

MATH180C: Introduction to Stochastic Processes II

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Today: Brownian Motion

> Q&A: December 2

Next: PK 8.2

This week:

- Homework 7 (due THURSDAY, December 3)
- HW6 regrades (until Wednesday, December 2, 11 PM)

BM as a Gaussian process

Def. Stochastic process $(X_t)_{t \geq 0}$ is called a Gaussian process if for any $0 \leq t_1 < t_2 < \dots < t_n$

$(X_{t_1}, \dots, X_{t_n})$ is a Gaussian vector, or equivalently for any $c_1, \dots, c_n \in \mathbb{R}$

$c_1 X_{t_1} + c_2 X_{t_2} + \dots + c_n X_{t_n}$ is a Gaussian r.v.

Recall that the distribution of a Gaussian vector is uniquely defined by its mean and covariance matrix.

Similarly, each Gaussian process is uniquely described by

$$\mu(t) = E(X_t) \quad \text{and} \quad \Gamma(s, t) = \text{Cov}(X_s, X_t) \geq 0$$

↑ covariance function

BM as a Gaussian process

Proposition BM $(B_t)_{t \geq 0}$ is a Gaussian process with
 $\mu(t) = 0$ and $\Gamma(s, t) = \min\{s, t\}$

Proof. For any $0 \leq t_1 < t_2 < \dots < t_n$, $B_{t_j} - B_{t_{j-1}}$ are indep.
Gaussian, thus $\sum_{i=1}^n c_i B_{t_i} = \sum_{i=1}^n c_i \sum_{j=1}^i (B_{t_j} - B_{t_{j-1}})$
is also Gaussian.

By definition $\mu(t) = E(B_t) = 0$ Let $s < t$.

$$\begin{aligned} \text{Then } \Gamma(s, t) &= \text{Cov}(B_s, B_t) \\ &= \text{Cov}(B_s, B_s + (B_t - B_s)) \\ &= \text{Cov}(B_s, B_s) + \text{Cov}(B_s, B_t - B_s) \\ &= s + 0 = s = \min\{s, t\} \end{aligned}$$



Some properties of BM

Proposition. Let $(B_t)_{t \geq 0}$ be a standard BM. Then

- (i) For any $s > 0$, the process $(B_{t+s} - B_s, t \geq 0)$ is a BM independent of $(B_u, 0 \leq u \leq s)$.
- (ii) The process $(-B_t, t \geq 0)$ is a BM
- (iii) For any $c > 0$, the process $(cB_{t/c^2}, t \geq 0)$ is a BM
- (iv) The process $(X_t)_{t \geq 0}$ defined by $X_0 = 0, X_t = tB_{1/t}$ for $t > 0$ is a BM.

Proof (i) Define $X_t = B_{t+s} - B_s$. Then $X_0 = 0$ and $X_{t_2} - X_{t_1} = B_{t_2+s} - B_{t_1+s}$
 \Rightarrow independent Gaussian increments, $E(X_{t_2} - X_{t_1}) = 0$, $\text{Var}(X_{t_2} - X_{t_1}) = t_2 - t_1$,

$(X_t)_{t \geq 0}$ has continuous paths $\Rightarrow (X_t)_{t \geq 0}$ is a BM

(iv) X_t is Gaussian, for $s < t$ $\text{Cov}(sB_{1/s}, tB_{1/t}) = st \min\{\frac{1}{s}, \frac{1}{t}\} = s$

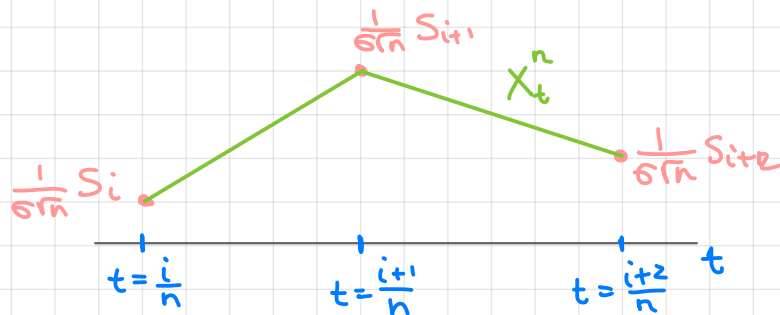
Proof of $\lim_{t \rightarrow 0} X_t = 0$ is more technical, thus omitted.

Construction of BM

BM can be constructed as a limit of properly rescaled random walks.

Let $\{\xi_k\}_{k=1}^{\infty}$ be a sequence of i.i.d. r.v.s, $E(\xi_i) = 0$, $\text{Var}(\xi_i) = \sigma^2 < \infty$. Denote $S_m = \sum_{k=1}^m \xi_k$ and define

$$X_t^n = \frac{1}{\sigma\sqrt{n}} (S_{[nt]} + (nt - [nt])\xi_{[nt]+1})$$



Theorem (Donsker) $(X_t^n)_{t \geq 0}$ converges in distribution to the standard BM.

Applying Donsker's theorem

Example Let $(\xi_i)_{i=1}^{\infty}$ be i.i.d. r.v. $P(\xi_i=1)=P(\xi_i=-1)=0.5$
 $E(\xi_i)=0$, $\text{Var}(\xi_i)=1$.

Denote $S_m := \sum_{i=1}^m \xi_i$, $S_0=0$. $(S_m)_{m \geq 0}$ is a Markov chain.

From the first step analysis of MC we know that for any $-a < 0 < b$ $P(S \text{ reaches } -a \text{ before } b) = \frac{b}{a+b}$

If X_t^n is the process interpolating S_m , then $\forall n$

$$\begin{aligned} P(X^n \text{ hits } -a \text{ before } b) &= P(S \text{ hits } -\sqrt{n}a \text{ before } \sqrt{n}b) \\ &= \frac{\sqrt{n}b}{\sqrt{n}a + \sqrt{n}b} = \frac{b}{a+b} \end{aligned}$$

$$\Rightarrow P(B \text{ hits } -a \text{ before } b) = \frac{b}{a+b}$$

$$\Rightarrow (\tilde{\xi}_i)_{i=1}^{\infty}, E(\tilde{\xi}_i)=0, \text{Var}(\tilde{\xi}_i)=1, P(\tilde{S} \text{ hits } -a \text{ before } b) \approx \frac{b}{a+b}$$

BM as a martingale

Let $(X_t)_{t \geq 0}$ be a continuous time stochastic process. We say that $(X_t)_{t \geq 0}$ is a martingale if $E(|X_t|) < \infty \quad \forall t \geq 0$ and $E(X_t | \{X_u, 0 \leq u \leq s\}) = X_s$ for all $s < t$.

Proposition Let $(B_t)_{t \geq 0}$ be a standard BM. Then

(i) $(B_t)_{t \geq 0}$ is a martingale

(ii) $(B_t^2 - t)_{t \geq 0}$ is a martingale (w.r.t. $(B_t)_{t \geq 0}$)

"Proof": $E(B_t | \{B_u, 0 \leq u \leq s\}) = E(B_s + B_t - B_s | \{B_u, 0 \leq u \leq s\}) = B_s + 0 = B_s$

$$\begin{aligned} E(B_t^2 - t | \{B_u, 0 \leq u \leq s\}) &= E(B_s^2 + 2B_s(B_t - B_s) + (B_t - B_s)^2 | \{B_u, 0 \leq u \leq s\}) \\ &= B_s^2 + 0 + t - s - t = B_s^2 - s \end{aligned}$$

Thm (Lévy) Let $(X_t)_{t \geq 0}$ be a continuous martingale such that $(X_t^2 - t)_{t \geq 0}$ is a martingale. Then $(X_t)_{t \geq 0}$ is BM.