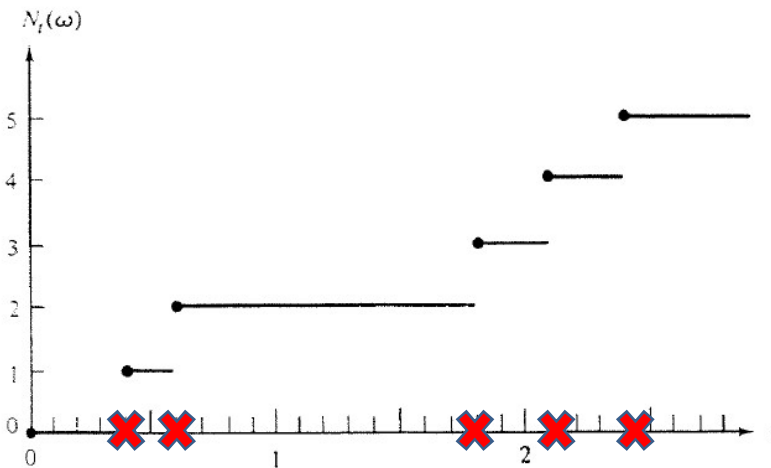


457.643 Structural Random Vibrations
In-Class Material: Class 07

II-1. Random Process (contd.)

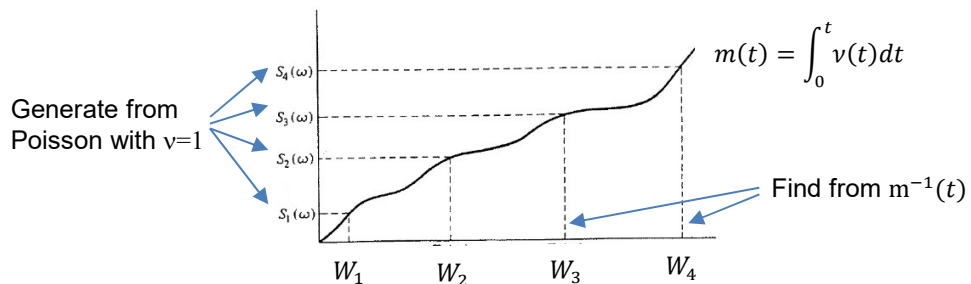
⊙ Artificial generation of Poisson process



- Actually, we aim to generate waiting times (arrival times) for $n = 1, 2, 3, \dots$, i.e. W_1, W_2, \dots
- For a homogeneous Poisson process, we can generate T_n using W_1 , and
- T_n and W_1 follows _____ distribution
- In Matlab®, one can generate _____ random variables using `exprnd(μ, M, N)`
 - 1) μ : mean = $1/\nu$
 - 2) M, N : size of the output matrix
- To generate non-homogeneous Poisson process, need to use a theorem,

$W_i = m^{-1}(S_i), i = 1, 2, \dots$, are arrival times of the non-homogeneous Poisson process with $m(t)$ when $S_i, i = 1, 2, \dots$, are arrival times of the homogeneous Poisson process with $\nu = 1$

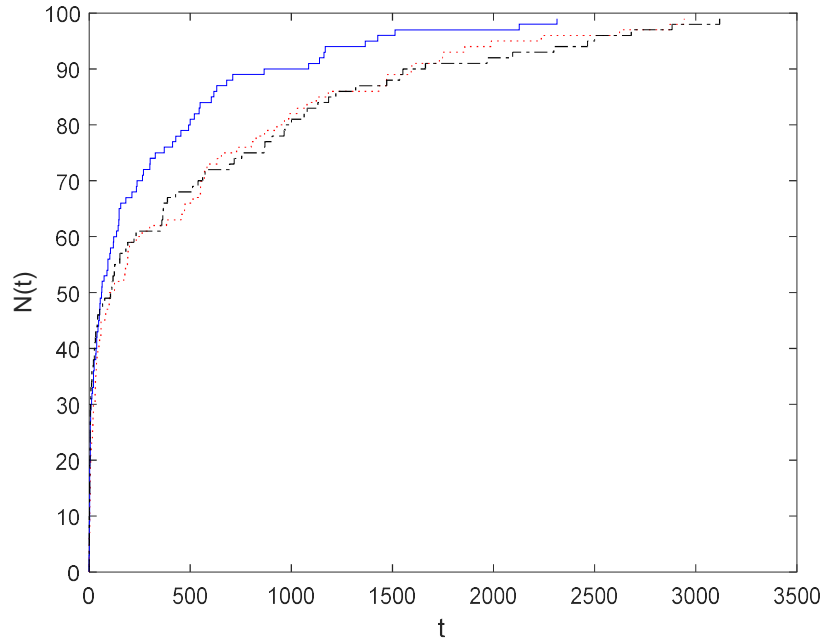
Cinlar, E. (1975). *Introduction to Stochastic Processes*, Dover Books (reprinted in 2013)



