always zero, that is:

$$\text{cov}(X, Y) = 0$$

if X and Y are independent

Theorem (19) Let X and Y be any two r.v.'s and let a and b be any two real numbers. Then

$$\text{var}(aX + bY) = a^2 \text{var}(X) + b^2 \text{var}(Y) + 2ab \text{cov}(X, Y)$$

Example (32)

If $\text{var}(X+Y) = 3$, $\text{var}(X-Y) = 1$, $E(X) = 1$ and $E(Y) = 2$. What is $E(XY)$?

$$\text{var}(X+Y) = \sigma_X^2 + \sigma_Y^2 + 2 \text{cov}(X, Y)$$

$$\text{var}(X-Y) = \sigma_X^2 + \sigma_Y^2 - 2 \text{cov}(X, Y)$$

$$\begin{aligned}
 \Rightarrow \text{cov}(X, Y) &= \frac{1}{4} [\text{var}(X+Y) - \text{var}(X-Y)] \\
 &= \frac{1}{4} [3 - 1] = \frac{1}{2}
 \end{aligned}$$

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

$$\frac{1}{2} = E(XY) - (1)(2)$$

$$\Rightarrow E(XY) = \frac{1}{2} + 2 = \frac{5}{2}$$

Theorem (20) If X and Y are independent r.v.'s with $E(X) = 0 = E(Y)$, then:

$$\text{Var}(XY) = \text{Var}(X) \cdot \text{Var}(Y)$$

Definition (36) Let X and Y be two r.v.'s. Then the correlation coefficient (ρ) between X and Y is given by

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

And, if X and Y are independent r.v.'s, then:

$$\rho = 0 \rightarrow \text{مستقلة}$$

Moment Generating Functions: العزوم موجدة دالة للمتغيرين العشوائيين

Definition (37) Let X and Y be two r.v.'s with joint density function $f(x, y)$. A real valued function

$M: \mathbb{R}^2 \rightarrow \mathbb{R}$, defined by

$$M(t_1, t_2) = E(e^{t_1 X + t_2 Y})$$

is called the joint moment generating function of X and Y if the expected values exists for all t_1 and t_2 where $-h < t_1 < h$ and $-K < t_2 < K$ for some positive h and K .

Remarks

(1) $M(t_1, 0) = M(t_1) = E e^{t_1 X}$, and $M(0, t_2) = M(t_2) = E e^{t_2 Y}$.

(2) $E(X^K) = \frac{\partial^K M(t_1, t_2)}{\partial t_1^K} \Big|_{(t_1=0, t_2=0)}$ for $K=1, 2, \dots$

$E(Y^K) = \frac{\partial^K M(t_1, t_2)}{\partial t_2^K} \Big|_{(t_1=0, t_2=0)}$ for $K=1, 2, \dots$

(3) $E(XY) = \frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} \Big|_{(t_1=0, t_2=0)}$.

(4) If X and Y are independent r.v.'s, then $M_{ax+by}(t) = M_X(at) M_Y(bt)$

Q10

Example (33) If the joint moment generating function of X and Y is:

$$M(t_1, t_2) = e^{(t_1 + 3t_2 + 2t_1^2 + 18t_2^2 + 12t_1t_2)}$$

what is the covariance of X and Y ?

Solution: $M(t_1, t_2) = e^{(t_1 + 3t_2 + 2t_1^2 + 18t_2^2 + 12t_1t_2)}$

$$\frac{\partial M}{\partial t_1} = (1 + 4t_1 + 12t_2) M(t_1, t_2)$$

$$\left. \frac{\partial M}{\partial t_1} \right|_{(0,0)} = M(0,0) = 1 \quad \text{✓}$$

$$\frac{\partial M}{\partial t_2} = (3 + 36t_2 + 12t_1) M(t_1, t_2)$$

$$\left. \frac{\partial M}{\partial t_2} \right|_{(0,0)} = 3 M(0,0) = 3 \quad \text{✓}$$

Hence; $\mu_X = 1$ and $\mu_Y = 3$

$$\frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} = \frac{\partial}{\partial t_2} \left(\frac{\partial M}{\partial t_1} \right)$$

$$= \frac{\partial}{\partial t_2} [M(t_1, t_2) (1 + 4t_1 + 12t_2)]$$

$$= (1 + 4t_1 + 12t_2) \frac{\partial M}{\partial t_2} + M(t_1, t_2) \cdot (12)$$

$$\frac{\partial^2 M(t_1, t_2)}{\partial t_1 \partial t_2} = (1)(3) + (1)(12) = 15 = \underline{E(XY)}$$

