

# 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. History

- Critical observation: **Robert Brown (1827)**, botanist, movement of pollen grains in water
- First (?) mathematical analysis of Brownian motion: **Louis Bachelier (1900)**, modeling stock market fluctuations
- Brownian motion in physics: **Albert Einstein (1905)** and **Marian Smoluchowski (1906)**, explained the phenomenon observed by Brown
- First rigorous construction of mathematical Brownian motion: **Norbert Wiener (1923)**

Brownian motion  $\stackrel{\uparrow}{=}$  Wiener process  
in mathematics

## Brownian motion. Motivation

- almost all interesting classes of stochastic processes contain Brownian motion: BM is a
  - martingale
  - Markov process
  - Gaussian process
  - Lévy process (independent stationary increments)
- BM allows explicit calculations, which are impossible for more general objects
- BM can be used as a building block for other processes
- BM has many beautiful mathematical properties

## 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  $s < t$   $B(t) - B(s)$  is a Gaussian random variable with mean 0 and variance  $\sigma^2(t-s)$

(iii) The increments of  $B$  are independent: if  $0 \leq t_0 < t_1 < \dots < t_n$  then  $\{B(t_i) - B(t_{i-1})\}_{i=1}^n$  are independent (Gaussian) rvs

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



## BM as a continuous time continuous space Markov process

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

Then  $(B_t)_{t \geq 0}$  is a Markov process with transition density

$$p_t(x, y) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2t}(x-y)^2}$$

Informal explanation: Independent stationary increments imply that  $(B_t)_{t \geq 0}$  is Markov with stationary transition density. Given  $B_s = x$ ,  $B_{t+s} = B_s + (B_{t+s} - B_s)$  information before time  $s$  is irrelevant.

$$\begin{aligned} P(B_{s+t} \leq u \mid B_s = x) &= P(B_s + (B_{t+s} - B_s) \leq u \mid B_s = x) \\ &= P(x + B_{t+s} - B_s \leq u) = \int_{-\infty}^u \frac{1}{\sqrt{2\pi t}} e^{-\frac{(y-x)^2}{2t}} dy \end{aligned}$$

## BM as a continuous time continuous space Markov process

Let  $t_1 < t_2 < \dots < t_n < \infty$ ,  $(a_i, b_i) \subset \mathbb{R}$ . Then

$$P(B_{t_1} \in (a_1, b_1), B_{t_2} \in (a_2, b_2)) =$$

$$= \int_{-\infty}^{+\infty} P(B_{t_1} \in (a_1, b_1), B_{t_2} \in (a_2, b_2) \mid B_{t_1} = x_1) p_{t_1}(0, x_1) dx_1,$$

$$= \int_{a_1}^{b_1} P(B_{t_2} \in (a_2, b_2) \mid B_{t_1} = x_1) p_{t_1}(0, x_1) dx_1,$$

$$= \int_{a_1}^{b_1} \int_{a_2}^{b_2} p_{t_2-t_1}(x_1, x_2) dx_2 p_{t_1}(0, x_1) dx_1,$$

More generally,

$$P(B_{t_1} \in (a_1, b_1), B_{t_2} \in (a_2, b_2), \dots, B_{t_n} \in (a_n, b_n))$$

$$= \int_{(a_1, b_1)} \dots \int_{(a_n, b_n)} p_{t_1}(0, x_1) p_{t_2-t_1}(x_1, x_2) \dots p_{t_n-t_{n-1}}(x_{n-1}, x_n) dx_1 \dots dx_n$$

# Diffusion equation. Transition semigroup. Generator

Let  $(X_t)_{t \geq 0}$  be a Markov process.

Suppose we want to know how the distribution of  $X_t$  evolves in time:

$$E(f(X_{t+s}) | X_s = x) = \int_{-\infty}^{+\infty} f(y) P_t^x(x, y) dy =: P_t f(x)$$

$\lim_{t \downarrow 0} \frac{P_t - Id}{t}$

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We call  $(P_t)_{t \geq 0}$  the transition semigroup  $[P_{s+t} f(x) = P_s(P_t f(x))]$

Proposition Let  $(P_t)_{t \geq 0}$  be the transition semigroup of BM.

Then (i) the "infinitesimal generator" of  $P(t)$  is given by

$$Q f(x) = \frac{1}{2} \frac{d^2}{dx^2} f(x)$$

(ii) density  $p_t$  satisfies  $\frac{\partial}{\partial t} p_t(x, y) = \frac{1}{2} \frac{\partial^2}{\partial x^2} p_t(x, y)$  [K backward]

(iii) density  $p_t$  satisfies  $\frac{\partial}{\partial t} p_t(x, y) = \frac{1}{2} \frac{\partial^2}{\partial y^2} p_t(x, y)$  [K forward]  
↑ diffusion equation, heat equation