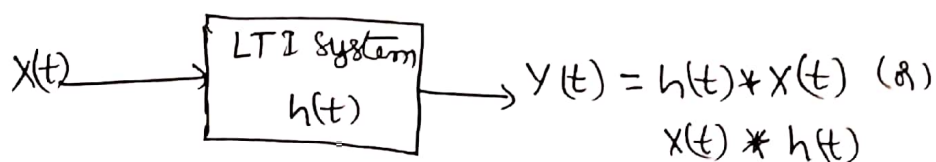


Linear Systems With Random InputsResponse of Linear Systems for random inputs: -

Consider a continuous LTI system with impulse response $h(t)$. Assume that the system is always causal and stable. When a continuous time random process $X(t)$ is applied to LTI system, the output response is also a continuous time random process $Y(t)$. If the random processes X and Y are discrete time signals, then the linear system is called a discrete time system.

System Response: - Let a random process $X(t)$ be applied to a continuous LTI system whose impulse response is $h(t)$, then the output response $Y(t)$ is also a random process. It can be expressed by the convolution integral, $Y(t) = h(t) * X(t)$.



i.e., the output response is $Y(t) = \int_{-\infty}^{\infty} h(\tau) \cdot X(t-\tau) d\tau$.

Mean and Mean-Squared Value of System Response: -

Consider the random process $X(t)$ is Wide Sense Stationary (WSS) process.

Mean Value of system response $= E[Y(t)] = \bar{Y}$

$$\begin{aligned}
 \bar{Y} &= E[Y(t)] = E[h(t) * X(t)] \\
 &= E\left[\int_{-\infty}^{\infty} h(\tau) \cdot X(t-\tau) d\tau\right] \\
 &= \int_{-\infty}^{\infty} h(\tau) \cdot E[X(t-\tau)] \cdot d\tau
 \end{aligned}$$

But $E[X(t-\tau)] = \bar{X} = \text{constant}$, since $X(t)$ is WSS.

$$\therefore \bar{Y} = E[Y(t)] = \bar{X} \cdot \int_{-\infty}^{\infty} h(\tau) \cdot d\tau$$

Also, if $H(\omega)$ is the Fourier transform of $h(t)$, then

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

at $\omega=0$, $H(0) = \int_{-\infty}^{\infty} h(t) dt$ is called the zero-frequency response of the system. Substituting this we get

$$\boxed{\bar{Y} = E[Y(t)] = \bar{X} H(0) \text{ is constant.}}$$

Thus the mean value of the system response (or) output response $Y(t)$ of a WSS random process is equal to the product of the mean value of the input process and the zero-frequency response of the system.

Mean Squared Value of system response $= E[Y^2(t)]$

$$\begin{aligned}
 E[Y^2(t)] &= E[(h(t) * X(t))^2] \\
 &= E[(h(t) * X(t)) \{h(t) * X(t)\}]
 \end{aligned}$$

$$\begin{aligned}
 E[Y^2(t)] &= E\left[\int_{-\infty}^{\infty} h(\sigma_1) X(t-\sigma_1) d\sigma_1 \cdot \int_{-\infty}^{\infty} h(\sigma_2) X(t-\sigma_2) d\sigma_2\right] \\
 &= E\left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} X(t-\sigma_1) \cdot X(t-\sigma_2) \cdot h(\sigma_1) \cdot h(\sigma_2) \cdot d\sigma_1 \cdot d\sigma_2\right] \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[X(t-\sigma_1) \cdot X(t-\sigma_2)] \cdot h(\sigma_1) \cdot h(\sigma_2) \cdot d\sigma_1 \cdot d\sigma_2
 \end{aligned}$$

Where σ_1 and σ_2 are shifts in time intervals.

If input $X(t)$ is a WSS random process, then

$$E[X(t-\sigma_1) X(t-\sigma_2)] = R_{XX}(\sigma_1 - \sigma_2)$$

$$\therefore E[Y^2(t)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{XX}(\sigma_1 - \sigma_2) \cdot h(\sigma_1) \cdot h(\sigma_2) \cdot d\sigma_1 \cdot d\sigma_2$$

This expression is independent of time t . And it represents the output power.

