



# **STA 4103/5107**

# **Computational Methods**

# **in Statistics II**

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Class 1  
January 10, 2017



## 5.4 Simulating Stochastic Processes



# Stochastic Process

- We can extend the ideas of simulating random variables to simulating stochastic processes.
- To start with, we briefly introduce the notion of **stochastic process**.
- So far we have dealt only with real-valued, i.e. scalar, random variables but now we extend to a collection of random variables.
- **Definition 15** A stochastic process is an indexed collection of random variables.
- Most often, the indexing variable is time.



# Continuous and Discrete Processes

- For example let  $X$  be a stochastic process, indexed by time, such that  $X_t$  is a real-valued random variable. For a fixed time  $t$ ,  $X_t$  is just a random variable.
- Each realization, or sample, of  $X_t$  is now a function of time, also called a sample path.
- In case the indexing variable  $t$  is continuous, the stochastic process is called a **continuous-time process**; else, it is called a **discrete-time process**.
- Similarly, if the space in which  $X_t$  takes values is continuous, then the process is called a **continuous-valued process**. Otherwise, it is called a **discrete-valued process**.



# Discrete-Time White Noise

- A **discrete white noise process** is defined as a process  $\varepsilon_t$  which is identically and independently distributed (i.i.d) with zero expectation and  $\sigma^2$  variance. Hence,

$$E(\varepsilon_t) = 0$$

$$E(\varepsilon_s \varepsilon_t) = \begin{cases} \sigma^2 & \text{if } s = t \\ 0 & \text{if } s \neq t \end{cases}$$

- In discrete-time models, a white noise process is often normally distributed (**Gaussian white noise**).
- A white noise process can also be distributed by any other distribution as long as the i.i.d. assumption is valid.



# Random Walk

- A **random walk** is one of the simplest examples of a discrete time, discrete state stochastic process.
- The process transitions only at times separated by a unit time  $T$ , and takes values only in the set of integer multiples of  $s$ .
- At every transition time  $nT$ , for  $n = 1, 2, \dots$ , toss a coin and depending on the outcome, add  $+s$  for a head or  $-s$  for a tail to the current value of the process.
- Assume we start with process at time zero with the process value being zero. At a transition time  $nT$ , the value of the process is

$$X(nT) = ks - (n-k)s$$

where the number of heads in  $n$  tosses is  $k$ .



# Probability Formula

- That is,

$$X(nT) = ms, \text{ where } m = (2k - n) \text{ or } k = (m + n)/2.$$

- Since  $k \in \{0, 1, 2, \dots, n\}$ , the possible values for  $m$  are  $\{-n, \dots, 0, \dots, n\}$ .
- Assuming a fair coin, the probability that  $X(nT) = ms$  is given by

$$P\{X(nT) = ms\} = \binom{n}{(m+n)/2} \left(\frac{1}{2}\right)^n$$

- Given this probability, one can analyze this random walk for any time  $nT$  (e.g. computing the mean and variance of  $X(nT)$ ).



## Sum of Random Variables

- Another interpretation of  $X(nT)$  is given by the following:

$$X(nT) = X_1 + X_2 + \dots + X_n,$$

where  $X_i$ s are i.i.d random variables.

- Each  $X_i$  takes on the value either  $+s$  or  $-s$  with equal probabilities; it is easy to show that  $E[X_i] = 0$  and  $\text{var}(X_i) = s^2$ .
- $X(nT)$  is a sum of independent and identically distributed random variables with mean zero and variance  $s^2$ , or **a sum of discrete white noise with variance  $s^2$** .
- Therefore,  $X(nT)$  is a random variable with mean zero and variance  $ns^2$ .





## Limiting Situation

- Setting  $s = \alpha T^{1/2}$  for some positive number  $\alpha$ . In the limit  $T \rightarrow 0$ , the random walk becomes a continuous time, continuous state process.
- Let  $t = nT$  denote the time index of the process;  $t$  is a continuous random variable in the limiting case.
- Consider the random process at time  $t$ .  $X(t)$ , through its relation with  $X(nT)$ , has the following statistics:

$$E[X(t)] = E[X(nT)] = 0$$

$$\text{var}(X(t)) = \text{var}(X(nT)) = ns^2 = \alpha^2 T t/T = \alpha^2 t.$$



# Wiener Process

- That is,  $X(t)$  is a continuous random variable with mean zero and variance  $\alpha^2 t$ .
- This limiting process is called the **Wiener process** (also called **Brownian motion**).
- **(Formal Definition) Wiener Process** is a continuous-time stochastic process  $X(t)$  for  $t \geq 0$  with  $X(0) = 0$  and such that
  - the increment  $X(t) - X(s)$  is Gaussian with mean 0 and variance  $\alpha^2(t - s)$  for any  $0 \leq s < t$ .
  - increments for non-overlapping time intervals are independent.
- For  $\alpha = 1$ , it is called the **standard Wiener process**.



# Properties

- 1. If  $(t_1, t_2)$  and  $(t_3, t_4)$  are non-overlapping intervals, then  $X(t_2) - X(t_1)$  and  $X(t_4) - X(t_3)$  are statistically independent random variables.
- 2. The increment  $X(t_2) - X(t_1)$  is independent of the value  $X(t_1)$  or any past value of the process. In fact, this increment is a Gaussian random variable with mean zero and variance  $\alpha^2(t_2 - t_1)$ .
- 3. For any two times  $t_1 < t_2$ , we have:

$$\begin{aligned} E[X(t_1)X(t_2)] &= E[X(t_1)(X(t_2) - X(t_1) + X(t_1))] \\ &= E[X(t_1)^2] + E[X(t_1)(X(t_2) - X(t_1))] \\ &= \alpha^2 t_1 + E[X(t_1)]E[X(t_2) - X(t_1)] \\ &= \alpha^2 t_1. \end{aligned}$$



## Example

- **Simulate a random walk**

For a given value of  $T$  and  $s = \alpha T^{1/2}$ , plot the sample paths of the  $X_t$  for  $\alpha = 1.0$  and  $T = 1, 0.1, 0.01$ , and  $0.001$ .

Choose the total number of steps  $n$  to be  $10/T$ .

So the total time is constant 10.