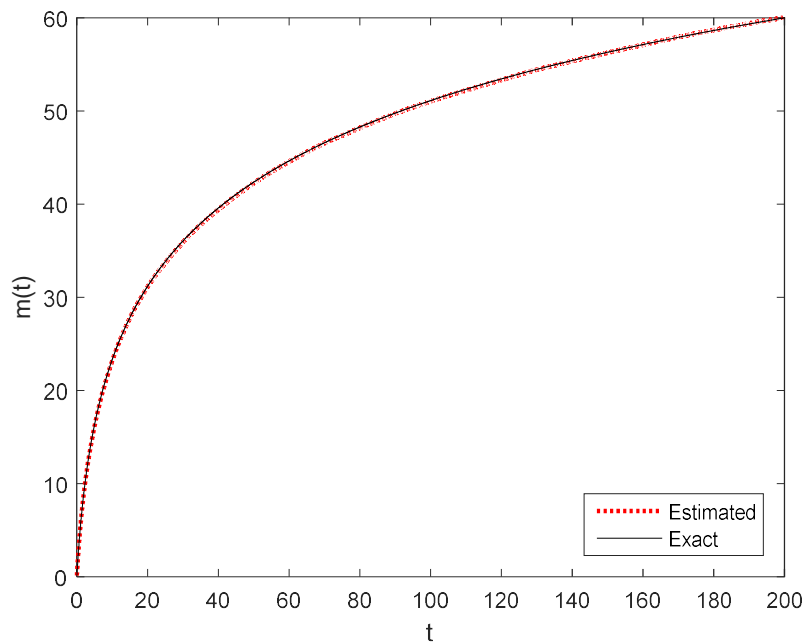
- **Example:** Generating NHPP with $m(t) = 13 \cdot \ln(0.5t + 1)$ (Byun et al. 2017)

Byun J., H.-M. Noh, J. Song (2017). [Reliability growth analysis of k-out-of-N systems using matrix-based system reliability method](#). *Reliability Engineering & System Safety*. Vol. 165, 410-421.

1) Three random samples:



2) Comparison between exact $m(t)$ and estimated one using 1,000 samples



** Check "[NHPOissonGenerationTest.m](#)" at eTL website for details

© **Normal (Gaussian) process (Read L&S 4.10)**

$X(t)$ is a Gaussian process

if, for any n , and any $\{t_1, t_2, \dots, t_n\}$,

the random variables $X(t_1), \dots, X(t_n)$ are _____

- The process is completely defined by specifying _____ for $\forall t$ and _____ for $\forall (t, s)$
- For a Gaussian process, being “weakly stationary” implies stationarity in the _____ sense
- Any linear function of Gaussian processes is a _____ process
 e.g. $\dot{X}(t)$ is Gaussian if $X(t)$ is Gaussian (why?)
 e.g. $X(t)$ and $\dot{X}(t)$ are _____
- Why useful?
 - 1) Convenient to handle
 - 2) _____ theorem
- Hard to justify Gaussian process assumption if
 - 1) the distribution is not symmetric, or
 - 2) _____ is not equal to 3
- Textbook focusing on non-Gaussian processes: M. Grigoriu (1995), *Applied Non-Gaussian Processes*

© **Jointly Gaussian processes**

$X_1(t), X_2(t), \dots, X_m(t)$ are jointly Gaussian processes

if, for any n , and any $\{t_1, t_2, \dots, t_n\}$, the random variables

$\{X_1(t_1), \dots, X_1(t_n), X_2(t_1), \dots, X_2(t_n), \dots, X_m(t_1), \dots, X_m(t_n)\}$ are

- The processes are completely defined by specifying $\mathbf{M}_X(t) = \{ \quad \quad \quad \}^T$ and $\Sigma_{(t,s)} = [\quad \quad]$

© **Other notable processes**

Wiener process: Coordinate of a particle doing _____ motion

- Increments $B(t_1) - B(t_2)$ follow zero-mean Gaussian and s.i. (non-overlapping)
- $E\{[B(t_2) - B(t_1)]^2\} = 2D|t_2 - t_1|$

Non-Gaussian processes:

- Memoryless nonlinear transformation $X(t) = g[Y(t)]$

- Nonlinear transformation with memory
e.g. Diffusion process $dX(t) = m[X(t), t]dt + \sigma[X(t), t]dB(t)$

Applications of Homogeneous and Non-homogeneous Poisson processes:

- Systems with repairable components

Der Kiureghian, A., O. Ditlevsen, and J. Song (2007). [Availability, reliability and downtime of systems with repairable components](#). *Reliability Engineering & System Safety*, Vol. 92, 231-242.

- Reliability growth of components during the test period

Byun J., H.-M. Noh, J. Song (2017). [Reliability growth analysis of k-out-of-N systems using matrix-based system reliability method](#). *Reliability Engineering & System Safety*. Vol. 165, 410-421.

Renewal processes: generalization of Poisson process

Pandey, M., and van der Weide, J.A.M. (2017). [Stochastic renewal process models for estimation of damage cost over the life-cycle of a structure](#). *Structural Safety*. Vol. 67, 27-38.

II-2. Stochastic Calculus

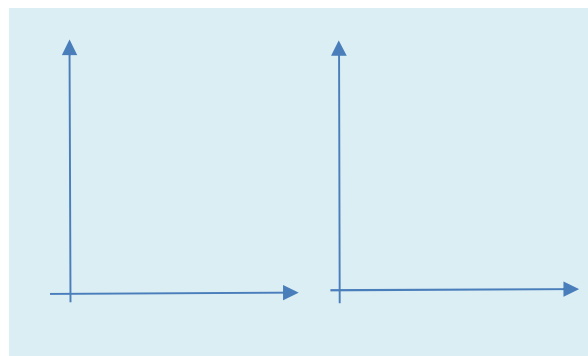
Lin, Y.K. (1967) *Probabilistic Theory of Structural Dynamics*, McGraw-Hill, New York, NY.

© Motivation

$$\frac{d}{dt} X(t) = \lim_{h \rightarrow 0} \frac{X(t+h) - X(t)}{h}$$

The conventional “limit” cannot be applied to random processes

Limit of a random process?



Need to consider the convergence of a sequence of random variables, i.e.

$$\lim_{n \rightarrow \infty} \{X_1, X_2, \dots, X_n\}$$



- ➔ Converging to the distribution of a random variable (not a particular value)
- ➔ “_____ Convergence”

© **Definitions of stochastic convergence**

- 1) Convergence with probability 1 (“almost sure” convergence)

$$P\left(\lim_{n \rightarrow \infty} X_n = X\right) =$$

- 2) Convergence in probability

$$\lim_{n \rightarrow \infty} P(|X_n - X| \geq \epsilon) = 0, \forall \epsilon > 0$$

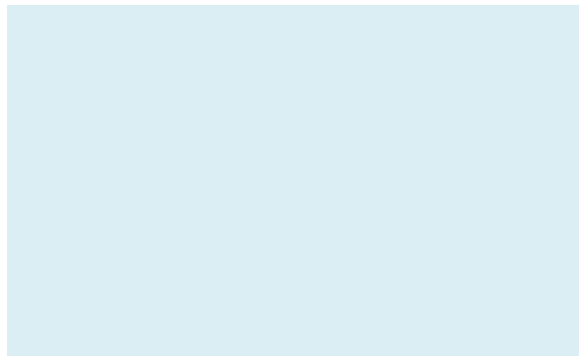
- 3) Convergence in distribution

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x)$$

- 4) ** Convergence in the mean square

$$\lim_{n \rightarrow \infty} E[|X_n - X|^2] = 0$$

→ requires $E[X^2] < \infty$, i.e. “**square-integrable**” process



Throughout this course, we use the fourth definition with the notation

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

to describe “**Limit In the Mean-square**”

© **Two theorems for limit in the mean square**

Theorem 1:

$$\text{If } \mathbf{l.i.m.}_{t \rightarrow t_0} X(t) = X \text{ and } \mathbf{l.i.m.}_{s \rightarrow s_0} Y(s) = Y, \text{ then } \lim_{t \rightarrow t_0, s \rightarrow s_0} E[X(t) \cdot Y(s)] = E[X \cdot Y]$$

Proof:

Using Theorem 1, we can show $\lim_{t \rightarrow t_0} E[X(t)] = E \left[\text{l. i. m.}_{t \rightarrow t_0} X(t) \right]$

Namely, $E[\cdot]$ and l. i. m. are c_____ or exchangeable

Proof:

Theorem 2:

$\text{l. i. m.}_{t \rightarrow t_0} X(t) = X$ $\phi_{XX}(t, s)$ is continuous at (t_0, t_0) no matter how (t, s) approaches (t_0, t_0)

- See Ex 4.9 in L&S
- Of course, for “second-order process”

Proof: