

Answer ALL Questions

Part A - (10 x 2 = 20 Marks)

1. The CDF of a continuous random variable is given by  $F(x) = \begin{cases} 0, & x < 0 \\ 1 - e^{-x/5}, & 0 \leq x < \infty \end{cases}$   
Find the PDF and mean of  $X$ .

The pdf of  $X$  is  $f(x) = \frac{d}{dx} F(x)$

$$\therefore f(x) = \begin{cases} 0, & x < 0 \\ \frac{1}{5} e^{-x/5}, & 0 \leq x < \infty. \end{cases}$$

$$\frac{d}{dx} (1 - e^{-x/5}) \\ = 0 - e^{-x/5} \times (-1/5)$$

Mean =  $E[X] = \begin{cases} \sum_i x_i p_i & \text{if } X \text{ is discrete RV} \\ \int_{-\infty}^{\infty} x f(x) dx & \text{if } X \text{ is continuous RV.} \end{cases}$

$$\therefore \text{Mean} = \int_0^{\infty} \frac{1}{5} x e^{-x/5} dx = \frac{1}{5} \left[ x \left( \frac{e^{-x/5}}{-1/5} \right) - (1) \left( \frac{e^{-x/5}}{(-1/5)^2} \right) \right]_0^{\infty} = \frac{1}{5} [0 - 0] - \left[ -\frac{1}{5} \right] \\ = \frac{1}{5} [25 + 5] = \frac{30}{5} = 6$$

2. Establish the memoryless property of the exponential distribution.

$$P(X > s+t / X > t) = P(X > s), \text{ for any } s, t > 0.$$

The pdf of  $X$  is  $f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0 \\ 0, & x < 0 \end{cases}$

$$\therefore P(X > k) = \int_k^{\infty} \lambda e^{-\lambda x} dx = \lambda \left[ \frac{e^{-\lambda x}}{-\lambda} \right]_k^{\infty} = \lambda \left[ 0 + \frac{e^{-\lambda k}}{\lambda} \right] = e^{-\lambda k}.$$

$$P(X > (s+t) / X > t) = \frac{P(X > s+t \cap X > t)}{P(X > t)} = \frac{P(X > s+t)}{P(X > t)} = \frac{e^{-\lambda(s+t)}}{e^{-\lambda t}} = \frac{e^{-\lambda s} \cdot e^{-\lambda t}}{e^{-\lambda t}} \\ = e^{-\lambda s} = P(X > s)$$

$\therefore P(X > s+t / X > t) = P(X > s) \therefore$  Exponential distribution possesses memoryless property.

3. Let  $X$  and  $Y$  be continuous random variables with joint probability density function  $f_{xy}(x,y) = \frac{x(x-y)}{8}$ ,  $0 < x < 2$ ,  $-x < y < x$  and  $f_{xy}(x,y) = 0$  elsewhere. Find  $f_{y/x}(y/x)$ .

$$f_{y/x}(y/x) = \frac{f(x,y)}{f(x)}, \quad f(x) = \text{Marginal density function of } x = \int_{-x}^x f(x,y) dy$$

$$f(x) = \int_{-x}^x \frac{x^2 - xy}{8} dy = \frac{1}{8} \left[ x^2 y - \frac{xy^2}{2} \right]_{-x}^x = \frac{1}{8} \left[ x^2(x+x) - \frac{1}{2} \left( \frac{x^2}{2} - x^2 \right) \right] = \frac{2x^3}{8} = \frac{x^3}{4}$$

$$\therefore f_{y/x}(y/x) = \frac{\frac{x(x-y)}{8}}{\frac{x^3}{4}} = \frac{x(x-y)}{8} \times \frac{4}{x^3} = \frac{x-y}{2x^2} //, \quad 0 < x < 2, \quad -x < y < x.$$

4. Find the acute angle between the two lines of regression, assuming the two lines of regression.

The angle between the regression lines is given by  $\tan \alpha = \frac{1 - r^2}{r} \cdot \frac{\sigma_x \sigma_y}{\sigma_x^2 + \sigma_y^2}$

5. Prove that a first order stationary process has a constant mean.

The first order density of a stationary process  $\{X(t)\}$  is independent of time  $t$  i.e.  $f(x, t+c) = f(x, t)$

$$\Rightarrow E[X(t+c)] = \mu = \text{a constant.}$$

i.e.  $E[X(t)]$  is a constant  $\Rightarrow$  a first order stationary R.P.

$\therefore$  A first order stationary process has a constant mean.

6. State the postulates of a Poisson process.

If  $X(t)$  represents the number of occurrences of certain event in  $(0, t)$  then the discrete random process  $\{X(t)\}$  is called the Poisson process, provided the following postulates are satisfied.

i)  $P[0 \text{ occurrence in } (t, t+\Delta t)] = 1 - \lambda \Delta t$

ii)  $P[1 \text{ occurrence in } (t, t+\Delta t)] = \lambda \Delta t$

iii)  $P[2 \text{ or more occurrences in } (t, t+\Delta t)] = 0$ .

iv)  $X(t)$  is independent of the number of occurrences of the event in any interval prior and after the interval  $(0, t)$ .

v) The probability that the event occurs a specified number of times in  $(t_0, t_0+t)$  depends only on  $t$ , but not on  $t_0$ .

7. The autocorrelation function of a stationary random process is  $R(\tau) = 16 + \frac{9}{1+16\tau^2}$ . Find the mean and variance of the process.

