

1. The CDF of a continuous RV 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 .

When $x < 0$, $F(x) = 0$.

$$\text{When } 0 \leq x < \infty, F(x) = \int_0^x (1 - e^{-x/5}) dx = \left[x - \frac{e^{-x/5}}{-1/5} \right]_0^x = \left[x + 5e^{-x/5} - 5 \right],$$

$$\therefore F(x) = \begin{cases} (x-5) + 5e^{-x/5}, & 0 < x < \infty \\ 0, & \text{otherwise.} \end{cases}$$

2. If X is a normal random variable with mean zero and variance σ^2 , find the PDF of $Y = e^X$.

$$X \text{ is a normal random variable, } f_X(x) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad -\infty < x < \infty$$

$$\text{Given } y = e^x \Rightarrow x = \log y \Rightarrow \frac{dx}{dy} = \frac{1}{y} \Rightarrow \left| \frac{dx}{dy} \right| = \frac{1}{y}.$$

$$f_Y(y) = f_X(x) \left| \frac{dx}{dy} \right| = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \cdot \frac{1}{y}, \quad 0 < y < \infty, x = \log y.$$

3. If the joint pdf of (X, Y) is $f_{XY}(x, y) = \begin{cases} e^{-(x+y)}, & x > 0, y > 0 \\ 0, & \text{otherwise} \end{cases}$, check whether X and Y are independent.

$$\text{Given } f_{X,Y}(x, y) = e^{-(x+y)} = e^{-x} \cdot e^{-y}$$

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy = \int_0^{\infty} e^{-x} \cdot e^{-y} dy = e^{-x} \left(\frac{e^{-y}}{-1} \right)_0^{\infty} = e^{-x} [-e^{-\infty} + e^0] = e^{-x}, x > 0$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx = \int_0^{\infty} e^{-x} \cdot e^{-y} dx = e^{-y} \left(\frac{e^{-x}}{-1} \right)_0^{\infty} = e^{-y} [-e^{-\infty} + e^0] = e^{-y}, y > 0.$$

$$f_{X,Y}(x, y) = e^{-(x+y)} = e^{-x} \cdot e^{-y} = f_X(x) f_Y(y). \therefore X \text{ and } Y \text{ are independent.}$$

4. The regression equations are $3x + 2y = 26$ and $6x + y = 31$. Find the correlation coefficient between x and y .

$$\text{Given } 3x + 2y = 26$$

$$2y = 26 - 3x$$

$$y = \frac{26}{2} - \frac{3x}{2}$$

$$b_{yx} = -3/2$$

$$6x + y = 31$$

$$6x = 31 - y$$

$$x = \frac{31}{6} - y/6$$

$$\therefore b_{xy} = -1/6$$

$$\therefore r = \sqrt{b_{yx} \cdot b_{xy}}$$

$$= \sqrt{-3/2 \times -1/6}$$

$$= \sqrt{3/12}$$

$$\boxed{r = 0.5}$$

5. When is a random process said to be mean ergodic?

A random process $\{x(t)\}$ is mean-ergodic if $\lim_{T \rightarrow \infty} \text{Var} \bar{x}_T = 0$,
 where \bar{x}_T is time average of $\{x(t)\}$ and μ is its mean.

6. If $\{x(t)\}$ is a normal process with $\mu(t) = 10$ and $C(t_1, t_2) = 16e^{-|t_1 - t_2|}$, find the variance of $x(10) - x(6)$.