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

$$= 15 - (1)(3)$$

$$= \underline{12}$$

كيفية Conditional Expectation

Definition (38). The conditional mean of X given $Y=y$ is defined as

$$\mu_{X/Y} = E(X/Y).$$

where

$$E(X/Y) = \begin{cases} \sum_{x \in R_X} x g(x/y) & \text{if } X \text{ is discrete} \\ \int_{-\infty}^{\infty} x g(x/y) dx & \text{if } X \text{ is continuous} \end{cases}$$

Similarly, the conditional mean of Y given $X=x$ is defined

$$E(Y/X) = \begin{cases} \sum_{y \in R_Y} y h(y/x) & \text{if } Y \text{ is discrete} \\ \int_{-\infty}^{\infty} y h(y/x) dy & \text{if } Y \text{ is continuous} \end{cases}$$

Definition (39). The expected value of the random variable $E(Y/X)$ is equal to the expected value of Y , that is

$$E_x(E_{y/x}(Y/X)) = E_y(Y).$$

Example (34). An insect lays Y number of eggs, where Y has a Poisson distribution with parameter (λ) . If the probability of each egg surviving is (p) , then on the average how many eggs will survive?

Sol: Let X denote the number of surviving eggs. Then, given that $Y=y$ (that is given that the insect has laid y eggs) the number variable X has a binomial (y, p) . Thus

$$X/Y \sim \text{Bin}(Y, p)$$

$$Y \sim \text{Poi}(\lambda)$$

Therefore, the expected number of survivors is given by:

$$\begin{aligned}
 E_x(X) &= E_y [E_{x/y}(X/Y)] \\
 &= E_y [PY] \\
 &= P E_y(Y) \\
 &= P\lambda
 \end{aligned}$$

Theorem (21)

Let X and Y be two r.v.s with mean μ_x and μ_y , and standard deviation σ_x and σ_y , respectively. If the conditional expectation of Y given $X=x$ is Linear in x , then

$$E(Y/X=x) = \mu_y + \rho \frac{\sigma_y}{\sigma_x} (x - \mu_x).$$

where ρ denotes the correlation coefficient of X and Y .

Ex
Example (35) Suppose X and Y are random variables with $E(Y/X=x) = -x+3$ and $E(X/Y=y) = -\frac{1}{4}y+5$. What is the correlation coefficient of X and Y ?

Sol: $\mu_y + \rho \frac{\sigma_y}{\sigma_x} (x - \mu_x) = -x + 3$

Therefore, equating the coefficients of x terms, we get:

$$\rho \frac{\sigma_y}{\sigma_x} = -1 \quad \dots (1)$$

Similarly, since

$$\mu_x + \rho \frac{\sigma_x}{\sigma_y} (y - \mu_y) = -\frac{1}{4}y + 5$$

We have $\rho \frac{\sigma_x}{\sigma_y} = -\frac{1}{4} \quad \dots (2)$

Multiplying (1) with (2), we get:

$$\rho \frac{\sigma_y}{\sigma_x} \cdot \rho \frac{\sigma_x}{\sigma_y} = (-1) \left(-\frac{1}{4}\right) \Rightarrow \rho^2$$

$$\rho^2 = \frac{1}{4} \Rightarrow \rho = \pm \frac{1}{2}$$

Since $\rho \frac{\sigma_y}{\sigma_x} = -1$ and $\frac{\sigma_y}{\sigma_x} > 0$, we get $\rho = -\frac{1}{2}$

Definition (40) Let X and Y be two r.v.'s with j.p.d.f. $f(x)$ and $f(y/x)$ be the conditional density of Y given $X=x$. The conditional variance of Y given $X=x$, denoted by $\text{Var}(Y/x)$ is denoted by:

$$\text{var}(Y/x) = E(Y^2/x) - [E(Y/x)]^2$$

Ex
Example (36) suppose the following j.p.d.f.:

$$f(x,y) = \begin{cases} e^{-y} & , 0 < x < y < \infty \\ 0 & , \text{o.w} \end{cases}$$

Compute $\text{Var}(Y/x)$.

