

# Brownian bridge

ICEF

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# Gaussian Process

**Def.** A Gaussian process  $X(t)$ ,  $t \geq 0$ , is a stochastic process that has the property that, for arbitrary times  $0 < t_1 < t_2 < \dots < t_n$ , the random variables  $X(t_1), X(t_2), \dots, X(t_n)$  are **jointly normally distributed**.

$$I(t) = \int_0^t \Delta^2 dW(s), \Delta - \text{non random}$$

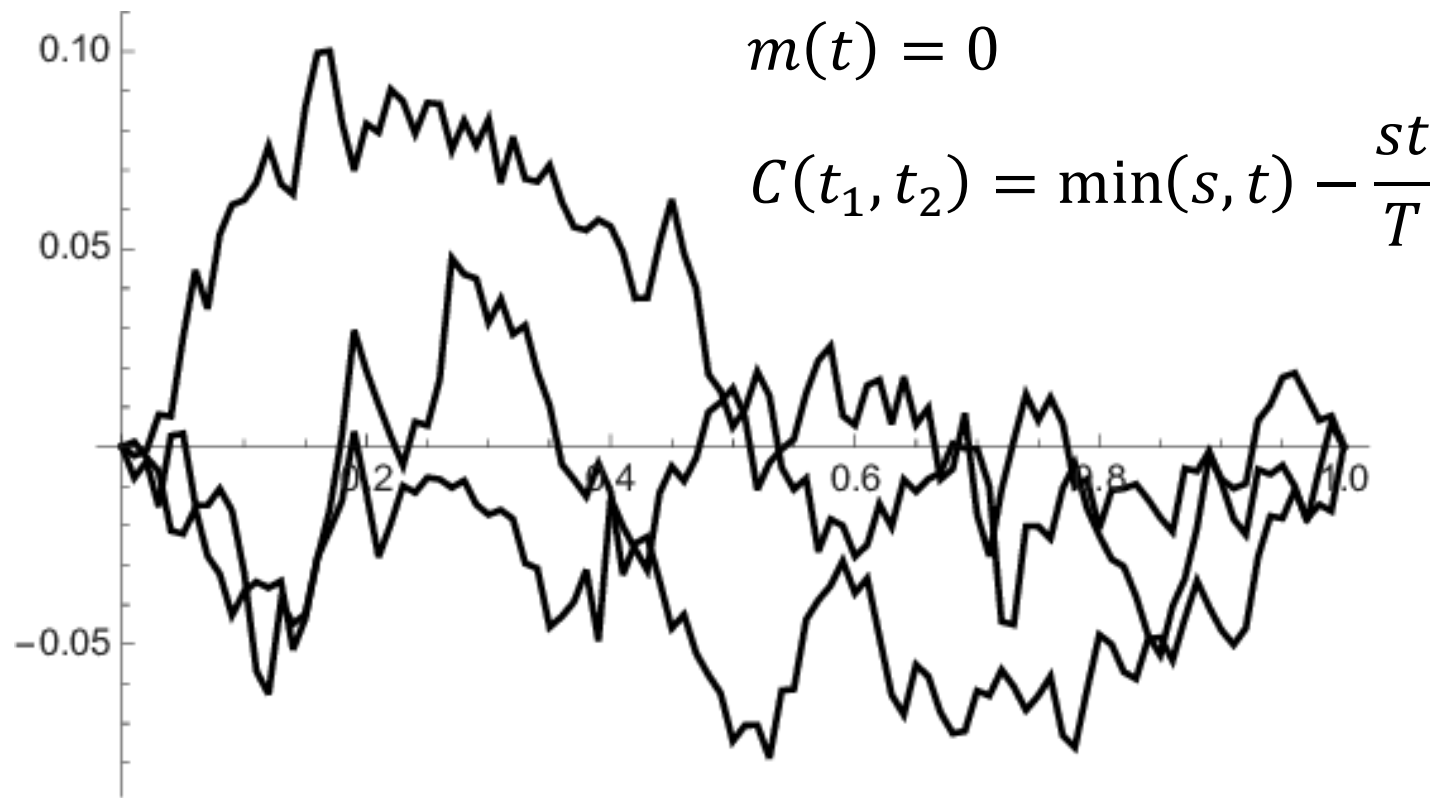
**Proof:** Shreve

$$C(t_1, t_2) = \int_0^{t_1} \Delta^2 ds, t_1 \leq t_2$$

**Proof:** on the board

# Brownian Bridge

**Define:**  $B(t) = W(t) - \frac{t}{T}W(T), 0 \leq t \leq T$



# Brownian Bridge

**Define:** Brownian bridge from a to b on  $[0, T]$

$$B^{a \rightarrow b}(t) = a + \frac{(b - a)t}{T} + B(t), 0 \leq t \leq T; a, b \in \mathbb{R}$$

$$m^{a \rightarrow b}(t) = a + \frac{(b - a)t}{T}, 0 \leq t \leq T$$

$$C^{a \rightarrow b}(t_1, t_2) = C(t_1, t_2) - \text{not affected}$$