Autocorrelation function of system response :—

The autocorrelation function of $Y(t)$ is

$$\begin{aligned}
 R_{YY}(t_1, t_2) &= E[Y(t_1) Y(t_2)] \\
 &= E\left[\{h(t_1) * X(t_1)\} \{h(t_2) * X(t_2)\}\right] \\
 &= E\left[\int_{-\infty}^{\infty} h(\sigma_1) \cdot X(t_1 - \sigma_1) d\sigma_1 \cdot \int_{-\infty}^{\infty} h(\sigma_2) X(t_2 - \sigma_2) d\sigma_2\right] \\
 &= E\left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} X(t_1 - \sigma_1) X(t_2 - \sigma_2) h(\sigma_1) h(\sigma_2) d\sigma_1 d\sigma_2\right] \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[X(t_1 - \sigma_1) X(t_2 - \sigma_2)] \cdot h(\sigma_1) h(\sigma_2) \cdot d\sigma_1 d\sigma_2
 \end{aligned}$$

We know that $E[X(t_1 - \tau_1) X(t_2 - \tau_2)] = R_{XX}(t_2 - t_1 + \tau_1 - \tau_2)$

If input $x(t)$ is WSS-RP, let the time difference $\tau = t_2 - t_1$, and $t = t_1$, then $E[X(t - \tau_1) X(t + \tau - \tau_2)] = R_{XX}(\tau + \tau_1 - \tau_2)$

$$\therefore R_{YY}(t, t + \tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{XX}(\tau + \tau_1 - \tau_2) h(\tau_1) h(\tau_2) \cdot d\tau_1 d\tau_2$$

$$\boxed{R_{YY}(\tau) = R_{XX}(\tau) * h(-\tau) * h(\tau)}$$

Two facts results from above expressions

1. $Y(t)$ is WSS, if $X(t)$ is WSS because $R_{YY}(\tau)$ does not depend on t and $E[Y(t)]$ is a constant.
2. $R_{YY}(\tau)$ is the twofold convolution of the input autocorrelation function with the network's impulse response.

Cross-correlation functions between input and output of the system : —

The cross-correlation function of $X(t)$ and $Y(t)$ is

$$\begin{aligned} R_{XY}(t, t + \tau) &= E[X(t) Y(t + \tau)] \\ &= E\left[X(t) \int_{-\infty}^{\infty} h(\tau_1) X(t + \tau - \tau_1) d\tau_1\right] \\ &= \int_{-\infty}^{\infty} E[X(t) X(t + \tau - \tau_1)] \cdot h(\tau_1) \cdot d\tau_1, \end{aligned}$$

If $X(t)$ is WSS, then

$$R_{XY}(\tau) = \int_{-\infty}^{\infty} R_{XX}(\tau - \tau_1) \cdot h(\tau_1) d\tau_1, \text{ which is the convolution } R_{XX}(\tau) \text{ with } h(\tau)$$

$$\boxed{R_{XY}(\tau) = R_{XX}(\tau) * h(\tau)}$$

(5)

A similar development shows that

$$R_{yx}(\tau) = \int_{-\infty}^{\infty} R_{xx}(\tau - \tau_1) h(\tau_1) d\tau_1$$

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

It is clear that the cross-correlation functions depend on τ and not on absolute time t . As a consequence of this fact $x(t)$ and $y(t)$ are jointly WSS, if $x(t)$ is WSS, because already $y(t)$ to be WSS.

Autocorrelation function and cross-correlation functions are seen to be related by

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

$$(a) \quad \boxed{R_{yy}(\tau) = R_{xy}(\tau) * h(\tau) \quad (b) \quad h(-\tau) * R_{xy}(\tau)}$$

Similarly $R_{yy}(\tau) = \int_{-\infty}^{\infty} R_{yx}(\tau - \tau_2) \cdot h(\tau_2) d\tau_2$

$$(a) \quad \boxed{R_{yy}(\tau) = R_{yx}(\tau) * h(\tau) \quad \& \quad h(\tau) * R_{yx}(\tau)}$$

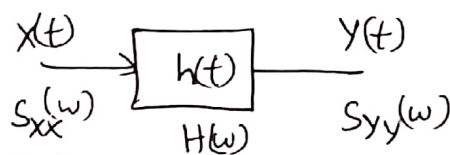
Spectral characteristics of System's Response : —

Consider that the random process $x(t)$ is a WSS-RP with the autocorrelation function $R_{xx}(\tau)$ applied through an LTI system. The o/p response $y(t)$ is also a WSS-RP and the processes $x(t)$ and $y(t)$ are jointly WSS. Now, we can obtain power spectral characteristics of the o/p process $y(t)$ by taking the Fourier transform of the correlation functions.

Power density spectrum of response : —

Consider that a random process $x(t)$ is applied on an LTI system having a transfer function $H(\omega)$. The o/p response is $y(t)$. If the power spectrum of the i/p process is $S_{xx}(\omega)$, then the power spectrum of the o/p response is given by

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



Proof:- Let $R_{yy}(\tau)$ be the autocorrelation of the o/p response $y(t)$. Then the power spectrum of the response is the F.T of $R_{yy}(\tau)$.

$$\therefore S_{yy}(\omega) = F[R_{yy}(\tau)]$$

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

$$\text{we know that } R_{yy}(\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(\tau + \tau_1 - \tau_2) \cdot h(\tau_1) h(\tau_2) d\tau_1 d\tau_2$$

$$\text{then } S_{yy}(\omega) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(\tau + \tau_1 - \tau_2) \cdot h(\tau_1) h(\tau_2) \cdot e^{-j\omega\tau} d\tau d\tau_1 d\tau_2$$

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

$$\therefore S_{yy}(\omega) = \int_{-\infty}^{\infty} h(\tau_1) \cdot e^{-j\omega\tau_1} d\tau_1 \cdot \int_{-\infty}^{\infty} h(\tau_2) \cdot e^{-j\omega\tau_2} d\tau_2 \cdot \int_{-\infty}^{\infty} R_{xx}(t) e^{-j\omega t} dt$$

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

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

##

Cross-power density spectrum of input and output:—

The cross-power density spectrum of i/p & o/p is

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

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

Proof:-

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

$$\text{We know that } R_{xy}(\tau) = \int_{-\infty}^{\infty} R_{xx}(\sigma - \tau_1) \cdot h(\tau_1) d\tau_1$$

$$\therefore S_{xy}(\omega) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(\sigma - \tau_1) \cdot h(\tau_1) \cdot e^{-j\omega\tau} \cdot d\sigma \cdot d\tau_1$$

$$\text{Let } \sigma - \tau_1 = t \Rightarrow d\sigma = dt \Rightarrow \tau = t + \tau_1$$

$$S_{xy}(\omega) = \int_{-\infty}^{\infty} h(\tau_1) \cdot e^{-j\omega\tau_1} \cdot d\tau_1 \cdot \int_{-\infty}^{\infty} R_{xx}(t) \cdot e^{-j\omega t} \cdot dt$$

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

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

slly

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

Ex 1:- A random process $x(t)$ is applied as input to a system whose impulse response is $h(t) = 3 u(t) \cdot t \cdot \exp(-8t)$. If $E[x(t)] = 2$, what is the mean value of the system response $y(t)$?

Sol:-

Mean value of the system response is $E[y(t)] = \bar{y}$

$$\begin{aligned}
 E[y(t)] &= E[h(t) * x(t)] \\
 &= E\left[\int_{-\infty}^{\infty} h(\tau) x(t-\tau) d\tau\right] \\
 &= \int_{-\infty}^{\infty} h(\tau) E[x(t-\tau)] \cdot d\tau \quad \because E[x(t-\tau)] = 2 \\
 &= 2 \int_{-\infty}^{\infty} 3 \cdot u(\tau) \cdot \tau \cdot e^{-8\tau} \cdot d\tau \\
 &= 6 \int_0^{\infty} \tau \cdot e^{-8\tau} d\tau \\
 &= 6 \left[\tau \cdot \frac{e^{-8\tau}}{-8} - 2\tau \cdot \frac{e^{-8\tau}}{(8)^2} + 2 \cdot \frac{e^{-8\tau}}{(8)^3} \right]_0^{\infty} \\
 &= 6 \left[0 - 0 + 2 \cdot \frac{(e^{-\infty} - e^0)}{(8)^3} \right] \\
 &= 6 \left(\frac{1}{8^3} \right) = \frac{3}{256} = \text{constant} \\
 &= =
 \end{aligned}$$

Ex 2:- Let $x(t)$ be a zero-mean WSS process with $R_{xx}(\tau) = e^{-|\tau|}$.

$x(t)$ is input to an LTI system with $|H(\omega)| = \begin{cases} \sqrt{1+\omega^2}, & |\omega| < 4\pi \\ 0, & \text{otherwise} \end{cases}$

Let $y(t)$ be the output.

a) find $E[y(t)]$, b) $E[y^2(t)]$, c) $R_{yy}(\tau)$

Sol: Note that $x(t)$ is WSS, $x(t)$ & $y(t)$ are jointly WSS, and therefore $y(t)$ is WSS.

$$\begin{aligned} \text{a) } E[y(t)] &= \bar{x} \cdot H(0) \\ &= 0 \cdot 1 \quad \because \bar{x} = 0, \quad H(\omega) = \sqrt{1 + \omega^2} \\ &= 0 \quad H(0) = 1 \end{aligned}$$

$$\text{c) } R_{yy}(\tau) = F^{-1}[S_{yy}(\omega)]$$

$$\text{where } S_{yy}(\omega) = S_{xx}(\omega) \cdot |H(\omega)|^2$$

$$\begin{aligned} S_{xx}(\omega) &= F[R_{xx}(\tau)] = \int_{-\infty}^{\infty} R_{xx}(\tau) \cdot e^{-j\omega\tau} d\tau \\ &= \int_{-\infty}^{\infty} e^{-|\tau|} \cdot e^{-j\omega\tau} d\tau \\ &= \int_{-\infty}^0 e^{\tau} \cdot e^{-j\omega\tau} d\tau + \int_0^{\infty} e^{-\tau} \cdot e^{-j\omega\tau} d\tau \\ &= \int_{-\infty}^0 e^{\tau(1-j\omega)} d\tau + \int_0^{\infty} e^{-\tau(1+j\omega)} d\tau \\ &= \left[\frac{e^{\tau(1-j\omega)}}{1-j\omega} \right]_{-\infty}^0 + \left[\frac{e^{-\tau(1+j\omega)}}{-(1+j\omega)} \right]_0^{\infty} \\ &= \frac{1}{1-j\omega} + \frac{1}{1+j\omega} = \frac{2}{1+\omega^2} \end{aligned}$$

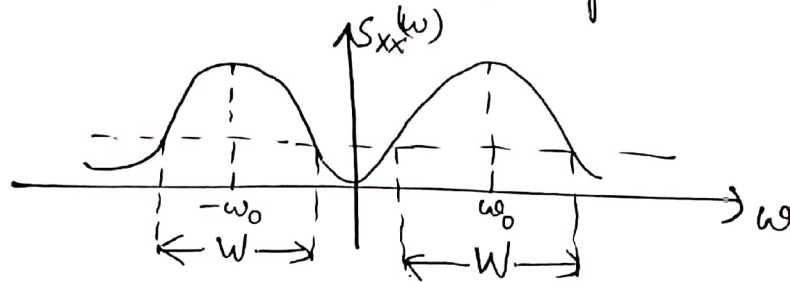
$$\therefore S_{yy}(\omega) = \frac{2}{1+\omega^2} \times 1+\omega^2 = 2, \quad |\omega| < 4\pi$$

$$R_{yy}(\tau) = F^{-1}[S_{yy}(\omega)] = \frac{1}{2\pi} \int_{-4\pi}^{4\pi} 2 \cdot e^{j\omega\tau} d\omega = 8 \operatorname{sinc}(4\pi\tau)$$

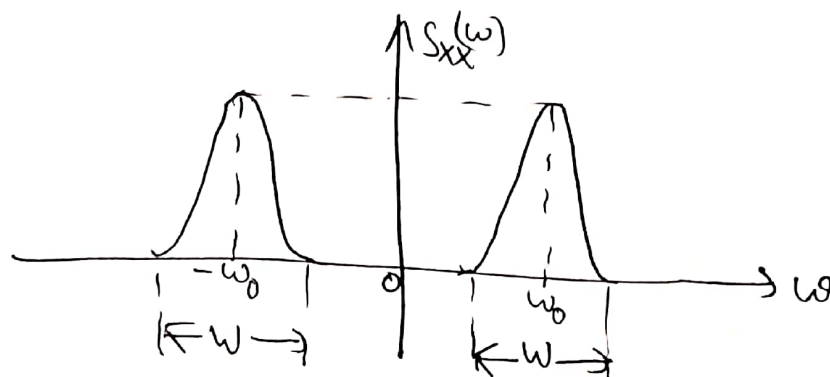
$$\text{b) } E[y^2(t)] = R_{yy}(0) = 8$$

