

ECE 863

Analysis of Stochastic Systems

Part III: Discrete- & Continuous- Time Random Processes

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ECE 863

- Reading assignments
 - Sections 6.1 - 6.5

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Random Processes

- Remember that a random variable $X(\zeta)$ is a mapping from a "random outcome" ζ to a "real number".
- A Random Process $X(t; \zeta)$ is a mapping from a "random outcome" ζ to a "function"
- A random process can be a function of time, space, etc.

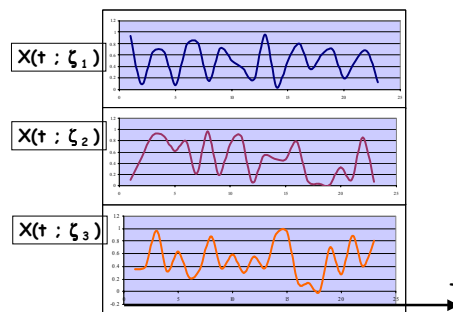
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Random Processes

$$X(t, \zeta) \quad \zeta \in S$$



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Random Processes

- For a given outcome $\zeta = \zeta_k$, the function $X(t, \zeta_k)$ is referred to as a "sample function"
- Therefore, a random process is a "collection of functions" (or a "family of functions") generated by the random outcome ζ

$$X(t, \zeta) \quad \zeta \in S$$

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Example: Sinusoidal Amplitude

- Let ζ be a uniformly distributed random variable in the interval $[-1, +1]$.

One can define the random process $X(t, \zeta)$:

$$X(t, \zeta) = \zeta \cos(t) \quad -\infty < t < \infty$$

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Example: Sinusoidal Amplitude

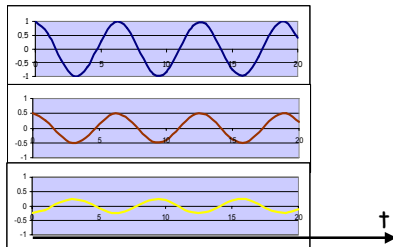
$$X(t, \zeta) = \zeta \cos(t) \quad -\infty < t < \infty$$

$$\zeta \in [-1, +1]$$

$$\zeta = 1$$

$$\zeta = 0.5$$

$$\zeta = -0.25$$



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Example: Sinusoidal Phase

- Let ζ be a uniformly distributed random variable in the interval $[0, 2\pi]$.

One can define the random process $Y(t, \zeta)$:

$$Y(t, \zeta) = \cos(t + \zeta)$$

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Example: Sinusoidal Phase

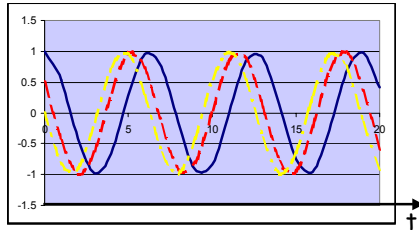
$$\zeta \in [0, 2\pi]$$

$$\zeta = 0$$

$$\zeta = \frac{\pi}{3}$$

$$\zeta = \frac{\pi}{2}$$

$$Y(t, \zeta) = \cos(t + \zeta)$$



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Random Processes

- The parameter (t) of the random process $X(t, \zeta)$ can be discrete or continuous

- When t is continuous, we have a "continuous-time" random process
- When t is discrete, we have a "discrete-time" random process

- Either case t is a member of an "index set" I :

$$X(t, \zeta) \quad t \in I \quad \zeta \in S$$

Source of randomness

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Random Processes

- At a given "time index" t , the random process $X(t, \zeta)$ represents a random variable
- As time changes, the random process generates "different" random variables:

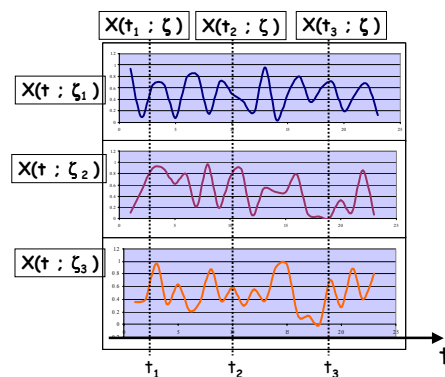
$$X(t_1; \zeta) \quad X(t_2; \zeta) \quad X(t_3; \zeta) \quad \dots$$

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Random Processes



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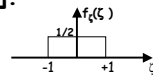
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Example: Sinusoidal Amplitude

- Let ζ be a uniformly distributed random variable in the interval $[-1, +1]$.

Therefore, $f_{\zeta}(\zeta) = 1/2$.



One can define the random process $X(t, \zeta)$:

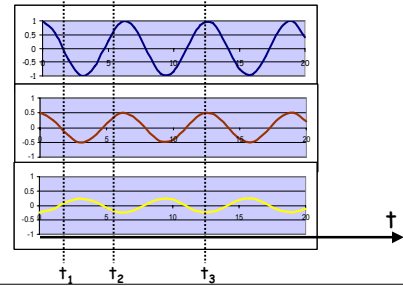
$$X(t, \zeta) = \zeta \cos(t) \quad -\infty < t < \infty$$

Find the probability density functions (pdf) of the random variable $X(t, \zeta)$

Example: Sinusoidal Amplitude

$$X(t, \zeta) = \zeta \cos(t) \quad -\infty < t < \infty$$

$$\zeta \in [-1, +1]$$

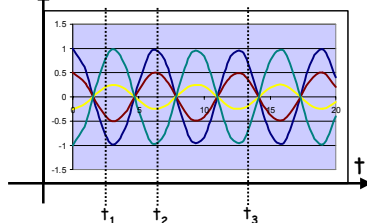


Example: Sinusoidal Amplitude

$$X(t, \zeta) = \zeta \cos(t)$$

$$\zeta \in [-1, +1]$$

For a given t_k , the RV $X(t_k, \zeta)$ varies within the interval:



$$X(t_k, \zeta) \in [-|\cos(t_k)|, |\cos(t_k)|]$$

Example: Sinusoidal Amplitude

- $X(t, \zeta)$ is a function of the random variable ζ .
- How would we find the pdf of $X(t, \zeta)$, given that we know the pdf of ζ ?
- Do you remember the Fundamental Theorem?

Example: Sinusoidal Amplitude

- Remember that we need to find the solution(s) to $X(t, \zeta)$. I.e. express in ζ terms x
- These solutions: ζ_1, ζ_2, \dots are functions of x
- Then, the Fundamental Theorem can be used as follows:

$$f_x(x) = \frac{f_\zeta(\zeta_1)}{|dx(t, \zeta_1)/d\zeta|} + \frac{f_\zeta(\zeta_2)}{|dx(t, \zeta_2)/d\zeta|} + \dots$$

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Example: Sinusoidal Amplitude

$$X(t, \zeta) = \zeta \cos(t) \quad \boxed{\zeta \in [-1, +1]}$$

Solving $x = \zeta \cos(t)$ leads to a single solution:

$$\boxed{\zeta = \frac{x}{\cos(t)}}$$

Taking the derivative of $x = \zeta \cos(t)$:

$$\boxed{\frac{dx}{d\zeta} = \cos(t)}$$

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Example: Sinusoidal Amplitude

$$X(t, \zeta) = \zeta \cos(t) \quad \boxed{\zeta \in [-1, +1]}$$

For a given t , the RV $X(t, \zeta)$ have a uniform distribution:

$$\boxed{f_{X(t, \zeta)}(x(t, \zeta)) = \frac{f_\zeta(\zeta)}{|\cos(t)|} = \frac{1}{2 |\cos(t)|}}$$

$$\boxed{X(t, \zeta) \in [-|\cos(t)|, |\cos(t)|]}$$

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Example: Sinusoidal Phase

- Let ζ be a uniformly distributed random variable in the interval $[0, 2\pi]$.

One can define the random process $Y(t, \zeta)$:

$$Y(t, \zeta) = \cos(t + \zeta)$$

Find the probability density functions (pdf) of the random variable $Y(t, \zeta)$

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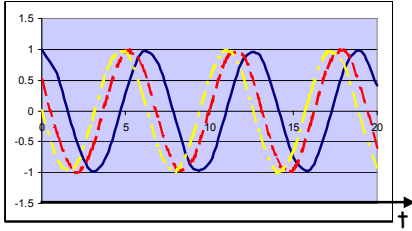
Example: Sinusoidal Phase

$$\zeta \in [0, 2\pi]$$

For a given t_k ,
the RV $Y(t_k, \zeta)$
varies within
the interval:

$$Y(t, \zeta) \in [-1, +1]$$

$$Y(t, \zeta) = \cos(t + \zeta)$$



Example: Sinusoidal Phase

- By using the fundamental theorem:

$$f_y(y(t, \zeta)) = \frac{f_\zeta(\zeta_1)}{|dy(t, \zeta_1)/d\zeta|} + \frac{f_\zeta(\zeta_2)}{|dy(t, \zeta_2)/d\zeta|} + \dots$$

we can show:

$$f_y(y(t, \zeta)) = \frac{1}{\pi\sqrt{1-y^2}} \quad |y| < 1$$

(See examples 6.4 and 3.28 in the book)

Probability Functions for RP

- Since a random process $X(t, \zeta)$ is a random variable at a given "time index" t , $X(t, \zeta)$ generates a sequence of random variables at a series of discrete time instances:

$$t_1, t_2 \dots t_k$$

$$X_1 = X(t_1, \zeta), X_2 = X(t_2, \zeta), \dots, X_k = X(t_k, \zeta)$$

Probability Functions for RP

- Therefore, we can define joint probability functions (cdf, pdf, pmf) of these "time samples" random variables:

$$X_1 = X(t_1, \zeta), X_2 = X(t_2, \zeta), \dots, X_k = X(t_k, \zeta)$$

- For example, the joint cdf:

$$F_{X_1, \dots, X_k}(x_1, x_2, \dots, x_k) = P[X_1 \leq x_1, X_2 \leq x_2, \dots, X_k \leq x_k]$$

Probability Functions for RP

- For discrete X_1, X_2, \dots, X_k , we can define the probability mass functions (pmf):

$$p_{X_1, \dots, X_k}(x_1, x_2, \dots, x_k) = P[X_1 = x_1, X_2 = x_2, \dots, X_k = x_k]$$

- Similarly, for continuous X_1, X_2, \dots, X_k , we can define the joint probability density functions:

$$f_{X_1 X_2 \dots X_k}(x_1, x_2, \dots, x_k)$$

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Moments of Random Processes

- Let $f_{X(t)}(x)$ be the pdf of the random variable $X(t)$. We can define the k th moment:

$$E[(X(t))^k] = \int_{-\infty}^{+\infty} x^k f_{X(t)}(x) dx$$

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Moments of Random Processes

- A special case of the k th moment is the mean:

$$m_X(t) = E[X(t)] = \int_{-\infty}^{+\infty} x f_{X(t)}(x) dx$$

- Therefore, in general, the mean and other moments could change with "time"

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Moments of Random Processes

- If $f_{X(t)}(x)$ is the pdf of the random variable $X(t)$, then the k th central moment:

$$E[(X(t) - m_X(t))^k] = \int_{-\infty}^{+\infty} (x - m_X(t))^k f_{X(t)}(x) dx$$

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Moments of Random Processes

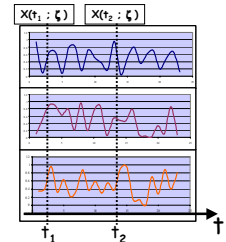
- A special case of the k th central moment is the variance:

$$\begin{aligned} \text{VAR}(X(t)) &= E[(X(t) - m_X(t))^2] \\ &= \int_{-\infty}^{+\infty} (x - m_X(t))^2 f_{X(t)}(x) dx \end{aligned}$$

Joint Moments of Random Proc.

- Two "time samples" $X(t_1)$ and $X(t_2)$ of the random process $X(t, \zeta)$ represent two random variables with a joint density function:

$$f_{X(t_1), X(t_2)}(x, y)$$



Joint Moments of Random Proc.

- The autocorrelation of the random process $X(t, \zeta)$ is defined as:

$$\begin{aligned} R_X(t_1, t_2) &= E[X(t_1)X(t_2)] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_{X(t_1), X(t_2)}(x, y) dx dy \end{aligned}$$

Joint Moments of Random Proc.

- The autocovariance of the random process $X(t, \zeta)$ is defined as:

$$\begin{aligned} C_X(t_1, t_2) &= \\ E[\{X(t_1) - m_X(t_1)\}\{X(t_2) - m_X(t_2)\}] \end{aligned}$$

What is $C_X(t_1, t_1)$?

Joint Moments of Random Proc.

- Evaluating the autocovariance at a single time sample t gives the variance at time t :

$$\begin{aligned}\text{VAR}(X(t)) &= E[(X(t) - m_x(t))^2] \\ &= C_x(t, t)\end{aligned}$$

Joint Moments of Random Proc.

- The autocorrelation and the autocovariance functions are related as follows:

$$C_x(t_1, t_2) = R_x(t_1, t_2) - m_x(t_1)m_x(t_2)$$

$$\begin{aligned}C_x(t, t) &= R_x(t, t) - m_x(t)m_x(t) \\ &= E[(X(t))^2] - (m_x(t))^2 \\ &= \text{VAR}(X(t))\end{aligned}$$

Joint Moments of Random Proc.

- The autocorrelation coefficient (or simply correlation coefficient) is defined as:

$$\rho_x(t_1, t_2) = \frac{C_x(t_1, t_2)}{\sqrt{C_x(t_1, t_1)}\sqrt{C_x(t_2, t_2)}}$$

$$\rho_x(t_1, t_2) = \frac{C_x(t_1, t_2)}{\sqrt{\text{VAR}(t_1)}\sqrt{\text{VAR}(t_2)}} \quad |\rho_x(t_1, t_2)| \leq 1$$

Example: Sinusoidal Amplitude

- Let ζ be a random variable, and let the random process $X(t, \zeta)$:

$$X(t, \zeta) = \zeta \cos(t)$$

Find the mean, variance, autocorrelation, and autocovariance of $X(t, \zeta)$

Example: Sinusoidal Amplitude

■ The mean:

$$m_x(t) = E[\zeta \cos(t)]$$

$$m_x(t) = E[\zeta] \cos(t)$$

Example: Sinusoidal Amplitude

■ The variance:

$$\begin{aligned} \text{VAR}(X(t)) &= E[(\zeta \cos(t))^2] - (E[\zeta] \cos(t))^2 \\ &= E[\zeta^2] \cos^2(t) - (E[\zeta])^2 \cos^2(t) \\ &= (E[\zeta^2] - (E[\zeta])^2) \cos^2(t) \end{aligned}$$

$$\text{VAR}(X(t)) = (\text{VAR}(\zeta)) \cos^2(t)$$

Example: Sinusoidal Amplitude

■ The autocorrelation function

$$R_x(t_1, t_2) = E[\zeta \cos(t_1) \zeta \cos(t_2)]$$

$$R_x(t_1, t_2) = E[\zeta^2] \cos(t_1) \cos(t_2)$$

Example: Sinusoidal Amplitude

■ The autocovariance function:

$$\begin{aligned} C_x(t_1, t_2) &= R_x(t_1, t_2) - m_x(t_1) m_x(t_2) \\ &= \{E[\zeta^2] - (E[\zeta])^2\} \cos(t_1) \cos(t_2) \end{aligned}$$

$$C_x(t_1, t_2) = \text{VAR}(\zeta) \cos(t_1) \cos(t_2)$$

Example: Sinusoidal Phase

- Let Θ be a uniform random variable over the interval $(-\pi, \pi)$, and let the random process $X(t, \Theta)$:

$$X(t) = \cos(t + \Theta)$$

Find the mean, variance, autocorrelation, and autocovariance of $X(t, \Theta)$

Example: Sinusoidal Phase

- The mean:

$$m_x(t) = E[\cos(t + \Theta)]$$

$$m_x(t) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos(t + \theta) d\theta$$

$$m_x(t) = 0$$

Example: Sinusoidal Phase

- The autocorrelation and autocovariance functions

$$C_x(t_1, t_2) = R_x(t_1, t_2) - m_x(t_1)m_x(t_2)$$

$$C_x(t_1, t_2) = R_x(t_1, t_2)$$

$$C_x(t_1, t_2) = E[\cos(t_1 + \Theta)\cos(t_2 + \Theta)]$$

Example: Sinusoidal Phase

- The autocorrelation and autocovariance functions

$$C_x(t_1, t_2) = E[\cos(t_1 + \Theta)\cos(t_2 + \Theta)]$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{1}{2} \{\cos(t_1 - t_2) + \cos((t_1 + t_2) + 2\theta)\} d\theta$$

$$C_x(t_1, t_2) = R_x(t_1, t_2) = \frac{1}{2} \cos(t_1 - t_2)$$

Example: Sinusoidal Phase

- The variance:

$$C_x(t_1, t_2) = R_x(t_1, t_2) = \frac{1}{2} \cos(t_1 - t_2)$$

$$C_x(t, t) = \text{VAR}(X(t))$$

$$\text{VAR}(X(t)) = \frac{1}{2}$$