Let $U = x(10) - x(6)$. Then U is also a random variable.

$$E(U) = E[x(10) - x(6)] = E[x(10)] - E[x(6)] = 10 - 10 = 0$$

(Mean of $x(t) = 10$
 $\therefore E[x(10)] = 10$
 $E[x(6)] = 10$)

$$\begin{aligned} E(U^2) &= E[x^2(10) + x^2(6) - 2x(10)x(6)] \\ &= E[x^2(10)] + E[x^2(6)] - 2E[x(10)x(6)] \\ &= E[x^2(10)] + E[x^2(6)] - 2\text{Cov}(10, 6) \end{aligned}$$

$$\begin{aligned} \therefore \text{Var}(U) &= E(U^2) - (E(U))^2 = E[x^2(10)] + E[x^2(6)] - 2\text{Cov}(10, 6) \\ &= \text{Cov}(10, 10) + \text{Cov}(6, 6) - 2\text{Cov}(10, 6) \quad \left(\because \text{Cov}(10, 10) = 16e^{-|10-10|} = 16 \right) \\ \text{Var}(U) &= 16 + 16 - 32e^{-|10-6|} = 32 - 32e^{-4} = 31.4139 \end{aligned}$$

7. The ACF of a stationary random process is $R_{xx}(\tau) = 25 + \frac{4}{1+6\tau^2}$. Find the mean and variance of the process.

$$\bar{x}^2 = \lim_{\tau \rightarrow \infty} R_{xx}(\tau) = \lim_{\tau \rightarrow \infty} 25 + \frac{4}{1+6\tau^2} = 25 \Rightarrow \boxed{\bar{x} = 5} \text{ (Mean)}$$

$$R_{xx}(0) = 25 + \frac{4}{1+0} = 29 \quad \therefore E(x^2(t)) = 29$$

$$\therefore \text{Var}(x(t)) = E(x^2(t)) - \{E(x(t))\}^2 = 29 - 5^2 = 4$$

8. Prove that for a WSS process $\{x(t)\}$, $R_{xx}(t_1, t_1 + \tau)$ is an even function of τ .

$$R_{xx}(\tau) = E[x(t) x(t + \tau)] \text{ Replacing } \tau \text{ by } -\tau \text{ in } R_{xx}(\tau)$$

$$R_{xx}(-\tau) = E[x(t) x(t - \tau)] \text{ put } t - \tau = t_1; t = t_1 + \tau$$

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

Hence $R_{xx}(-\tau) = R_{xx}(\tau)$. i.e. $R_{xx}(t_1, t_1 + \tau)$ is an even function of τ .

9. State any two properties of a linear time-invariant system.

1) If the input $x(t)$ and its output $y(t)$ are related by $y(t) = \int_{-\infty}^{\infty} h(u) x(t-u) du$ then the system is a linear time-invariant system.

2) If $\{x(t)\}$ is a WSS process and if $y(t) = \int_{-\infty}^{\infty} h(u) x(t-u) du$, then

a) $R_{xy}(\tau) = R_{xx}(\tau) * h(\tau)$

b) $R_{yy}(\tau) = R_{xy}(\tau) * h(-\tau)$ where $*$ denotes the convolution.

c) $R_{yy}(\tau) = R_{xx}(\tau) * h(\tau) * h(-\tau)$

d) $S_{xy}(\omega) = S_{xx}(\omega) H(\omega)$ and

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

10. If $\{x(t)\}$ and $\{y(t)\}$ in the system $y(t) = \int_{-\infty}^{\infty} h(u) x(t-u) du$ are WSS processes, how are their auto correlation functions related.

$$R_{yy}(t, t+\tau) = E[y(t) \cdot y(t+\tau)]$$

$$= E \left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u_1) x(t-u_1) h(u_2) x(t+\tau-u_2) du_1 du_2 \right]$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u_1) h(u_2) E[x(t-u_1) x(t+\tau-u_2)] du_1 du_2$$

Since $\{x(t)\}$ is a WSS process, $R_{yy}(t, t+\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u_1) h(u_2) R_{xx}(\tau) du_1 du_2$

$R_{yy}(t, t+\tau) =$ a function of τ . Hence the output $\{y(t)\}$ is also a WSS process.

PART B - (5 x 16 = 80 marks).

11) a) (i) The probability function of an infinite discrete distribution is given by

$$P(X=j) = \frac{1}{2^j} \quad (j=1, 2, 3, \dots) \text{ Find}$$

(1) Mean of X

(2) $P(X \text{ is even})$ and

(3) $P(X \text{ is divisible by } 3)$.

$$(1) \text{ Mean} = E[X] = \sum_j x_j p(X=j)$$

$$= 1 \cdot \frac{1}{2^1} + 2 \cdot \frac{1}{2^2} + 3 \cdot \frac{1}{2^3} + 4 \cdot \frac{1}{2^4} + \dots$$

$$= \frac{1}{2} (1 + 2 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2^2} + 4 \cdot \frac{1}{2^3} + \dots)$$

$$= \frac{1}{2} (1 + 2(\frac{1}{2}) + 3(\frac{1}{2})^2 + 4(\frac{1}{2})^3 + \dots)$$

$$= \frac{1}{2} (1 + 2x + 3x^2 + 4x^3 + \dots) = (1-x)^{-2}$$

$$= \frac{1}{2} (\frac{1}{2})^{-2} = \frac{1}{2} \frac{1}{(\frac{1}{2})^2} = \frac{1}{2} \cdot \frac{1}{(\frac{1}{4})} = \frac{1}{2} \times 4 = 2.$$

$$\text{Mean} = E[X] = 2.$$

$$\text{or } \frac{1}{2} (\frac{1}{2})^2 = \frac{1}{2} (2)^2 = \frac{1}{2} \times 4 = 2.$$

$$(2) P(X \text{ is even}) = P(X=2, 4, 6, 8, \dots)$$

$$= P(X=2) + P(X=4) + P(X=6) + P(X=8) + \dots$$

$$= \frac{1}{2^2} + \frac{1}{2^4} + \frac{1}{2^6} + \frac{1}{2^8} + \dots$$

$$= \frac{1}{2^2} (1 + \frac{1}{2^2} + \frac{1}{2^4} + \frac{1}{2^6} + \dots)$$

$$= \frac{1}{4} (1 + \frac{1}{2^2} + (\frac{1}{2^2})^2 + (\frac{1}{2^2})^3 + \dots)$$

$$1 + x + x^2 + \dots = (1-x)^{-1}$$

$$= \frac{1}{4} [1 - \frac{1}{4}]^{-1} = \frac{1}{4} (\frac{3}{4})^{-1} = \frac{1}{4} (\frac{4}{3})^1 = \frac{1}{4} \times \frac{4}{3} = \frac{1}{3}.$$

$$\frac{1}{2^3 \times 2^3} = 2^{-6} = \frac{1}{2^6}$$

$$2^6 \cdot 2^2 = 2^{6+2} = 2^8$$

$$(3) P(X \text{ is divisible by } 3) = P(X=3, 6, 9, 12, \dots)$$

$$= P(X=3) + P(X=6) + P(X=9) + \dots$$

$$= (\frac{1}{2^3} + \frac{1}{2^6} + \frac{1}{2^9} + \dots) = \frac{1}{2^3} + \frac{1}{2^3 \cdot 2^3} + \frac{1}{2^3 \cdot 2^3 \cdot 2^3} + \dots$$

$$= \frac{1}{2^3} (1 + \frac{1}{2^3} + \frac{1}{2^3 \cdot 2^3} + \dots) = \frac{1}{2^3} [1 + (\frac{1}{2^3}) + (\frac{1}{2^3})^2 + \dots]$$

$$P(x \text{ is divisible by } 3) = \frac{1}{2^3} \left[1 - \frac{1}{2^3} \right]^{-1} = \frac{1}{8} \left(\frac{7}{8} \right)^{-1} = \frac{1}{8} \left(\frac{8}{7} \right)^{-1} = \frac{1}{7}.$$

(i) A continuous RV X has the pdf $f(x) = \begin{cases} \frac{k}{1+x^2}, & -\infty < x < \infty \\ 0, & \text{elsewhere} \end{cases}$. Find

(1) the value of k

(2) Distribution function of X .

(3) $P(X \geq 0)$.

Since $f(x)$ is pdf we have $\int_{-\infty}^{\infty} f(x) dx = 1$.

$$k \int_{-\infty}^{\infty} \frac{1}{1+x^2} dx = 1 \Rightarrow k \left(\tan^{-1} x \right)_{-\infty}^{\infty} = 1 \Rightarrow k [\tan^{-1}(\infty) - \tan^{-1}(-\infty)] = 1.$$

$$k \left[\left(\frac{\pi}{2} \right) + \frac{\pi}{2} \right] = 1 \Rightarrow \boxed{k = \frac{1}{\pi}} \quad \text{since } \tan^{-1}(\infty) = \frac{\pi}{2} \text{ \& } \tan^{-1}(-\infty) = -\frac{\pi}{2}.$$

$$\therefore f(x) = \begin{cases} \frac{1}{\pi} \cdot \frac{1}{1+x^2}, & -\infty < x < \infty \\ 0, & \text{otherwise.} \end{cases}$$

$$F(x) = \int_{-\infty}^x f(x) dx = \frac{1}{\pi} \int_{-\infty}^x \frac{1}{1+x^2} dx = \frac{1}{\pi} \left(\tan^{-1} x \right)_{-\infty}^x = \frac{1}{\pi} [\tan^{-1} x - \tan^{-1}(-\infty)]$$

$$F(x) = \frac{1}{\pi} [\tan^{-1} x + \frac{\pi}{2}]$$

$$P(X > 0) = 1 - P(X \leq 0) = 1 - F(0) = 1 - \frac{1}{\pi} \left[0 + \frac{\pi}{2} \right] = 1 - \frac{1}{2} = \frac{1}{2}.$$

11) b) Let X and Y be independent normal variates with mean 45 and 44 and standard deviation 2 and 1.5 respectively. What is the probability that randomly chosen values of X and Y differ by 1.5 or more.

12) a) i) The joint probability density function of random variable x and y is given by $f(x, y) = \begin{cases} \frac{8xy}{9}, & 1 \leq x \leq y \leq 2 \\ 0, & \text{otherwise} \end{cases}$ Find the conditional density function of x and y .

Conditional density function of x given $Y=y$ is $f(x|y) = \frac{f(x, y)}{f(y)}$

$$f(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_1^y \frac{8}{9} xy dx = \frac{8}{9} y \left(\frac{x^2}{2} \right)_1^y = \frac{4}{9} y(y^2 - 1), 1 < y < 2.$$

$$f(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_x^2 \frac{8}{9} xy dy = \frac{8}{9} x \left(\frac{y^2}{2} \right)_x^2 = \frac{4}{9} x(4 - x^2), 1 < x < 2$$

$$\therefore f(x|y) = \frac{\frac{8}{9} xy}{\frac{4}{9} (y^2 - 1)} = \frac{2x}{y^2 - 1}, 1 < x < y < 2.$$

$$\text{and } f(y|x) = \frac{f(x, y)}{f(x)} = \frac{\frac{8}{9} xy}{\frac{4}{9} x(4 - x^2)} = \frac{2y}{4 - x^2}, 1 < x < y < 2.$$

ii) The jpdf of the 2D RV (X, Y) is $f(x, y) = \begin{cases} 2 - x - y, & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{otherwise} \end{cases}$

Find the correlation coefficient between x and y .

$$\rho(x, y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)} \sqrt{\text{Var}(Y)}} \quad \text{Cov}(X, Y) = E(XY) - E(X)E(Y) = \frac{1}{6} - \frac{5}{12} \cdot \frac{5}{12} = -\frac{1}{144}$$

$$f(x) = \frac{3}{2} - x$$

$$f(y) = \frac{3}{2} - y$$

$$E(X) = \int_{-\infty}^{\infty} x f(x) dx = \int_0^1 x \left(\frac{3}{2} - x \right) dx = \frac{5}{12}$$

$$E(Y) = \int_{-\infty}^{\infty} y f(y) dy = \int_0^1 y \left(\frac{3}{2} - y \right) dy = \frac{5}{12}$$

$$E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_0^1 x^2 \left(\frac{3}{2} - x \right) dx = \frac{1}{4} = E(Y^2)$$

$$\text{Var}(X) = E(X^2) - (E(X))^2 = \frac{1}{4} - \left(\frac{5}{12} \right)^2 = \frac{11}{144} \Rightarrow \sigma_x = \frac{\sqrt{11}}{12} \text{ \& } \sigma_y = \frac{\sqrt{11}}{12}$$