By properties of Auto correlation function,

$$\bar{x}^2 = \lim_{\tau \rightarrow \infty} R_{xx}(\tau) = \lim_{\tau \rightarrow \infty} \left( 16 + \frac{9}{1+16\tau^2} \right) = 16 \Rightarrow \bar{x} = 4. \Rightarrow \mu_1' = 4.$$

$$E[x^2(t)] = \lim_{\tau \rightarrow 0} R_{xx}(\tau) = \lim_{\tau \rightarrow 0} \left( 16 + \frac{9}{1+16\tau^2} \right) = 16 + \frac{9}{1} = 25 \Rightarrow \mu_2' = 25$$

$$\therefore \text{Mean} = E[x(t)] = \bar{x} = \mu = \mu_1' = 4.$$

$$\text{Variance} = \mu_2' - \mu_1'^2 = 25 - 4^2 = 25 - 16 = 9.$$

Prove that for a WSS process  $\{x(t)\}$ ,  $R_{xx}(t, t+\tau)$  is an even function of  $\tau$ .

$$\text{By definition, } R(\tau) = E[x(t)x(t+\tau)]$$

$$\text{Put } t+\tau = t_1 \Rightarrow t = t_1 - \tau.$$

$$\begin{aligned} \therefore R(\tau) &= E[x(t_1 - \tau)x(t_1)] \\ &= E[x(t_1) \cdot x(t_1 + \tau)] \end{aligned}$$

$$R(\tau) = R(-\tau)$$

$\therefore$  The autocorrelation function is an even function of  $\tau$ .

1. Find the system Transfer function, if a Linear Time Invariant system has an impulse function  $H(t) = \begin{cases} \frac{1}{2c}, & |t| \leq c \\ 0, & |t| > c. \end{cases}$

The system transfer function  $H(\omega) = F[H(t)] = F\left[\frac{1}{2c}\right]$

$$H(\omega) = \int_{-\infty}^{\infty} H(t)e^{j\omega t} dt = \int_{-c}^c \frac{1}{2c} dt = \frac{1}{2c} \int_{-c}^c dt = \frac{1}{2c} [t]_{-c}^c = \frac{1}{2c} (c - (-c))$$

$$\therefore H(\omega) = \frac{1}{2c}(2c) = 1.$$

10. Define white noise.

A noise process whose power spectral density is independent of the operating frequency is called white noise.

PART B

11. a) The probability density function of a random variable  $x$  is given by

$$f_x(x) = \begin{cases} x, & 0 < x < 1 \\ k(2-x), & 1 \leq x \leq 2 \\ 0, & \text{otherwise.} \end{cases}$$

- (1) Find the value of  $k$
- (2) Find  $p(0.2 < x < 1.2)$
- (3) What is  $p[0.5 < x < 1.5 / x > 1]$
- (4) Find the distribution function of  $f(x)$ .

(1) By total probability,  $\int_{-\infty}^{\infty} f(x) dx = 1$  i.e.  $\int_{-\infty}^0 0 dx + \int_0^1 x dx + \int_1^2 k(2-x) dx = 1$

$$\text{i.e. } \left( x^2/2 \right)_0^1 + k \left[ 2x - x^2/2 \right]_1^2 = 1$$
$$\frac{1}{2} + k \left[ (4 - \frac{1}{2}) - (2 - \frac{1}{2}) \right] = 1 \Rightarrow \frac{1}{2} + k \left[ 4 - 2 - \frac{1}{2} + \frac{1}{2} \right] = 1$$
$$\frac{1}{2} + 2k = 1 \Rightarrow 2k = 1 - \frac{1}{2} \Rightarrow 2k = \frac{1}{2} \Rightarrow \boxed{k = \frac{1}{4}}$$

(2)  $p(0.2 < x < 1.2)$

$$= \int_{0.2}^{1.2} f(x) dx = \int_{0.2}^1 x dx + \int_1^{1.2} \frac{1}{4}(2-x) dx$$
$$= \left[ x^2/2 \right]_{0.2}^1 + \frac{1}{4} \left[ 2x - x^2/2 \right]_1^{1.2}$$
$$= \left[ \frac{1}{2} - \frac{(0.2)^2}{2} \right] + \frac{1}{4} \left[ \left\{ 2(1.2) - \frac{(1.2)^2}{2} \right\} - \left\{ 2 - \frac{1}{2} \right\} \right]$$
$$= \frac{1}{2} - 0.02 - \frac{1}{4} \left[ 2.4 - 0.72 - 2 + \frac{1}{2} \right]$$
$$= \frac{1}{2} - 0.02 - 0.6 + 0.18 + \frac{1}{2} - \frac{1}{8} = \frac{1 - 0.62 + 0.18 - \frac{1}{8}}{0.56 - 0.125}$$

$p(0.2 < x < 1.2) = 0.435$

(3)  $p(0.5 < x < 1.5 / x > 1) = \frac{p[0.5 < x < 1.5 \cap x > 1]}{p[x > 1]}$

$$= \frac{p(1 < x < 1.5)}{p(x > 1)}$$

$$P(1 < x < 1.5) = \int_1^{1.5} \frac{1}{4} (2-x) dx = \frac{1}{4} \left[ 2x - \frac{x^2}{2} \right]_1^{1.5}$$

$$= \frac{1}{4} \left[ 2(1.5-1) - \frac{1}{2} (1.5^2 - 1^2) \right] = \frac{1}{4} \left[ 2(0.5) - \frac{1}{2} (1.25) \right]$$

$$P(1 < x < 1.5) = 0.09375$$

$$P(x > 1) = \int_1^{\infty} f(x) dx = \int_1^2 \frac{1}{4} (2-x) dx = \frac{1}{4} \left[ 2x - \frac{x^2}{2} \right]_1^2$$

$$= \frac{1}{4} \left[ 2(2-1) - \frac{1}{2} (4-1) \right] = \frac{1}{4} \left[ 2 - \frac{3}{2} \right] = 0.125$$

$$\therefore P(0.5 < x < 1.5 | x > 1) = \frac{0.09375}{0.125} = 0.75$$

(4) The distribution function of  $X = F(x) = \int_{-\infty}^x f(x) dx$ .