# Brownian bridge as a stochastic integral

## Issues:

- The Brownian bridge  $B(t)$  is not adapted to the filtration  $\mathcal{F}(t)$  generated by  $W(t)$
- $\mathbb{E}[B^2(t)] = t - \frac{t^2}{T}$  contradicts properties of Ito integral of a Gaussian process

## Define:

$$Y(t) = \begin{cases} (T - t) \int_0^t \frac{1}{T-s} dW(s), & 0 \leq t < T \\ 0, & t = T \end{cases} \text{ - Gaussian process}$$

**Proof:** on the board, pdf of Brownian bridge and Kolmogorov second equation

$$m^Y(t) = 0$$

$$c^Y(t_1, t_2) = \min(t_1, t_2) - \frac{t_1 t_2}{T}, \forall t_1, t_2 \in [0, T]$$

# Brownian bridge as a stochastic integral

$$dY(t) = -\frac{Y(t)}{T-t}dt + dW(t), 0 \leq t \leq T$$

## Transition density of a Brownian bridge from 0 to 0

$$\frac{p(T - t_n, x_n, b)}{p(T, a, b)} \prod_{j=1}^n p(t_j - t_{j-1}, x_{j-1}, x_j), \text{ where } p(\tau, x, y) = \frac{1}{\sqrt{2\pi\tau}} \exp\left\{-\frac{(y-x)^2}{2\tau}\right\}$$

