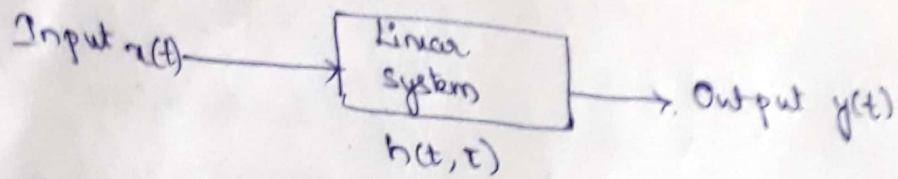


Unit-IV (part 2)

Linear Systems - fundamentals:

①



→ A system can be represented as

$$y(t) = L[x(t)] \quad \text{--- ①}$$

where L is an operator representing the action of the system on $x(t)$.

→ A system is said to be linear if its response to a sum of inputs $x_n(t)$, $n=1, 2, \dots, N$, is equal to the sum of responses taken separately.

$$\begin{aligned} \therefore y(t) &= L\left[\sum_{n=1}^N \alpha_n x_n(t)\right] = \sum_{n=1}^N \alpha_n L[x_n(t)] \\ &= \sum_{n=1}^N \alpha_n y_n(t) \quad \text{--- ②} \end{aligned}$$

where α_n are arbitrary constants.

→ From the properties of the impulse fn, we have.

$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau \quad \text{--- ③}$$

→ By substituting eq. ③ in eq. ①, we obtain.

$$\begin{aligned} y(t) &= L[x(t)] = L\left[\int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau\right] \\ &= \int_{-\infty}^{\infty} x(\tau) \underbrace{L[\delta(t-\tau)]}_{h(t, \tau)} d\tau \quad \text{--- ④} \end{aligned}$$

→ Now, define a new fn. $h(t, \tau)$ as the impulse resp. of the linear system, i.e.

$$L[\delta(t-\tau)] = h(t, \tau).$$

eq. (4) becomes

$$y(t) = \int_{-\infty}^{\infty} x(\tau) h(t, \tau) d\tau. \quad \text{--- (5)}$$

which shows that the response of a general linear system is completely determined by its impulse response.

Mean & Mean Squared value of System Response:

→ To find the mean of the sys. response, assume $x(t)$ is WSS, we have.

$$\begin{aligned} E[y(t)] &= E\left[\int_{-\infty}^{\infty} h(\tau) x(t-\tau) d\tau\right] \\ &= \int_{-\infty}^{\infty} h(\tau) \cdot E[x(t-\tau)] d\tau \\ &= \bar{x} \cdot \int_{-\infty}^{\infty} h(\tau) d\tau = \bar{y} = \text{const.} \end{aligned}$$

→ This expression indicates that the mean value of $y(t)$ equals the mean value of $x(t)$ times the area under the impulse resp. if $x(t)$ is WSS.

→ For mean-squared value of $y(t)$,

$$\begin{aligned} E[y^2(t)] &= E\left[\int_{-\infty}^{\infty} h(\tau_1) x(t-\tau_1) d\tau_1 \cdot \int_{-\infty}^{\infty} h(\tau_2) x(t-\tau_2) d\tau_2\right] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[x(t-\tau_1) x(t-\tau_2)] \cdot h(\tau_1) h(\tau_2) d\tau_1 d\tau_2 \end{aligned}$$

→ If we assume, the i/p is WSS, then

$$E[x(t-\tau_1) x(t-\tau_2)] = R_{xx}(\tau_1 - \tau_2)$$

$$\therefore \bar{y}^2 = E[y^2(t)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(\tau_1 - \tau_2) \cdot h(\tau_1) h(\tau_2) d\tau_1 d\tau_2$$

Auto correlation fn. of Response

Let $x(t)$ be WSS. The auto correlation fn. of $y(t)$ is

$$\begin{aligned} R_{yy}(t, t+\tau) &= E[y(t)y(t+\tau)] \\ &= E\left[\int_{-\infty}^{\infty} h(\tau_1) x(t-\tau_1) d\tau_1 \cdot \int_{-\infty}^{\infty} h(\tau_2) x(t+\tau-\tau_2) d\tau_2\right] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[x(t-\tau_1)x(t+\tau-\tau_2)] h(\tau_1) h(\tau_2) d\tau_1 d\tau_2 \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(\tau+\tau_1-\tau_2) h(\tau_1) h(\tau_2) d\tau_1 d\tau_2 \\ &= R_{yy}(\tau) \end{aligned}$$

Note

$\therefore y(t)$ is WSS if $x(t)$ is WSS, because $R_{yy}(\tau)$ does not depend on t & $E[y(t)]$ is a const.

$\rightarrow R_{yy}(\tau)$ is the two fold convolution of the l/p auto correlation fn. with the n/w.s impulse response. i.e.

