

Lecture 6: Gaussian Processes

Stochastic Process

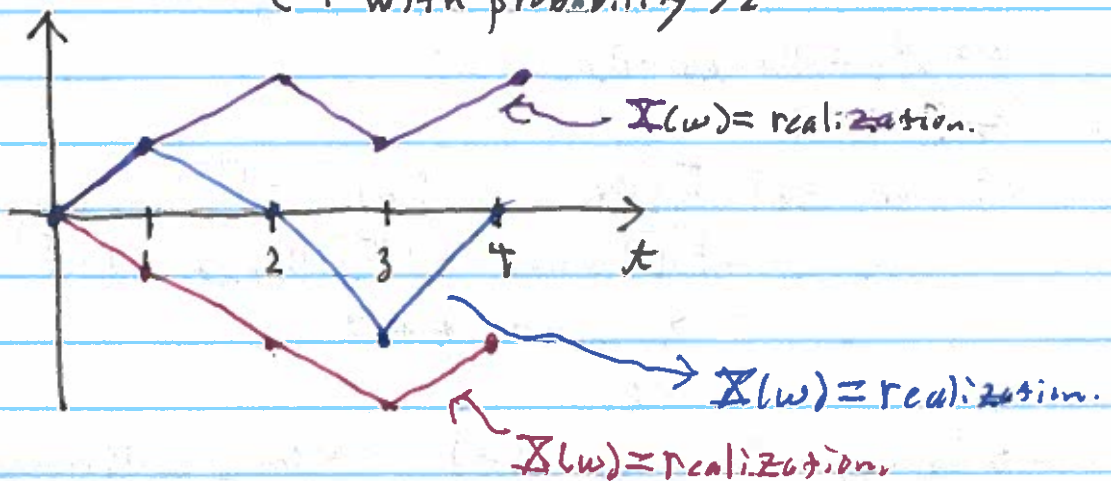
$X = (X_t, t \in \mathcal{T})$, on (Ω, \mathcal{F}, P) ,
where \mathcal{T} is discrete or continuous, e.g.
 $\mathcal{T} = \{0, 1, 2, 3, \dots\}$ or $\mathcal{T} = [0, \infty)$.

Examples

1. $\mathcal{T} = \{0, 1, 2, 3, \dots\}$

$\Rightarrow X_0 = 0$

$$X_{i+1} - X_i = \begin{cases} 1 & \text{with probability } 1/2 \\ -1 & \text{with probability } 1/2 \end{cases}$$



2. $\mathcal{T} = [0, \infty)$

$X_0 = 0$

$$X_{t+\Delta t} - X_t \sim N(0, \sigma^2 \Delta t)$$

Gaussian i.i.d. mean 0 and variance $\sigma^2 \Delta t$.

Definition - A Gaussian process $X = (X_t, t \in \mathbb{R})$ is a stochastic process whose finite-dimensional distributions are Gaussian, i.e., for all $n \in \mathbb{N}$ and any t_1, \dots, t_n , $(X_{t_1}, \dots, X_{t_n})$ is Gaussian.

The distribution is determined by

$$m(t) = \mathbb{E}[X_t]$$

$$C(s, t) = \text{Cov}[X_s, X_t].$$

Examples:

1. Brownian Motion:

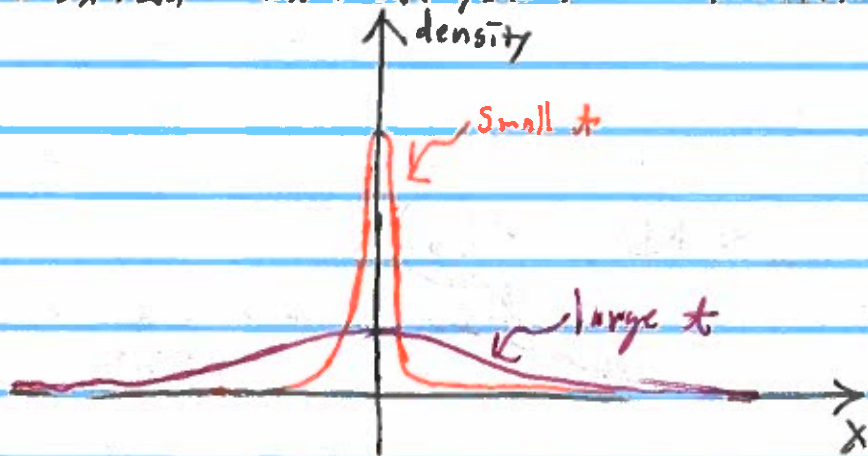
$(B_t, t \geq 0)$ is the Gaussian process with mean $m(t) = \mathbb{E}[B(t)] = 0$
and $\text{Cov}(B_s, B_t) = \mathbb{E}[B_s B_t] = \min(s, t)$.

