

Joint Distribution.

Probability distribution fn will be,

$$F_X(x) = P\{X \leq x\}.$$

$$F_Y(y) = P\{Y \leq y\}$$

$$F_{XY}(x, y) = P\{X \leq x ; Y \leq y\}.$$

For discrete random variables X and Y ,

$$F_{XY}(x, y) = \sum_{n=1}^N \sum_{m=1}^M P(x_n, y_m) u(x - x_n) u(y - y_m).$$

Joint density fn.

$$f_{XY}(x, y) = \frac{d^2 F_{XY}(x, y)}{dx dy}$$

for discrete variables

$$f_{XY}(x, y) = \sum_{n=1}^N \sum_{m=1}^M P(x_n, y_m) \delta(x - x_n) \delta(y - y_m)$$

Properties:

$$(i) f_{XY}(x, y) \geq 0$$

$$(ii) \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = 1 \rightarrow \text{Valid}$$

$$(iii) F_{XY}(x, y) = \int_{-\infty}^y \int_{-\infty}^x f_{XY}(\xi_1, \xi_2) d\xi_1 d\xi_2.$$

$$(iv) f_X(x) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(\xi_1, \xi_2) d\xi_1 d\xi_2.$$



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$$f_Y(y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x_1, x_2) dx_1 dx_2$$

$$(v) \quad P[x_1 < X \leq x_2; y_1 < Y \leq y_2] =$$

$$\int_{y_1}^{y_2} \int_{x_1}^{x_2} f_{XY}(x, y) dx dy.$$

$$(vi) \quad f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy.$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(x, y) dx$$

Conditional distribution & density

$$F_X(x|B) = P[X \leq x|B].$$

$$f_X(x|B) = \frac{P[X \leq x \cap B]}{P[B]}.$$

Point Conditioning.

$$f_X(x|Y=y_k) = \frac{\sum_{i=1}^N P(x_i, y_k)}{P(y_k)} u(x - x_i).$$

$$f_X(x|Y=y_k) = \frac{\sum_{i=1}^N P(x_i, y_k)}{P(y_k)} \delta(x - x_i).$$

Internal Conditioning

$$f_X(x | y_a < Y \leq y_b) = \frac{\int_{y_a}^{y_b} f_{XY}(x, y) dy}{\int_{y_1}^{y_2} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy}$$

Statistical Independence

$$P(A \cap B) = P(A) \cdot P(B).$$

for two events A, B defined.

$$A = \{X \leq x\}; B = \{Y \leq y\}.$$

$$P[X \leq x, Y \leq y] = P(X \leq x) \cdot P(Y \leq y)$$

$$F_{XY}(x, y) = F_X(x) \cdot F_Y(y).$$

(Joint) marginal of x marginal of y

Conditional distribution for independent variables.

$$F_X(x | Y \leq y) = \frac{F_{XY}(x, y)}{F_Y(y)}$$

Joint moments about the origin.

$$E(x^n y^k) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x^n y^k f_{XY}(x, y) dx dy = m_{n, k}$$

so, $m_{1,1} = E[xy] \rightarrow$ cov b/w x and y .

Joint Central Moments

$$\mu_{mn} = E[(X-\bar{X})^n (Y-\bar{Y})^m] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x-\bar{x})^n (y-\bar{y})^m f_{XY}(x,y) dx dy$$

Here,

$$\mu_{20} = E[(X-\bar{X})^2] = \sigma_X^2$$

$$\mu_{02} = E[(Y-\bar{Y})^2] = \sigma_Y^2$$

} Second Order Central Moments

$$C_{XY} = \mu_{11} = E[(X-\bar{X})(Y-\bar{Y})] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x-\bar{x})(y-\bar{y}) f_{XY}(x,y) dx dy$$

covariant
of X and Y

Normalized Second order moment

$$\rho = \mu_{11} / \sqrt{\mu_{20} \mu_{02}} = C_{XY} / \sigma_X \sigma_Y$$

$$\rho = E \left[\frac{(X-\bar{X})}{\sigma_X} \cdot \frac{(Y-\bar{Y})}{\sigma_Y} \right] ; -1 \leq \rho \leq 1$$

Here, ρ refers to correlation coefficient.

for n random variables.

$$\mu_{n_1 n_2 \dots n_n} = E[(X_1 - \bar{X}_1)^{n_1} (X_2 - \bar{X}_2)^{n_2} \dots (X_n - \bar{X}_n)^{n_n}]$$
$$= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} (x_1 - \bar{x}_1)^{n_1} (x_2 - \bar{x}_2)^{n_2} \dots f_{X_1 X_2 \dots X_n}(x_1, \dots, x_n) dx_1 dx_2 \dots dx_n$$



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Joint characteristic fcn.

$$\Phi_{XY}(\omega_1, \omega_2) = E \left[e^{j\omega_1 x + j\omega_2 y} \right]; \omega_1, \omega_2 \rightarrow \text{real } \omega.$$

$$\Phi_{XY}(\omega_1, \omega_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) e^{j\omega_1 x + j\omega_2 y} dx dy.$$

from the inverse fourier transform, we get,

$$f_{XY}(x, y) = \frac{1}{(2\pi)^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \Phi_{XY}(\omega_1, \omega_2) e^{-j\omega_1 x - j\omega_2 y} d\omega_1 d\omega_2$$

Ex. (1) Solⁿ

$$f_{xy}(x, y) = \begin{cases} kx, & 0 < x < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\int_0^1 \int_0^y f_{xy}(x, y) dx dy = 1$$

$$\int_0^1 \int_0^y kx dx dy \Rightarrow k \int_0^1 \left(\frac{x^2}{2} \right)_0^y dy = 1$$

$$k \left(\frac{y^3}{6} \right)_0^1 = 1$$

$$\Rightarrow \frac{k}{6} = 1 \Rightarrow k = 6.$$

$$\text{So, } f_{xy}(x, y) = \begin{cases} 6x, & 0 < x < y < 1 \\ 0, & \text{otherwise} \end{cases}$$

to get marginal Pdf;