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

$$\rho(x, y) = \frac{-1/144}{\frac{\sqrt{11}}{12} \cdot \frac{\sqrt{11}}{12}} = -\frac{1}{11}$$

13a) Show that the random process $X(t) = A \cos(\omega t + \theta)$ is wide sense stationary, if A and ω are constants and θ is uniformly distributed in $(0, 2\pi)$. $\theta \sim U(0, 2\pi) \therefore f(\theta) = \frac{1}{2\pi}, (0, 2\pi)$.

$$E[X(t)] = E[A \cos(\omega t + \theta)] = A \int_0^{2\pi} \cos(\omega t + \theta) \cdot f(\theta) d\theta = \frac{A}{2\pi} \int_0^{2\pi} \cos(\omega t + \theta) d\theta$$

$$= \frac{A}{2\pi} \left[\frac{\sin(\omega t + \theta)}{1} \right]_0^{2\pi} = \frac{A}{2\pi} [\sin \omega t - \sin \omega t] = 0, \text{ a constant.}$$

$$R_{XX}(t, t+\tau) = E[X(t)X(t+\tau)] = E[A^2 \cos(\omega t + \theta) \cdot \cos(\omega t + \omega\tau + \theta)]$$

$$= \frac{A^2}{2} E[\cos(2\omega t + \omega\tau + 2\theta) + \cos(-\omega\tau)]$$

$\cos A \cos B = \frac{1}{2} [\cos(A+B) + \cos(A-B)]$

$$= \frac{A^2}{2} \cos \omega\tau + \frac{A^2}{2} E[\cos(\omega t + \omega\tau + 2\theta)]$$

$$= \frac{A^2}{2} \cos \omega\tau + \frac{A^2}{2} \int_0^{2\pi} \cos(\omega t + \omega\tau + 2\theta) \cdot f(\theta) d\theta$$

$$= \frac{A^2}{2} \cos \omega\tau + \frac{A^2}{4\pi} \left[\sin(\omega t + \omega\tau + 2\theta) \right]_0^{2\pi}$$

$$R_{XX}(t, t+\tau) = \frac{A^2}{2} \cos \omega\tau, \text{ a fn of } \tau \quad (\because \sin(\omega t + \omega\tau + 2\pi) = \sin(\omega t + \omega\tau))$$

$\therefore X(t)$ is WSS process.

b) State the postulates of a poisson process and derive the probability distribution. Also prove that the sum of two independent poisson process is a poisson process.

Refer:

14) a) The ACF of a random process is given by $R(\tau) = \begin{cases} \alpha^2, & |\tau| > \epsilon \\ \alpha^2 + \frac{\alpha}{\epsilon} (1 - \frac{|\tau|}{\epsilon}), & |\tau| \leq \epsilon \end{cases}$
Find the power Spectral density of the process.