$$\Rightarrow Z_t = B_{t+\Delta t} - B_t$$

$$\Rightarrow \mathbb{E}[Z_t] = 0$$

$$\begin{aligned} \Rightarrow \text{Var}[Z_t] &= \mathbb{E}[(B_{t+\Delta t} - B_t)^2] - \mathbb{E}[B_{t+\Delta t} - B_t]^2 \\ &= \mathbb{E}[B_{t+\Delta t}^2] - 2\mathbb{E}[B_{t+\Delta t} B_t] + \mathbb{E}[B_t^2] \\ &= t + \Delta t - 2t + t \\ &= \Delta t. \end{aligned}$$

$$\Rightarrow B_{t+\Delta t} = B_t + N(0, \Delta t) \rightarrow \text{Brownian Motion}$$



2. Brownian Motion with Drift

$$\text{Let } X_t = \sigma B_t + \nu t.$$

$$\Rightarrow \mathbb{E}[X_t] = \nu t$$

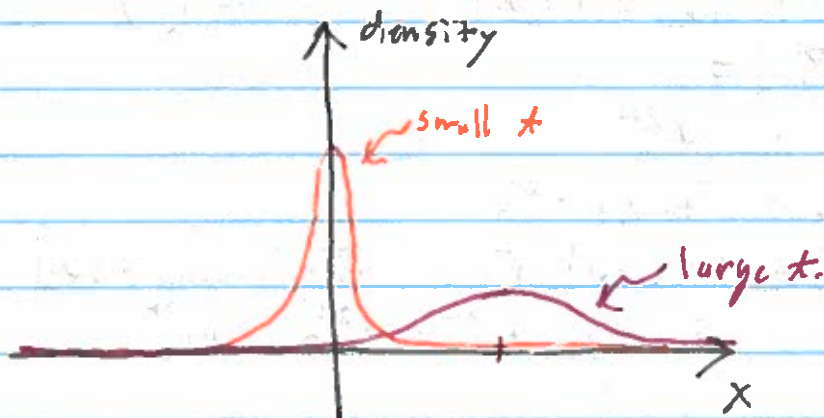
$$\begin{aligned} \text{Cov}[X_t, X_s] &= \text{Cov}[\sigma B_t + \nu t, \sigma B_s + \nu s] \\ &= \sigma^2 \text{Cov}(B_t, B_s) + \nu \sigma \text{Cov}(t, B_s) \\ &\quad + \nu \sigma \text{Cov}(B_t, s) + \nu^2 \text{Cov}(s, t) \end{aligned}$$

$$* \text{Cov}(t, B_s) = \mathbb{E}[t B_s] - \mathbb{E}[t] \mathbb{E}[B_s] = 0$$

$$\Rightarrow \text{Cov}[X_t, X_s] = \sigma^2 \min(s, t).$$

Therefore,

$$\text{Var}(X_t) = \sigma^2 t.$$



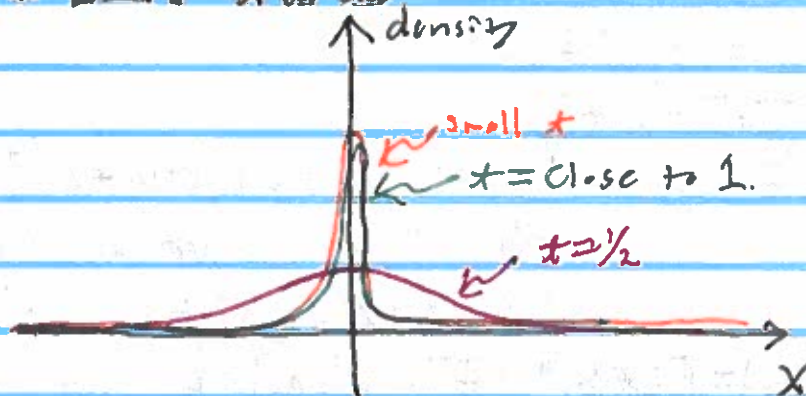
3. Brownian Bridge

$$Z_t = B_t - t B_1, \quad t \in [0, 1]$$

$$\Rightarrow \mathbb{E}[Z_t] = 0$$

$$\begin{aligned} \Rightarrow \text{Cov}[Z_x, Z_s] &= \text{Cov}[B_x - x B_1, B_s - s B_1] \\ &= \text{Cov}[B_x, B_s] + s x \text{Cov}[B_1, B_1] \\ &\quad - x \text{Cov}[B_1, B_s] - s \text{Cov}[B_x, B_1] \\ &= \min(s, x) + s x - x - s \\ &= \begin{cases} s(1-x) & \text{if } s \leq x \\ x(1-s) & \text{if } x \leq s \end{cases} \end{aligned}$$

$$\Rightarrow \text{Var}[Z_t] = t(1-t)$$



4. Fractional Brownian Motion

Let $0 < H < 1$

Define: $\text{Cov}(Y_s, Y_t) = \frac{1}{2}(s^{2H} + t^{2H} - |t-s|^{2H})$

$$\mathbb{E}[Y_t] = 0$$

$$\Rightarrow \text{Var}[Y_t] = t^{2H}$$

$$\Rightarrow \text{Let } Z_t = Y_{t+\Delta t} - Y_t$$

$$\begin{aligned} \Rightarrow \text{Var}[Z_t] &= \text{Cov}[Y_{t+\Delta t} - Y_t, Y_{t+\Delta t} - Y_t] \\ &= (t+\Delta t)^{2H} + t^{2H} - |\Delta t|^{2H} \end{aligned}$$