When  $0 < x < 1$ ,  $F(x) = \int_{-\infty}^x f(x) dx = \int_{-\infty}^0 f(x) dx + \int_0^x f(x) dx = 0 + \int_0^x x dx = \frac{x^2}{2}$

When  $1 \leq x < 2$ ,  $F(x) = \int_{-\infty}^x f(x) dx = \int_{-\infty}^0 f(x) dx + \int_0^1 f(x) dx + \int_1^x f(x) dx$ .

$$F(x) = 0 + \int_0^1 x dx + \int_1^x \frac{1}{4} (2-x) dx = \left( \frac{x^2}{2} \right)_0^1 + \frac{1}{4} \left[ 2x - \frac{x^2}{2} \right]_1^x = \frac{1}{2} + \frac{1}{4} \left[ (2x - \frac{x^2}{2}) - (2 - \frac{1}{2}) \right]$$

$$F(x) = \frac{1}{2} + \frac{1}{4} (2x - \frac{x^2}{2}) - \frac{1}{4} (2 - \frac{1}{2}) = \frac{1}{2} + \frac{1}{4} (2x - \frac{x^2}{2}) - \frac{1}{4} + \frac{1}{8}$$

$$F(x) = \frac{1}{4} (2x - \frac{x^2}{2}) + \frac{1}{8}$$

When  $x > 2$ ,  $F(x) = \int_{-\infty}^x f(x) dx = \int_{-\infty}^0 f(x) dx + \int_0^1 f(x) dx + \int_1^2 \frac{1}{4} (2-x) dx + \int_2^{\infty} f(x) dx$ .

$$F(x) = 0 + \int_0^1 x dx + \int_1^2 (2-x) dx + 0 = 1.$$

$$\therefore F(x) = \begin{cases} \frac{x^2}{2}, & 0 < x \leq 1 \\ \frac{1}{4} (2x - \frac{x^2}{2}) + \frac{1}{8}, & 1 \leq x < 2, \\ 0, & x > 2. \end{cases}$$

1) b) i) Derive the m.g.f of Poisson distribution and hence or otherwise deduce its mean and variance.

The probability mass function of Poisson distribution is

$$P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!}, x=0,1,2,\dots$$

The moment generating function of the Poisson distribution is

$$M_X(t) = E[e^{tx}] = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-\lambda} \lambda^x}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{(e^t \lambda)^x}{x!} = e^{-\lambda} e^{\lambda e^t} = e^{\lambda(e^t-1)}$$

To find mean and variance using MGF

$$\text{Mean} = E(X) = \mu_1' = \left[ \frac{d}{dt} M_X(t) \right]_{t=0} = \left[ \frac{d}{dt} (e^{\lambda(e^t-1)}) \right]_{t=0}$$

$$\mu_1' = \left[ e^{-\lambda} e^{\lambda e^t} \cdot \lambda e^t \right]_{t=0} = e^{-\lambda} e^{\lambda} \cdot \lambda = \lambda, \text{ i.e. } \boxed{E(X) = \lambda}$$

$$\mu_2' = E(X^2) = \left[ \frac{d^2}{dt^2} M_X(t) \right]_{t=0} = \left[ \frac{d^2}{dt^2} (e^{-\lambda} e^{\lambda e^t}) \right]_{t=0}$$

$$= \left[ e^{-\lambda} e^{\lambda e^t} (\lambda e^t)^2 + e^{-\lambda} e^{\lambda e^t} \lambda e^t \right]_{t=0} = e^{-\lambda} e^{\lambda} \lambda^2 + e^{-\lambda} e^{\lambda} \lambda = \lambda^2 + \lambda$$

$$\text{Variance} = E(X^2) - [E(X)]^2 = \lambda + \lambda^2 - \lambda^2 = \lambda \Rightarrow \boxed{\text{Var}(X) = \lambda}$$

ii) The marks obtained by a number of students in a certain subject are assumed to be normally distributed with mean 65 and standard deviation 5. If 3 students are selected at random from this set, what is the probability that exactly 2 of them will have marks over 70?

$$\text{Given } \mu = 65, \sigma = 5 \quad z = \frac{x - \mu}{\sigma} = \frac{x - 65}{\sigma = 5}$$

$$\text{when } x = 70, z = \frac{70 - 65}{5} \Rightarrow z = 1.$$

$$\therefore P(X > 70) = P(Z > 1) = 0.5 - P(0 < Z < 1) = 0.5 - 0.3413 = \underbrace{0.1587}_P$$

$$\therefore q = 1 - p = 0.8413$$

$\therefore P(\text{No. of students i.e. 2 of them scoring more than 70})$

$$= P(Y=2) = n C_y p^y q^{n-y}, n=3.$$

$$P(Y=2) = 3 C_2 p^2 q^1 = 3 \times (0.1587)^2 (0.8413) = \underline{\underline{0.06358}}$$

(2) (i) If  $x$  and  $y$  are independent poisson random variables with respective parameters  $\lambda_1$  and  $\lambda_2$ . Calculate the conditional distribution of  $x$ , given that  $x+y=n$ .

Given  $x$  and  $y$  are independently and poisson distributed random variables.

$$P(X=r / X+Y=n) = \frac{P(X=r \cap X+Y=n)}{P(X+Y=n)} = \frac{P(X=r) \cdot P(X+Y=n-X)}{P(X+Y=n)} \quad (\because X=r)$$

$$= \frac{P(X=r \cap Y=n-r)}{P(X+Y=n)} = \frac{P(X=r) \cdot P(Y=n-r)}{P(X+Y=n)}$$

$$= \frac{\left( \frac{e^{-\lambda_1} \lambda_1^r}{r!} \right) \left( \frac{e^{-\lambda_2} \lambda_2^{n-r}}{(n-r)!} \right)}{e^{-(\lambda_1+\lambda_2)} (\lambda_1+\lambda_2)^n} = \frac{e^{-(\lambda_1+\lambda_2)} \lambda_1^r \lambda_2^{n-r} n!}{e^{-(\lambda_1+\lambda_2)} (\lambda_1+\lambda_2)^n r! (n-r)!}$$

$$= \frac{n!}{r! (n-r)!} \left( \frac{\lambda_1}{\lambda_1+\lambda_2} \right)^r \left( \frac{\lambda_2}{\lambda_1+\lambda_2} \right)^{n-r} \quad \left( \begin{array}{l} \text{since } X+Y \text{ is a poisson} \\ \text{variate with parameter} \\ \lambda_1+\lambda_2 \end{array} \right)$$

$= {}^n C_r p^r q^{n-r}$  which is a pmf of a binomial distribution.

where  $p = \frac{\lambda_1}{\lambda_1+\lambda_2}$ ,  $q = \frac{\lambda_2}{\lambda_1+\lambda_2}$  and  ${}^n C_r = \frac{n!}{r! (n-r)!}$ .

(ii) The regression equation of  $x$  on  $y$  is  $3y - 5x + 108 = 0$ . If the mean value of  $y$  is 44 and the variance of  $x$  is  $\frac{9}{16}$ th of the variance of  $y$ . Find the mean value of  $x$  and the correlation coefficient.

Given the line of regression of  $x$  on  $y$  is  $5x = 3y + 108$

$$\text{i.e. } x = \frac{3}{5}y + \frac{108}{5}$$

i.e.  $x = 0.6y + 21.6 \Rightarrow b_{yx} = 0.6$ .

Also given  $\bar{y} = 44$  and  $\text{Var}(x) = \frac{9}{16} \text{Var}(y)$ .

$$\therefore b_{yx} = r \frac{\sigma_y}{\sigma_x} = r \frac{\sigma_y}{\sqrt{\frac{9}{16} \sigma_y^2}} = r \frac{4\sigma_y}{3\sigma_y}$$

$$\Rightarrow b_{yx} = 4r \Rightarrow r = \frac{b_{yx}}{4}$$

$$\rho = \sqrt{b_{xy} \times b_{yx}}$$

$$\frac{b_{yx}}{4} = \sqrt{0.6 \times b_{yx}}$$

$$\frac{(b_{yx})^2}{16} = 0.6 b_{yx}$$

$$b_{yx} = 16 \times 0.6 \Rightarrow \boxed{b_{yx} = 9.6}$$

$$\therefore \rho = \sqrt{9.6 \times 0.6} =$$

12. b) (i) If  $x$  and  $y$  are independent random variables with density function  $f_x(x) = \begin{cases} 1, & 1 \leq x \leq 2 \\ 0, & \text{otherwise} \end{cases}$  and  $f_y(y) = \begin{cases} y/6, & 2 \leq y \leq 4 \\ 0, & \text{otherwise} \end{cases}$ , find the density function of  $z = xy$ .

Given  $x$  and  $y$  are independent random variables

$$\therefore f_{x,y} = f_x(x) \cdot f_y(y) = \begin{cases} 1 \cdot y/6, & 1 \leq x \leq 2, 2 \leq y \leq 4 \\ 0, & \text{otherwise} \end{cases}$$

$$= \begin{cases} \frac{y}{6}, & 2 \leq y \leq 4, 1 \leq x \leq 2 \\ 0, & \text{otherwise} \end{cases}$$

Given  $z = xy$  let  $w = y$

$$x = \frac{z}{y} = \frac{z}{w} \text{ and } y = w$$

$$J = \begin{vmatrix} \partial x / \partial z & \partial x / \partial w \\ \partial y / \partial z & \partial y / \partial w \end{vmatrix} = \begin{vmatrix} \frac{1}{w} & -\frac{z}{w^2} \\ 0 & 1 \end{vmatrix} = \frac{1}{w}$$

The joint PDF of  $z$  and  $w$  is given by

$$f_{z,w}(z,w) = f_{x,y}\left(\frac{z}{w}, w\right) |J| = \frac{y}{6} \cdot \frac{1}{w} = \begin{cases} \frac{z}{6} \cdot \frac{1}{w}, & 1 \leq \frac{z}{w} \leq 2 \\ & \text{and } 2 \leq w \leq 4 \\ 0, & \text{other} \end{cases}$$



$$f(z, w) = \begin{cases} \frac{1}{6}, & 2 \leq w \leq 4 \text{ \& } 1 \leq z \leq 2w \\ 0, & \text{otherwise} \end{cases}$$

Range:

$$2 \leq y \leq 4 \Rightarrow 2 \leq w \leq 4$$

$$1 \leq x \leq 2 \Rightarrow 1 \leq \frac{z}{w} \leq 2 \Rightarrow 1 \leq z \leq 2w$$

b) ii) The life time of a particular variety of electric bulb may be considered as a random variable with mean 1200 hours and standard deviation 250 hours. Using central limit theorem, find the probability that the average life time of 60 bulbs exceeds 1250 hours. Let  $\bar{x}$  denote the mean life-time of 60 bulbs.

Given  $n=60$ , and  $\mu=1200$ ,  $\sigma=250$ .

$\bar{x}$  follows normal distribution with mean  $\mu$  and variance  $\frac{\sigma^2}{n}$ .

To find the probability that average life-time of 60 bulbs exceeds 1250 hours:

$$P(\bar{x} \geq 1250) = P\left(\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \geq \frac{1250 - \mu}{\sigma/\sqrt{n}}\right) = P\left(z \geq \frac{1250 - 1200}{250/\sqrt{60}}\right)$$

$$= P(z \geq 1.55) = P(0 \leq z \leq \infty) - P(0 \leq z \leq 1.55)$$

$$= 0.5 - 0.4394$$

$$P(\bar{x} \geq 1250) = \underline{\underline{0.0606}}$$

13) a) i) A random process  $X(t)$  defined by  $X(t) = A \cos t + B \sin t$ ,  $-\infty < t < \infty$ , where  $A$  and  $B$  are independent random variables each of which takes a value  $-2$  with probability  $1/3$  and a value  $1$  with probability  $2/3$ . Show that  $X(t)$  is wide-sense stationary.

$\{X(t)\}$  is WSS if  $E[X(t)] = \text{constant}$  and  $R_{XX}(t, t+\tau)$  depends only on  $\tau$ .

