

# MATH180C: Introduction to Stochastic Processes II

<https://mathweb.ucsd.edu/~ynemish/teaching/180c>

Today: Brownian motion

Next: PK 8.1- 8.2

Week 9:

- homework 7 (due Friday, June 2)

## Brownian motion. Definition

Def. **Brownian motion** with diffusion coefficient  $\sigma^2$  is a continuous time stochastic process  $(B_t)_{t \geq 0}$  satisfying

- (i)  $B(0) = 0$ ,  $B(t)$  is continuous as a function of  $t$
- (ii) For all  $0 \leq s < t < \infty$   $B(t) - B(s)$  is a Gaussian random variable with mean 0 and variance  $\sigma^2(t-s)$
- (iii) The increments of  $B$  are independent: if  $0 = t_0 < t_1 < \dots < t_n$  then  $\{B(t_i) - B(t_{i-1})\}_{i=1}^n$  are independent (Gaussian) r.v.s.

$\sigma^2 = 1 \leftarrow$  standard BM

## 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  $c_1, \dots, c_n \in \mathbb{R}$

$c_1 X_{t_1} + \dots + c_n X_{t_n}$  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 is a Gaussian process with  
 $(B_t)_{t \geq 0}$   $\mu(t) = 0$  and  $\Gamma(s, t) = \min\{s, t\} = s \wedge t$

Proof. For any  $0 \leq t_1 < t_2 < \dots < t_n$ ,  $B_{t_j} - B_{t_{j-1}}$  are indep.

Gaussian, thus  $\sum_{i=1}^n c_i B_{t_i} = \sum_{i=1}^n c_i \sum_{j=1}^i (B_{t_j} - B_{t_{j-1}}) = \sum_{j=1}^n \sum_{i=j}^n c_i (B_{t_j} - B_{t_{j-1}})$   
is also Gaussian.

By definition  $\mu(t) = E(X_t) = 0$ . 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  $(\underbrace{B_{t+s} - B_s}_{X_t}, t \geq 0)$  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  $(c B_{\frac{t}{c^2}}, t \geq 0)$  is a BM

(iv) The process  $(X_t)_{t \geq 0}$  defined by  $X_0 = 0$ ,  $X_t = t B_{\frac{1}{t}}$  for  $t > 0$  is a BM.   
  $\downarrow$   
continuous on  $(0, +\infty)$

Proof (i) Define  $X_t = B_{t+s} - B_s$ . Then  $X_0 = 0$ ,  $X_{t_2} - X_{t_1} = B_{s+t_2} - B_{s+t_1}$

$\Rightarrow$  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  $s < t$   $\text{Cov}(s B_{\frac{1}{s}}, t B_{\frac{1}{t}}) = st \min\{\frac{1}{s}, \frac{1}{t}\} = s \wedge t$

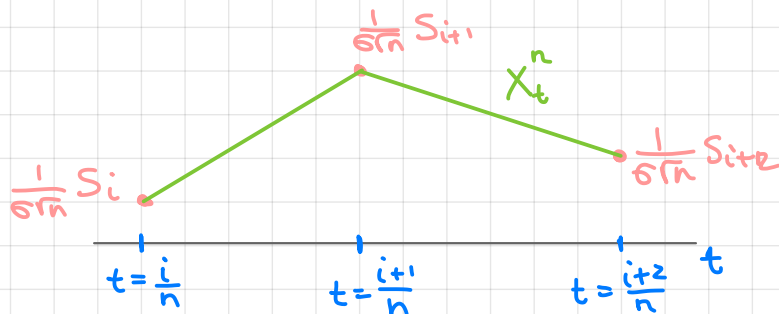
Proof of  $\lim_{t \rightarrow 0} X_t = 0$  is more technical, thus omitted.

## 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  $S_m = \sum_{i=1}^m \xi_i$  and define

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



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  $-a < 0 < b$

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

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

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

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

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