$$f(x) = \int_x^1 f(x, y) dy = \int_x^1 6x dy = 6x [y]_x^1$$

$$f_x(x) = 6x(1-x); \quad 0 < x < 1$$

$$\text{Similarly for } f_y(y) = \int_0^y f(x, y) dx$$

$$= \int_0^y 6x dx$$

$$= \frac{6x^2}{2} \Big|_0^y$$

$$f_y(y) = 3y^2; \quad 0 < y < 1$$

$$E(X) = \int_0^1 x f(x) dx$$

$$= \int_0^1 6(x - x^3) dx$$

$$= 6 \left[\frac{x^2}{2} - \frac{x^4}{4} \right]_0^1$$

$$E(X) = 6 \left[\frac{1}{2} - \frac{1}{4} \right]$$

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

$$E(Y) = \int_0^1 y f(y) dy$$

$$= \int_0^1 y \cdot 3y^2 dy$$

$$= 3 \left[\frac{y^3}{3} \right]_0^1 = \frac{1}{3}$$

$$E(XY) = \int_0^1 \int_0^3 xy f(x,y) dx dy$$

$$= \int_0^1 \int_0^y 6x^2y dx dy$$

$$E[XY] = 6 \int_0^1 y \left[\frac{x^3}{3} \right]_0^y dy$$

$$= \frac{6}{3} \int_0^1 y^4 dy$$

$$= \frac{2}{5}$$



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

$$= \frac{2}{5} - \left(\frac{1}{2}\right)\left(\frac{3}{4}\right)$$

$$= \frac{2}{5} - \frac{3}{8}$$

$$= \frac{1}{40}$$

$$\text{Cov}(X, Y) = 1/40$$

Q. 2) Let $z = ax + (1-a)y$; $0 \leq a \leq 1$

$$\text{Var}(z) = \text{Var}[ax + (1-a)y]$$

$$\text{Var}(z) = a^2 \text{Var}(X) + (1-a)^2 \text{Var}(Y)$$

$$\text{Var}(z) = a^2 \sigma_1^2 + (1-a)^2 \sigma_2^2$$

When variance is minimum,

$$\frac{d \text{Var}(z)}{da} = 0$$

$$\frac{d}{da} [a^2 \sigma_1^2 + (1-a)^2 \sigma_2^2] = 0$$

$$2a \sigma_1^2 - 2(1-a) \sigma_2^2 = 0$$

$$a \sigma_1^2 - \sigma_2^2 + a \sigma_2^2 = 0$$

$$a(\sigma_1^2 + \sigma_2^2) = \sigma_2^2$$

$$a = \frac{\sigma_2^2}{\sigma_1^2 + \sigma_2^2}$$

$$\begin{aligned}\text{var}^*(z) &= a^2 (\sigma_1^2 - \sigma_2^2) + \sigma_2^2 \\ &= \frac{\sigma_2^4}{(\sigma_1^2 + \sigma_2^2)^2} (\sigma_1^2 - \sigma_2^2) + \sigma_2^2\end{aligned}$$

$$\begin{aligned}\text{var}^*(z) &= \sigma_1^4 \sigma_2^2 + 3 \sigma_1^2 \sigma_2^4 \\ &\quad \sigma_1^4 + 2 \sigma_1^2 \sigma_2^2 + \sigma_2^4.\end{aligned}$$

Q4) $\phi_X(-it) = M_X(t).$

$$Z = ax + by + c, \quad c \neq 0.$$

$$M_Z(t) = M_{ax+by+c}(t) = M_{ax}(t) \cdot M_{by}(t) \cdot e^{ct}.$$

$$\rightarrow \exp \left[(a\mu_1 + b\mu_2)t + \frac{1}{2} (a^2\sigma_1^2 + b^2\sigma_2^2)t^2 \right].$$

mean & var = 0, 1

$$M_Z(t) = e^{\frac{1}{2}(a^2 + b^2)t^2}.$$

$$\phi_Z(-it) = e^{\frac{1}{2}(a^2 + b^2)(-it)^2}.$$

$$= e^{-\frac{1}{2}(a^2 + b^2)t^2} \quad \text{--- (1)}$$

Z is also a random variable.

$$\phi_Z(it) = e^{i\mu t - \frac{1}{2}\sigma^2 t^2}$$

$$E(Z) = 0; \text{var}(Z) = a^2 + b^2.$$

$$Z \sim N(0, a^2 + b^2) \quad \text{--- (3)}$$



$$Q.5) f_2(z) = \int_{z-i\infty}^z f_1(x) f_1(2-x) dx.$$

$$F_2(z) = \int_{y=-\infty}^z f_2(y) dy.$$

Case 1: $0 \leq z < 1$

$$f_1(z) = \int_0^z 1 \cdot 1 dx = z.$$

$$F_2(z) = \int_0^z y dy = \frac{z^2}{2}$$

Case 2: $1 \leq z \leq 2$.

$$f_1(z) = \int_{z-1}^1 1 \cdot 1 dx = 2-z.$$

$$F_2(z) = \int_0^1 y dy + \int_1^z (2-y) dy.$$

$$f_2(z) = 2z - \frac{z^2}{2} - 1.$$

Q. 4.6.3-

Random Variables X and Y have respective density function

$$f_X(x) = \frac{1}{a} [u(x) - u(x-a)],$$

$$f_Y(y) = be^{-by}.$$

where $a > 0$ & $b > 0$. Find & sketch the density fcn $W = X + Y$ if X and Y are statistically indep.

Soln

$$f_W(w) = \int_{-\infty}^{\infty} f_Y(y) f_X(w-y) dy.$$

$$= \int_0^{\infty} be^{-by} \frac{1}{a} [u(w-y) - u(w-y-a)] dy.$$

$$= \int_0^w \frac{b}{a} e^{-by} dy - \int_0^{w-a} \frac{b}{a} e^{-by} dy.$$

$$= 0 \text{ for } w \leq a.$$



Case (i): $0 < w \leq a$.

$$f_w(w) = \int_0^w \frac{b}{a} e^{-by} dy = \frac{1}{a} [1 - e^{-bw}].$$

Case (ii)

$a < w$

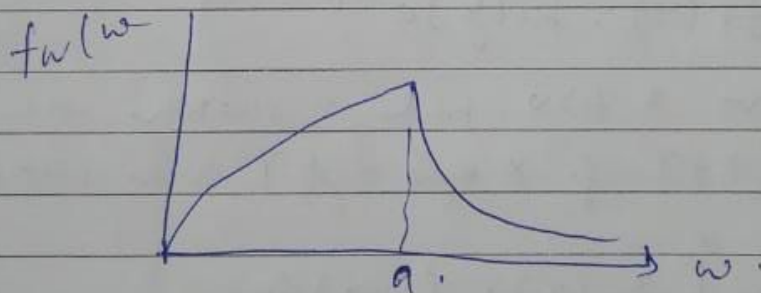
$$f_w(w) = \frac{1}{a} [1 - e^{-bw}] - \frac{b}{a} \int_0^{w-a} e^{-by} dy.$$

$$= \frac{e^{-bw}}{a} [e^{ba} - 1].$$

$$f_w(w) = 0; \quad w \leq 0.$$

$$= \frac{1}{a} e^{-bw} (e^{ba} - 1), \quad a < w$$

$$= \frac{1}{a} e^{-bw} (e^{ba} - 1), \quad a < w$$



B. 4. 6. 4.

Random variables X and Y have respectively density functions

$$f_X(x) = 0.18(x-1) + 0.28(x-2) + 0.48(x-3) + 0.38(x-4).$$

$$f_Y(y) = 0.48(y-5) + 0.58(y-6) + 0.18(y-7).$$

Find and sketch the density function $W = X + Y$ if X and Y are independent.

solⁿ $f_W(w) = \int_{-\infty}^{\infty} f_Y(y) f_X(w-y) dy$

$$= \int_{-\infty}^{\infty} 0.4 \delta(y-5) + 0.5 \delta(y-6) + 0.1 \delta(y-7) \{$$

$$[0.1 \delta(w-y-1) + 0.2 \delta(w-y-2) + 0.4 \delta(w-y-3)$$

$$+ 0.3 \delta(w-y-4)] dy.$$

each integral is of form

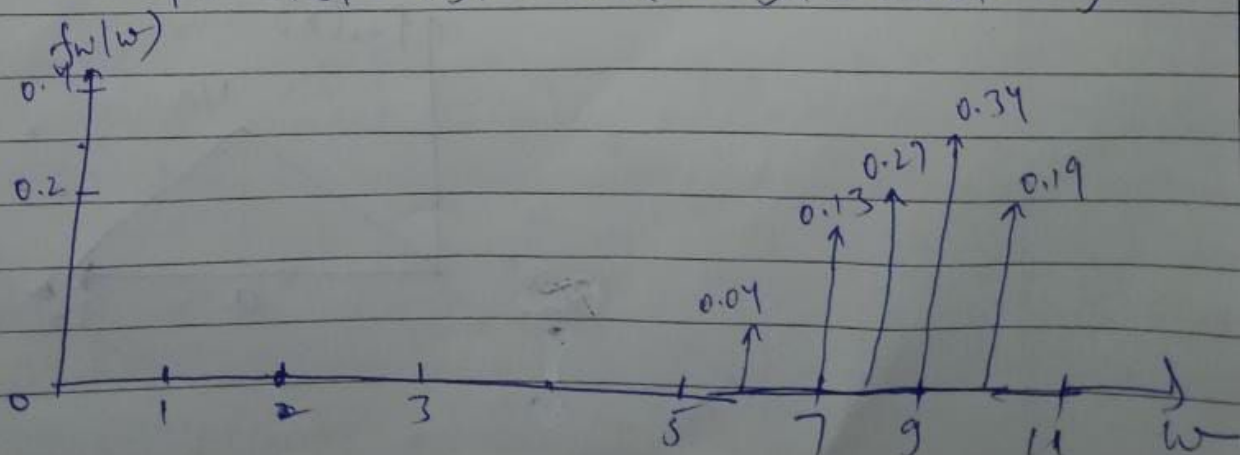
$$K \int_{-\infty}^{\infty} \delta(y-y_K) \delta(w-y-x_m) dy \quad \text{where}$$