Since  $A$  and  $B$  are discrete R.V.s which assumes values  $-2$  and  $1$  with probability  $\frac{1}{3}$  respectively.

$$\therefore \text{Mean } E(A) = \sum a_i p(a_i) = (-2)\left(\frac{1}{3}\right) + 1\left(\frac{2}{3}\right) = 0 = E(B).$$

$$E(A^2) = \sum a_i^2 p(a_i) = (-2)^2\left(\frac{1}{3}\right) + (1)^2\left(\frac{2}{3}\right) = \frac{4}{3} + \frac{2}{3} = 2 = E(B^2)$$

Also given  $A$  and  $B$  are independent R.V.s.

$$\therefore E(AB) = E(A) \cdot E(B) = 0.$$

Since  $x(t) = A \cos t + B \sin t$

$$\begin{aligned} \text{(i)} \quad E[x(t)] &= E[A \cos t] + E[B \sin t] \\ &= \cos t E(A) + \sin t E(B) = \cos t (0) + \sin t (0) \\ &= 0, \text{ a constant.} \end{aligned}$$

$$\begin{aligned} \text{(ii)} \quad \text{By definition, } R_{xx}(t, t+\tau) &= E[x(t) \cdot x(t+\tau)] = E[(A \cos t + B \sin t) \cdot \\ &\quad (A \cos(t+\tau) + B \sin(t+\tau))] \\ &= E[A^2 \cos t \cos(t+\tau)] + E[AB \cos t \sin(t+\tau) + B \sin t \cos(t+\tau)] \\ &\quad + E[B^2 \sin t \sin(t+\tau)] \\ &= E(A^2) \cos t \cos(t+\tau) + E(AB) \sin(2t+\tau) + E(B^2) \sin t \sin(t+\tau) \\ &= 2 \cos t \cos(t+\tau) + 0 + 2 \sin t \sin(t+\tau) \quad (\because E(AB) = E(A)E(B) = 0) \\ &= 2 [\cos t \cos(t+\tau) + \sin t \sin(t+\tau)] \\ &= 2 \cos(t+\tau-t) \end{aligned}$$

$R(\tau) = 2 \cos \tau$ , which depends on  $\tau$  alone.

$\therefore \{x(t)\}$  is WSS process.

13) a) i) A random process has sample functions of the form  $x(t) = A \cos(\omega t + \theta)$ , where  $\omega$  is constant,  $A$  is a random variable with mean zero and variance 1 and  $\theta$  is a random variable that is uniformly distributed between 0 and  $2\pi$ . Assume that random variables  $A$  and  $\theta$  are independent. Is  $x(t)$  is mean-ergodic process?



13) b) (i) If  $\{x(t)\}$  is a Gaussian process with  $\mu(t) = 10$  and  $C(t_1, t_2) = 16e^{-|t_1 - t_2|}$   
 find the probability that

(1)  $x(10) \leq 8$

(2)  $|x(10) - x(6)| \leq 4$ .

$\{x(t)\}$  is a Normal process with parameter with mean = 10

WKT  $C(t_1, t_2) = R_{xx}(t_1, t_2) - E[x(t_1)]E[x(t_2)]$

$\therefore C(t, t) = R_{xx}(t, t) - \{E[x(t)]\}^2 = E[x^2(t)] - (E[x(t)])^2$   
 $= \text{Variance}(x(t)) = 16e^{-|t-t|} = 16 \Rightarrow \sigma = 4$ .

Case (i)  $P[x(10) \leq 8] = P\left[\frac{x(10) - \text{mean}}{\sigma} \leq \frac{8 - 10}{4}\right] = P\left[z \leq -\frac{0.5}{1}\right]$   
 $= 0.5 - P(0 \leq z \leq 0.5) = 0.5 - 0.1915 = 0.3085$

Case (ii) Let  $X = x(10) - x(6) \therefore E[X(t)] = E[x(10)] - x(6) = 10 - 10 = 0$

$C(t_1, t_2) = \text{Var}[x(t_1)] + \text{Var}[x(t_2)] - 2\text{Cov}(x(t_1), x(t_2))$

$C(t, t) = 16 + 16 - 2\text{Cov}(t, t) = 16 + 16 - 2(16e^{-|10-6|}) = 31.4$

$\sigma = 5.6 \therefore P[|x(10) - x(6)| \leq 4] = P[|X(t)| \leq 4] = P(-4 \leq X \leq 4)$   
 $= P(-0.714 \leq z \leq 0.714) = 2P(0 < z < 0.714) = 0.522$ .

b) (ii) Prove that the interval between two successive occurrences of a Poisson process with parameter  $\lambda$  has an exponential distribution with mean  $\frac{1}{\lambda}$ .

Let  $x(t)$  denote the number of occurrences in a time interval  $t$ .  
 Let  $T$  be the interval between two successive occurrences  $E_i$  and  $E_{i+1}$ .  
 $T$  is a continuous random variable.

Let the event  $E_i$  occurs at the time instant  $t_i$ .

$P(T > t) = P[E_{i+1} \text{ did not occur in } (t_i, t_i + t)]$

$= P[\text{no event occurs in } (t_i, t_i + t)]$

$= P[x(t) = 0] = \frac{e^{-\lambda t} (\lambda t)^0}{0!} = e^{-\lambda t}$

$\Rightarrow P(T > t) = e^{-\lambda t}$

$\therefore$  Cumulative distribution function  $F(t) = P(T \leq t) = 1 - P(T > t) = 1 - e^{-\lambda t}$ .  
 Probability density function  $f(t) = \frac{d}{dt} F(t)$ .

r.e  $f(t) = F'(t) = 0 - (-\lambda)e^{-\lambda t} = \lambda e^{-\lambda t}$  which is the PDF of exponential function.

For exponential distribution, mean =  $\frac{1}{\lambda}$ .

Hence proved.

14) a) i) The power spectral density function of a zero mean WSS process  $x(t)$  is given by  $S(\omega) = \begin{cases} 1, & |\omega| < \omega_0 \\ 0, & \text{otherwise} \end{cases}$ . Find  $R(\tau)$  and show that  $x(t)$  and  $x(t + \frac{\pi}{\omega_0})$  are uncorrelated.

Given  $E[x(t)] = 0$ . By definition  $R(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S(\omega) e^{j\omega\tau} d\omega = \frac{1}{2\pi} \int_{-\omega_0}^{\omega_0} e^{j\omega\tau} d\omega$

$$R(\tau) = \frac{1}{2\pi} \int_{-\omega_0}^{\omega_0} (\cos\omega\tau + j\sin\omega\tau) d\omega = \frac{2}{2\pi} \int_0^{\omega_0} \cos\omega\tau d\omega + 0.$$

$$= \frac{1}{\pi} \left[ \frac{\sin\omega\tau}{\tau} \right]_0^{\omega_0} = \frac{1}{\pi} \left[ \frac{\sin\omega_0\tau}{\tau} - 0 \right]$$

$$\therefore R(\tau) = \frac{\sin\omega_0\tau}{\pi\tau}$$

$$C[x(t)x(t + \frac{\pi}{\omega_0})] = E[x(t)x(t + \frac{\pi}{\omega_0})] - E[x(t)]E[x(t + \frac{\pi}{\omega_0})] \\ = R_{xx}(\frac{\pi}{\omega_0}) - 0 = R_{xx}(\frac{\pi}{\omega_0}) \quad \text{since } E[x(t)] = 0.$$

$$\text{But } R_{xx}(\tau) = \frac{\sin\omega_0\tau}{\pi\tau}$$

$$\therefore R_{xx}(\frac{\pi}{\omega_0}) = \frac{\sin\omega_0(\frac{\pi}{\omega_0})}{\pi(\frac{\pi}{\omega_0})} = \frac{\sin\pi}{\pi^2/\omega_0} = 0.$$

$$\therefore C[x(t)x(t + \frac{\pi}{\omega_0})] = 0.$$

$\therefore x(t)$  and  $x(t + \frac{\pi}{\omega_0})$  are uncorrelated.

ii) The ACF of a WSS process is given by  $R(\tau) = \alpha^2 e^{-2\lambda|\tau|}$  determine the PSD of the process.

$$S(\omega) = \int_{-\infty}^{\infty} R(\tau) e^{-j\omega\tau} d\tau = \int_{-\infty}^{\infty} \alpha^2 e^{-2\lambda|\tau|} e^{-j\omega\tau} d\tau$$

$$= \alpha^2 F[e^{-2\lambda|\tau|}]$$

$$= \alpha^2 \frac{2(2\lambda)}{4\lambda^2 + \omega^2} = \frac{4\alpha^2\lambda}{4\lambda^2 + \omega^2}.$$

14) b) i) State and prove Wiener Khintchine theorem.

If  $X_T(\omega)$  is the Fourier transform of the truncated random process defined as  $X_T(t) = \begin{cases} X(t), & -T \leq t \leq T \\ 0, & \text{otherwise} \end{cases}$  where  $X(t)$  is a real WSS process with PSD function  $S_{XX}(\omega)$  then  $S_{XX}(\omega) = \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} E \{ |X_T(\omega)|^2 \} \right]$

Given that  $X_T(\omega) = F[X_T(t)] = \int_{-\infty}^{\infty} X_T(t) e^{-j\omega t} dt = \int_{-T}^T X(t) e^{-j\omega t} dt$

$$|X_T(\omega)|^2 = X_T(\omega) \overline{X_T(\omega)} = X_T(\omega) X_T(-\omega) = \int_{-T}^T X(t_1) e^{-j\omega t_1} dt_1 \int_{-T}^T X(t_2) e^{+j\omega t_2} dt_2$$

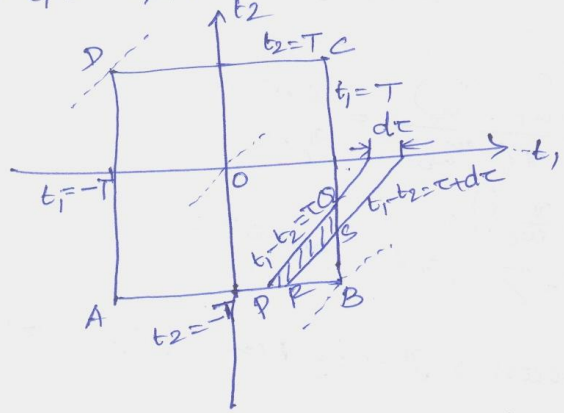
$$= \int_{-T}^T \int_{-T}^T X(t_1) X(t_2) e^{-j\omega(t_1 - t_2)} dt_1 dt_2$$

$$\therefore E \{ |X_T(\omega)|^2 \} = \int_{-T}^T \int_{-T}^T E[X(t_1) X(t_2)] e^{-j\omega(t_1 - t_2)} dt_1 dt_2$$

$$= \int_{-T}^T \int_{-T}^T R_{XX}(t_1 - t_2) e^{-j\omega(t_1 - t_2)} dt_1 dt_2 \quad \text{Since } R_{XX}(t_1, t_2) = E[X(t_1) X(t_2)]$$

$$= \int_{-T}^T \int_{-T}^T \phi(t_1 - t_2) dt_1 dt_2 \quad \text{where } \phi(t_1 - t_2) = R_{XX}(t_1 - t_2) e^{-j\omega(t_1 - t_2)}$$

We evaluate the double integral over the area of the square ABCD bounded by  $t_1 = -T, +T$  and  $t_2 = -T, +T$  as shown in the figure.



Let us divide the area of the square into a number of strips like PQRS, PQ is given by  $t_1 - t_2 = \tau$  and RS given by  $t_1 - t_2 = \tau + d\tau$ .

When PQRS is at the initial position D,  $t_1 - t_2 = -2T$ . When PQRS is at the final position B,  $t_1 - t_2 = 2T$ . Hence when  $\tau$  varies from  $-2T$  to  $2T$ , the area ABCD is covered.

