

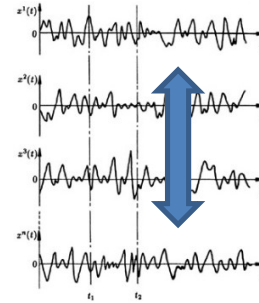
457.643 Structural Random Vibrations
In-Class Material: Class 03

II-1. Random Process (contd.)

◎ **“Average” of random process**

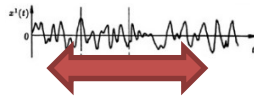
(a) “Ensemble” average: average over the ensemble

$$E[X(t)] = \lim_{n \rightarrow \infty} \frac{x_1(t) + x_2(t) + \dots + x_n(t)}{n} = \int_{-\infty}^{\infty} p(x) dx$$

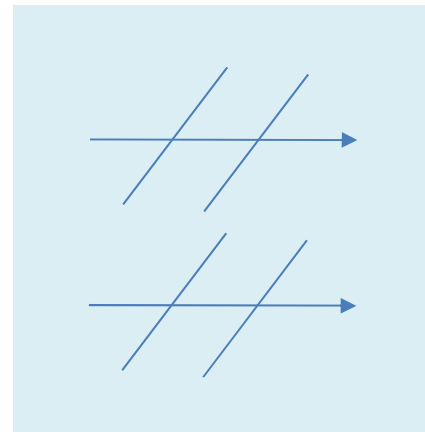


(b) “Temporal” average (for a specific time history)

$$\langle X(t) \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x(t) dt$$



- ◆ Temporal average is another r_____ v_____



◎ **Specification of a random process**

(a) By probabilistic distribution function

- ◆ $f_X(x, t)$: 1st order “m_____” PDF
- ◆ $f_{X(t_1)X(t_2)}(x_1, t_1; x_2, t_2)$: 2nd order joint PDF
- ◆ \vdots
- ◆ $f_{X(t_1)\dots X(t_n)}(x_1, t_1; \dots; x_n, t_n)$: nth order joint PDF

Theoretically, need the _____th order joint PDF for complete description of a random process

(b) By characteristic function

- ◆ $M_{X(t)}(\theta, t)$: 1st order characteristic function
- ◆ \vdots
- ◆ $M_{X(t_1)\dots X(t_n)}(\theta_1, t_1; \dots; \theta_n, t_n)$: nth order joint characteristic function

(c) By moment functions (i.e. partial descriptors)

→ most common (because of lack of i_____)

- ◆ $E[X(t)] = \mu_X(t)$ or $\mu(t)$: _____ function
- ◆ $E[X(t_1)X(t_2)] = \phi_{XX}(t_1, t_2)$ or $\phi(t_1, t_2)$: auto _____ function
- ◆ $E\{[X(t_1) - \mu(t_1)][X(t_2) - \mu(t_2)]\} =$ _____ : auto _____ function

(d) By a function of random variables

- ◆ $X(t) = At + B$
- ◆ $X(t) = \sum_{i=1}^n A_i \cos(\omega_i t + \theta_i)$

(e) Others: random pulses, log-characteristic function, cumulants, ARMA, etc.

◎ **First & second order moment functions**

$E[X(t)] = \mu_X(t)$ or $\mu(t)$: (E_____) mean function

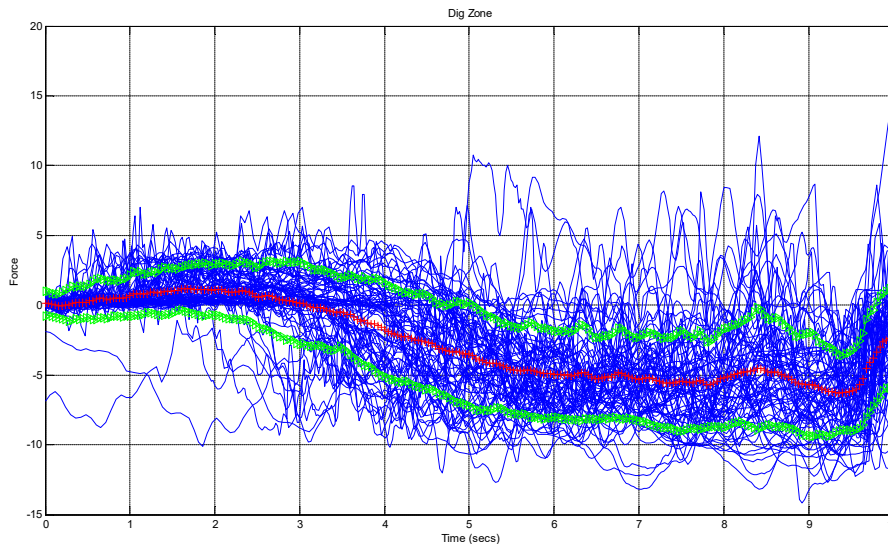
$E[X(t_1)X(t_2)] = \phi_{XX}(t_1, t_2)$ or $\phi(t_1, t_2)$: *Auto-correlation* function

⋮

$E\{[X(t_1) - \mu(t_1)][X(t_2) - \mu(t_2)]\} = \phi_{XX}(t_1, t_2) - \mu(t_1)\mu(t_2)$
 $= \kappa_{XX}(t_1, t_2)$: *Auto-covariance* function

$\sigma_X(t) = \sqrt{\text{_____}}$: Standard deviation function

$\rho_{XX}(t_1, t_2) = \text{_____}$: *Auto-correlation-coefficient* function



Example: 77 force time histories during “digging” tasks and their moment functions

Note:

If $\mu_X(t) = 0$ (zero-mean process),

$$\phi_{XX}(t_1, t_2) = \kappa_{XX}(t_1, t_2)$$

One can transform a random process to a zero-mean process by

$$Y(t) = X(t) - \mu_X(t)$$

Why?

