MATH180C: Introduction to Stochastic Processes II

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Today: Brownian Motion

> Q&A: November 30

Next: PK 8.1-8.2

This week:

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

Brownian motion. History

fluctuations

- Critical observation: Robert Brown (1827), botanist,
 movement of pollen grains in water
- First (?) mathematical analysis of Brownian motion: Louis Bachelier (1900), modeling stock market
- · 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 = Wiener process
 in mathematics

Brownian motion. Motivation

- almost all interesting classes of stochastic processes
 contain Brownian motion: BM is a
 - martingale
 - Markou 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 62 is a continuous time stochastic process (Bt)t20 satisfying

(i) B(o)=0, B(t) is continuous as a function of t

(ii) For all 0≤ sct c∞ B(t)-B(s) is a Gaussian random variable with mean 0 and variance 6°(t-s)

variable with mean 0 and variance o (t-s)

(iii) The increments of B are independent: if 0 = to < ti < -- < tn

then {B(ti)-B(ti-1)};_, are independent (Gaussian) r.v.s.

62=1 < standard BM

BM as a continuous time continuous space Markov process Recall: continuous time discrete space MC (Xt)tzo is characterized by the transition probability tunction $P(j(t) = P(X_{sit} = j | X_{s} = i)$ ((X+)+20 has stationary transition probability functions) In particular, P(Xs+eA | Xs = i) = Z Pij (+) In the continuous state space case the transition probabilities are described by the transition density (ii) $P(X_{s+t} \in A \mid X_s = x) = \int_A P_t(x,y) dy$ for any $x \in \mathbb{R}$, $A \subset \mathbb{R}$ 2 density of X_{s+t} given $X_s = x$

BM as a continuous time continuous space Markov process

Propotition. Let (Bt) +20 be a standard BM.

Then (Bt)t20 is a Markov process with transition density $P_{t}(x,y) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2t}} (y-x)^{2}$

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

P(Bs+6 = u | Bs = x) = P(Bs + (Bt+s - Bs) = u | Bs = 2)

= P (x + Bt+s-Bs &u) = 5 1 1 2 1 dy

BM as a continuous time continuous space Markov process Let tictz c... etn co (ai, bi) c IR. Then P(Bt, E(a, b)), Bt2 E(az, b2)) = = \ P(Bt, \(\epsilon(a, b_1), Bt; \(\epsilon(a, b_2) \) \ Bt, \(=\pi_1\) \ Pt, \(o_1\pi_1\) \ dx, = $\int P(B_{t_2} + (a_2, b_2) | B_{t_1} = x_1) P_{t_1}(o_1x_1) dx_1$ $= \int_{\alpha_1} Pt_1(o_1x_1) \left(\int_{\alpha_2} Pt_2-t_1(x_1,x_2) dx_2 \right) dx_1$ More generally, P(Bt, e(a, b1), Bt e (a2, b2), ..., Btn e (an, bn)) = $\int -\int P_{t_1}(o_1x_1) P_{t_2-t_1}(x_1,x_2) - P_{t_1-t_1}(x_{n-1},x_n) dx_1 - dx_n$ (a, b,) x ... x (an, bn)

Diffusion equation. Transition semigroup. Generator

Let (Xt) teo be a Markon process. Suppose we want to know how the distribution of Xt evolves in time:

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

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

Proposition Let (Pt)t20 be the transition semigroup of BM.

Then (i) the infinitesimal generator of P(t) is given by $Qf(z) = \frac{1}{2} \frac{dz}{dx} f(z)$

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