Sol
Solution:

$$f_1(x) = \int_{-\infty}^{\infty} f(x,y) dy = \int_x^{\infty} e^{-y} dy = e^{-x}$$

$$h(y/x) = \frac{f(x,y)}{f_1(x)} = \frac{e^{-y}}{e^{-x}} = e^{-(y-x)}, y > x$$

$$E(Y/x) = \int_{-\infty}^{\infty} y h(y/x) dy = \int_x^{\infty} y e^{-(y-x)} dy$$

Let $z = y - x$, then:

$$E(Y/x) = \int_0^{\infty} (z+x) e^{-z} dz$$

$$= \int_0^{\infty} z e^{-z} dz + x \int_0^{\infty} e^{-z} dz$$

$$= \Gamma(2) + x \Gamma(1) = x + 1$$

$$E(Y^2/x) = \int_0^{\infty} y^2 h(y/x) dy = \int_x^{\infty} y^2 e^{-(y-x)} dy$$

$$= \int_0^{\infty} (z+x)^2 e^{-z} dz, \text{ where } z = y - x$$

$$= x^2 \int_0^{\infty} e^{-z} dz + \int_0^{\infty} z^2 e^{-z} dz + 2x \int_0^{\infty} z e^{-z} dz$$

$$= x^2 \Gamma(1) + \Gamma(3) + 2x \Gamma(2)$$

$$= x^2 + 2 + 2x = (1+x)^2 + 1$$

$$\therefore \text{var}(Y/X) = E(Y^2/X) - [E(Y/X)]^2$$

$$= (1+x)^2 + 1 - (1+x)^2$$

$$= \underline{\underline{1}}$$

Theorem (22) Let X and Y be two r.v's with mean μ_X and μ_Y , and standard deviation σ_X and σ_Y respectively. If the conditional expectation of Y given $X=x$ is linear in x , then;

$$E_x[\text{var}(Y/X)] = (1-\rho^2) \text{var}(Y)$$

Ex
Example (37). Let $E(Y/X=x) = 2x$ and $\text{Var}(Y/X=x) = 4x^2$, and let $X \sim U(0,1)$. what is the variance of Y ?

Solution

If $E(Y/X=x)$ is linear function of x , then

$$E(Y/X=x) = \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X)$$

and;

$$E_x[\text{var}(Y/X)] = \sigma_Y^2 (1-\rho^2)$$

we are given that;

$$\mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X) = 2x$$

Hence, equating the coefficient of x terms, we get

$$\rho \frac{\sigma_Y}{\sigma_X} = 2$$

which is

$$\underline{\underline{\rho = 2 \frac{\sigma_X}{\sigma_Y}}}$$

Further, we are given that $\text{Var}(Y/X=x) = 4x^2$.

Since $X \sim U(0,1)$, we get the density of X to be $f(x)=1$ on the interval $(0,1)$. Therefore,

$$E_x[\text{var}(Y/X)] = \int_{-\infty}^{\infty} \text{var}(Y/X=x) f(x) dx$$

$$= \int_0^1 4x^2 dx = \frac{4}{3}$$

$X \sim U(0,1)$
 $f(x) = 1$
 $\int_0^1 1 dx = 1$

By Theorem :

$$\begin{aligned}\frac{4}{3} &= E_x [\text{var}(Y/X)] \\ &= \sigma_y^2 (1 - \rho^2) \\ &= \sigma_y^2 (1 - 4 \frac{\sigma_x^2}{\sigma_y^2}) \\ &= \sigma_y^2 - 4 \sigma_x^2\end{aligned}$$

Hence;

$$\sigma_y^2 = \frac{4}{3} + 4 \sigma_x^2$$

Since $X \sim U(0,1)$, the variance of X is given by $\sigma_x^2 = \frac{1}{12}$.
Therefore, the variance of Y is given by:

$$\begin{aligned}\sigma_y^2 &= \frac{4}{3} + \frac{4}{12} \\ &= \frac{5}{3}\end{aligned}$$

Example (38) Let $E(X/Y=y) = 3y$ and $\text{var}(X/Y=y) = 2$ and let Y have the p.d.f. $f(y) = e^{-y}$, $y > 0$. Find $\text{var}(X)$.

Sol: By previous Theorem:

$$\text{var}(X/Y=y) = \sigma_x^2 (1 - \rho^2) = 2 \quad \dots (2)$$

and

$$\mu_x + \rho \frac{\sigma_x}{\sigma_y} (y - \mu_y) = 3y$$

$$\Rightarrow \rho = 3 \frac{\sigma_y}{\sigma_x}$$

Hence from (2), we get $E_y [\text{var}(X/Y)] = 2$ and thus

$$\sigma_x^2 (1 - 9 \frac{\sigma_y^2}{\sigma_x^2}) = 2$$

which is:

$$\sigma_x^2 = 9 \sigma_y^2 + 2$$

Now;

$$E(Y) = \int_0^\infty y f(y) dy = \int_0^\infty y e^{-y} dy = 1$$

$$E(Y^2) = \int_0^\infty y^2 f(y) dy = \int_0^\infty y^2 e^{-y} dy = 2$$

$$\Rightarrow \sigma_y^2 = E(Y^2) - [E(Y)]^2 = 2 - 1 = 1$$

$$\Rightarrow \sigma_x^2 = 9 \sigma_y^2 + 2 = 9(1) + 2 = 11$$

Transformation of Random Variables and their distribution.

There are several methods for finding the probability density function of a transformation r.v's. Some of these methods are:

- (1) Distribution function method (c.d.f.)
- (2) Transformation method
- (3) Convolution method, and
- (4) Moment generating function method.

(1) Distribution Function Method: (c.d.f.)

Example (39) A box is to be constructed so that the height is (4) inches and its base is X inches. If $X \sim N(0, 1)$. what is the distribution of the volume of the box?

Sol: The volume is an r.v., since X is an r.v. The r.v. V is given by $V = 4X^2$.

$$G(v) = P(V \leq v) = P(4X^2 \leq v)$$

$$= P\left(-\frac{1}{2}\sqrt{v} \leq X \leq \frac{1}{2}\sqrt{v}\right)$$

$$= \int_{-\frac{1}{2}\sqrt{v}}^{\frac{1}{2}\sqrt{v}} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = 2 \int_0^{\frac{1}{2}\sqrt{v}} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

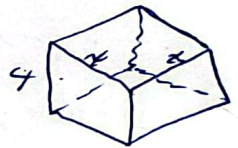
$$g(v) = \frac{d}{dv} G(v) = \frac{d}{dv} \left(2 \int_0^{\frac{1}{2}\sqrt{v}} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \right)$$

$$= 2 \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{1}{2}\sqrt{v}\right)^2} \cdot \left(\frac{1}{2}\right) \frac{d\sqrt{v}}{dv} + 0$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{8}v} \cdot \frac{1}{2\sqrt{v}}$$

$$= \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{2\sqrt{v}} e^{-v/8}$$

$$\Rightarrow V \sim \text{Gam}\left(\frac{1}{2}, 8\right)$$



Some useful Examples:

Example (40): Suppose $Y \sim U(0,1)$ and $Y = \frac{1}{4}X^2$. what is the Probability function of X

Sol: By using cdf:

$$\begin{aligned} F(x) &= P(X \leq x) = P(X^2 \leq x^2) \\ &= P\left(\frac{1}{4}X^2 \leq \frac{1}{4}x^2\right) \\ &= P\left(Y \leq \frac{1}{4}x^2\right) \\ &= \int_0^{x^2/4} f(y) dy = \int_0^{x^2/4} 1 dy \\ &= [y]_0^{x^2/4} = \frac{x^2}{4} \end{aligned}$$

$$f(x) = \frac{\partial}{\partial x} F(x) = \begin{cases} \frac{x}{2}, & 0 \leq x \leq 2 \\ 0, & \text{o.w} \end{cases}$$

Hint: $0 \leq Y \leq 1 \Rightarrow 0 \leq \frac{1}{4}x^2 \leq 1 \Rightarrow 0 \leq x^2 \leq 4$
 $\Rightarrow 0 \leq x \leq 2$

Exa
Example (41): Let $f(x) = \frac{e^{-x}}{(1+e^{-x})^2}$, $x \in \mathbb{R}$.

Find the pdf of $Y = \frac{1}{1+e^{-x}}$.

Sol By using cdf:

$$\begin{aligned} F(y) &= P(Y \leq y) = P\left(\frac{1}{1+e^{-x}} \leq y\right) \\ &= P\left(1+e^{-x} \geq \frac{1}{y}\right) \\ &= P\left(e^{-x} \geq \frac{1}{y} - 1\right) = P\left(-x \geq \ln\left(\frac{1}{y} - 1\right)\right) \\ &= P\left(X \leq -\ln\left[\frac{1}{y} - 1\right]\right) = P\left(X \leq -\ln\left(\frac{1-y}{y}\right)\right) \\ &= P\left(X \leq \ln\left[\frac{1-y}{y}\right]^{-1}\right) = P\left(X \leq \ln \frac{y}{1-y}\right) \\ &= \int_{-\infty}^{\ln \frac{y}{1-y}} f(x) dx = \int_{-\infty}^{\ln \frac{y}{1-y}} \frac{e^{-x}}{(1+e^{-x})^2} dx \\ &= \int_{-\infty}^{\ln \frac{y}{1-y}} (1+e^{-x})^{-2} (e^{-x}) dx \end{aligned}$$

$$\begin{aligned}
 \therefore F(y) &= - \left[\frac{1 + e^{-x}}{-2+1} \right]_{-\infty}^{\ln \frac{y}{1-y}} = \left[1 + e^{-x} \right]_{-\infty}^{\ln \frac{y}{1-y}} \\
 &= \left[\left(1 + e^{-\ln \frac{y}{1-y}} \right)^{-1} - \left(1 + e^{-\infty} \right)^{-1} \right] \\
 &= \left[\left(1 + e^{\ln \left(\frac{y}{1-y} \right)^{-1}} \right)^{-1} - \left(\frac{1}{1+e^{\infty}} \right) \right] \quad \begin{matrix} \infty = \infty \\ e^{\infty} = 0 \end{matrix} \\
 &= \left[\left(1 + \left(\frac{y}{1-y} \right)^{-1} \right)^{-1} - 0 \right] = \left[1 + \frac{1-y}{y} \right]^{-1} \\
 &= \left[\frac{y+1-y}{y} \right]^{-1} = \left[\frac{1}{y} \right]^{-1} = y
 \end{aligned}$$

$$\therefore F(y) = y$$

Hint $-\infty < x < \infty \Rightarrow$ if $x \rightarrow -\infty \Rightarrow y = \frac{1}{1+e^{\infty}} = 0$
if $x \rightarrow \infty \Rightarrow y = \frac{1}{1+e^{-\infty}} = 1$

$$\Rightarrow 0 < y < 1$$

$$\therefore f(y) = \frac{\partial}{\partial y} F(y) = \begin{cases} 1 & , 0 < y < 1 \\ 0 & , 0 \text{ or } 1 \end{cases}$$

$$\Rightarrow Y \sim U(0, 1)$$

Example (42) If $X \sim \text{Gam}(\alpha=1, \beta=2)$. Find the pdf. of $Y = e^X$.

Sol: By using cdf:

$$F(y) = P(Y \leq y) = P(e^X \leq y) = P(X \leq \ln y)$$

$$= \int_0^{\ln y} f(x) dx = \int_0^{\ln y} \frac{1}{2} e^{-\frac{x}{2}} dx$$

$$= -e^{-x/2} \Big|_0^{\ln y} = -e^{-\frac{1}{2} \ln y} + e^0$$

$$= -e^{\ln y^{-1/2}} + 1 = \left(1 - \frac{1}{\sqrt{y}} \right)$$

$$g(y) = \frac{\partial}{\partial y} F(y) = \frac{d}{dy} \left(1 - \frac{1}{\sqrt{y}} \right) = \frac{1}{2\sqrt{y}}$$

Hint: $0 < X < \infty \Rightarrow$ if $X \rightarrow 0 \Rightarrow y = e^0 = 1$
if $y \rightarrow \infty \Rightarrow y = e^{\infty} = \infty$

$$\therefore g(y) = \begin{cases} \frac{1}{2\sqrt{y}} & , y > 1 \\ 0 & , 0 \text{ or } \infty \end{cases}$$

Ex. 43. Example (43): Let $X \sim N(\mu, \sigma^2)$
 Show that $\left(\frac{X-\mu}{\sigma}\right)^2 \sim \chi^2_{(1)}$

Sol: by using cdf: let $\left(\frac{X-\mu}{\sigma}\right)^2 = W$

$$F(w) = P(W \leq w)$$

$$= P\left(\left(\frac{X-\mu}{\sigma}\right)^2 \leq w\right)$$

$$= P\left(-\sqrt{w} \leq \frac{X-\mu}{\sigma} \leq \sqrt{w}\right)$$

$$= P\left(-\sqrt{w} \leq Z \leq \sqrt{w}\right) = \int_{-\sqrt{w}}^{\sqrt{w}} f(z) dz$$

$$f(w) = \frac{d}{dw} F(w)$$

$$= \frac{d}{dw} \int_{-\sqrt{w}}^{\sqrt{w}} f(z) dz$$

$$= f(\sqrt{w}) \cdot \frac{d}{dw}(\sqrt{w}) - f(-\sqrt{w}) \cdot \frac{d}{dw}(-\sqrt{w})$$

$$= f(\sqrt{w}) \cdot \frac{1}{2\sqrt{w}} - f(-\sqrt{w}) \cdot \frac{-1}{2\sqrt{w}}$$

$$= \frac{1}{\sqrt{2\pi}} e^{-w/2} \cdot \frac{1}{2\sqrt{w}} + \frac{1}{\sqrt{2\pi}} e^{-w/2} \cdot \frac{1}{2\sqrt{w}}$$

$$= 2 \cdot \frac{1}{\sqrt{2\pi}} e^{-w/2} \cdot \frac{1}{2\sqrt{w}} = \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sqrt{w}} e^{-w/2}$$

$$\therefore f(w) = \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sqrt{w}} e^{-w/2} = \begin{cases} \frac{1}{2^{1/2} \Gamma_{\frac{1}{2}}} w^{1/2-1} e^{-w/2} & w > 0 \\ 0 & , 0 \cdot w \end{cases}$$

$\Rightarrow W \sim \chi^2_{(1)}$

(2) Transformation Method

نظريات التحويل

① (a) Univariate Case

Theorem (23) Let X be a continuous r.v. with p.d.f. $f(x)$. Let $y = T(x)$ be an increasing (or decreasing) function. Then the density function of the r.v. $Y = T(X)$ is given by:

$$g(y) = \left| \frac{dx}{dy} \right| f(w(y))$$

where $x = w(y)$ is the inverse function of $T(x)$.

Example (44). Let $y = -\ln X$. If $X \sim U(0, 1)$, then what is the density function of Y where nonzero?

Solution:

$$y = T(x) = -\ln x \quad \text{أذن}$$

$$x = e^{-y}$$

$$\frac{dx}{dy} = -e^{-y}$$

الطريق
من x إلى y
من x إلى y

$$g(y) = \left| \frac{dx}{dy} \right| f(w(y))$$

$$= e^{-y} \cdot 1$$

$$= \begin{cases} e^{-y} & , y > 0 \\ 0 & , y \leq 0 \end{cases}$$

$$\Rightarrow Y \sim \exp(1).$$

② (b) Bivariate Case

Theorem (24). Let X and Y be two continuous r.v.s with j.p.d.f. $f(x, y)$. Let $u = P(X, Y)$ and $v = Q(X, Y)$ be functions of X and Y . If the functions $P(X, Y)$ and $Q(X, Y)$ have single valued inverses, say $X = R(u, v)$ and $Y = S(u, v)$, then the joint density $g(u, v)$ of u and v is given by:

$$g(u, v) = |J| f(R(u, v), S(u, v))$$

where

$$J = \det \begin{pmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{pmatrix}$$

Ex 45
Example (45) Let X and Y have the j.p.d.f.:

$$f(x, y) = \begin{cases} 8xy & , 0 < x < y < 1 \\ 0 & , \text{o.w.} \end{cases}$$

what is the joint density of $u = \frac{x}{y}$ and $v = y$?