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

Cross correlation fns of l/p & o/p:

$\rightarrow$ The cross correlation fn. of $x(t)$ & $y(t)$ is

$$\begin{aligned} R_{xy}(t, t+\tau) &= E[x(t)y(t+\tau)] \\ &= E\left[x(t) \cdot \int_{-\infty}^{\infty} h(\tau_1) x(t+\tau-\tau_1) d\tau_1\right] \\ &= \int_{-\infty}^{\infty} E[x(t)x(t+\tau-\tau_1)] h(\tau_1) d\tau_1 \quad \text{--- (1)} \end{aligned}$$

$\rightarrow$ If $x(t)$ is WSS,

$$\therefore R_{xy}(\tau) = \int_{-\infty}^{\infty} R_{xx}(\tau-\tau_1) h(\tau_1) d\tau_1 \quad \text{--- (2)}$$

which is the convolution of $R_{xx}(t)$ with $h(t)$. (4)

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

Similarly,

$$R_{yx}(t) = \int_{-\infty}^{\infty} R_{xx}(t-\tau) h(-\tau) d\tau \quad \text{--- (3)}$$

$$= R_{xx}(t) * h(-t).$$

→ from eqs (2) & (3);

It is clear that the cross correlation fns depend on τ and not on absolute time t .

$\therefore x(t)$ & $y(t)$ are jointly WSS, if $x(t)$ is WSS, because $y(t)$ already is WSS.

→

$$R_{yy}(t) = \int_{-\infty}^{\infty} R_{xy}(t+\tau_1) h(\tau_1) d\tau_1,$$

$$\therefore \boxed{R_{yy}(t) = R_{xy}(t) * h(-t)}$$

→ Similarly,

$$R_{yy}(t) = \int_{-\infty}^{\infty} R_{yx}(t-\tau_2) h(\tau_2) d\tau_2.$$

$$\boxed{R_{yy}(t) = R_{yx}(t) * h(t)}$$

Gaussian Random Processes

Consider a continuous RP, and define N RVs

$x_1 = x(t_1), x_2 = x(t_2), \dots, x_i = x(t_i), \dots, x_N = x(t_N)$. corresponding to N time instants $t_1, \dots, t_i, \dots, t_N$.

→ If, for any $N = 1, 2, \dots$ and any times $t_1, t_2, \dots, t_N$, these RVs are jointly gaussian, i.e. they have a joint density, then the process is called Gaussian.

→ The joint density is given by.

$$f_x(x_1, \dots, x_N; t_1, \dots, t_N) = \frac{\exp\left\{-(1/2) [x-\bar{x}]^T [C_x]^{-1} [x-\bar{x}]\right\}}{\sqrt{(2\pi)^N |C_x|}}$$

where
→ $[x-\bar{x}]$ & $[C_x]$ are matrices.

→ The mean values $\bar{x}_i$ of $x_i = x(t_i)$ are

$$\bar{x}_i = E[x_i] = E[x(t_i)]$$

→ The elements of the covariance matrix $[C_x]$ are

$$\begin{aligned} c_{ik} &= C_{x_i x_k} = E[(x_i - \bar{x}_i)(x_k - \bar{x}_k)] \\ &= E\left[\{x(t_i) - E[x(t_i)]\} \{x(t_k) - E[x(t_k)]\}\right] \\ &= C_{xx}(t_i, t_k) \rightarrow \text{the auto covariance of } x(t_i) \text{ \& } x(t_k). \end{aligned}$$

→ WKT

$$C_{xx}(t_i, t_k) = R_{xx}(t_i, t_k) - E[x(t_i)]E[x(t_k)]$$

→ If the gaussian process is not stationary the mean & auto covariance fns depend on absolute time.

For a WSS process, the mean will be const.

$$\bar{X}_t = E[X(t)] = \bar{X} \text{ (const.)}$$

while the auto covariance & auto correlation fns. will depend only on time diff.s & not absolute time.

$$C_{XX}(t_i, t_k) = C_{XX}(t_k - t_i)$$

$$R_{XX}(t_i, t_k) = R_{XX}(t_k - t_i)$$

Poisson Random Process:

→ It is an example of Discrete Random Process.

→ It describes the no. of times that some event has occurred as a fn. of time, where the events occur at random times.

→ Examples:

1. Arrival of a customer at a bank.
2. Super market check-out register
3. failure of some component in a system.
4. Emission of an electron from the surface of photo detector etc.

→ A single event occurs at a random time & the process amounts to count the no. of such occurrences with time. For this reason, the process is also known as poisson counting process.

→ To define the Poisson RP, we require two conditions.

1. The event occurs only once in any small interval of time.
2. The occurrence times be statistically independent so that the no. that occurs in ^{any} given time interval is independent of the no. in any other non-overlapping time interval.

A consequence of the 2 conditions is that the no. of event occurrences in any finite interval of time is described by the poisson distribution where the avg. rate of occurrences is denoted by λ .

→ The prob. of exactly 'k' occurrences over a time interval (0, t) is

$$p[x(t) = k] = \frac{(\lambda t)^k e^{-\lambda t}}{k!}, \quad k = 0, 1, 2, \dots$$

→ ∴ The prob. density of the no. of occurrences is

$$f_x(x) = \sum_{k=0}^{\infty} \frac{(\lambda t)^k e^{-\lambda t}}{k!} \delta(x - k)$$

→ The mean of the Poisson RP is

$$\begin{aligned} E[x(t)] &= \int_{-\infty}^{\infty} x \cdot f_x(x) dx = \int_{-\infty}^{\infty} x \sum_{k=0}^{\infty} \frac{(\lambda t)^k e^{-\lambda t}}{k!} \delta(x - k) dx \\ &= \sum_{k=0}^{\infty} \frac{k (\lambda t)^k e^{-\lambda t}}{k!} = \lambda t \end{aligned}$$

→ The mean squared value is

$$\begin{aligned} E[x^2(t)] &= \int_{-\infty}^{\infty} x^2 f_x(x) dx = \sum_{k=0}^{\infty} \frac{k^2 (\lambda t)^k e^{-\lambda t}}{k!} \\ &= \lambda t [1 + \lambda t] \end{aligned}$$

→ The variance is also λt

→ The joint probability density f_j is given by

$$f_x(x_1, x_2) = \sum_{k_1=0}^{\infty} \sum_{k_2=0}^{\infty} p(k_1, k_2) \delta(x_1 - k_1) \delta(x_2 - k_2)$$

Unit-5 Random Processes - Spectral Characteristics

①

The spectral description of a deterministic waveform is obtained by Fourier Transforming the waveform.

→ ∴ The spectral properties of a deterministic signal $x(t)$ are contained in its F.T $X(\omega)$ given by.

$$X(\omega) = \int_{-\infty}^{\infty} x(t) \cdot e^{-j\omega t} dt.$$

→ The f.t. $X(\omega)$ is the spectrum of $x(t)$, has the unit of volts per hertz. when $x(t)$ is a voltage.

→ It describes the way in which relative signal voltage is distributed with freq.

→ The F.T can be considered to be a voltage density spectrum applicable to $x(t)$. Both the amplitudes & phases of the frequencies present in $x(t)$ are described by $X(\omega)$.

→ If $X(\omega)$ is known then $x(t)$ can be recovered by means of the Inverse Fourier Transform.

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) \cdot e^{j\omega t} d\omega.$$

Power Density Spectrum:

→ For a RP $x(t)$, let $x_T(t)$ be defined as a sample f.t. $x(t)$ that exists between $-T$ and T , i.e.

$$x_T(t) = \begin{cases} x(t) & -T < t < T \\ 0 & \text{elsewhere} \end{cases}$$

→ The Fourier Transform $X_T(\omega)$ is given by.

$$X_T(\omega) = \int_{-T}^T x_T(t) \cdot e^{-j\omega t} dt = \int_{-T}^T x(t) \cdot e^{-j\omega t} dt.$$

Energy contained in $x(t)$ in the interval $(-T, T)$ is

$$E(T) = \int_{-T}^T x^2(t) dt = \int_{-T}^T x^2(t) dt.$$

→ By Parseval's theorem,

$$E(T) = \int_{-T}^T x^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X_T(\omega)|^2 d\omega. \quad \text{--- (1)}$$

→ By dividing eq. (1) by $2T$, we obtain the avg. power

$P(T)$ in $x(t)$ over the interval $(-T, T)$:

$$P(T) = \frac{1}{2T} \int_{-T}^T x^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{|X_T(\omega)|^2}{2T} d\omega. \quad \text{--- (2)}$$

We observe that $\frac{|X_T(\omega)|^2}{2T}$ is a power density spectrum because the power results through its integration.

→ $P(T)$ is actually a random variable w.r.to the RP.

By taking the expected value in eq. (2), we can obtain an avg. power P_{xx} for the Random Process.

$$\therefore P_{xx} = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T E[x^2(t)] dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} \frac{E[|X_T(\omega)|^2]}{2T} d\omega$$

↓
extending to all T values.

--- (3)

→ From eq. (3), we observed that

① The avg. power P_{xx} in a RP $x(t)$ is given by the time avg. of its second moment.

$$P_{xx} = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T E[x^2(t)] dt = A\{E[x^2(t)]\}$$

for a WSS process, $E[x^2(t)] = \overline{x^2} = \text{const.}$

$$\& \quad P_{xx} = \overline{x^2}$$

(3)

P_{xx} can be obtained by a freq. domain integration.

The power density spectrum for the RP is

$$S_{xx}(\omega) = \lim_{T \rightarrow \infty} \frac{E[|X_T(\omega)|^2]}{2T}$$

& the power is given by

$$P_{xx} = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) d\omega$$

Prob: Consider the RP $x(t) = A_0 \cos(\omega_0 t + \theta)$ where A_0 & ω_0 are real const.s & θ is a RV uniformly distributed on the interval $(0, \pi/2)$. Find the avg. power P_{xx} in $x(t)$.

Sol:

$$P_{xx} = A \{E[x^2(t)]\}$$

$$E[x^2(t)] = E[A_0^2 \cos^2(\omega_0 t + \theta)]$$

$$= E \left[\frac{A_0^2}{2} + \frac{A_0^2}{2} \cos(2\omega_0 t + 2\theta) \right]$$

$$\cos 2\theta = 2\cos^2 \theta - 1$$

$$= \frac{A_0^2}{2} + \frac{A_0^2}{2} \int_0^{\pi/2} \cos(2\omega_0 t + 2\theta) \cdot \frac{1}{\pi/2} d\theta$$

$$= \frac{A_0^2}{2} - \frac{A_0^2}{\pi} \sin(2\omega_0 t)$$

$$\left[\frac{\sin(2\omega_0 t + 2\theta)}{2} \right]_0^{\pi/2}$$

$$A \{E[x^2(t)]\} = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \left[\frac{A_0^2}{2} - \frac{A_0^2}{\pi} \sin(2\omega_0 t) \right] dt$$

$$= \frac{A_0^2}{2}$$

$$\therefore P_{xx} = A \{E[x^2(t)]\} = A_0^2/2$$

Properties of the Power Density Spectrum:

1. $S_{xx}(\omega) \geq 0$
2. $S_{xx}(-\omega) = S_{xx}(\omega)$; $x(t)$ real.
3. $S_{xx}(\omega)$ is real.
4. $\frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) d\omega = A\{E[x^2(t)]\}$
5. $S_{\dot{x}\dot{x}}(\omega) = \omega^2 S_{xx}(\omega)$

It says that the power density spectrum of the derivative $\dot{x}(t) = \frac{d}{dt}x(t)$ is ω^2 times the power spectrum of $x(t)$.

6. $\frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) \cdot e^{j\omega t} d\omega = A[R_{xx}(t, t+\tau)]$

$$\therefore S_{xx}(\omega) = \int_{-\infty}^{\infty} A[R_{xx}(t, t+\tau)] \cdot e^{-j\omega \tau} d\tau.$$

It states that the power density spectrum & the time avg. of the auto correlation fn. form a FT pair.

→ If $x(t)$ is at least WSS,

$$A[R_{xx}(t, t+\tau)] = R_{xx}(\tau)$$

then prop. ⑥ indicates that the power spectrum & the auto correlation fn. form a FT pair.

Thus

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

$$\& \quad R_{xx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) \cdot e^{j\omega \tau} d\omega.$$

for a WSS Process.

(5)

Bandwidth of the power Density spectrum:

Assume that $x(t)$ is a low pass process. i.e. its spectral components are clustered near $\omega=0$ & have decreasing magnitudes at higher ~~freqs.~~ freqs. Such processes are also called Base band.

- Then the area of $f_{xx}(\omega)$ is not necessarily unity, $f_{xx}(\omega)$ has characteristics similar to the prob. density fn.
- The normalized power spectrum is a measure of its spread that call as rms bandwidth, which we denote ω_{rms} (rad/sec).
- Since $f_{xx}(\omega)$ is an even fn. for a real process, its 'mean value' is 'zero' & its standard deviation is the sq. root of its second moment.
- ∴ The rms (root mean squared) Bandwidth is given by

$$\omega_{rms}^2 = \frac{\int_{-\infty}^{\infty} \omega^2 \cdot f_{xx}(\omega) d\omega}{\int_{-\infty}^{\infty} f_{xx}(\omega) d\omega}$$

- If the process $x(t)$ is real, $f_{xx}(\omega)$ will be real & have even symmetry about $\omega=0$, We define a mean freq. $\bar{\omega}_0$ by

$$\bar{\omega}_0 = \frac{\int_0^{\infty} \omega f_{xx}(\omega) d\omega}{\int_0^{\infty} f_{xx}(\omega) d\omega}$$

& rms band width by

$$\omega_{rms}^2 = \frac{4 \int_0^{\infty} (\omega - \bar{\omega}_0)^2 \cdot f_{xx}(\omega) d\omega}{\int_0^{\infty} f_{xx}(\omega) d\omega}$$

(6)

Relationship b/w Power Spectrum & Auto Correlation fn.

The Inverse F.T of the power density spectrum is the time avg. of the auto correlation fn. i.e.

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} f_{xx}(\omega) \cdot e^{j\omega t} d\omega = A [R_{xx}(t, t+T)]$$

This expression will now be proved.

→ We know that

$$|x_T(\omega)|^2 = \underline{x_T(\omega) \cdot x_T^*(\omega)}$$

$$f_{xx}(\omega) = \lim_{T \rightarrow \infty} \frac{E[|x_T(\omega)|^2]}{2T}$$

from the def. of $x_T(\omega) = \begin{cases} x(t) & -T < t < T \\ 0 & \text{elsewhere} \end{cases}$

$$f_{xx}(\omega) = \lim_{T \rightarrow \infty} E \left[\frac{1}{2T} \int_{-T}^T \underbrace{x(t_1)}_{x_T(\omega)} \cdot e^{j\omega t_1} dt_1 \cdot \int_{-T}^T \underbrace{x(t_2)}_{x_T^*(\omega)} \cdot e^{-j\omega t_2} dt_2 \right]$$

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T E[x(t_1)x(t_2)] \cdot e^{-j\omega(t_2-t_1)} dt_2 dt_1$$

$$\text{Wk1, } E[x(t_1)x(t_2)] = R_{xx}(t_1, t_2) \quad -T < (t_1, t_2) < T$$

$$\therefore f_{xx}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xx}(t_1, t_2) \cdot e^{-j\omega(t_2-t_1)} dt_1 dt_2 \quad \text{--- (1)}$$

→ Take I.F.T of eqn. (1)

This I.F.T will exist since the power spectrum is real, non-negative & absolutely integrable.

∴ The Inverse F.T is

$$\begin{aligned} & \frac{1}{2\pi} \int_{-\infty}^{\infty} f_{xx}(\omega) \cdot e^{j\omega t} d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xx}(t_1, t_2) \cdot e^{-j\omega(t_2-t_1)} dt_2 dt_1 \cdot e^{j\omega t} d\omega \end{aligned}$$

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xx}(t, t_1) \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{j\omega(T-t_1+t)} d\omega dt_1 dt$$

$\downarrow$
 $2\pi \delta(t-t_1+T)$
 $= 2\pi \delta(t_1-t-T)$

$$\therefore \frac{1}{2\pi} \int_{-\infty}^{\infty} f_{xx}(\omega) d\omega = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xx}(t, t_1) \underbrace{\delta(t_1-t-T)}_{\substack{\text{exists only at} \\ t_1=t+T}} dt_1 dt$$

$$\therefore = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_{xx}(t, t+T) dt$$

$$\therefore \frac{1}{2\pi} \int_{-\infty}^{\infty} f_{xx}(\omega) d\omega = A[R_{xx}(t, t+T)]$$

which indicates the Inverse F.T of the PDS is the time Avg. of the process auto correlation.

$$\Rightarrow f_{xx}(\omega) = \int_{-\infty}^{\infty} A[R_{xx}(t, t+T)] \cdot e^{-j\omega T} dT$$

Show that $f_{xx}(\omega)$ & $A[R_{xx}(t, t+T)]$ form a F.T pair.

$$\therefore A[R_{xx}(t, t+T)] \longleftrightarrow f_{xx}(\omega)$$

* for the important case, where $x(t)$ is atleast WSS.

$A[R_{xx}(t, t+T)] = R_{xx}(T)$ & we get.

$$\boxed{\begin{aligned} \therefore f_{xx}(\omega) &= \int_{-\infty}^{\infty} R_{xx}(T) \cdot e^{-j\omega T} dT \quad \rightarrow \textcircled{A} \\ \textcircled{B} \quad R_{xx}(T) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} f_{xx}(\omega) \cdot e^{j\omega T} d\omega \quad \rightarrow \textcircled{B} \end{aligned}}$$

$$R_{xx}(T) \longleftrightarrow f_{xx}(\omega)$$

The expressions $\textcircled{A}$ & $\textcircled{B}$ are usually called the

* Wiener-Khinchine relations.

8

Find the power spectrum for the RP $x(t) = A \cos(\omega_0 t + \theta)$ at has the auto correlation fn.

$R_{xx}(t) = \frac{A_0^2}{2} \cos \omega_0 t$. where A_0 & ω_0 are constants.

Sol:

$$f_{xx}(\omega) = \text{FT}[R_{xx}(t)]$$

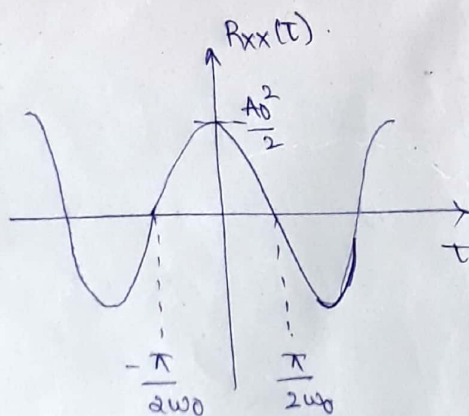
$$\text{We can write as } R_{xx}(t) = \frac{A_0^2}{2} \left[\frac{e^{j\omega_0 t} + e^{-j\omega_0 t}}{2} \right]$$

$$f_{xx}(\omega) = \int_{-\infty}^{\infty} \frac{A_0^2}{4} (e^{j\omega_0 t} + e^{-j\omega_0 t}) e^{j\omega t} dt$$

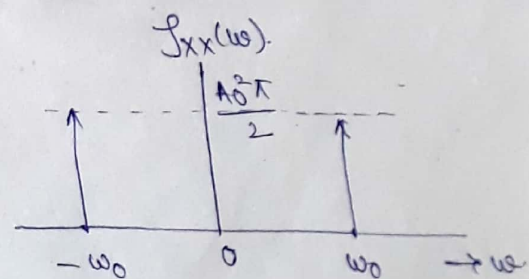
$$= \frac{A_0^2}{4} \left[\int_{-\infty}^{\infty} e^{j(\omega + \omega_0)t} dt + \int_{-\infty}^{\infty} e^{-j(\omega - \omega_0)t} dt \right]$$

$$= \frac{A_0^2}{4} \left[2\pi \delta(\omega + \omega_0) + 2\pi \delta(\omega - \omega_0) \right]$$

$$f_{xx}(\omega) = \frac{A_0^2 \pi}{2} \left[\delta(\omega + \omega_0) + \delta(\omega - \omega_0) \right]$$



(a) auto correlation fn.



(b) power spectrum density.

(9)

For a WSS RP $x(t)$, the auto correlation fn. is given by

$$R_{xx}(\tau) = \begin{cases} A_0 [1 - (|\tau|/T)] & -T \leq \tau \leq T \\ 0 & \text{elsewhere.} \end{cases}$$

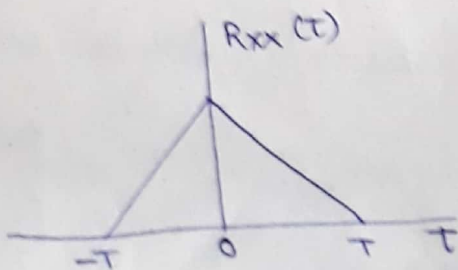
where $T > 0$ & A_0 are constants. Find the power spectrum.

sol:

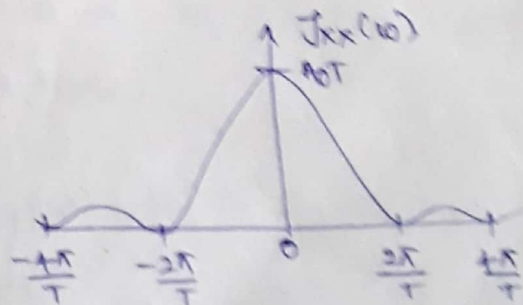
$$\begin{aligned} S_{xx}(\omega) &= \int_{-\infty}^{\infty} R_{xx}(\tau) \cdot e^{-j\omega\tau} d\tau \\ &= A_0 \int_{-T}^0 [1 + \tau/T] e^{-j\omega\tau} d\tau + A_0 \int_0^T [1 - \tau/T] e^{-j\omega\tau} d\tau \end{aligned}$$

$$= A_0 \cdot T \operatorname{sinc}^2\left(\frac{\omega T}{2}\right)$$

FT [Triangular fn] = sinc fn.



a) auto correlation fn.



b) power spectrum fn.

Cross Power Density Spectrum & Its Properties:

consider a real RP $w(t)$ given by the sum of two other real processes $x(t)$ & $y(t)$

$$\therefore w(t) = x(t) + y(t) \quad \text{--- (1)}$$

→ The auto correlation of $w(t)$ is

$$\begin{aligned} R_{ww}(t, t+\tau) &= E[w(t)w(t+\tau)] \\ &= E[\{x(t) + y(t)\}\{x(t+\tau) + y(t+\tau)\}] \\ &= R_{xx}(t, t+\tau) + R_{xy}(t, t+\tau) + R_{yx}(t, t+\tau) \\ &\quad + R_{yy}(t, t+\tau) \quad \text{--- (2)} \end{aligned}$$

→ take time avg. of both sides of eq. (2) & F.T, the resulting expression we have

$$\begin{aligned} J_{ww}(\omega) &= J_{xx}(\omega) + J_{yy}(\omega) + F\{A[R_{xy}(t, t+\tau)]\} \\ &\quad + F\{A[R_{yx}(t, t+\tau)]\} \quad \text{--- (3)} \end{aligned}$$

$J_{xx}(\omega) \rightarrow$ power spectrum of $x(t)$

$J_{yy}(\omega) \rightarrow$ power spectrum of $y(t)$.

& $F\{A[R_{xy}(t, t+\tau)]\}$ & $F\{A[R_{yx}(t, t+\tau)]\}$ are.

Cross power density spectrum.

Cross power density spectrum:

→ For two real RPs $x(t)$ & $y(t)$, define $x_T(t)$ & $y_T(t)$ as ensemble member, we

$$x_T(t) = \begin{cases} x(t) & -T < t < T \\ 0 & \text{elsewhere} \end{cases} \quad \& \quad y_T(t) = \begin{cases} y(t) & -T < t < T \\ 0 & \text{elsewhere} \end{cases}$$

(11)

the F.T.s are denoted by.

$$x_T(t) \longleftrightarrow X_T(\omega)$$

$$y_T(t) \longleftrightarrow Y_T(\omega).$$

→ The cross power $P_{xy}(T)$ in the two processes with in the interval $(-T, T)$ by.

$$\begin{aligned} P_{xy}(T) &= \frac{1}{2T} \int_{-T}^T x_T(t) y_T(t) dt \\ &= \frac{1}{2T} \int_{-T}^T x(t) y(t) dt. \end{aligned}$$

→ using Parseval's theorem, we have.

$$P_{xy}(T) = \frac{1}{2T} \int_{-T}^T x(t) y(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{X_T^*(\omega) Y_T(\omega)}{2T} d\omega \quad \text{--- (1)}$$

→ This cross power is a random quantity, since its value will vary depending on which ensemble member is considered.

→ The Avg. cross power, denoted $\bar{P}_{xy}(T)$, by taking the expected value in eq. (1). The result is.

$$\bar{P}_{xy}(T) = \frac{1}{2T} \int_{-T}^T R_{xy}(t, t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{E[X_T^*(\omega) Y_T(\omega)]}{2T} d\omega$$

→ finally, we form the total avg. cross power P_{xy} by letting $T \rightarrow \infty$:

$$P_{xy} = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_{xy}(t, t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} \frac{E[X_T^*(\omega) Y_T(\omega)]}{2T} d\omega$$

→ The integrand involving ω can be defined as the cross-power density spectrum, i.e.

$$f_{xy}(\omega) = \lim_{T \rightarrow \infty} \frac{E[X_T^*(\omega) Y_T(\omega)]}{2T}$$

The cross power formula is

$$P_{xy} = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xy}(\omega) d\omega$$

→ The ^{other} cross power density spectrum is given by.

$$S_{yx}(\omega) = \lim_{T \rightarrow \infty} \frac{E[Y_T^*(\omega) \cdot X_T(\omega)]}{2T}$$

→ Cross power is given by.

$$P_{yx} = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{yx}(\omega) d\omega = P_{xy}^*$$

→ Total power $P_{xy} + P_{yx}$ can be interpreted as the additional power two processes are capable of generating, over & above their individual powers due to the fact that they are not orthogonal.

Properties of Cross Power Density Spectrum:

1. $S_{xy}(\omega) = S_{yx}(-\omega) = S_{yx}^*(\omega)$
2. $\text{Re}[S_{xy}(\omega)]$ & $\text{Re}[S_{yx}(\omega)]$ are even fns of ω .
3. $\text{Im}[S_{xy}(\omega)]$ & $\text{Im}[S_{yx}(\omega)]$ are odd fns of ω .
4. $S_{xy}(\omega) = 0$ & $S_{yx}(\omega) = 0$ if $x(t)$ & $y(t)$ are orthogonal.
5. If $x(t)$ & $y(t)$ are uncorrelated & have constant means $\bar{x}$ & $\bar{y}$.

$$S_{xy}(\omega) = S_{yx}(\omega) = 2\pi \bar{x} \bar{y} \delta(\omega)$$

$$A[R_{xy}(t, t+\tau)] \longleftrightarrow S_{xy}(\omega)$$

$$A[R_{yx}(t, t+\tau)] \longleftrightarrow S_{yx}(\omega)$$

For the case, of jointly WSS processes,

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

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

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

$$R_{yx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{yx}(\omega) \cdot e^{j\omega\tau} d\omega$$

prob: For the given cross power spectrum defined as

$$S_{xy}(\omega) = \begin{cases} a + j\frac{b\omega}{W} & -W < \omega < W \\ 0 & \text{elsewhere} \end{cases}$$

find the cross correlation fn.

Sol:

$$R_{xy}(\tau) = \frac{1}{2\pi} \int_{-W}^W \left(a + j\frac{b\omega}{W} \right) \cdot e^{j\omega\tau} d\omega$$

$$= \frac{a}{2\pi} \int_{-W}^W e^{j\omega\tau} d\omega + j\frac{b}{2\pi W} \int_{-W}^W \omega e^{j\omega\tau} d\omega$$

$$= \frac{a}{2\pi} \left[\frac{e^{j\omega\tau}}{j\tau} \right]_{-W}^W + j\frac{b}{2\pi W} \left\{ e^{j\omega\tau} \left[\frac{\omega}{j\tau} - \frac{1}{(j\tau)^2} \right] \right\}_{-W}^W$$

$$= \frac{1}{\pi W \tau^2} \left[(aW\tau - b) \sin(W\tau) + bW\tau \cos(W\tau) \right]$$

(13)

Relationship b/w Cross power spectrum and cross correlation fn:

We have to show that

$$S_{xy}(\omega) = \int_{-\infty}^{\infty} \left\{ \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_{xy}(t, t_1) dt \right\} e^{-j\omega t} dt$$

→ We know that

$$x_T(\omega) = \int_{-T}^T x(t) \cdot e^{-j\omega t} dt$$

$$y_T(\omega) = \int_{-T}^T y(t_1) \cdot e^{-j\omega t_1} dt_1$$

Consider

→ $x_T^*(\omega) \cdot y_T(\omega) = \int_{-T}^T x(t) \cdot e^{j\omega t} dt \cdot \int_{-T}^T y(t_1) \cdot e^{-j\omega t_1} dt_1 \quad \text{--- (1)}$

