

CM26-230525Thu

Announcements

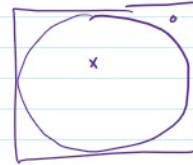
1. Final Exam will be taken on Wednesday 6/7 19:00-22:00 at LG 104.
2. Topics of Voluntary Projects must be discussed with the Lecturer before you take the final exam.
3. Handouts, annotated handouts are available at <http://cisl.postech.ac.kr/class/eece574/schedule.htm> . Passwords are ...
4. Lecture videos are available at cisl.postech YouTube channel. Clickable time stamps, please.
5. HWs will be assigned. Check the PLMS.

Corrections; Clarifications; Common mistakes, misunderstandings, misconceptions, myth; Frequently Asked Questions; Conventional wisdom

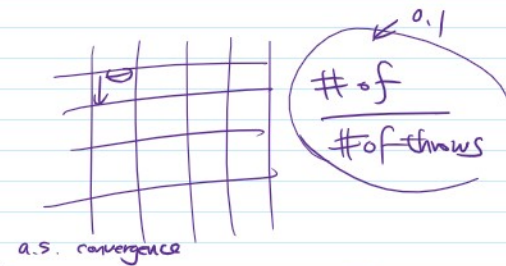
$$\limsup_{n \rightarrow \infty} x_n \quad \limsup_{n \rightarrow \infty} \liminf_{m \rightarrow \infty} x_n \quad \liminf_{n \rightarrow \infty} \limsup_{m \rightarrow \infty} x_n$$

Overview w/ Chapter numbers from the main textbook by Papoulis, (2)-(7) from Peebles

- Pa 9.1 Definitions
- Pe 6 Random Process--Temporal Characteristics
- Pa 7.4 Stochastic Convergence and Limit Theorems
- Pa 9.2 Systems with Stochastic Inputs
- Pa 9.3 The Power Spectrum
- Pe 7 Random Process--Spectral Characteristics
- Pe 8 Linear Systems with Random Inputs
- Pe 9 Optimal Linear Systems



$$P(\liminf_{n \rightarrow \infty} E_n^c) = 1$$



$$P(\limsup_{n \rightarrow \infty} E_n) = 0$$

Review

1. a limit point of a sequence of real numbers
2. the $\limsup$ and the $\liminf$ of a sequence of real numbers
3. the $\limsup$ and the $\liminf$ of a sequence of events
4. the Borel-Cantelli lemma to prove a convergence w.p.1
5. Monte-Carlo integration/simulation
6. Summary of 4 modes of the convergence of a sequence of random variables

$$\int_0^1 g(x) dx = \int_0^1 g(x) f(x) dx = E\{g(x)\}$$

Preview

1. (Pa 9.5) Stochastic Calculus
2. (Pa 9.1) Systems with stochastic inputs

$$\text{Buffon's needle} \approx \frac{g(x_1) + \dots + g(x_n)}{n}$$

GENERAL CONCEPTS

9 - 1

Chapter 9 GENERAL CONCEPTS

Handout by Joon Ho Cho
based on A. Papoulis & S. U. Pillai, Probability, Random Variables and Stochastic Processes, 4th ed.

EECE 574, Spring 2023

9.1 Definitions

9.5 APPENDIX 9A Continuity, Differentiation, Integration

9.2 Systems with Stochastic Inputs

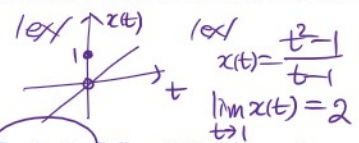
9.3 The Power Spectrum

9.4 Spectral Characteristics of System Response

9.5 Bandpass Processes

9.6 Sampling of Processes

9.7 Discrete-Time Processes



Def. $x: \mathbb{R} \rightarrow \mathbb{R}$ is continuous at t_0

$$\text{if } \lim_{t \rightarrow t_0} x(t) = x(\lim_{t \rightarrow t_0} t) = x(t_0)$$

Def. $x: \mathbb{R} \rightarrow \mathbb{R}$ is continuous

if x is " " for all $t \in \mathbb{R}$

Def. Given $x: \mathbb{R} \rightarrow \mathbb{R}$, we say

$$\lim_{t \rightarrow t_0} x(t) = c$$

$x(t)$ is well approximated by c
for sufficiently close t to t_0 .

if $\forall \epsilon > 0, \exists \delta(\epsilon) > 0$ s.t.

$$|t - t_0| < \delta(\epsilon) \Rightarrow |x(t) - c| < \epsilon.$$

$$\text{ex/ } x(t) = \frac{1}{t}$$

$\lim_{t \rightarrow 0} x(t)$ does Not exist.

GENERAL CONCEPTS

9 - 3

9.5 Continuity, Differentiation, Integration

9.5 Continuity, Differentiation, Integration

Q1: What is $X(t_0)$ given a RP $X(t)$? At. a RV.

Q2: What is

$$\lim_{t \rightarrow t_0} X(t)$$

?

- Since a RP is a collection of RVs parameterized by t , it means $\lim_{t \rightarrow t_0} X(t, s)$
- If the limit exists in some sense, then it must be a RV.
- We need to use the notion of the convergence of a sequence of RVs to a RV.

≅ "What is $X(t)$ for sufficiently close t to t_0 ?"

$$X(t, s)$$

What do we mean by the continuity

$$\lim_{t \rightarrow t_0} X(t) = X(t_0)$$

?

- a.s.
- in the p th mean
- in probability
- in distribution

$$\lim_{t \rightarrow t_0} X(t, s) = X(t_0, s)$$

4 modes of the convergence of a sequence of RVs

We mainly consider the notion of the limit in the mean-square sense, i.e., we are interested in the case with

$$\text{l.i.m.}_{t \rightarrow t_0} X(t) = X(t_0)$$

$$\text{l.i.m.}_{t \rightarrow t_0} X(t) = X(t_0)$$

$$P(X(t_0) < \alpha)$$

$$\simeq P(\lim_{t \rightarrow t_0} X(t) < \alpha) \simeq \lim_{t \rightarrow t_0} P(X(t) < \alpha)$$

$$P(X(t_0 + \epsilon) < \alpha)$$

for s. small ϵ ! -

→ Def. $E[|X(t_0 + \epsilon) - X(t_0)|^2] \rightarrow 0$ as $\epsilon \rightarrow 0$

Mean-Square Calculus

- From Chapter 9 of Stark and Woods, *Probability, Random Processes, and Estimation Theory for Engineers*, 2nd ed.
- Def. A random process $X(t)$ is **continuous in the mean-square sense at the point t** if

→ Def. $E[|X(t + \epsilon) - X(t)|^2] \rightarrow 0$, as $\epsilon \rightarrow 0$. $\forall t$

- Def. A random process $X(t)$ is **mean-square (m.-s.) continuous** if $X(t)$ is continuous in the mean-square sense for all t .

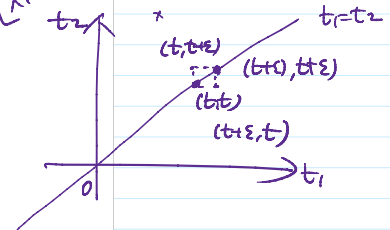
Theorem. The random process $X(t)$ is m.-s. continuous at t iff $R_{XX}(t_1, t_2)$ is continuous at the point $t_1 = t_2 = t$.
proof.

$$E[|X(t + \epsilon) - X(t)|^2] = R_{XX}(t + \epsilon, t + \epsilon) - R_{XX}(t + \epsilon, t) - R_{XX}(t, t + \epsilon) + R_{XX}(t, t)$$

Clearly the right-hand side goes to zero as $\epsilon \rightarrow 0$ if the deterministic function R_{XX} is continuous at $t_1 = t_2 = t$. The only if part is somewhat complicated and will be omitted.

Corollary. A WSS random process is m.-s. continuous for all t iff $R_{XX}(\tau)$ is continuous at $\tau = 0$.

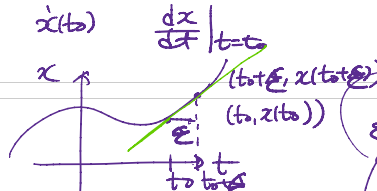
$$\simeq E[X(t_0)X(t_0)]$$



(S) Given a CT signal $x(t)$, what do we mean by $\dot{x}(t)$

$$\lim_{m \rightarrow \infty} \frac{m}{m+n}$$

$$\dot{x}(t_0) \leftarrow \frac{dx}{dt} \bigg|_{t=t_0}$$



$$\lim_{\epsilon \rightarrow 0} \frac{x(t_0 + \epsilon) - x(t_0)}{(t_0 + \epsilon) - t_0} \simeq \dot{x}(t_0)$$

Def. The random process $X(t)$ has a **mean-square derivative at t** if the mean-square limit of $[X(t + \epsilon) - X(t)]/\epsilon$ exists as $\epsilon \rightarrow 0$.

Theorem. A random process $X(t)$ with autocorrelation function $R_{XX}(t_1, t_2)$ has a m.-s. derivative at time t iff $\partial^2 R_{XX}(t_1, t_2) / \partial t_1 \partial t_2$ exists at $t_1 = t_2 = t$.

Theorem. If a random process $X(t)$ with mean function $\mu_X(t)$ and correlation function $R_{XX}(t_1, t_2)$ has an m.-s. derivative $X'(t)$, then the mean and correlation function of $X'(t)$ are given by

$$E\left[\frac{X(t_0 + \epsilon) - X(t_0)}{\epsilon}\right] \rightarrow 0 \text{ as } \epsilon \rightarrow 0, \forall t$$

$$E[X'(t)] = \mu_{X'}(t) = \frac{d\mu_X(t)}{dt}$$

and

$$R_{X'X'}(t_1, t_2) = \partial^2 R_{XX}(t_1, t_2) / \partial t_1 \partial t_2$$

Theorem. The m.-s. derivative of a WSS random process $X(t)$ exists at t if the

Def.

$$\frac{X(t_0 + \epsilon) - X(t_0)}{\epsilon}$$

Def. $\dot{x}(t)$ is the derivative of $x(t)$

Def.

$$\dot{x}(t_0) = \lim_{\epsilon \rightarrow 0} \frac{x(t_0 + \epsilon) - x(t_0)}{\epsilon}$$

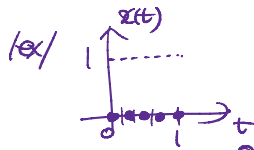
$$\dot{x}(t_0) = \text{l.i.m.}_{\epsilon \rightarrow 0} \frac{x(t_0 + \epsilon) - x(t_0)}{\epsilon}$$

and

$$R_{X',X'}(t_1, t_2) = \frac{\partial^2 R_{XX}(t_1, t_2)}{\partial t_1 \partial t_2}$$

- Theorem. The m.-s. derivative of a WSS random process $X(t)$ exists at t if the autocorrelation function $R_{XX}(\tau)$ has derivatives up to order two at $\tau = 0$.

$$\dot{X}(t_0) = \dot{X}(t_0) = \lim_{\varepsilon \rightarrow 0} \frac{X(t_0 + \varepsilon) - X(t_0)}{\varepsilon}$$



$$x(t) = \begin{cases} 1, & t \in \mathbb{Q} \cap [0, 1] \\ 0, & t \in \mathbb{Q}^c \cap [0, 1] \end{cases}$$

$$Q. \int_0^1 x(t) dt \quad 0 = t_0 < t_1^{(n)} < t_2^{(n)} \dots < t_n^{(n)} = 1$$

$$(15) \quad \int_{T_1}^{T_2} x(t) dt \quad \int_0^1 x(t) dt = \int_0^1 x(t) dt = \lim_{n \rightarrow \infty} \sum_{i=1}^n \Delta_i^{(n)} \max_{t \in I_i^{(n)}} \{x(t)\}$$

$$\lim_{n \rightarrow \infty} \sum_{i=1}^n \Delta_i^{(n)} \min_{t \in I_i^{(n)}} \{x(t)\}$$

the Lebesgue integral.