y_K and x_m are numbers & K is a constant

the
 $w = x_m + y_K$ integral equals
 $K \delta(w - y_K - x_m)$. Hence,

$$f_W(w) = 0.04 \delta(w-6) + 0.08 \delta(w-7) + 0.16 \delta(w-8) \\
+ 0.12 \delta(w-9) + 0.05 \delta(w-7) + 0.10 \delta(w-8) \\
+ 0.20 \delta(w-9) + 0.15 \delta(w-10) + 0.01 \delta(w-8) \\
+ 0.02 \delta(w-9) + 0.04 \delta(w-10) + 0.03 \delta(w-11).$$

$$f_W(w) = 0.04 \delta(w-6) + 0.13 \delta(w-7) + 0.27 \delta(w-8) \\
+ 0.34 \delta(w-9) + 0.19 \delta(w-10) + 0.03 \delta(w-11)$$





Q.4.6.7. Three statistically independent random variables X_1, X_2 and X_3 all have the same density fnn,

$$f_{X_i}(x_i) = \frac{1}{a} [u(x_i) - u(x_i - a)] \quad i=1,2,3.$$

Find and sketch the density fnn of $Y = X_1 + X_2 + X_3$ if $a > 0$ is constant.

soln

$$W_1 = X_1 + X_2$$

$$\text{then } f_{W_1}(w) = \int_{-\infty}^{\infty} f_{X_1}(x) \cdot f_{X_2}(w_1 - x) dx$$

$$= \int_{-\infty}^{\infty} \frac{1}{a^2} [u(x) - u(x-a)] [u(w_1 - x) - u(w_1 - x - a)] dx$$

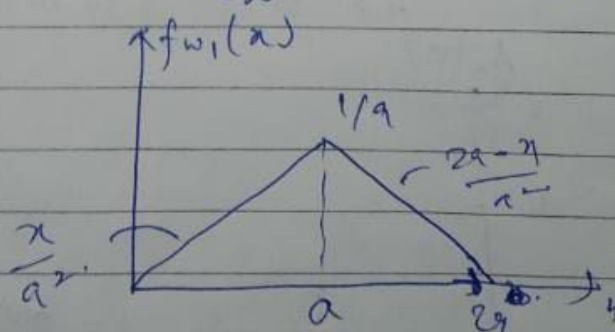
$$f_{W_1}(w_1) = 0, \quad w_1 \leq 0$$

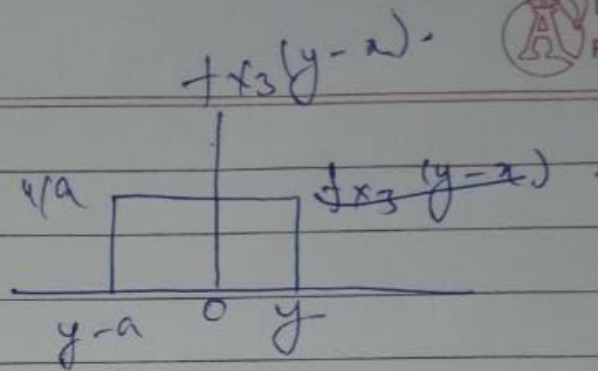
$$= w_1 / a^2, \quad 0 < w_1 \leq a.$$

$$= (2a - w_1) / a^2, \quad a < w_1 \leq 2a.$$

$$= 0, \quad 2a < w_1$$

$$\text{let } Y = W_1 + X_3, \quad f_Y(y) = \int_{-\infty}^{\infty} f_{W_1}(x) f_{X_3}(y-x) dx$$





$$f_Y(y) = 0, y < 0.$$

$$f_Y(y) = \int_0^y \frac{x}{a^3} dx = y^2 / 2a^3, 0 < y < a.$$

$$f_Y(y) = \int_{y-a}^a \frac{x}{a^3} dx + \int_a^y \frac{(2a-x)}{a^3} dx.$$

$$= \frac{-2y^2 + 6ay - 3a^2}{2a^3}$$

$$f_Y(y) = \int_{y-a}^{2a} \frac{(2a-x)}{a^3} dx.$$

$$= \frac{(y-3a)^2}{2a^3}, 2a < y < 3a.$$

$$f_Y(y) = 0, 3a < y$$