Band pass, Band-Limited and Narrowband processes and their Properties :-

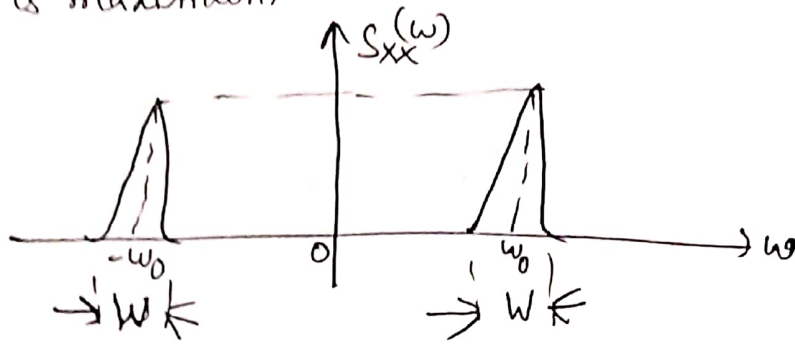
Band pass random processes :- A random process $x(t)$ is called a band pass process, if its power spectral density $S_{xx}(\omega)$ has significant components within a bandwidth 'W' that does not include $\omega=0$. But in practice, the spectrum may have a small amount of power spectrum at $\omega=0$. The spectrum components outside the band 'W' are very small and can be neglected.



Band-limited random process :- A random process $x(t)$ is said to be band-limited, if its power spectrum components are zero outside the frequency band of width 'W' that does not include $\omega=0$. The power density spectrum of the band-limited band pass is



Narrowband random processes:— A band limited random process is said to be a narrowband process, if the bandwidth 'W' is very small compared to the band centre frequency, i.e. $W \ll \omega_0$ where W = bandwidth and ω_0 is the frequency at which the power spectrum is maximum.



Representation of a narrowband process:— For any arbitrary WSS random process $N(t)$, the quadrature form of narrowband process can be represented as $N(t) = X(t) \cos \omega_0 t - Y(t) \sin \omega_0 t$. where $X(t)$ and $Y(t)$ are called the in-phase & quadrature phase components of $N(t)$.

They can be expressed as

$$X(t) = A(t) \cdot \cos[\theta(t)]$$

$$Y(t) = A(t) \cdot \sin[\theta(t)]$$

and the relationship between the processes $A(t)$ & $\theta(t)$ are given by

$$A(t) = \sqrt{X^2(t) + Y^2(t)} \quad \&$$

$$\theta(t) = \tan^{-1} \left(\frac{Y(t)}{X(t)} \right)$$

Properties of Band-limited Random processes:

Let $N(t)$ be any band-limited WSS-RP with zero mean value and a power spectral density, $S_{NN}(\omega)$. If the RP is represented by

$$N(t) = X(t) \cos \omega_0 t - Y(t) \sin \omega_0 t.$$

1. If $N(t)$ is WSS, then $X(t)$ & $Y(t)$ are jointly WSS.
2. If $N(t)$ has zero mean. i.e. $E[N(t)] = 0$, then $E[X(t)] = E[Y(t)] = 0$.
3. The mean-square values of the processes are equal
i.e. $E[N^2(t)] = E[X^2(t)] = E[Y^2(t)]$.
4. Both processes $X(t)$ & $Y(t)$ have the same autocorrelation functions.
i.e. $R_{XX}(\tau) = R_{YY}(\tau)$

5. The cross-correlation functions of $X(t)$ & $Y(t)$ are given by
 $R_{YX}(\tau) = -R_{XY}(\tau)$. If the processes are orthogonal, then
 $R_{XY}(\tau) = R_{YX}(\tau) = 0$

6. Both $X(t)$ & $Y(t)$ have the same power spectral densities.

$$S_{YY}(\omega) = S_{XX}(\omega) = \begin{cases} S_N(\omega - \omega_0) + S_N(\omega + \omega_0) & , |\omega| \leq \omega_0 \\ 0 & , \text{otherwise} \end{cases}$$

7. The cross-power spectrums are $S_{XY}(\omega) = -S_{YX}(\omega)$

8. If $N(t)$ is a Gaussian-RP, then $X(t)$ & $Y(t)$ are jointly Gaussian

9. The relationship between autocorrelation & power spectrum $S_{NN}(\omega)$ is

$$R_{XX}(\tau) = \frac{1}{\pi} \int_0^\infty S_{NN}(\omega) \cdot \cos[(\omega - \omega_0)\tau] d\omega = R_{YY}(\tau)$$

$$R_{XY}(\tau) = \frac{1}{\pi} \int_0^\infty S_{NN}(\omega) \sin[(\omega - \omega_0)\tau] d\omega = -R_{YX}(\tau)$$

10. If $N(t)$ is zero-mean Gaussian and its PSD, $S_{NN}(\omega)$ is symmetric about $\pm \omega_0$, then $X(t)$ & $Y(t)$ are S.I.