- For a complex-valued random process,

$$\phi_{XX}(t_1, t_2) = E[X(t_1)X^*(t_2)]$$

$$\kappa_{XX}(t_1, t_2) = E[(X(t_1) - \mu(t_1))(X^*(t_2) - \mu^*(t_2))]$$

Note that $\phi_{XX}(t, t)$ and $\kappa_{XX}(t, t)$ are always _____-valued.

- More than one random process involved

$$\phi_{XY}(t_1, t_2) = E[X(t_1)Y^*(t_2)] : \text{_____ correlation function}$$

$$\kappa_{XY}(t_1, t_2) = E[(X(t_1) - \mu(t_1))(Y^*(t_2) - \mu^*(t_2))] : \text{_____ covariance function}$$

$$\rho_{XY}(t_1, t_2) = \text{_____} : \text{_____ correlation coefficient function}$$

- Importance of 1st and 2nd order moment functions

- 1) Most of the time, 1st and 2nd order moment functions are all one can get from data
- 2) For Gaussian, 1st and 2nd order moment functions are all you need for a complete description.
- 3) Using Chebyshev bounds, one can get upper bound estimate on the probability using moments

$$P(|Z| > b) \leq \frac{E[|Z|^c]}{b^c}$$

e.g. $c = 2, Z = X - \mu_X$

$$P(|X - \mu_X| > b) \leq \frac{E[|X - \mu_X|^2]}{b^2} = \text{_____}$$

◎ **Five important properties of $\phi_{XY}(t_1, t_2)$ and $\kappa_{XY}(t_1, t_2)$**

- 1) **Hermitian** ("Symmetric" for a real random process)

$$\phi_{XY}(t_1, t_2) =$$

$$\kappa_{XY}(t_1, t_2) =$$

- 2) **Boundedness**

$$\text{Schwarz inequality } |E[XY]| \leq \sqrt{E[X^2]E[Y^2]}$$

$$\text{Thus, } |\phi_{XY}(t_1, t_2)| \leq \sqrt{\phi_{XX}(t_1, t_1)\phi_{YY}(t_2, t_2)}$$

Also, $|\phi_{XX}(t_1, t_2)| \leq \sqrt{\phi_{XX}(t_1, t_1)\phi_{XX}(t_2, t_2)}$

Similarly, $|\kappa_{XY}(t_1, t_2)| \leq \sqrt{\kappa_{XX}(t_1, t_1)\kappa_{YY}(t_2, t_2)} = \sqrt{\sigma_X^2(t_1)\sigma_Y^2(t_2)}$

Note:

If $E[X^2(t)]$ is bounded ($< \infty$) for $\forall t$,

$|\phi_{XX}(t, s)| < \infty$

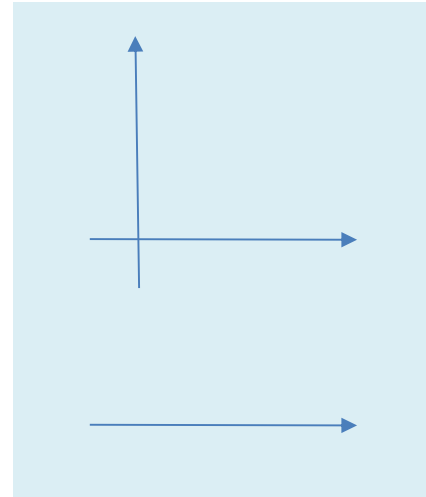
If $\sigma_X^2(t)$ is bounded ($< \infty$) for $\forall t$,

$|\phi_{XX}(t, s)| < \infty$

$X(t)$ is a “_____” random process

if _____ is always finite

(Check L&S p.121. Later we will confirm that this means PSD exists)



3) **Non-negative Definiteness**

For an arbitrary function $h(t)$,

$$\sum_{i=1}^n \sum_{j=1}^n \phi_{XX}(t_i, t_j) h(t_i) h^*(t_j) \geq 0$$

Proof:

$$\begin{aligned} \text{(LHS)} &= \{h(t_1) \dots h(t_n)\} [\phi_{XX}(t_i, t_j)]_{n \times n} \{h^*(t_1) \dots h^*(t_n)\}^T \\ &= \mathbf{h}^T E[\mathbf{X}\mathbf{X}^T] \mathbf{h}^* \\ &= E[\mathbf{h}^T \mathbf{X}\mathbf{X}^T \mathbf{h}^*] \\ &= E[YY^*] \\ &= E[\dots] \geq 0 \end{aligned}$$

Why is this property important?

Fourier transform of non-negative definite function is _____
 (Lin 1967, p.42 – Bochner’s theorem)

Lin, Y.K. (1967) *Probabilistic Theory of Structural Dynamics*, McGraw-Hill, New York, NY.

Note: However,
 $\phi_{XY}(t_1, t_2)$: NOT non-negative definite

$\therefore E[XY]$ can be _____
 \therefore Cross PSD can be _____