→ The cross power density spectrum, is given by

$$S_{xy}(\omega) = \lim_{T \rightarrow \infty} \frac{E [x_T^*(\omega) \cdot y_T(\omega)]}{2T}$$

from eq. (1)

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} E \left[\int_{-T}^T x(t) \cdot e^{j\omega t} dt \cdot \int_{-T}^T y(t_1) \cdot e^{-j\omega t_1} dt_1 \right]$$

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xy}(t, t_1) \cdot e^{-j\omega(t_1 - t)} dt \cdot dt_1$$

→ Apply Inverse F.T both sides,

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xy}(\omega) e^{j\omega t} d\omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xy}(t, t_1) \cdot e^{-j\omega(t_1 - t)} dt \cdot dt_1 \cdot e^{j\omega t} d\omega$$

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xy}(t, t_1) \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{j\omega(T - t_1 + t)} d\omega dt_1 dt$$

$$= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \int_{-T}^T R_{xy}(t, t_1) \cdot \delta(t_1 - T - t) dt_1 dt$$

↓
defined only at $t_1 = t + T$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} J_{xy}(\omega) \cdot e^{j\omega t} d\omega = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_{xy}(t, t+\tau) dt \quad \text{--- (2)}$$

which is valid for $-T < t+\tau < T$.

$\therefore$ eq. (2) indicates that the cross power spectrum and the time avg. cross correlation fn. form a F.T pair.

$\rightarrow$ For jointly WSS, the cross correlation fn. $R_{xy}(\tau)$ can be found from $J_{xy}(\omega)$ since its time average is just $R_{xy}(\tau)$.

Problem: Let the cross-correlation fn. of two processes $x(t)$ and $y(t)$ be

$$R_{xy}(t, t+\tau) = \frac{AB}{2} \left\{ \sin(\omega_0 \tau) + \cos[\omega_0(2t+\tau)] \right\}$$

where A, B and ω_0 are constants. Find the cross-power spectrum.

Sol.

$$J_{xy}(\omega) = \text{F.T} \left\{ A [R_{xy}(t, t+\tau)] \right\}$$

$$\begin{aligned} A [R_{xy}(t, t+\tau)] &= \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T R_{xy}(t, t+\tau) dt \\ &= \frac{AB}{2} \sin(\omega_0 \tau) + \frac{AB}{2} \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T \underbrace{\cos(\omega_0(2t+\tau))}_{\text{0}} dt \\ &= \frac{AB}{2} \sin(\omega_0 \tau) \end{aligned}$$

$$\therefore J_{xy}(\omega) = \text{F.T} \left\{ \frac{AB}{2} \sin \omega_0 \tau \right\}$$

$$= -j \frac{\pi AB}{2} \left[\delta(\omega - \omega_0) - \delta(\omega + \omega_0) \right]$$

Spectral Characteristics of System Response

(15)

power Density spectrum of Response:

→ The PSD $S_{YY}(\omega)$ of the response of a linear time-invariant system having a transfer fn. $H(\omega)$ is given by.

$$S_{YY}(\omega) = S_{XX}(\omega) \cdot |H(\omega)|^2 \quad \text{--- (1)}$$

where $S_{XX}(\omega)$ is the PSD of input $x(t)$.

$|H(\omega)|^2$ is the power transfer fn. of the system.

→ WKT,

$$S_{YY}(\omega) = \int_{-\infty}^{\infty} R_{YY}(\tau) \cdot e^{-j\omega\tau} d\tau \quad \text{--- (2)}$$

$$\& R_{YY}(\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{XX}(\tau + \tau_1 - \tau_2) h(\tau_1) h(\tau_2) d\tau_1 d\tau_2 \quad \text{--- (3)}$$

Substitute eq. (3) in eq. (2)

We have

$$S_{YY}(\omega) = \int_{-\infty}^{\infty} h(\tau_1) \int_{-\infty}^{\infty} h(\tau_2) \int_{-\infty}^{\infty} R_{XX}(\tau + \tau_1 - \tau_2) \cdot e^{-j\omega\tau} d\tau_2 d\tau_1$$

$$\text{let } \tau = \tau + \tau_1 - \tau_2 \Rightarrow d\tau = d\tau$$

$$\therefore S_{YY}(\omega) = \int_{-\infty}^{\infty} h(\tau_1) e^{j\omega\tau_1} d\tau_1 \int_{-\infty}^{\infty} h(\tau_2) \cdot e^{-j\omega\tau_2} d\tau_2 \int_{-\infty}^{\infty} R_{XX}(\tau) \cdot e^{-j\omega\tau} d\tau$$

$$S_{YY}(\omega) = H^*(\omega) \cdot H(\omega) \cdot S_{XX}(\omega) = \underline{S_{XX}(\omega) \cdot |H(\omega)|^2}$$

avg. power is $P_{YY} = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{YY}(\omega) \cdot |H(\omega)|^2 d\omega$

Cross Power Density Spectrum of Input & Output:

The f.t.s of the cross correlation f.d. may be written as

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

$$S_{yx}(\omega) = S_{xx}(\omega) \cdot H(-\omega)$$