Now  $dt_1 dt_2 =$  elemental area of the  $t_1 t_2$  plane = area of PQRS.  
 when  $\tau > 0$ , Area of PQRS = Area of triangle PBCQ - Area of  $\Delta$  PBS  
 $= \frac{1}{2} (2T - \tau)^2 - \frac{1}{2} (2T - \tau - d\tau)^2$   
 $= (2T - \tau) d\tau.$

$$BQ = PB = T - (\tau - T) = \begin{cases} 2T - \tau, & \tau \geq 0 \\ 2T + \tau, & \tau < 0. \end{cases}$$

For all  $\tau$  we have Area of PQRS =  $dt_1 dt_2 = (2T - |\tau|) d\tau.$

$$\therefore E[|X_T(\omega)|^2] = \int_{-2T}^{2T} \phi(\tau) (2T - |\tau|) d\tau$$

$$\therefore \frac{1}{2T} E[|X_T(\omega)|^2] = \int_{-2T}^{2T} \phi(\tau) \left(1 - \frac{|\tau|}{2T}\right) d\tau$$

$$\begin{aligned} \lim_{T \rightarrow \infty} \frac{1}{2T} E[|X_T(\omega)|^2] &= \lim_{T \rightarrow \infty} \int_{-2T}^{2T} \phi(\tau) d\tau - \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-2T}^{2T} |\tau| \phi(\tau) d\tau \\ &= \int_{-\infty}^{\infty} \phi(\tau) d\tau, \text{ assuming } \int_{-\infty}^{\infty} \phi(\tau) |\tau| d\tau \text{ is bounded} \\ &= \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-j\omega\tau} d\tau \end{aligned}$$

$$\therefore \lim_{T \rightarrow \infty} E[|X_T(\omega)|^2] = S_{XX}(\omega). \text{ By the definition of PSD}$$

\*) b) i) The cross-power spectrum of real random process  $\{x(t)\}$  and  $\{y(t)\}$  given by  $S_{xy}(\omega) = \begin{cases} a + bj\omega, & |\omega| < 1 \\ 0, & \text{elsewhere} \end{cases}$  find the cross correlation function.

$$R_{xy}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xy}(\omega) e^{j\omega\tau} d\omega = \int_{-1}^1 (a + jb\omega) e^{j\omega\tau} d\omega$$

$$= \frac{1}{2\pi} \int_{-1}^1 (a + jb\omega) \cos \omega\tau d\omega + \frac{j}{2\pi} \int_{-1}^1 (a + jb\omega) \sin \omega\tau d\omega$$

$$= \frac{1}{2\pi} \int_{-1}^1 a \cos \omega\tau d\omega + \frac{j}{2\pi} b \int_{-1}^1 \omega \cos \omega\tau d\omega + \frac{j}{2\pi} a \int_{-1}^1 \sin \omega\tau d\omega + \frac{j^2 b}{2\pi} \int_{-1}^1 \omega \sin \omega\tau d\omega$$

$$\begin{aligned}
&= \frac{2}{2\pi} \int_0^1 a \cos \omega \tau \, d\omega + 0 + 0 - \frac{2b}{2\pi} \int_0^1 \omega \sin \omega \tau \, d\omega \\
&= \frac{1}{\pi} \left[ a \frac{\sin \omega \tau}{\tau} - b \left\{ \omega \left( -\frac{\cos \omega \tau}{\tau} \right) - 1 \left( -\frac{\sin \omega \tau}{\tau^2} \right) \right\} \right] \\
&= \frac{1}{\pi} \left[ \frac{a \sin \tau}{\tau} + \frac{b \cos \tau}{\tau} - \frac{b \sin \tau}{\tau^2} \right]
\end{aligned}$$

$$R_{xy}(\tau) = \frac{1}{\pi \tau^2} [ (a\tau - b) \sin \tau + b \cos \tau ]$$

15a) i) Consider a system with transfer function  $\frac{1}{1+j\omega}$ . An input signal with auto correlation function  $m\delta(\tau) + m^2$  is fed as input to the system. Find the mean and mean-square value of the output.

Given system transfer function  $H(\omega) = \frac{1}{1+j\omega}$   
 Input auto correlation  $R_{xx}(\tau) = m\delta(\tau) + m^2$

$$\therefore S_{xx}(\omega) = F[m\delta(\tau) + m^2] = m F[\delta(\tau)] + F[m^2] = m + 2\pi m^2 \delta(\omega)$$

$$\begin{aligned}
\text{WKT } S_{yy}(\omega) &= |H(\omega)|^2 S_{xx}(\omega) = H(\omega) \cdot \overline{H(\omega)} \cdot S_{xx}(\omega) \\
&= \frac{1}{1+j\omega} \cdot \frac{1}{1-j\omega} [m + 2\pi m^2 \delta(\omega)] \\
S_{yy}(\omega) &= \frac{1}{1+\omega^2} [m + 2\pi m^2 \delta(\omega)] \\
&=
\end{aligned}$$

Mean of the output  $E[y(t)] = \bar{y} = H(0) \cdot \bar{x}$

$$\text{Now } \bar{x}^2 = \lim_{T \rightarrow \infty} \frac{1}{T} R_{xx}(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} [m\delta(\tau) + m^2]$$

$$\bar{x}^2 = m^2 \Rightarrow \boxed{\bar{x} = m} \text{ and } H(0) = \frac{1}{0+1} = 1 \text{ and.}$$

$$\bar{y} = H(0) \cdot \bar{x}$$

$$= m \cdot 1$$

$$\boxed{\bar{y} = m}$$

Mean square value of the output  $= R_{yy}(0)$

$$R_{yy}(\tau) = F^{-1}[S_{yy}(\omega)] = F^{-1} \left[ \frac{1}{1+\omega^2} (m + 2\pi m^2 \delta(\omega)) \right]$$



$$\begin{aligned}
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{1}{1+\omega^2} (m + 2\pi m^2 \delta(\omega)) e^{j\omega\tau} d\omega \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{k e^{j\omega\tau}}{1+\omega^2} d\omega + \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{2\pi m^2 \delta(\omega) e^{j\omega\tau}}{1+\omega^2} d\omega \\
&= \frac{k}{2\pi} \int_{-\infty}^{\infty} \frac{1}{1+\omega^2} e^{j\omega\tau} d\omega + k^2 \int_{-\infty}^{\infty} \frac{\delta(\omega)}{1+\omega^2} e^{j\omega\tau} d\omega \\
&= \frac{k}{2\pi} \pi e^{-|\tau|} + k^2
\end{aligned}$$

Hence  $R_{yy}(\tau) = k/2 e^{-|\tau|} + k^2$ .

Mean square value  $\bar{y}^2 = R_{xx}(0) = k/2 + k^2$ .

(ii) An LTI system has an impulse response  $h(t) = e^{-\beta t} u(t)$ . Find the output autocorrelation function  $R_{yy}(\tau)$  corresponding to an input  $x(t)$ .

We know that  $H(\omega) = F[h(t)] = \int_0^{\infty} e^{-\beta t} e^{-j\omega t} dt = \int_0^{\infty} e^{-(\beta + j\omega)t} dt$

$$= - \int \frac{e^{-(\beta + j\omega)t}}{\beta + j\omega} \Big|_0^{\infty} = \frac{1}{\beta + j\omega}$$

$$|H(\omega)|^2 = \frac{1}{\beta^2 + \omega^2}$$

The output power density spectrum  $S_{yy}(\omega) = |H(\omega)|^2 S_{xx}(\omega)$

$$S_{yy}(\omega) = \frac{1}{\beta^2 + \omega^2} S_{xx}(\omega) = \frac{1}{2\beta} \left( \frac{2\beta}{\beta^2 + \omega^2} \right) S_{xx}(\omega)$$

Taking inverse Fourier transform on both sides of the above expression,

$$R_{yy}(\tau) = \frac{1}{2\beta} F^{-1} \left[ \frac{2\beta}{\beta^2 + \omega^2} \right] * R_{xx}(\tau) = \frac{1}{2\beta} e^{-\beta/|\tau|} * R_{xx}(\tau) = \frac{1}{2\beta} \int_0^{\infty} e^{-\beta/|u|} R_{xx}(\tau - u) du$$

15) bi) A linear system is described by the impulse response  $h(t) = \frac{1}{RC} e^{-t/RC} u(t)$ . Assume an input process whose ACF is  $A\delta(\tau)$ . Find the mean and ACF of the output process.

$$\text{let } \beta = \frac{1}{RC}$$

$$\therefore h(t) = \left( \frac{1}{RC} e^{-t/RC} \right) u(t) = \beta e^{-\beta t} u(t)$$

$$\text{we know that } S_{yy}(\omega) = |H(\omega)|^2 S_{xx}(\omega)$$

$$\therefore |H(\omega)|^2 = H(\omega) \overline{H(\omega)}$$

$$H(\omega) = \int_{-\infty}^{\infty} h(t) e^{-j\omega t} dt = \int_{-\infty}^{\infty} \beta e^{-\beta t} u(t) e^{-j\omega t} dt$$

$$= \int_{-\infty}^0 0 dt + \int_0^{\infty} \beta e^{-\beta t} (1) e^{-j\omega t} dt, \text{ using the definition of } u(t) = \begin{cases} 1, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

$$= \beta \int_0^{\infty} e^{-(\beta + j\omega)t} dt = \beta \left[ \frac{e^{-(\beta + j\omega)t}}{-(\beta + j\omega)} \right]_0^{\infty}$$

$$= \beta \left[ e^{-\infty} + \frac{e^0}{\beta + j\omega} \right] = \frac{\beta}{\beta + j\omega}$$

$$\therefore H(\omega) = \frac{\beta}{\beta + j\omega} \quad \text{and } H^*(\omega) = \overline{H(\omega)} = \frac{\beta}{\beta - j\omega} \quad ; |H(\omega)|^2 = \frac{\beta^2}{\beta^2 + \omega^2}$$

$$\therefore S_{yy}(\omega) = |H(\omega)|^2 S_{xx}(\omega)$$

$$S_{yy}(\omega) = \frac{\beta^2}{\beta^2 + \omega^2} S_{xx}(\omega) \quad ; \quad S_{xx}(\omega) = F[R_{xx}(\tau)] = F[A\delta(\tau)] = A$$

$$\therefore S_{yy}(\omega) = \frac{\beta^2}{\beta^2 + \omega^2} A$$

$$\therefore R_{yy}(\tau) = F^{-1} \left[ \frac{A\beta^2}{\beta^2 + \omega^2} \right] = A\beta^2 F^{-1} \left[ \frac{1}{\omega^2 + \beta^2} \right] = A\beta^2 \frac{1}{2\beta} e^{-\beta|\tau|}$$

$$\therefore R_{yy}(\tau) = \underline{\underline{\frac{A\beta}{2} e^{-\beta|\tau|}}}$$

15) b) ii) If  $\{N(t)\}$  is a band-limited white noise such that

$$S_{NN}(\omega) = \begin{cases} \frac{N_0}{2}, & |\omega| < \omega_B \\ 0, & \text{elsewhere.} \end{cases} \quad \text{find the autocorrelation function.}$$

By definition, the PSD is given by

$$S_{NN}(\omega) = \begin{cases} N_0/2, & |\omega| < \omega_B \\ 0, & \text{elsewhere.} \end{cases}$$

The ACF is given by the inverse Fourier transform

$$\therefore R_{NN}(\tau) = \frac{1}{2\pi} \int_{-\omega_B}^{\omega_B} \frac{N_0}{2} e^{j\omega\tau} d\omega = \frac{N_0}{4\pi} \int_{-\omega_B}^{\omega_B} e^{j\omega\tau} d\omega$$

$$= \frac{N_0}{4\pi} \left[ \frac{e^{j\omega\tau}}{j\tau} \right]_{-\omega_B}^{\omega_B} = \frac{N_0}{4\pi} \left[ \frac{e^{j\omega_B\tau} - e^{-j\omega_B\tau}}{j\tau} \right]$$

$$= \frac{N_0}{2\pi\tau} \left[ \frac{e^{j\omega_B\tau} - e^{-j\omega_B\tau}}{2j} \right]$$

$$R_{NN}(\tau) = \frac{N_0}{2\pi} \left( \frac{\sin\omega_B\tau}{\tau} \right).$$