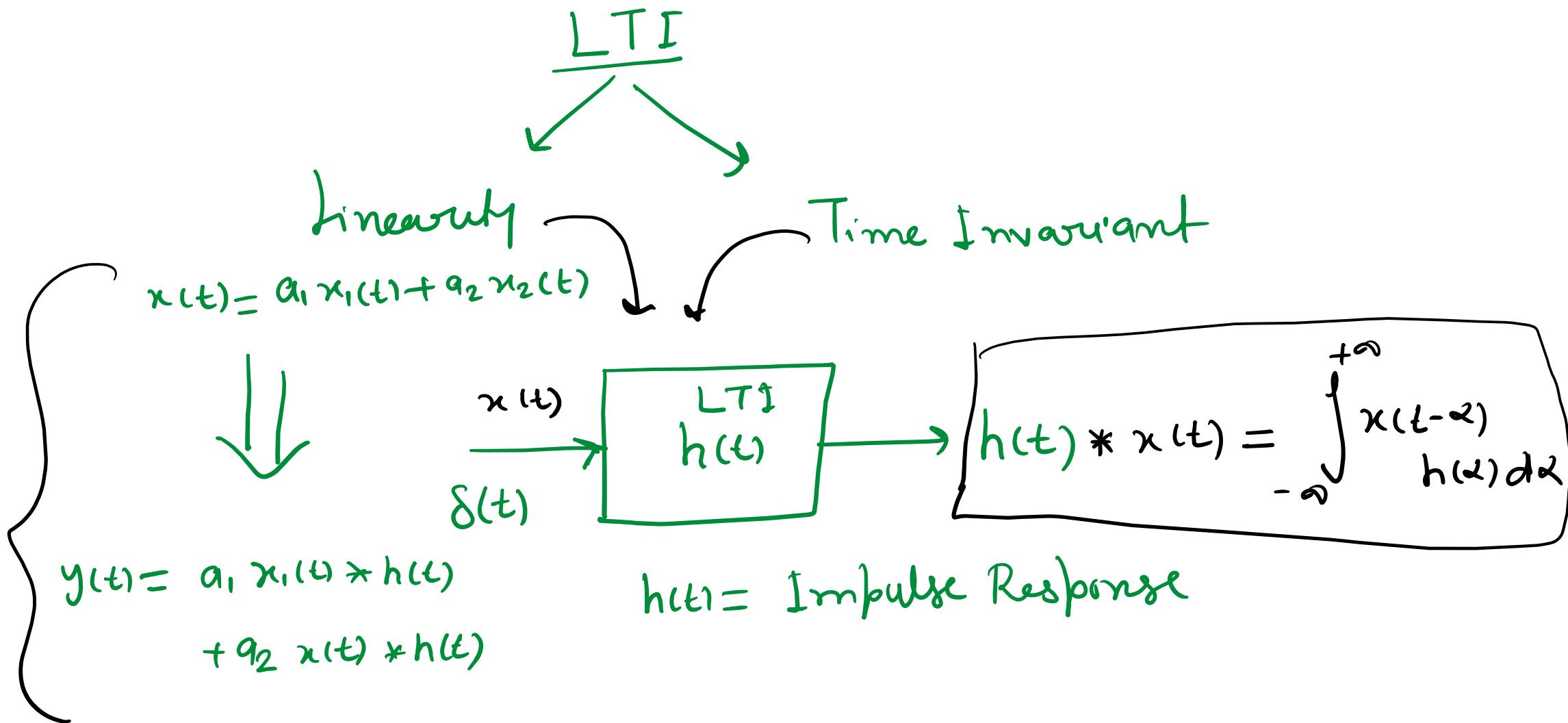


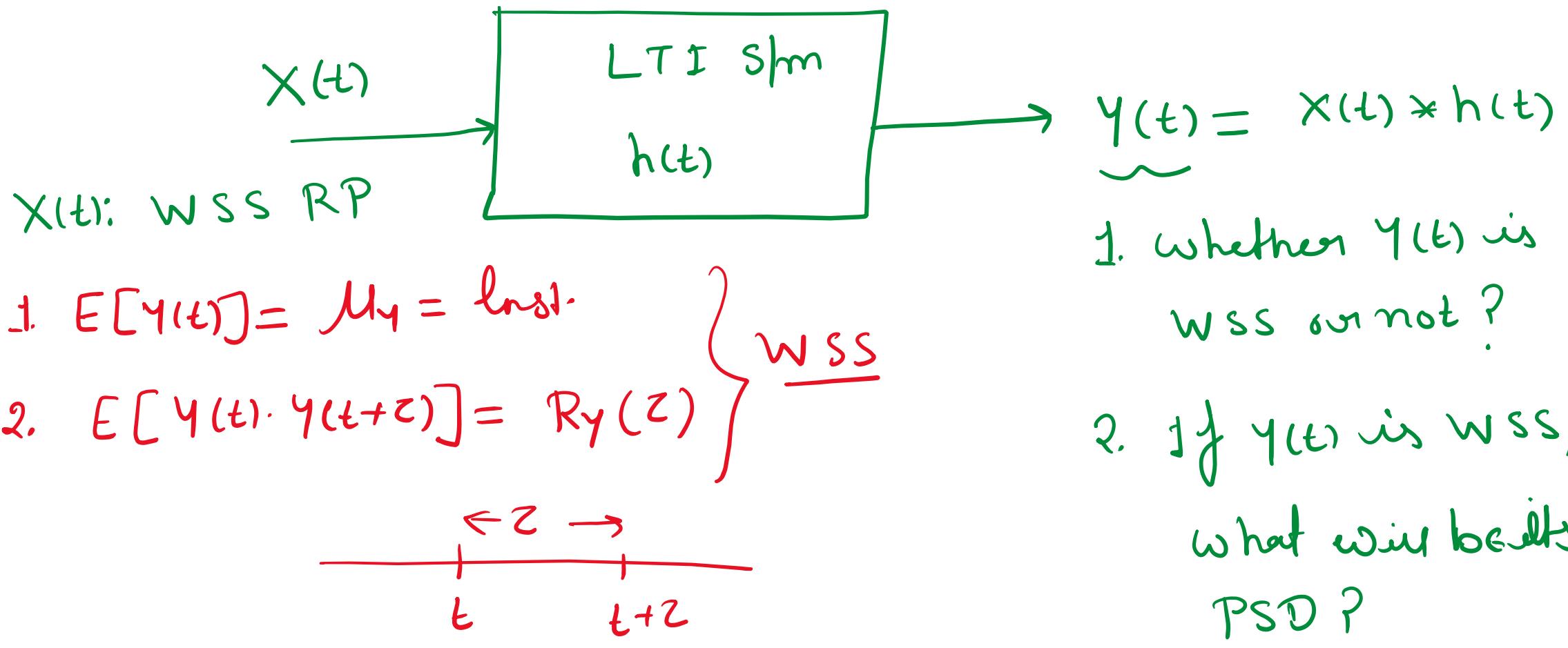
MA 203

Transmission of WSS RP Through LTI System

Liner Time Invariant (LTI) System:



Transmission of WSS random process through LTI system:



Mean:-

$$\mu_y(t) = E[y(t)] = E[x(t) * h(t)]$$

$$= E \left[\int_{-\infty}^{+\infty} x(t-\omega) \underbrace{h(\omega)}_{\text{constant}} d\omega \right]$$

$$= \int_{-\infty}^{+\infty} \underbrace{E[x(t-\omega)]}_{x(t) \text{ is WSS}} h(\omega) d\omega$$

$$E[x(t-\omega)] = \mu_x = \text{const}$$

$$= \int_{-\infty}^{+\infty} \mu_x h(\omega) d\omega$$

It is NOT the function of time
instant $t \Rightarrow$ const

Autocorrelation:

$$y(t) = \int_{-\infty}^{+\infty} x(t-\alpha) h(\alpha) d\alpha = x(t) * h(t)$$


$$y(t+z) = \int_{-\infty}^{+\infty} x(t+z-\beta) h(\beta) d\beta$$

$$\begin{aligned} R_y(t, t+z) &= E \left[y(t) \cdot y(t+z) \right] = E \left[\left\{ \int_{-\infty}^{+\infty} x(t-\alpha) h(\alpha) d\alpha \right\} \left\{ \int_{-\infty}^{+\infty} x(t+z-\beta) h(\beta) d\beta \right\} \right] \\ &= E \left[\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x(t-\alpha) x(t+z-\beta) h(\alpha) h(\beta) d\alpha d\beta \right] \end{aligned}$$

$$E[Y(t)Y(t+\tau)] = E\left[\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \{x(t-\alpha) \underbrace{x(t+\tau-\beta)}_{t_2}\} h(\alpha) \cdot h(\beta) d\alpha d\beta\right]$$

$x(t)$ is WSS

$$E[x(t_1)x(t_2)] = R_x(t_2-t_1)$$

$$t_1 = t - \alpha$$

$$t_2 = t + \tau - \beta.$$

$$t_2 - t_1 = t + \tau - \beta - (t - \alpha)$$

$$= \underline{\tau + \alpha - \beta}$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} E\left\{ \underbrace{x(t-\alpha)}_{t_1} \underbrace{x(t+\tau-\beta)}_{t_2} \right\} h(\alpha) h(\beta) d\alpha d\beta.$$

$$\hookrightarrow R_x(t_2 - t_1) = R_x(\tau + \alpha - \beta)$$

$$E[Y(t)Y(t+\tau)] = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} R_x(\tau - \beta + \alpha) h(\alpha) h(\beta) d\alpha d\beta$$



This part depends only on time shift τ

\Rightarrow It does not depend on time t .

\Rightarrow Therefore, $Y(t)$ is a WSS RP.

\Rightarrow 1. $E[Y(t)] = \mu_y = \text{const.} \Rightarrow Y(t) \text{ is a WSS R.P.}$

2. $E[Y(t)Y(t+\tau)] = \underline{R_y(\tau)}$



$$R_y(z) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} R_x(z - \beta + \alpha) h(\beta) \cdot h(\alpha) d\alpha d\beta$$

$$\text{Put } \bar{\alpha} = -\alpha$$

$$d\bar{\alpha} = -d\alpha$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} R_x(z - \beta - \bar{\alpha}) h(\beta) \underbrace{h(-\bar{\alpha})}_{\text{Put } \bar{h}(\bar{\alpha}) = h(-\bar{\alpha})} d\beta d(+\bar{\alpha})$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} R_x(z - \underline{\beta} - \bar{\alpha}) \underline{h(\beta)} \underline{\bar{h}(\bar{\alpha})} d\beta d\bar{\alpha} \quad \checkmark$$

$$R_y(z) = \underline{R_x(z) * h(z) * \bar{h}(z)}$$

$$\left\{ \begin{array}{l} \because h(-\omega) \\ = \underline{\bar{h}(\omega)} \end{array} \right.$$

$$R_y(z) = R_x(z) * h(z) * h(-z) \quad \textcircled{1}$$

Convolution

Fourier transform of $\textcircled{1}$

$$\begin{aligned} S_y(f) &= \int_{-\infty}^{+\infty} R_y(z) e^{-j2\pi fz} dz \\ &= \mathcal{F.T.}(R_x(z)) \end{aligned}$$

$$\text{F.T.} \left\{ R_y(z) \right\} = \text{F.T.} \left\{ R_x(z) * h(z) * h(-z) \right\}$$

using Property of F.T.

$$\Rightarrow S_y(f) = \text{F.T.} \left\{ R_x(z) \right\} \cdot \text{F.T.} \left\{ h(z) \right\} \text{F.T.} \left\{ h(-z) \right\}$$

$$\Rightarrow S_y(f) = S_x(f) \cdot \underline{H(f)} \cdot \underline{H^*(f)}$$

$$S_y(f) = S_x(f) \cdot |H(f)|^2$$

$$H(f) = \int_{-\infty}^{+\infty} h(t) e^{-j2\pi ft} dt$$

Let $\underline{h(t)} \leftrightarrow \underline{H(f)}$

$\underline{h(-z)} \leftrightarrow \underline{H^*(f)}$

PSD of R.P.

$y(t)$
G/P PSD

$$S_y(f) = \underline{S_x(f)} \cdot \underline{|H(f)|^2}$$

F.T. of $h(t)$

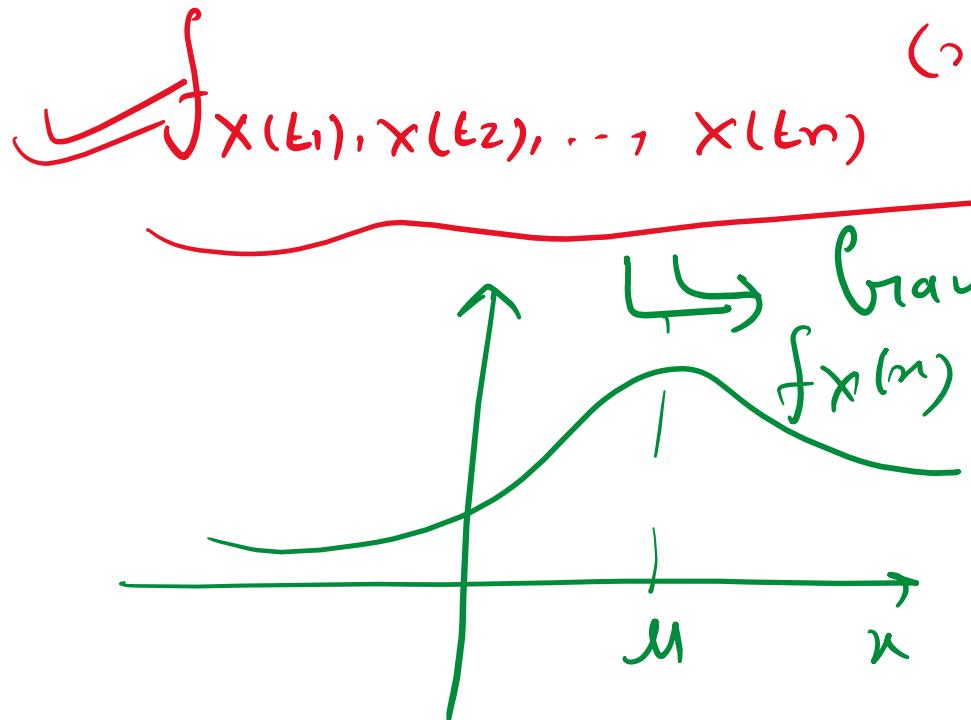
PSD of R.P. $x(t)$
J/P PSD

Gaussian Random Process:

Suppose $\underline{X(t)}$

$$t_1 \quad t_2 \quad \dots \quad t_n$$

$$X(t_1) \quad X(t_2) \quad \dots \quad X(t_n)$$



$$(x_1, x_2, \dots, x_n)$$

Gaussian R.V. are following Gaussian distribution

$\Rightarrow X(t)$ is a Gaussian R.P., then each R.V. $X(t_1), X(t_2), \dots, X(t_n)$ must be Gaussian.

White Noise:- A R.P. is white if it is WSS and

(white R.P.)

$$R_x(z) = \frac{\eta}{2} \underline{\delta(z)}$$

White Random Process

1. It is WSS

$$\rightarrow E[x] = \mu_x = \text{const.}$$

$$\rightarrow E[x(t)x(t+z)] = R_x(z)$$

2. $R_x(z) = \frac{\eta}{2} \underline{\delta(t)} ; \eta = \text{const.}$

$\delta(t) = \text{Impulse}$

$$R_x(z) = \frac{n}{2} \delta(z)$$

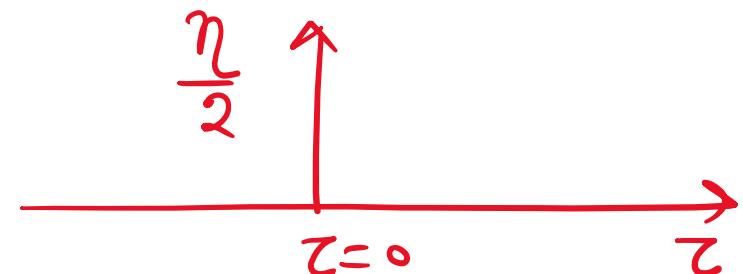
$\Rightarrow \underline{\text{PSD}}$

$$S_x(f) = \int_{-\infty}^{+\infty} R_x(z) e^{-j2\pi fz} dz$$

$$= \int_{-\infty}^{+\infty} \frac{n}{2} \delta(z) e^{-j2\pi fz} dz$$

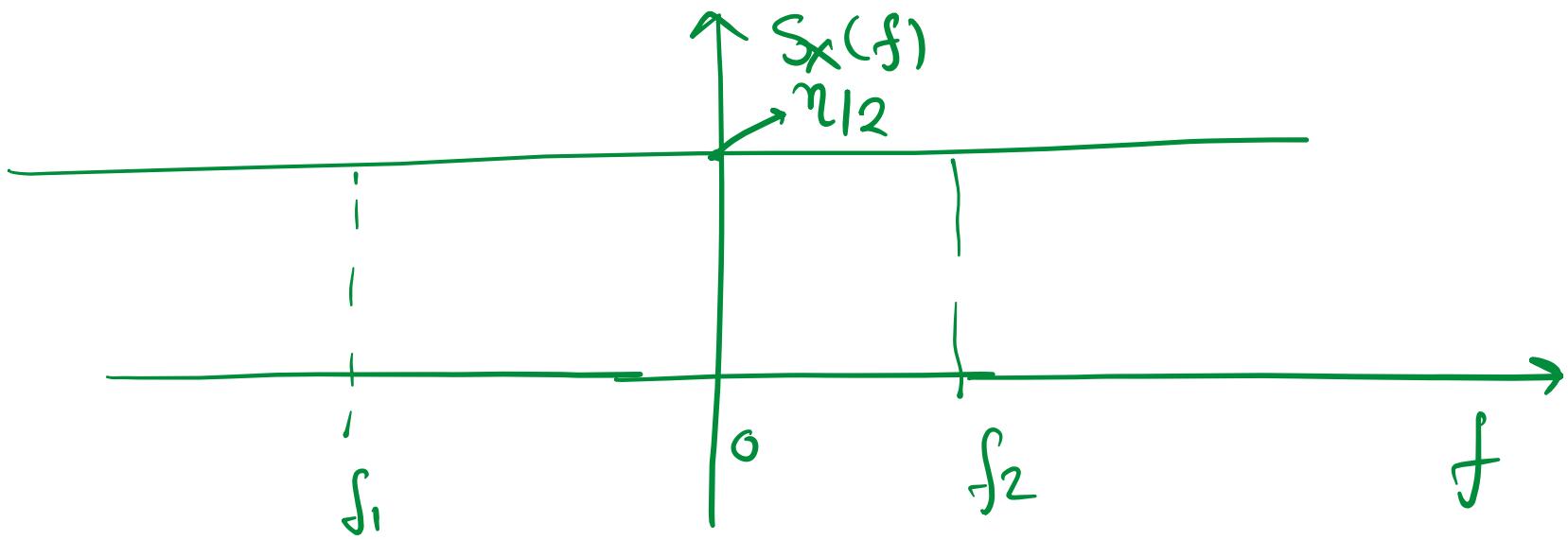
$$= \int_{-\infty}^{+\infty} \frac{n}{2} \delta(z) \cdot e^0 dz = \frac{n}{2} \int_{-\infty}^{+\infty} \delta(z) dz$$

$$= \frac{n}{2}$$



$$R_x(z) = 0, z \neq 0$$

$$S(z)\delta = 1$$



\Rightarrow Distribution of Power corresponding to each frequency component is same

\Rightarrow Therefore, it is known as white Random Process.

White Gaussian R.P.

- $x(t)$

white

&
Gaussian

$x(t)$ is white

Gaussian R.P.

→ 1. W.S.S.

- 2. $R_x(z) = \frac{\gamma}{2} \delta(z)$

$$f_{x(t_1), x(t_2), \dots, x(t_n)}^{(n_1, n_2, \dots, n_n)}$$

Must be Gaussian

