

**MA 203**

Stochastic Process

**Example 1:** Suppose someone gives you a coin and claims that this coin is biased; that it lands on heads only 48% of the time. You decide to test the coin for yourself. If you want to be 95% confident that this coin is indeed biased, how many times must you toss the coin? ( $\epsilon = 0.02$ .)

Sol:-

$$P \left\{ \left| \frac{S_n}{n} - \mu \right| < \epsilon \right\} > 1 - \frac{\sigma}{n \epsilon^2}$$

$$P \left\{ \left| \frac{S_n - n\mu}{n} \right| < 0.02 \right\} > 0.95$$

$$P \left\{ \left| \frac{S_n - n\mu}{\sqrt{n} \sigma} \right| < \frac{0.02 \sqrt{n}}{\sigma} \right\} > 0.95$$

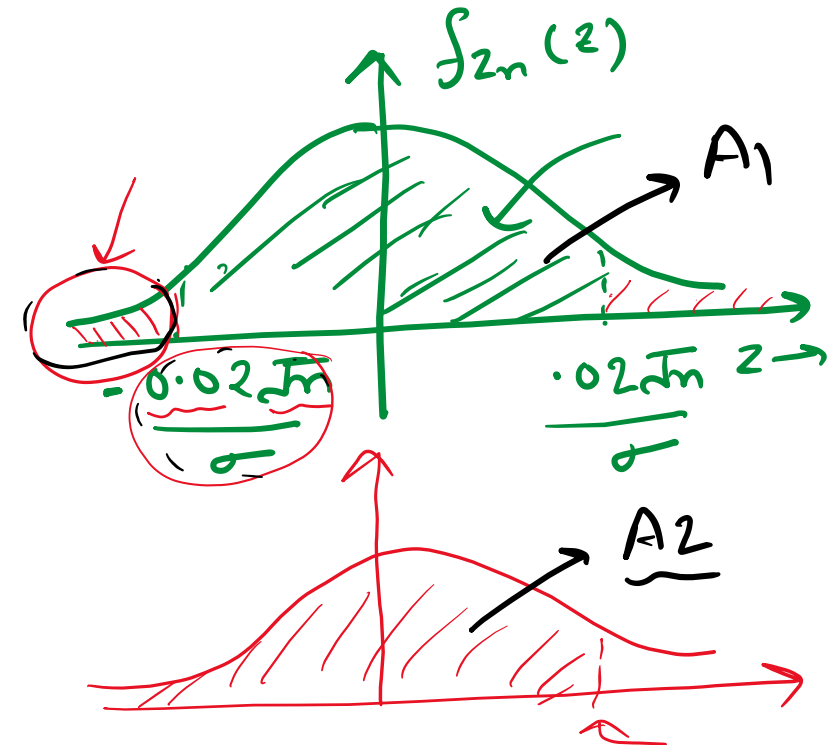
$$P \left\{ |Z_n| < \frac{0.02 \sqrt{n}}{\sigma} \right\} > 0.95$$

$$\approx P \left\{ Z_n < \frac{0.02 \sqrt{n}}{\sigma} \right\} > 0.95$$

$$\mu = p = 0.48$$

$$\sigma^2 = p(1-p)$$

$$= 0.48 \times 0.52$$



$$1 - 2Q\left(\frac{0.02\sqrt{n}}{\sigma}\right) \geq 0.95$$

$$\geq 0.95$$

$\Rightarrow$

$n =$

$$\frac{0.02\sqrt{n}}{\sigma} \geq 1.645$$

$$\sqrt{n} \geq \frac{1.645 \times \sigma}{0.02}$$

$$n \geq \left\{ \frac{1.645 \times \sigma}{0.02} \right\}^2$$

$$\boxed{n = 1692} \Rightarrow \underline{\text{CLT}}$$

$$\boxed{\phi(n) = 1 - Q(n)}$$


$$P(Z_n \leq z) = \Phi(z) = \int_{-\infty}^z e^{-t^2/2} dt$$

$$\begin{aligned} \sigma^2 &= p(1-p) \\ &= .48 \times .52 \\ &= \end{aligned} \left\{ \begin{aligned} &\Phi^{-1}(0.95) \\ &= 1.645 \end{aligned} \right.$$

$$\text{WLLN} \Rightarrow \boxed{n = 12480}$$

# **Stochastic Process**

## A Review of Last Class

1. A stochastic process is defined as a function of two arguments  $X(s, t)$ ,  $s \in S$  and  $t \in T$ .
2. A stochastic process maps each sample point to a waveform.
3. If we sample a stochastic process  $X(s, t)$  at time instants  $t_1, t_2, \dots, t_n$ , we get RVs  $X(t_1), X(t_2), \dots, X(t_n)$ , respectively.  