Solution:

$$u = \frac{x}{y}$$

$$v = y$$

$$\Rightarrow \begin{cases} x = uv \\ y = v \end{cases}$$

$$\frac{\partial x}{\partial u} = v$$

$$\frac{\partial x}{\partial v} = u$$

$$\frac{\partial y}{\partial u} = 0$$

$$\frac{\partial y}{\partial v} = 1$$

$$|J| = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} v & u \\ 0 & 1 \end{vmatrix} = v$$

The joint density function of u and v is:

$$g(u, v) = |J| f(R(u, v), S(u, v))$$

$$= v \times 8(uv)v$$

$$= 8uv^3 f(uv, v)$$

$$= 8uv^3 \times v$$

Note that, since

$$0 < \frac{x}{y} < y < 1 \Rightarrow 0 < uv < v < 1$$

The last inequalities yield:

$$0 < uv < v \Rightarrow 0 < u < 1$$

$$0 < v < 1 \Rightarrow 0 < v < 1$$

Thus,

$$g(u, v) = \begin{cases} 8uv^3 & , 0 < u < 1 ; 0 < v < 1 \\ 0 & , \text{o.w.} \end{cases}$$

Ex 46
Example (46) Let each of the independent r.v's X and Y have the density function:

$$f(x) = \begin{cases} e^{-x} & , 0 < x < \infty \\ 0 & , \text{o.w.} \end{cases}$$

what is the j.p.d.f. of $U = X$ and $V = 2X + 3Y$ and the domain on which this density is positive?

Solution: $u = x$
 $v = 2x + 3y \Rightarrow \begin{cases} x = u \\ y = \frac{1}{3}v - \frac{2}{3}u \end{cases}$

$$|J| = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 1 & 0 \\ -\frac{2}{3} & \frac{1}{3} \end{vmatrix} = \frac{1}{3}$$

The joint function of u and v is $g(u, v)$

$$g(u, v) = |J| f(R(u, v), S(u, v))$$

$$= \frac{1}{3} e^{-\frac{u}{3}} e^{-\frac{1}{3}v + \frac{2}{3}u}$$

$$= \begin{cases} \frac{1}{3} e^{-\frac{(u+v)}{3}} & 0 < u < \infty \\ & 0 < v < \infty \end{cases}$$

Further, since $v = 2u + 3y$ and $3y > 0$

$$\Rightarrow v > 2u \Rightarrow 0 < 2u < v < \infty$$

$$\therefore g(u, v) = \begin{cases} \frac{1}{3} e^{-\frac{(u+v)}{3}} & , 0 < 2u < v < \infty \\ 0 & , \text{o.w.} \end{cases}$$

Ex 47

Example (47) Let X and Y be independent r.v's, each with density function:

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2}, \quad -\infty < x < \infty$$

Let $u = \frac{x}{y}$, and $v = y$. what is the joint density of u and v ? what is the density of u ?

Solution:

$$\begin{cases} u = \frac{x}{y} \\ v = y \end{cases} \Rightarrow \begin{cases} x = uv \\ y = v \end{cases}$$

$$|J| = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} v & u \\ 0 & 1 \end{vmatrix} = v$$

$$f(x, y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} + \frac{1}{\sqrt{\pi}} e^{-\frac{y^2}{2}} \quad R(u, v) \leq K = uv$$

The joint density function of u and v is

$$\begin{aligned} g(u, v) &= |J| f(R(u, v), S(u, v)) \\ &= |v| \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}v^2} \\ &= \begin{cases} \frac{|v|}{2\pi} e^{-\frac{1}{2}v^2(u^2+1)} & -\infty < u < \infty \\ 0 & -\infty < v < \infty \end{cases} \end{aligned}$$

$$S(u, v) = y = v$$

ملاحظة
السؤال

The marginal $g_1(u)$ of u is given by:

$$\begin{aligned} g_1(u) &= \int_{-\infty}^{\infty} g(u, v) dv = \int_{-\infty}^{\infty} \frac{|v|}{2\pi} e^{-\frac{1}{2}v^2(u^2+1)} dv \\ &= \int_{-\infty}^0 (-v) \frac{1}{2\pi} e^{-\frac{1}{2}v^2(u^2+1)} dv + \int_0^{\infty} v \frac{1}{2\pi} e^{-\frac{1}{2}v^2(u^2+1)} dv \\ &= \frac{1}{2\pi} \left(\frac{1}{2} \right) \left[\frac{2}{u^2+1} e^{-\frac{1}{2}v^2(u^2+1)} \right]_0^{\infty} + \frac{1}{2\pi} \left(\frac{1}{2} \right) \left[\frac{-2}{u^2+1} e^{-\frac{1}{2}v^2(u^2+1)} \right]_0^{\infty} \end{aligned}$$

ملاحظة
2V(u^2+1)

$$f(-v) = \frac{1}{2\pi} \cdot \frac{1}{u^2+1} + \frac{1}{2\pi} \cdot \frac{1}{u^2+1}$$

$$\frac{1}{u^2+1} = \begin{cases} \frac{1}{\pi(u^2+1)} & -\infty < x < \infty \\ 0 & \text{o.w} \end{cases} \Rightarrow u \sim \text{Cauchy}(0, 0)$$

Theorem (25)

Let the joint density of the r.v's X and Y be $f(x, y)$. Then the probability density functions $X+Y$, XY , and $\frac{X}{Y}$ are given by:

$$h_{X+Y}(v) = \int_{-\infty}^{\infty} f(u, v-u) du$$

$$h_{XY}(v) = \int_{-\infty}^{\infty} \frac{1}{|u|} f(u, \frac{v}{u}) du$$

$$h_{\frac{X}{Y}}(v) = \int_{-\infty}^{\infty} |u| f(u, vu) du$$

In addition, if the r.v's X and Y are independent and have the m.p.d.f's $f(x)$ and $g(y)$ respectively, then we have

$$h_{X+Y}(z) = \int_{-\infty}^{\infty} g(y) f(z-y) dy$$

$$h_{XY}(z) = \int_{-\infty}^{\infty} \frac{1}{|y|} g(y) f\left(\frac{z}{y}\right) dy$$

$$h_{\frac{X}{Y}}(z) = \int_{-\infty}^{\infty} |y| g(y) f(zy) dy$$

3. Convolution Method for Sums of Random Variables

Definition (41) Let f and g be two real-valued functions. The convolution of f and g denoted by $f * g$, is defined as

$$\begin{aligned} (f * g)(z) &= \int_{-\infty}^{\infty} f(z-y) g(y) dy \\ &= \int_{-\infty}^{\infty} g(z-x) f(x) dx \end{aligned}$$

Hint: (1) $f * g = g * f$

(2) If X and Y be two independent r.v's with p.m.f $f(x)$ and $g(y)$. Then, we get:

$$h(z) = \int_{-\infty}^{\infty} f(z-y) g(y) dy$$

Exo Example (48) what is the probability density of the sum of two independent r.v's, each of which is gamma with $\alpha = 1$ and $\beta = 1$?

Solution: Let $Z = X + Y$, where $X \sim \text{Gam}(1, 1)$ and $Y \sim \text{Gam}(1, 1)$ with the following densities:

$$f(x) = \begin{cases} e^{-x} & , x > 0 \\ 0 & , x \leq 0 \end{cases}$$

$$g(y) = \begin{cases} e^{-y} & , y > 0 \\ 0 & , y \leq 0 \end{cases}$$

since X and Y are independent, the $f(z)$ can be obtained by the convolution method. Note that the sum $Z = X + Y$ is between

(0) and (∞), and $0 < x < z$. Hence $h(z)$ is given by:

$$h(z) = (f * g)(z)$$

$$= \int_{-\infty}^{\infty} f(\underbrace{z-x}) g(x) dx \quad \begin{matrix} z = x+y \\ y = z-x \end{matrix}$$

$$= \int_0^z f(z-x) g(x) dx$$

$$= \int_0^z e^{-(z-x)} e^{-x} dx \quad \begin{matrix} g(y) = g(z-x) \\ \Rightarrow f(z-x) \end{matrix}$$

$$= \int_0^z e^{-z} dx = x e^{-z} \Big|_0^z = z e^{-z}$$

$$= \frac{1}{\Gamma(2)} z^{2-1} e^{-z}, \quad z > 0$$

Hence $Z \sim \text{Gam}(1, 2)$. Thus, if $X \sim \text{Gam}(1, 1)$ and $Y \sim \text{Gam}(1, 1)$ then $X+Y \sim \text{Gam}(1, 2)$.

Example (49) what is the Probability density of the sum of two independent r.v's, each of which is Standard Normal?

Solution: Let $Z = X + Y$, where $X \sim N(0, 1)$ and $Y \sim N(0, 1)$. Hence, the density of X is given by:

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

Similarly, the density of Y is given by:

$$g(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}$$

Since X and Y are independent, the density of Z can be obtained by the method of convolution. The sum $Z = x + y$ is between $(-\infty)$ and (∞) .

Hence, the density Function of Z is given by: