

MA 203

Ergodic Random Process

1. Stationary R.P. \Rightarrow Strict-Sense Stationary RP

$x(t)$: CDF is time invariant

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

$$F_{x(t_1), x(t_2), \dots, x(t_n)}(n_1, n_2, \dots, n_m)$$

$$= F_{x(t_1+\Delta), x(t_2+\Delta), \dots, x(t_n+\Delta)}(n_1, n_2, \dots, n_m)$$

$\forall n$

2. WSS: —

$$1. E[x(t)] = M_x$$

$$2. R_x(t_1, t_2) = R_x(t_1 - t_2)$$

Ensemble mean $\Rightarrow E[x(t)] = \int_{-\infty}^{+\infty} A \cos(\omega_0 t + \phi) f_\phi(\phi) d\phi$

$$x(t) = A \cos(\omega_0 t + \phi)$$

↑
R.V.

Ensemble auto correlation:

$$\Rightarrow E[x(t) x(t+z)] = \int_{-\infty}^{+\infty} x(t) \cdot x(t+z) f_\phi(\phi) d\phi$$

Time-Average:-

$x(t)$: RP

$$\langle \underline{M_x} \rangle_T = \frac{1}{2T} \int_{-T}^{+T} \underline{x(t)} dt$$

$$x(t) = A \cos(\omega_0 t + \phi)$$

$$\underbrace{\langle M_x(t) \rangle_T}_{\text{g(B)}} = \underline{g(B)}$$

$$E[\langle M_x \rangle_T] = \underline{M_x}$$

Time-Autocorrelation: —

Time-Averaged Autocorrelation

$$\langle R_x(z) \rangle_T = \frac{1}{2T} \int_{-T}^{+T} x(t) \cdot x(t+z) dt$$

Mean Ergodic RP:

A random process is known as mean ergodic RP if

$$\langle M_x \rangle_T \rightarrow M_x \text{ as } T \rightarrow \infty$$

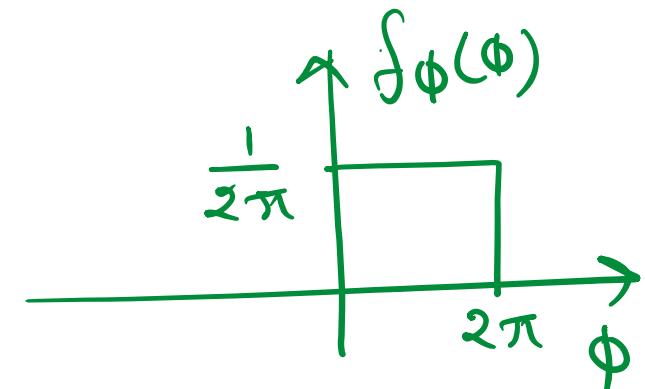
Autocorrelation Ergodic RP:-

$$\langle R_x(z) \rangle_T \xrightarrow{T \rightarrow \infty} \underline{R_x(z)}$$

Ex:- $x(t) = A \cos(\omega_0 t + \phi)$;

where A & ω_0 are const.

$$\phi \rightarrow RV$$



Sol:- $E[x(t)] = 0$

$$E[x(t)x(t+z)] = R_x(z) = \frac{A^2}{2} \cos \omega_0 z$$

Time-Average Mean

$$\begin{aligned}\langle M_x \rangle_T &= \frac{1}{2T} \int_{-T}^{+T} x(t) dt \\ &= \frac{1}{2T} \int_{-T}^{+T} A \cos(\omega_0 t + \phi) dt \\ &= \frac{A}{2T} \left\{ \frac{\sin(\omega_0 t + \phi)}{\omega_0} \right\} \Big|_{-T}^{+T}\end{aligned}$$

$$= \frac{A}{T\omega_0} \sin \omega_0 t$$

$$\lim_{T \rightarrow \infty} \langle M_x \rangle_T = \lim_{T \rightarrow \infty} \underbrace{\frac{A}{T\omega_0}}_{\text{Underline}} \underbrace{\sin \omega_0 t}_{\text{Underline}} = 0$$

$$\underbrace{\langle \mu_x \rangle_T}_{\text{as } T \rightarrow \infty} \longrightarrow \underbrace{E[x(t)] = \mu_x}_{\text{as } T \rightarrow \infty}$$

$\Rightarrow x(t)$ is a mean Ergodic R.P.