Let $\lambda = \alpha$ & $\epsilon = \lambda$ we get

$$R(\tau) = \begin{cases} \alpha^2, & |\tau| > \lambda \\ \alpha^2 + \frac{\alpha}{\lambda} (1 - \frac{|\tau|}{\lambda}), & |\tau| \leq \lambda. \end{cases}$$

The spectral density function is given by

$$S_{xx}(\omega) = \int_{-\infty}^{\infty} R_{xx}(\tau) e^{-j\omega\tau} d\tau$$

$$= \int_{-\infty}^{-\lambda} R_{xx}(\tau) e^{-j\omega\tau} d\tau + \int_{-\lambda}^{+\lambda} R_{xx}(\tau) e^{j\omega\tau} d\tau + \int_{\lambda}^{\infty} R_{xx}(\tau) e^{-j\omega\tau} d\tau$$

$$= \int_{-\infty}^{-\lambda} \alpha^2 e^{-j\omega\tau} d\tau + \int_{-\lambda}^{\lambda} \left(\alpha^2 + \frac{\alpha}{\lambda} \left(1 - \frac{|\tau|}{\lambda} \right) \right) e^{-j\omega\tau} d\tau + \int_{\lambda}^{\infty} \alpha^2 e^{-j\omega\tau} d\tau$$

$$= \int_{-\infty}^{\infty} \alpha^2 e^{-j\omega\tau} d\tau + \frac{\alpha}{\lambda} \int_{-\lambda}^{\lambda} \left(1 - \frac{|\tau|}{\lambda} \right) e^{-j\omega\tau} d\tau$$

$$= F(\alpha^2) + \frac{2\alpha}{\lambda} \int_0^{\lambda} \left(1 - \frac{\tau}{\lambda} \right) \cos \omega\tau d\tau$$

$$= F(\alpha^2) + \frac{2\alpha}{\lambda} \left[\left(1 - \frac{\tau}{\lambda} \right) \left(\frac{\sin \omega\tau}{\omega} \right) - \left(-\frac{\tau}{\lambda} \right) \left(-\frac{\cos \omega\tau}{\omega^2} \right) \right]_0^{\lambda}$$

$$= \frac{2\alpha}{\lambda^2 \omega^2} 2 \sin^2 \left(\frac{\omega\lambda}{2} \right) + F(\alpha^2)$$

$$= 4\alpha \frac{\sin^2(\lambda\omega/2)}{\lambda^2 \omega^2} + F(\alpha^2)$$

$$S_{xx}(\omega) = \frac{4\alpha \sin^2(\lambda\omega/2)}{\lambda^2 \omega^2} + 2\pi \alpha^2 \delta(\omega)$$

$$\left(\begin{array}{l} \because F(\alpha^2) = \alpha^2 \int_{-\infty}^{\infty} e^{j\omega\tau} d\tau \\ \delta(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-j\omega\tau} d\tau \\ 2\pi \delta(\omega) = \int_{-\infty}^{\infty} e^{-j\omega\tau} d\tau \end{array} \right)$$

(ii) Given the power spectral density of a continuous process as $S_{xx}(\omega) = \frac{\omega^2 + 9}{\omega^4 + 15\omega^2 + 4}$. Find the mean square value of the process.

$$R_{xx}(\tau) = F^{-1} \left[\frac{\omega^2 + 9}{\omega^4 + 15\omega^2 + 4} \right] \text{ consider } \frac{\omega^2 + 9}{\omega^4 + 15\omega^2 + 4} = \frac{u + 9}{u^2 + 15u + 4}, u = \omega^2$$

$$\frac{u + 9}{(u+1)(u+4)} = \frac{A}{u+1} + \frac{B}{u+4} \Rightarrow \boxed{u+9 = A(u+4) + B(u+1)}$$

$$\text{put } u = -4, \quad B = -5/3$$

$$\text{put } u = -1, \quad A = 8/3$$

$$\therefore \frac{u+9}{(u+1)(u+4)} = \frac{8/3}{u+1} - \frac{5/3}{u+4}$$

$$\begin{aligned} \therefore R_{xx}(z) &= F^{-1} \left[\frac{8/3}{\omega^2+1} - \frac{5/3}{\omega^2+4} \right] \\ &= \frac{8/3}{\omega^2+1} \times \frac{1}{2} F^{-1} \left[\frac{2 \times 1}{\omega^2+1} \right] - \frac{5/3}{\omega^2+4} \times \frac{1}{2} F^{-1} \left[\frac{2 \times 2}{\omega^2+4} \right] \end{aligned}$$

$$R_{xx}(z) = \frac{8}{6} e^{-|z|} - \frac{5}{12} e^{-2|z|}$$

$$\text{Mean square value } R_{xx}(0) = 8/6 - 5/12 = 11/12 //$$

14) b) state and prove Wiener-Khinchine theorem.