4. We can define stochastic process as a time varying random variable.
5. Stochastic Process/ Random Process/ Chance process
6. Notation:  $X(t)$   $\equiv$   $X(s, t)$

**Parameter Space:** The set  $\mathcal{T}$  is called parameter space where  $t \in \mathcal{T}$  may denote time, length, or any other quantity.

**Continuous-Time Random Process:** If the index set  $\mathcal{T}$  is continuous,  $\{X(t), t \geq 0\}$  is called a continuous time process.

**Example:** Suppose  $X(t) = A \cos(\omega_0 t + \Phi)$ , where  $A$  and  $\omega_0$  are constant and  $\Phi$  is uniformly distributed between 0 and  $2\pi$ .

$$\underline{X(t)} \quad ; \quad \underline{t \geq 0}$$

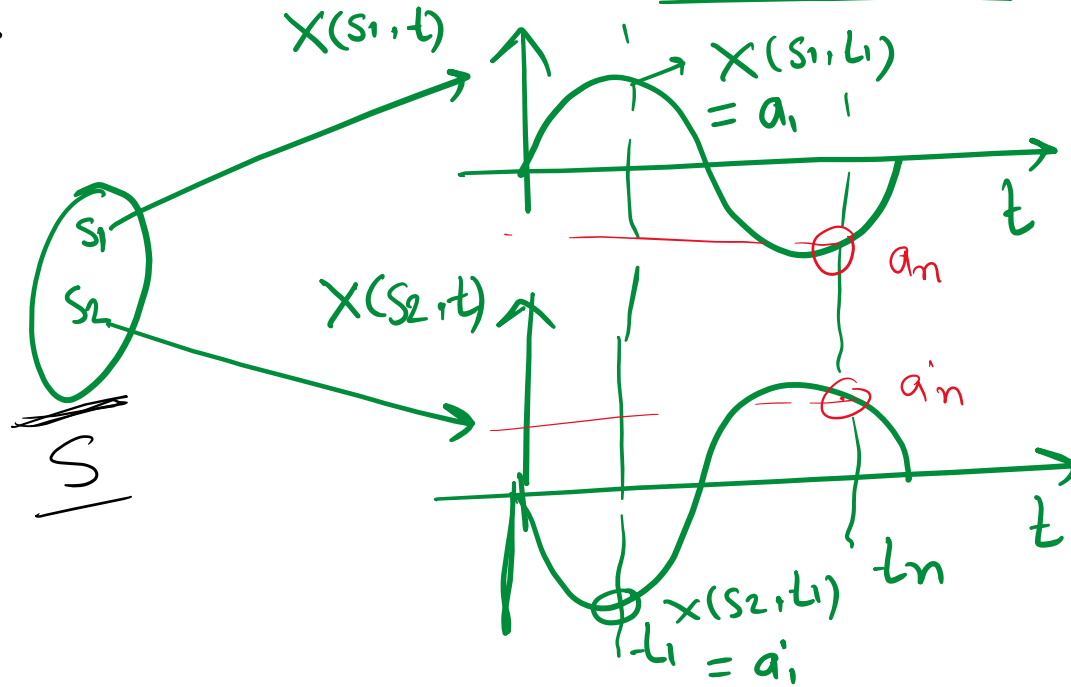
**Discrete Time Random Process:** If the index set  $\mathcal{T}$  is a countable set,  $\{X(t), t \geq 0\}$  is called a discrete-time process. Such a random process can be represented as  $\{X[n], n \in \mathbb{Z}\}$  and called a random sequence.

**Example:** Suppose  $X[n] = A \cos(\omega_0 n + \Phi)$ , where  $A$  and  $\omega_0$  are constant and  $\Phi$  is uniformly distributed between 0 and  $2\pi$ .

$$\underline{X[n]} \quad ; \quad n = -2, -1, 0, 1, 2, 3, \dots$$

**State Space:** The set  $\Phi$  is the set of all possible values of  $X(t)$  for all  $t$  and is called the state space where  $X(t): S \rightarrow A_t$ ,  $A_t \subseteq \mathbb{R}$ ,  $\Phi = \bigcup A_t$ .

The state is the value taken by  $X(t)$  at a time  $t$ , and set of all such states is called the state space.



$$A_{t_1} = \{a_1, a'_1\}$$

$$\vdots$$

$$A_{t_n} = \{a_n, a'_n\}$$

$$\Phi = \bigcup_{t_i \in \eta} A_{t_i}$$

**Discrete-State Process:** A random process is discrete-state if state-space is finite or countable.

if  $\Phi$  contains countable / countably infinite elements

**Continuous-State Process:** A random process is continuous-state if state-space is uncountable.

if  $\Phi$  contains uncountable element

## Probabilistic Structure of a Random Process:

$$\begin{array}{ccccccc} X(t), & t_1 & t_2 & t_3 & \dots & t_n \\ & \underline{X(t_1)} & \underline{X(t_2)} & \underline{X(t_3)} & \dots & \underline{X(t_n)} \end{array}$$

### Mean of a RP:

Mean of the RP  $X(t)$  at time  $t = \underline{\mu_X(t)} = \underline{E[X(t)]}$   $\begin{array}{c} t \\ t_1, t_2 \end{array}$

Auto-correlation Function: Auto-correlation function of the RP  $X(t)$  at times  $t_1$  and  $t_2$  is defined as

$$\begin{array}{ccc} t_1 & t_2 & \text{Self} \\ \underline{X(t_1)} & \underline{X(t_2)} & \nearrow \end{array} \quad \underline{R_X(t_1, t_2)} = E[\underline{X(t_1)} \cdot \underline{X(t_2)}]$$
$$\begin{array}{c} t_1 = t_2 \\ R_X(t_1, t_1) = E[X^2(t_1)] \end{array}$$



**Auto-Covariance Function:** The autocovariance function of RP  $X(t)$  at time instants  $t_1$  and  $t_2$  is defined by

$$C_X(t_1, t_2) = E[(X(t_1) - \mu_X(t_1))(X(t_2) - \mu_X(t_2))]$$

$$t_1 = t_2$$
$$C_X(t_1, t_1) = E[(X(t_1) - \mu_X(t_1))^2] = \text{Var}(X(t_1))$$

**Correlation Coefficient:** The correlation-coefficient of RP  $X(t)$  at time instants  $t_1$  and  $t_2$  is defined by

$$\rho_X(t_1, t_2) = \frac{C_X(t_1, t_2)}{\sqrt{C_X(t_1, t_1)C_X(t_2, t_2)}} =$$

$\text{Var}(X(t_1)) \quad \text{Var}(X(t_2))$

**Example 1:**  $X(t) = A \cos(\omega_0 t + \Phi)$  where  $A$  and  $\omega_0$  are constants and  $\Phi$  is uniformly distributed between 0 and  $2\pi$ . Find the mean and autocorrelation of  $X(t)$ .

Sol:-

$$\underline{X(t)} \rightarrow$$

$$X(t_1) = A \cos(\omega_0 t_1 + \Phi)$$

$$X(t_2) = A \cos(\omega_0 t_2 + \Phi)$$

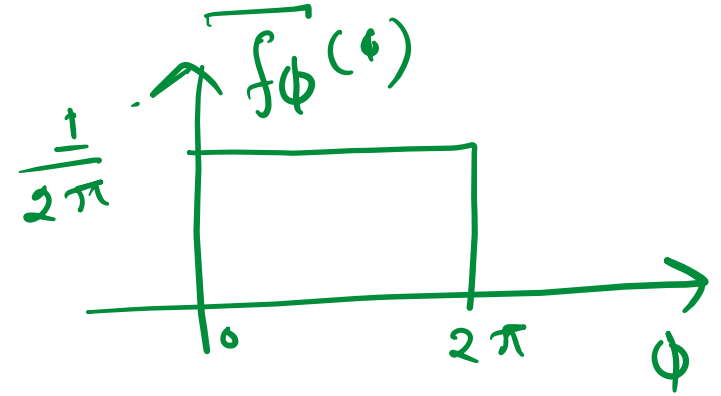
$$\underline{X(t)} = A \cos(\omega_0 t + \Phi)$$

↓

$t = 1 \text{ sec.}$

$$\underline{X(1)} = A \cos(\omega_0 \cdot 1 + \Phi) = \underline{g(\Phi)}$$

where  $\Phi$  is a random variable



$$\boxed{\begin{aligned} Y &= g(X) \\ E[Y] &= \int_{-\infty}^{+\infty} g(x) \cdot f_X(x) dx \end{aligned}}$$

$$E[x(t)] = \int_{-\infty}^{+\infty} \underbrace{A \cos(\omega_0 t + \phi)}_{g(\phi)} \cdot \underbrace{f_\phi(\phi)}_{\text{pdf}} \cdot d\phi$$

$$= \int_0^{2\pi} A \cos(\omega_0 t + \phi) \frac{1}{2\pi} d\phi$$

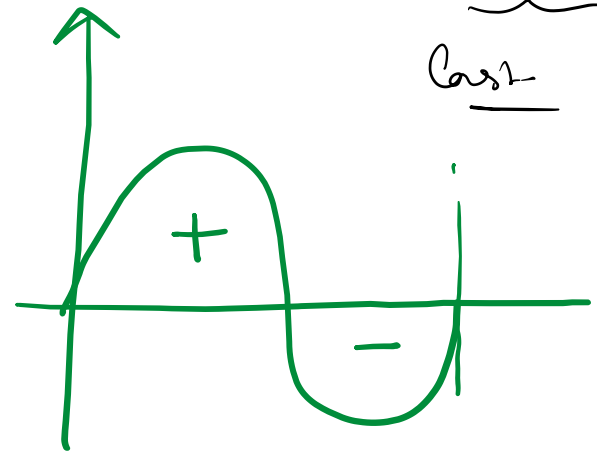
$$= \frac{A}{2\pi} \int_0^{2\pi} \cos(\underbrace{\omega_0 t + \phi}_{\text{phase}}) d\phi$$

→

$$= \frac{A}{2\pi} \cdot 0 = 0$$

$$\cos(\phi + \theta_0)$$

$$\theta_0 = \underbrace{\omega_0 t}_{\text{const}}$$



$$R_x(t_1, t_2) = E[X(t_1) X(t_2)]$$

$$= E\left[ \underbrace{A \cos(\omega_0 t_1 + \phi)} \cdot \underbrace{A \cos(\omega_0 t_2 + \phi)} \right]$$

$$= \frac{A^2}{2} E\left[ 2 \cos(\omega_0 t_1 + \phi) \cos(\omega_0 t_2 + \phi) \right]$$

$$\left\{ 2 \cos A \cos B = \cos(A-B) + \cos(A+B) \right\}$$

$$R_x(t_1, t_2) = \frac{A^2}{2} E\left[ \underbrace{\cos(\omega_0(t_1 - t_2))} + \cos\{\omega_0(t_1 + t_2) + 2\phi\} \right]$$

$$= \frac{A^2}{2} \cos \omega_0(t_1 - t_2) + \frac{A^2}{2} E\left[ \cos\left(\underbrace{\omega_0(t_1 + t_2)}_{+ 2\phi}\right) \right]$$

$$= \frac{A^2}{2} \cos \omega_0(t_1 - t_2) + \underline{B}$$

$$B = \frac{A^2}{2} \int_0^{2\pi} \cos(\omega_0(t_1+t_2) + 2\phi) \cdot \underbrace{\frac{1}{2\pi}}_{f_\phi(\phi)} d\phi$$

$$= \frac{A^2}{4\pi} \int_0^{2\pi} \cos(\omega_0(t_1+t_2) + 2\phi) d\phi$$

$$R_x(t_1, t_2) = \frac{A^2}{2} \cos \omega_0(t_1 - t_2)$$