$$R_x(z) = \frac{A^2}{2} \cos(\omega_0 z)$$

$$\begin{aligned} \langle R_x(z) \rangle_T &= \frac{1}{2T} \int_{-T}^{+T} \{ A \cos(\omega_0 t + \phi) \} \{ A \cos(\omega_0(t+z) + \phi) \} dt \\ &= \frac{1}{2T} \cdot \frac{A^2}{2} \int_{-T}^{+T} \{ 2 \cos(\omega_0 t + \phi) \cos(\omega_0(t+z) + \phi) \} dt \\ &= \frac{A^2}{4T} \int_{-\infty}^{+\infty} \{ \cos^2(\omega_0 z) + \cos(\omega_0(2t+z) + 2\phi) \} dt \end{aligned}$$

$$\begin{aligned}
 &= \frac{A^2}{4T} \int_{-T}^{+T} \cos \omega_0 z dt + \frac{A^2}{4T} \int_{-T}^{+T} \cos \{\omega_0(2t+z) + 2\phi\} dt \\
 &= \frac{A^2}{2 \cdot 2T} \left\{ \cos \omega_0 z \right\} \Big|_{-T}^{+T} + \frac{A^2}{4T} \cdot \frac{\sin (\omega_0(2T+z) + 2\phi)}{2\omega_0} \Big|_{-T}^{+T}
 \end{aligned}$$

$$= \frac{A^2 \cos \omega_0 z}{2} + \frac{A^2 \sin (\omega_0(2T+z))}{4\omega_0 T}$$

$$\lim_{T \rightarrow \infty} \langle R_x(z) \rangle_T = \lim_{T \rightarrow \infty} \left\{ \frac{A^2 \cos \omega_0 z}{2} + \underbrace{\frac{A^2 \sin (\omega_0(2T+z))}{4\omega_0 T}}_{\text{circled}} \right\}$$

$$= \frac{A^2 \cos \omega_0 z}{2} = \underline{R_x(z)} = E[X(t) X(t+z)]$$

$X(t)$ is Ergodic in Autocorrelation.

So, $x(t)$ is Ergodic in mean
.. in AutoCorrelation.

$$\underline{\langle \mu_x \rangle_T}$$

$$\underline{x(t)}$$

$$\underline{\langle \mu_x \rangle_T} = \frac{1}{2T} \int_{-T}^{+T} x(t) dt$$

$$\begin{aligned} E[\underline{\langle \mu_x \rangle_T}] &= \frac{1}{2T} \int_{-T}^{+T} E[x(t)] dt \\ &= \frac{1}{2T} \int_{-T}^{+T} \mu_x dt = \mu_x \end{aligned} \quad \left\{ \begin{array}{l} \because E[x(t)] \\ = \mu_x \end{array} \right.$$

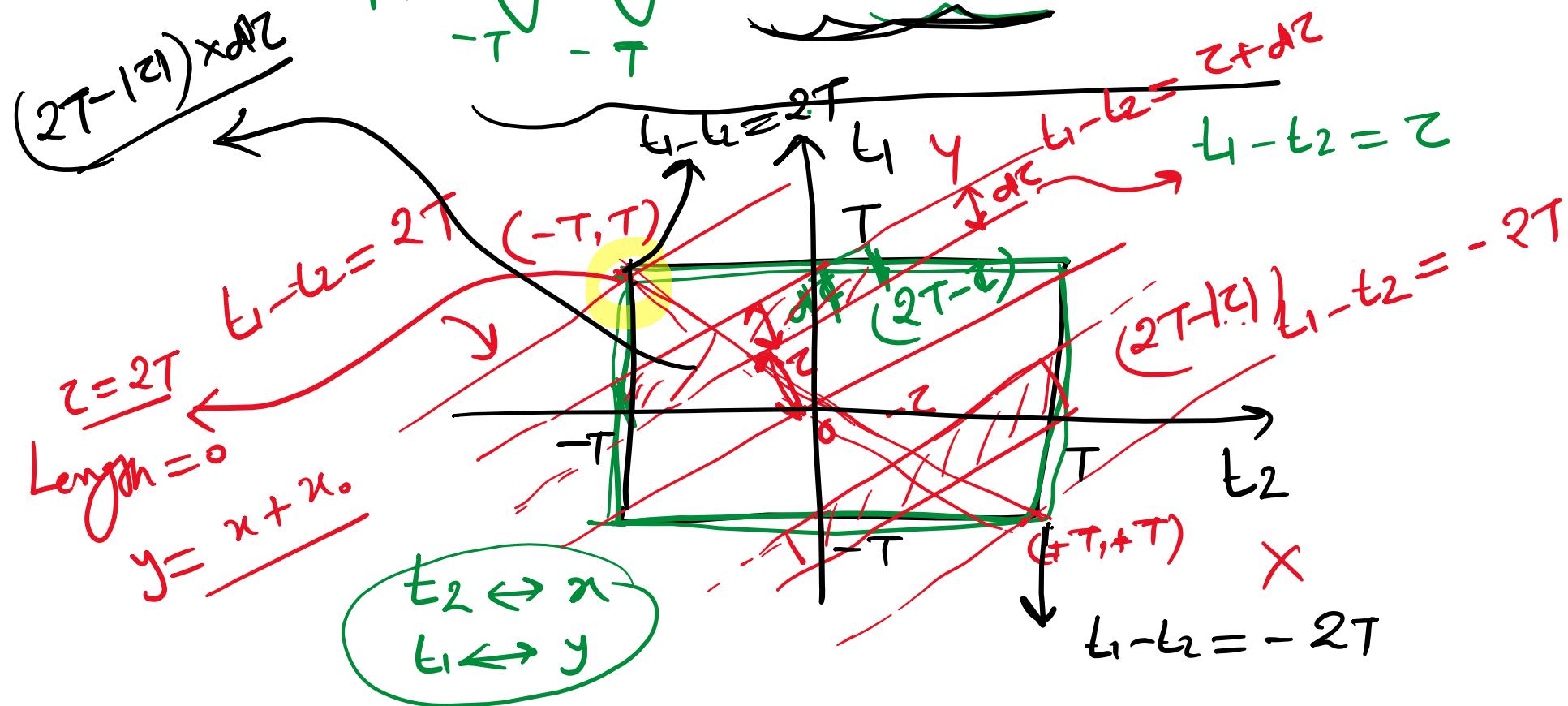
Variance of $\langle M_x \rangle_T$

$$\begin{aligned} E[(\langle M_x \rangle_T - \mu_x)^2] &= E \left[\left\{ \frac{1}{2T} \int_{-T}^{+T} x(t) dt - \mu_x \right\}^2 \right] \\ &= E \left[\left\{ \frac{1}{2T} \int_{-T}^{+T} \underbrace{\{x(t) - \mu_x\}}_{dt} \right\}^2 \right] \\ &= \frac{1}{4T^2} E \left\{ \int_{-T}^{+T} \underbrace{\{x(t) - \mu_x\}_{dt}}_{dt} \right\}^2 \\ &= \frac{1}{4T^2} \int_{-T}^{+T} \int_{-T}^{+T} E \left\{ \{x(t_1) - \mu_x\} \{x(t_2) - \mu_x\} \right. \\ &\quad \left. dt_1 dt_2 \right\} \end{aligned}$$

$$= \frac{1}{4T^2} \int_{-T}^{+T} \int_{-T}^{+T} E \left\{ (X(t_1) - \mu_x) (X(t_2) - \mu_x) \right\} dt_1 dt_2$$

$\left\{ \begin{array}{l} E[(x-\mu_x)(y-\mu_y)] \\ = C_x(x,y) \end{array} \right.$

$$= \frac{1}{4T^2} \int_{-T}^{+T} \int_{-T}^{+T} C_x(t_1 - t_2) dt_1 dt_2$$



In case of
WSS

$\left\{ \begin{array}{l} C_x(t_1, t_2) \\ = C_x(t_1 - t_2) \end{array} \right.$

$\left\{ \begin{array}{l} R_x(t_1, t_2) \\ = R_x(t_1 - t_2) \end{array} \right.$

Only for WSS RP

$$= \frac{1}{4T^2} \int_{-2T}^{+2T} C_x(z) \cdot (2T - |z|) dz$$

$$= \frac{1}{4T^2} \cdot 2T \cdot \int_{-2T}^{+2T} C_x(z) \left(1 - \frac{|z|}{2T} \right) dz$$

$$\boxed{\text{Var}(\langle \mu_x \rangle_T) = \frac{1}{2T} \int_{-2T}^{+2T} C_x(z) \cdot \left(1 - \frac{|z|}{2T} \right)^2 dz}$$