(c)

Refer: Apr/May 2011 Q. No 14 b (i)

(ii) The cross power spectrum of real random processes $\{x(t)\}$ and $\{y(t)\}$ is given by $S_{xy}(\omega) = \begin{cases} a+bj\omega, & |\omega| < 1 \\ 0, & \text{elsewhere} \end{cases}$ Find the cross correlation function

Refer: Apr/May 2011 Q. No 14 b (ii)

15) a) i) If the input to a time invariant, stable, linear system is a WSS process. Prove that the output will also be a WSS process.

$$\text{The I/P and O/P are related by } y(t) = \int_{-\infty}^{\infty} h(\omega) x(t-u) du \quad \text{--- (1)}$$

$$\Rightarrow E[y(t)] = \int_{-\infty}^{\infty} h(\omega) E[x(t-u)] du \quad \text{--- (2)}$$

Since $\{x(t)\}$ is a WSS process, mean is a constant i.e. $E[x(t-u)]$ is a constant

$$E[y(t)] = E[x(t-u)] \int_{-\infty}^{\infty} h(\omega) du = \bar{x}_u \int_{-\infty}^{\infty} h(\omega) du = \text{a finite constant, independent of } t$$

$\therefore E[y(t)] = \text{a constant}$.

$$\text{By definition } R_{yy}(t, t+\tau) = E[y(t) \cdot y(t+\tau)] = E \left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\omega_1) x(t-u_1) h(\omega_2) x(t+\tau-u_2) du_1 du_2 \right]$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\omega_1) h(\omega_2) E[x(t-u_1) x(t+\tau-u_2)] du_1 du_2$$

= a fn of τ since $\{x(t)\}$ is a WSS process

Hence, the output $\{y(t)\}$ is also a WSS process.

a) (i) Let $x(t)$ be a wss process which is the input to a linear time invariant system with unit impulse $h(t)$ and output $y(t)$, then prove that $S_{yy}(\omega) = |H(\omega)|^2 S_{xx}(\omega)$.

WKT $R_{xy}(\tau) = R_{xx}(\tau) * h(\tau)$ (Q.15 b)(i) proof.

Taking Fourier Transform, we have

$$\begin{aligned} F[R_{xy}(\tau)] &= F[R_{xx}(\tau) * h(\tau)] \\ &= F[R_{xx}(\tau)] F[h(\tau)] \end{aligned}$$

By convolution theorem on Fourier transform,

$$\begin{aligned} S_{xy}(\omega) &= S_{xx}(\omega) F[h(\tau)] \\ S_{xy}(\omega) &= S_{xx}(\omega) H(\omega) \quad \text{--- (1)} \end{aligned}$$

Similarly Applying Fourier Transform on $R_{yy}(\tau) = R_{xy}(\tau) * h(-\tau)$ we get

$$S_{yy}(\omega) = S_{xy}(\omega) \overline{H(\omega)} \quad \text{--- (2)}$$

From (1) and (2), $S_{yy}(\omega) = S_{xx}(\omega) H(\omega) \cdot \overline{H(\omega)} = \underline{S_{xx}(\omega) |H(\omega)|^2}$

b) (i) For a input-output linear system $(x(t), h(t), y(t))$, derive $R_{xy}(\tau)$ & $R_{yy}(\tau)$.

(i) Given $y(t) = \int_{-\infty}^{\infty} h(u) x(t-u) du$.

WKT $R_{xy}(\tau) = E[x(t)y(t+\tau)] = E\left[x(t) \int_{-\infty}^{\infty} h(u) x(t+\tau-u) du\right]$
 $= \int_{-\infty}^{\infty} E[x(t)x(t+\tau-u)] h(u) du = \int_{-\infty}^{\infty} R_{xx}(\tau-u) h(u) du$

$\therefore R_{xy}(\tau) = R_{xx}(\tau) * h(\tau)$.

(ii) $R_{yy}(\tau) = E[y(t)y(t+\tau)] = E\left[\int_{-\infty}^{\infty} h(u) x(t-u) \cdot \int_{-\infty}^{\infty} h(u_1) x(t+\tau-u_1) du\right] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[x(t-u)x(t+\tau-u_1)] h(u) h(u_1) du du_1$

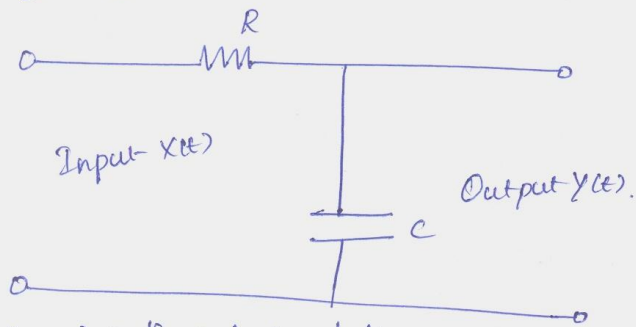
put $t-u = u_1$ when $u = -\infty, u_1 = -\infty$
 $t = u + u_1$
 $t+\tau = u + u_1 + \tau$ $u = \infty, u_1 = \infty$

$\therefore R_{yy}(\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[x(u_1)y(u_1+u+\tau)] h(u) h(u_1) du du_1 = \int_{-\infty}^{\infty} R_{xy}(\tau+u) h(u) du$

$= \int_{-\infty}^{\infty} R_{xy}(\tau-\alpha) h(-\alpha) (-d\alpha)$ (Taking $u = -\alpha$)

$= \int_{-\infty}^{\infty} R_{xy}(\tau-\alpha) h(-\alpha) d\alpha \quad \therefore \boxed{R_{yy}(\tau) = R_{xy}(\tau) * h(-\tau)}$

156) A white Gaussian noise $x(t)$ with zero mean and spectral density $\frac{N_0}{2}$ is applied to a low-pass RC filter shown in the figure.



Determine the autocorrelation of the output $y(t)$.

Answer:

The transfer function of the filter is, $H(f) = \frac{1}{1 + i2\pi fRC}$

$$\therefore H(\omega) = \frac{1}{1 + i\omega RC}, \quad \omega = 2\pi f.$$

Let Output $y(t)$ be $N(t)$.

The power spectral density of the noise $N(t)$ appearing at the low pass RC filter output is given by

$$S_{NN}(\omega) = S_{WW}(\omega) |H(\omega)|^2$$

$$= \frac{N_0}{2} \frac{1}{1 + (\omega RC)^2}$$

$$= \frac{N_0}{2} \frac{\beta^2}{\beta^2 + \omega^2}, \quad \beta = \frac{1}{RC}$$

$$S_{NN}(\omega) = \frac{\beta N_0}{2} \cdot \frac{\beta}{\beta^2 + \omega^2}$$

Taking Inverse Fourier transform, we get the ACF

$$R_{NN}(\tau) = F^{-1} \left[\frac{\beta N_0}{2} \cdot \frac{\beta}{\beta^2 + \omega^2} \right]$$

$$= \frac{\beta N_0}{4} F^{-1} \left[\frac{2\beta}{\beta^2 + \omega^2} \right]$$

$$= \frac{\beta N_0}{4} e^{-\beta|\tau|} \quad \left(\text{since } F(e^{-\alpha|\tau|}) = \frac{2\alpha}{\alpha^2 + \omega^2} \right)$$

$$R_{NN}(\tau) = \frac{N_0}{4RC} e^{-\frac{|\tau|}{RC}} \quad \text{and } \beta = \frac{1}{RC}.$$