**Proof:** Shreve

# Kolmogorov-Smirnov test

**Gist:** Non-parametric of the equality of one-dimensional distributions

**Define:** K-S statistic:

$$D_n = \sup |F_n(x) - F(x)| \text{ or}$$

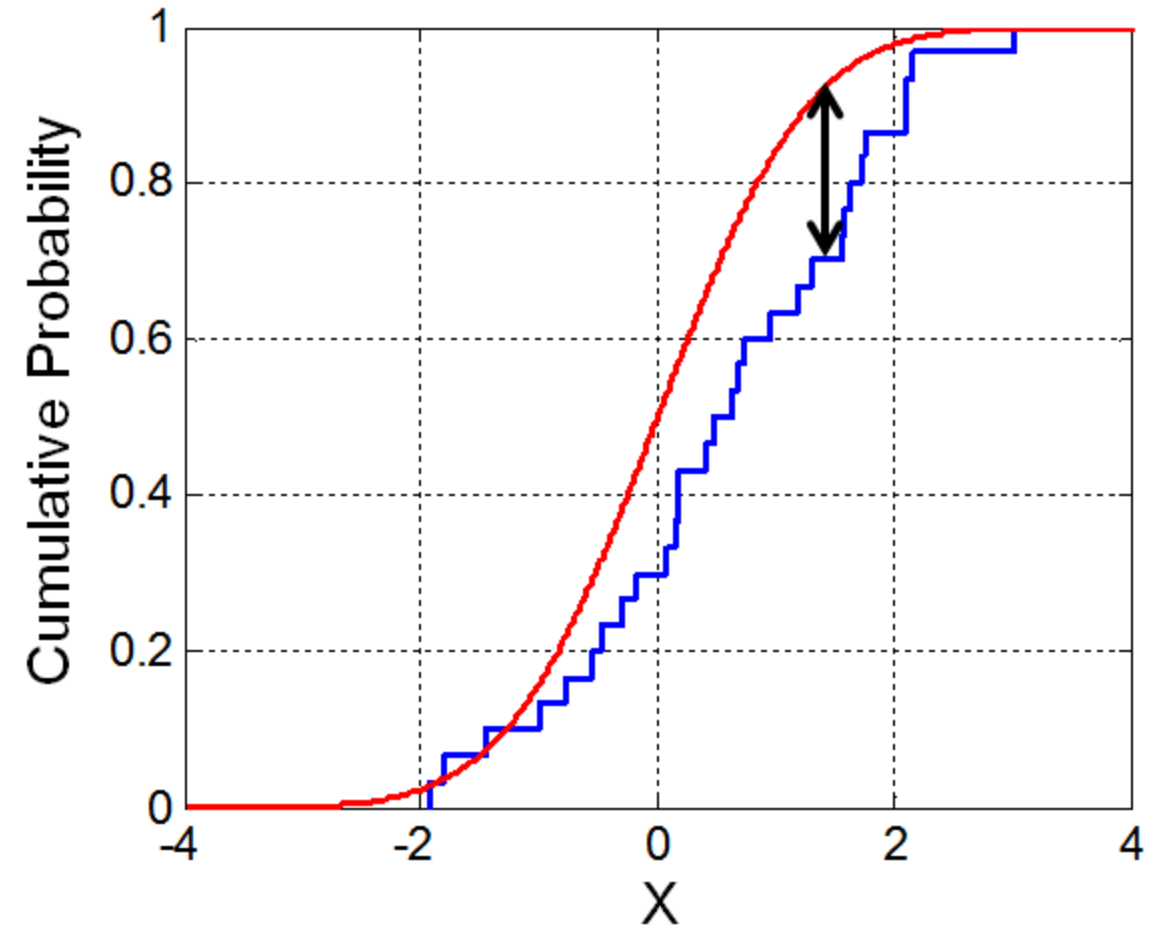
$$D_{n,m} = \sup |F_{1,n}(x) - F_{2,m}(x)|$$

Rate of  $D_n$  convergence to 0 as  $n \rightarrow \infty$ :

$$K = \sup |B(t)|, \text{ where } t \in [0,1]$$

&  $B(t)$  – Brownian Bridge from 0 to 0

**Proof:** Kolmogorov, 1933



# Maximum value of a Brownian Motion

Define:

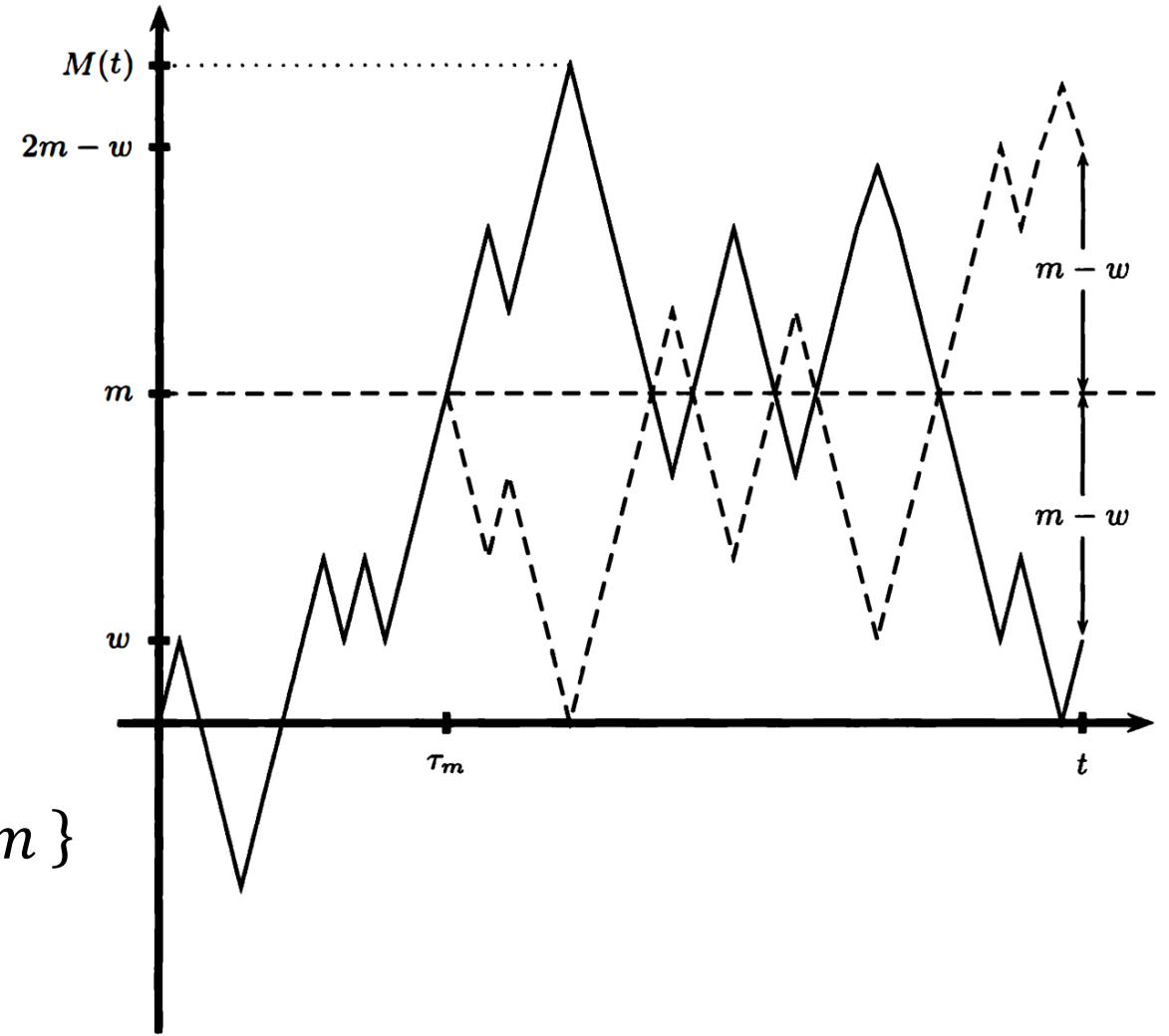
$$M(t) = \max[W(s)], \text{ where } 0 \leq s \leq t$$

Reflection path equality:

$$\mathbb{P}\{\tau_m \leq t, W(t) \leq w\} = \mathbb{P}\{W(t) \geq m\}$$

Consider:  $M(t) \geq m$  iff  $\tau_m \leq t$  >>

$$\mathbb{P}\{M(t) \geq m, W(t) \leq w\} = \mathbb{P}\{W(t) \geq m\}$$





# Maximum value of a Brownian Motion

Joint distribution:

$$f_{M(t), W(t)}(m, w) = \frac{2(2m - w)}{t\sqrt{2\pi t}} \exp\left\{-\frac{(2m - w)^2}{2t}\right\}$$

Proof: on the board

Conditional distribution:

$$f_{M(t)|W(t)}(m|w) = \frac{2(2m - w)}{t} \exp\left\{-\frac{2m(m - w)}{t}\right\}$$

Proof: on the board

# Maximum value of a Brownian Bridge

Define:

$$M^{a \rightarrow b}(T) = \max B^{a \rightarrow b}(t), 0 \leq t \leq T$$

pdf:

$$f_{M^{a \rightarrow b}(T)}(y) = \frac{2(2y - b - a)}{T} \exp \left\{ -\frac{2}{T} (y - a)(y - b) \right\}, \text{ where } y > \max(a, b)$$

**Proof:** on the board, start point and end point condition

# Other applications

## Information processes:

Brownian bridges disturbed by transformed and time scaled Levy processes

$$\eta_T^T = W(t) - \frac{t}{T}W(T) + \frac{t}{T}X(T-t) + \sigma tH(t),$$

Where  $X(t)$  is a Levy process and  $H(t)$  - the flow of information concerning a random cash flow at T.

## Optimal stopping issues:

$$V = \sup \left| \mathbb{E}[B^{a \rightarrow b}(t)] \right|, 0 \leq t \leq 1 - \text{starting point}$$

## To be continued:

Brody, Hughston, Macrina, 2007

Hoyle, Hughston, Macrina, 2010

Ekstrom, Vaicenavicius, 2009