
BROWNIAN MOTION AND STOCHASTIC CALCULUS

Master class 2015-2016

1. GAUSSIAN VECTORS

- (a) Let ξ be a (real-valued) Gaussian variable with mean μ and variance σ^2 . Compute the characteristic function $\varphi(z) = \mathbb{E}[\exp(iz\xi)]$, $z \in \mathbb{R}$.
- (b) Let $\xi = (\xi_1, \dots, \xi_d)$ be a Gaussian vector with mean $\mu = (\mu_1, \dots, \mu_d) \in \mathbb{R}^d$ and covariance matrix $G = (G_{jk})_{j,k=1}^d \in \mathbb{R}^{d \times d}$. Prove that the matrix G is positive definite, i.e. $\lambda^\top G \lambda = \sum_{j,k=1}^d \lambda_j G_{jk} \lambda_k > 0$ for all $\lambda = (\lambda_1, \dots, \lambda_d) \in \mathbb{R}^d$ except $\lambda = 0$.
- (c) Let $\xi = (\xi_1, \dots, \xi_d)$ be a Gaussian vector with mean $\mu \in \mathbb{R}^d$ and covariance matrix $G \in \mathbb{R}^{d \times d}$. Compute the characteristic function $\varphi(z) = \mathbb{E}[\exp(iz^\top \xi)]$, $z \in \mathbb{R}^d$. *Hint:* write $G = U^\top \Lambda U$, where U is an orthogonal matrix and Λ is diagonal.
- (d) Check that if ξ_1, \dots, ξ_d are independent Gaussian variables, then $\xi = (\xi_1, \dots, \xi_d)$ is a Gaussian vector. For any matrix $U \in \mathbb{R}^{d \times d}$ check that $U\xi$ is also a Gaussian vector. What can be said about their covariance matrices?
- (e) Let $\xi = (\xi_1, \dots, \xi_d)$ be a Gaussian vector. Prove that its components ξ_1, \dots, ξ_d are independent if and only if the covariance matrix G is diagonal. Is it true that two Gaussian variables are independent if and only if their covariance is zero?

2. FOURIER SERIES

- (a) Prove that both families $(\sqrt{2} \cos(\pi n t))_{n \geq 0}$ and $(\sqrt{2} \sin(\pi n t))_{n \geq 1}$ are orthonormal bases in $L^2[0, 1]$. *Hint:* Use the fact that $(e^{i\pi n t})_{n \in \mathbb{Z}}$ is an orthogonal basis in $L^2[-1, 1]$.
- (b) For all $s, t \in [0, 1]$, prove the following identity:

$$\sum_{n=1}^{+\infty} \frac{2 \sin(\pi n s) \sin(\pi n t)}{\pi^2 n^2} = \min\{s, t\} - st.$$

- (c) (*) Note that the identity given above can be also derived from the identity

$$\sum_{n=1}^{+\infty} \frac{\cos(\pi n t)}{\pi^2 n^2} = \frac{t^2}{4} - \frac{|t|}{2} + \frac{1}{6}, \quad |t| \leq 1,$$

which follows (by integration) from the *Poisson summation formula*

$$\sum_{n=-\infty}^{+\infty} e^{i\pi n t} = 2 \sum_{m=-\infty}^{+\infty} \delta_{2m}(t)$$

(this should be understood in terms of Schwartz distributions).

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3. GREEN'S FUNCTION OF THE LAPLACIAN ON $[0, 1]$.

- (a) Prove that the eigenfunctions and eigenvalues of the Dirichlet boundary value problem

$$-f'' = \lambda f, \quad f(0) = f(1) = 0.$$

are given by $f_n(t) = \sqrt{2} \sin(\pi n t)$ and $\lambda_n = \pi^2 n^2$ with $n \geq 1$. Find eigenfunctions of the similar problem with Neumann boundary conditions $f'(0) = f'(1) = 0$.

- (b) *Green's function* $G(s, t)$ of the Laplacian $f \mapsto -f''$ with Dirichlet boundary conditions is defined to be the kernel of the inverse operator, i.e. the unique function G such that $-f'' = g$ and $f(0) = f(1) = 0$ imply $f(t) = \int_0^1 G(s, t)g(s)ds$. Prove that

$$G(s, t) = \min\{s, t\} - st, \quad s, t \in [0, 1].$$

- (c) Prove that

$$G(s, t) = \sum_{n=1}^{+\infty} \frac{2 \sin(\pi n s) \sin(\pi n t)}{\pi^2 n^2}, \quad s, t \in [0, 1].$$

4. POISSON PROCESS

Recall that we defined the *Poisson process* $(N_t)_{t \in [0, +\infty)}$ of intensity $\lambda > 0$ by

$$N_t := \min\{n : \xi_0 + \dots + \xi_n \geq t\},$$

where ξ_0, ξ_1, \dots is a sequence of i.i.d. *exponential variables* with density $\lambda e^{-\lambda x}$, $x \in [0, +\infty)$. (Also recall that $(N_t)_{t \in [0, +\infty)}$ is a process with independent increments due to the memoryless property of the exponential variable.)

- (a) Prove that the increments $N_{t+s} - N_t$ are stationary and have Poisson distribution with parameter λs , i.e. $\mathbb{P}[N_{t+s} - N_t = n] = e^{-\lambda s} \cdot (\lambda s)^n / n!$, $n \geq 0$.
- (b) Assume that $\lambda^{(1)}, \lambda^{(2)} > 0$ and $\lambda = \lambda^{(1)} + \lambda^{(2)}$. Let $N_t^{(1)}$ and $N_t^{(2)}$ be two independent Poisson processes of intensities $\lambda^{(1)}$ and $\lambda^{(2)}$. Prove that the process $N_t := N_t^{(1)} + N_t^{(2)}$ is a Poisson processes of intensity λ .
- (c) Let $(N_t)_{t \geq 0}$ be a Poisson processes of intensity $\lambda > 0$, and let $p \in (0, 1)$. Let us color every jump point of N_t white or blue independently with probabilities p and $1 - p$, respectively. Prove that the collections of white and blue points define jumps of two Poisson processes of intensities λp and $\lambda(1 - p)$, respectively.
- (d) (*) Prove that a counting process $(N_t)_{t \in [0, +\infty)}$ (i.e., a non-decreasing integer-valued right-continuous process with $N_0 = 0$) is a Poisson process of intensity $\lambda > 0$ if and only if for all $0 < t_1 < \dots < t_k$ and $0 \leq n_1 \leq \dots \leq n_k$, one has

$$\mathbb{P}_\lambda(N_{t_k+\delta} - N_{t_k} = 0 \mid N_{t_j} = n_j, 1 \leq j \leq k) = 1 - \lambda\delta + o(\delta),$$

$$\mathbb{P}_\lambda(N_{t_k+\delta} - N_{t_k} = 1 \mid N_{t_j} = n_j, 1 \leq j \leq k) = \lambda\delta + o(\delta),$$

$$\mathbb{P}_\lambda(N_{t_k+\delta} - N_{t_k} \geq 2 \mid N_{t_j} = n_j, 1 \leq j \leq k) = o(\delta), \quad \text{as } \delta \rightarrow 0.$$

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5. MISCELLANEOUS

- (a) Let B_t be the standard Brownian motion on $[0, \infty)$. Check that the process $(1-t)B_{\frac{t}{1-t}}$ is a Brownian bridge on $[0, 1]$.
- (b) Let \tilde{B}_t be the standard Brownian bridge on $[0, 1]$. Check that the process $(1+t)\tilde{B}_{\frac{t}{1+t}}$ is a standard Brownian motion on $[0, \infty)$.
- (c) Prove that

$$(a + \frac{1}{a})^{-1} \cdot e^{-\frac{a^2}{2}} < \int_a^{+\infty} e^{-\frac{x^2}{2}} dx < a^{-1} \cdot e^{-\frac{a^2}{2}}.$$

- (d) Prove that for any dyadic rationals $s = p2^{-m}$ and $t = q2^{-s}$ one has

$$\sum_{n=1}^{+\infty} \sum_{k=1, k \text{ odd}}^{2^n-1} g_{k,n}(s)g_{k,n}(t) = \min\{s, t\} - st,$$

where the functions $g_{k,n}(x) = \int_0^x f_{k,n}(y)dy$ are the primitives of the *Haar functions*

$$f_{k,n} = 2^{-\frac{n+1}{2}} \cdot (\chi_{[(k-1)2^{-n}, k2^{-n})} - \chi_{[k2^{-n}, (k+1)2^{-n})}).$$

- (e) (*) Prove that the Haar functions $f_{k,n}(t)$ form a *complete family* in $L^2([0, 1])$.

6. MEASURABILITY

Let \mathcal{A} be a σ -algebra on a space Ω and $(X_t)_{t \in [0, T]}$ be a family of mappings $X_t : \Omega \rightarrow \mathbb{R}$. Let $X : \Omega \times [0, T] \rightarrow \mathbb{R}$ denote the mapping $(\omega, t) \mapsto X_t$. By $\mathcal{B}(M)$ we will denote the Borel σ -algebra on a metric space M . Prove that the following statements are equivalent:

- (a1) for each $t \in [0, T]$ the mapping $X_t : (\Omega, \mathcal{A}) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$ is measurable;
- (a2) the mapping $X : (\Omega, \mathcal{A}) \rightarrow (\mathbb{R}^{[0, T]}, \mathcal{B}(\mathbb{R})^{\otimes [0, T]})$ is measurable.

Assume that for all $\omega \in \Omega$ the function $\mathcal{X} = \mathcal{X}(\omega) : t \mapsto X_t$ is continuous on $[0, T]$ and let $\mathcal{C}([0, T])$ denote the (Banach) space of real-valued continuous functions on $[0, T]$. Prove that the following statements are equivalent:

- (a) the mapping $X : (\Omega, \mathcal{A}) \rightarrow (\mathbb{R}^{[0, T]}, \mathcal{B}(\mathbb{R})^{\otimes [0, T]})$ is measurable;
- (b) the mapping $X : (\Omega \times [0, T], \mathcal{A} \otimes \mathcal{B}([0, T])) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$ is measurable;
- (c) the mapping $\mathcal{X} : (\Omega, \mathcal{A}) \rightarrow (\mathcal{C}([0, T]), \mathcal{B}(\mathcal{C}([0, T])))$ is measurable.

Without the continuity assumption, check that (b) implies (a) but not vice versa.

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7. MAXIMUM PROCESS

Theorem (Bachelier). *Let $(B_t)_{t \geq 0}$ be a standard Brownian motion and $M_t := \max_{s \in [0, t]} B_s$. Then for each (fixed) $t \geq 0$ one has*

$$M_t \stackrel{(d)}{=} M_t - B_t \stackrel{(d)}{=} |B_t|.$$

(a) Could it be true that, say, $(M_t)_{t \in [0, 1]} \stackrel{(d)}{=} (|B_t|)_{t \in [0, 1]}$?

(b) Let $x \geq 0$, $y \leq x$ and $\tau = \inf\{t : B_t = x\}$. Using reflection principle show that

$$\mathbb{P}[M_1 \geq x, B_1 \leq y] = \mathbb{P}[B_{\min\{1, \tau\}} - (B_1 - B_{\min\{1, \tau\}}) \geq 2x - y].$$

(c) Show that the joint distribution of the pair (M_1, B_1) is given by the measure

$$-2p'(2x - y) dx dy, \quad x \geq 0, y \leq x,$$

where $p(t) = (2\pi)^{-1/2} \exp(-t^2/2)$ is the standard normal density.

(d) Deduce the theorem for $t = 1$ and use the scaling invariance to treat the general case.

8. LAW OF THE ITERATED LOGARITHM

Theorem (Khinchin). *Let $(B_t)_{t \geq 0}$ be a Brownian motion. Then we have a.s.*

$$\limsup_{t \rightarrow 0} \frac{B_t}{\sqrt{2t \log \log(1/t)}} = 1.$$

(a) Show that $\limsup_{t \rightarrow 0} (2t \log \log(1/t))^{-1/2} B_t \stackrel{(d)}{=} \limsup_{t \rightarrow \infty} (2t \log \log t)^{-1/2} B_t$.

(b) Let $M_t = \sup_{s \in [0, t]} B_s$. Use Bachelier's theorem to show that

$$\mathbb{P}[M_t > ut^{1/2}] \sim (2/\pi)^{1/2} u^{-1} e^{-u^2/2} \quad \text{as } u \rightarrow \infty.$$

(c) Show that $\limsup_{t \rightarrow \infty} (2t \log \log t)^{-1/2} B_t \leq 1$ almost surely.

(d) Show that for $r > 1$

$$\limsup_{t \rightarrow \infty} \frac{B_t - B_{t/r}}{\sqrt{2t \log \log t}} \geq (1 - r^{-1})^{1/2} \quad \text{a.s.}$$

(e) Show that for $r > 1$

$$\limsup_{t \rightarrow \infty} \frac{B_t}{\sqrt{2t \log \log t}} \geq (1 - r^{-1})^{1/2} - r^{-1/2} \quad \text{a.s.}$$

(f) Show that $\limsup_{t \rightarrow \infty} (2t \log \log t)^{-1/2} B_t \geq 1$ almost surely.

9. IDENTITIES FOR RANDOM WALKS. LAST RETURN TO 0 AND RUNNING MAXIMUM.

Theorem (last return to 0, Feller). Let $(S_m)_{m \geq 0}$ be a simple symmetric random walk in \mathbb{Z} and $\sigma_n := \max\{k \leq n : S_{2k} = 0\}$. Then

$$\mathbb{P}(\sigma_n = k) = u_k u_{n-k}, \quad k = 0, \dots, n,$$

where $u_k = \mathbb{P}(S_{2k} = 0) = 2^{-2k} \binom{2k}{k}$.

- (a) Formulate a discrete version of the reflection principle for the Brownian motion.
- (b) Show that $\mathbb{P}(\sigma_n = k) = u_k \cdot \mathbb{P}(M_{2(n-k)-1} = 0)$, where $M_m = \max_{k \leq m} S_k$.
- (c) Note that $1 - \mathbb{P}(M_{2m-1} = 0) = \mathbb{P}(M_{2m-1} \geq 1, S_{2m-1} \geq 1) + \mathbb{P}(M_{2m-1} \geq 1, S_{2m-1} \leq -1)$.
- (d) Using the reflection principle for the simple random walk, prove that

$$\mathbb{P}(M_{2m-1} = 0) = \mathbb{P}(S_{2m-1} = 1) = u_m.$$

Theorem (running maximum times, Sparre-Andersen). Let $(S_m)_{m \geq 0}$ be a random walk in \mathbb{R} based on a symmetric diffuse (i.e. absolutely continuous w.r.t. Lebesgue measure) distribution, put $M_n := \max_{k \leq n} S_k$, and write $\tau_n := \min\{k \geq 0 : S_k = M_n\}$. Then

$$\mathbb{P}(\tau_n = k) = u_k u_{n-k}, \quad k = 0, \dots, n,$$

where u_k are the same as in the previous theorem.

- (a) Show that $\mathbb{P}[\tau_k = 0] = \mathbb{P}[\tau_k = k]$ for all $k \geq 0$.
- (b) Show that $\mathbb{P}[\tau_n = k] = v_k v_{n-k}$ for all $0 \leq k \leq n$, where $v_k = \mathbb{P}[\tau_k = k]$.
- (c) By induction show that $v_k = u_k$.
- (d) (*) Is it true that the processes $(\tau_n)_{n \geq 0}$ and $(\sigma_n)_{n \geq 0}$ are identically distributed?

10. IDENTITIES FOR RANDOM WALKS. SOJOURNS AND MAXIMA.

Theorem (sojourns and maxima, Sparre-Andersen). Let ξ_1, \dots, ξ_n be i.i.d. (more generally, exchangeable) random variables and $S_m = \sum_{k=1}^m \xi_k$ for $0 \leq m \leq n$. Then,

$$\#\{1 \leq m \leq n : S_m > 0\} \stackrel{(d)}{=} \min\{k : S_k = \max_{0 \leq m \leq n} S_m\}.$$

- (a) This is a deterministic statement. Given some values $\xi_1, \dots, \xi_n \in \mathbb{R}$, let us construct a permutation $\beta : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$ by the following algorithm applied consecutively to $k = n, k = n - 1, \dots, k = 1$:

- if $S_k \leq 0$, denote by $\beta(k)$ the maximal available index in $\{1, \dots, n\}$;
- if $S_k > 0$, denote by $\beta(k)$ the minimal available index in $\{1, \dots, n\}$.

Let $S_m^{(\beta)} = \sum_{k=1}^m \xi_{\beta(k)}$ for $0 \leq m \leq n$. Prove that

$$\#\{1 \leq m \leq n : S_m > 0\} = \min\{k : S_k^{(\beta)} = \max_{0 \leq m \leq n} S_m^{(\beta)}\}.$$

- (b) Prove that $(\xi_{\beta(1)}, \dots, \xi_{\beta(n)}) \stackrel{(d)}{=} (\xi_1, \dots, \xi_n)$ and deduce the theorem.

Hint: re-write the condition $S_k \leq 0$ as $\xi_{k+1} + \dots + \xi_n \geq S_n = S_n^{(\beta)}$.

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11. REFLECTED BROWNIAN MOTION. THE PROCESSES $(M_t - B_t)_{t \geq 0}$ AND $(|B_t|)_{t \geq 0}$.

- (a) Let $(S_m)_{m \geq 0}$, $(S'_m)_{m \geq 0}$ be two independent simple symmetric random walks in \mathbb{Z} started at the origin and $\tilde{S}'_m := S'_m + \frac{1}{2}$ for $m \geq 0$. Let $M_m := \max_{k \leq m} S_k$ and

$$L'_m := \# \left\{ 1 \leq k \leq m : \tilde{S}'_k = -\tilde{S}'_{k-1} \in \left\{ \pm \frac{1}{2} \right\} \right\}$$

denote the number of steps before time m when the trajectory $(\tilde{S}'_m)_{m \geq 0}$ crosses the horizontal line. Prove that the processes

$$(M_m - S_m, M_m)_{m \geq 0} \quad \text{and} \quad (|S'_m| - \frac{1}{2}, L'_m)_{m \geq 0}$$

are identically distributed. *Hint:* note that both processes can be described as the (identically distributed) random walks in $\mathbb{Z}_+ \times \mathbb{Z}_+$.

- (b) Let $(B_t)_{t \geq 0}$ be a standard Brownian motion and $M_t := \max_{s \in [0, t]} B_s$. Using Donsker's theorem, prove that

$$(M_{t_1} - B_{t_1}, \dots, M_{t_p} - B_{t_p}) \stackrel{(d)}{=} (|B_{t_1}|, \dots, |B_{t_p}|)$$

for all $0 \leq t_1 \leq \dots \leq t_p$.

- (c) Conclude that the processes $(M_t - B_t)_{t \geq 0}$ and $(|B_t|)_{t \geq 0}$ are identically distributed. *Hint:* The mapping $f(t) \mapsto \max_{s \in [0, t]} f(s)$ is continuous in $C([0, T])$ for each $T > 0$.
- (d) (*) Denote by $(\mathcal{F}_t^{(1)})_{t \geq 0}$ and $(\mathcal{F}_t^{(2)})_{t \geq 0}$ the filtrations generated by these two processes. Is it true that $\mathcal{F}_t^{(1)} = \mathcal{F}_t$, where the filtration $(\mathcal{F}_t)_{t \geq 0}$ is generated by $(B_t)_{t \geq 0}$ itself? (In other words, can one reconstruct B_t from $M_t - B_t$?) Is it true that $\mathcal{F}_t^{(2)} = \mathcal{F}_t$?

12. UNIFORM LAWS FOR THE BROWNIAN BRIDGE

Theorem. Let $(B_t)_{t \in [0, 1]}$ be a Brownian bridge on $[0, 1]$ and $M := \max_{s \in [0, 1]} B_s$. Then the following random variables are both $U(0, 1)$, i.e. uniformly distributed on $[0, 1]$:

$$\tau_1 = \lambda \{t \in [0, 1] : B_t > 0\}, \quad \tau_2 = \inf\{t : B_t = M\}.$$

- (a) Given $u \in [0, 1]$, define $B_t^u := B_{(u+t)} - B_u$, where $(x) := x - \lfloor x \rfloor$. Show that the process $(B_t^u)_{t \in [0, 1]}$ is distributed as a Brownian bridge.
- (b) Let $\tau_2^u := \inf\{t : B_t^u = M - B_u\}$ and let u be uniformly distributed over $[0, 1]$. Apply Fubini's theorem to show that $\mathbb{P}(\tau_2 \leq t) = \int_0^1 \mathbb{P}(\tau_2^u \leq t) du$ is also $U(0, 1)$.
- (c) Using Donsker's theorem and Exercise 10, show that τ_1 and τ_2 have the same law.
- (d) (*) Let $(B_t)_{t \geq 0}$ be a standard Brownian motion, $(\tilde{B}_t)_{t \in [0, 1]}$ be a Brownian bridge. For each $T < 1$ prove that the distributions of the processes $(B_t)_{t \in [0, T]}$ and $(\tilde{B}_t)_{t \in [0, T]}$ are mutually absolutely continuous. *Hint:* prove that the processes $(B_t - tT^{-1}B_T)_{t \in [0, T]}$ and $(\tilde{B}_t - tT^{-1}\tilde{B}_T)_{t \in [0, T]}$ are identically distributed (as a re-scaled Brownian bridge).

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13. EXIT TIME FROM $[-a, a]$ AND TIME SPENT IN $[0, -a]$ DURING A DOWNCROSSING

- (a) Let $a > 0$ and $\tau_{\pm a} := \inf\{t \geq 0 : |B_t| = a\}$, where $(B_t)_{t \geq 0}$ is a standard Brownian motion (started from 0). Show that

$$\mathbb{E}[\exp(-\mu\tau_{\pm a})] = 1/\cosh(a\sqrt{2\mu}), \quad \mu \geq 0.$$

Compute the expectations $\mathbb{E}[\tau_a]$ and $\mathbb{E}[\tau_a^2]$. Is it true that $\mathbb{E}[\exp(\theta\tau_{\pm a})] < +\infty$ for some $\theta > 0$? What is the optimal upper bound for such θ 's?

Proposition (removing of negative excursions). *Let $(B_t)_{t \geq 0}$ be a standard Brownian motion, $s(t) := \lambda(\{s' \in [0, t] : B_{s'} \geq 0\})$ and $t(s) := \inf\{t \geq 0 : s(t) \geq s\}$. Then*

$$(B_{t(s)})_{s \geq 0} \stackrel{(d)}{=} (|B_s|)_{s \geq 0}.$$

- (b) Check a similar statement for simple random walks (this is trivial).
(c) Prove that for each $T > 0$ one has $\lim_{\varepsilon \downarrow 0} \lambda(t \in [0, T] : |B_t| \leq \varepsilon) = 0$ almost surely.
(d) Let $s_f(t)$ and $t_f(s)$ be defined as in the proposition via a continuous function $f = f(t)$ instead of B_t . Check that the mapping $(f(t))_{t \in [0, T]} \mapsto (f(t(s \wedge s(T))))_{s \in [0, T]}$ is continuous (in the $C([0, T])$ metric) at almost every Brownian motion trajectory $(B_t)_{t \in [0, T]}$.
(e) Prove the proposition using (b) and Donsker's invariance principle.
(f) Let $\tau_{-a} := \inf\{t \geq 0 : B_t = -a\}$. Prove that $\lambda(s \in [0, \tau_{-a}] : 0 \geq B_s \geq -a) \stackrel{(d)}{=} \tau_{\pm a}$.

14. RECURRENCE/TRANSIENCY OF THE D-DIMENSIONAL BROWNIAN MOTION

Let $d \geq 2$ and $A(r, R) := \{x \in \mathbb{R}^d : r < \|x\| < R\}$. Let $(B_t^x)_{t \geq 0}$ denote a standard d -dimensional Brownian motion started from x and $\tau_{r, R}^x = \tau_{r, R}^x := \inf\{t > 0 : B_t^x \notin A(r, R)\}$.

- (a) For $x \in A(r, R)$, prove that

$$\mathbb{P}^x[|B_{\tau_{r, R}^x}^x| = r] = \begin{cases} (\log R/\|x\|) \cdot (\log R/r)^{-1} & \text{if } d = 2, \\ (\|x\|^{2-d} - R^{2-d}) \cdot (\|r\|^{2-d} - R^{2-d})^{-1} & \text{if } d \geq 3. \end{cases}$$

- (b) Prove that $\mathbb{P}[\exists t > 0 : B_t^x = 0] = 0$ for all $x \in \mathbb{R}^d$.
(c) Let $d = 2$ and $\|x\| > r > 0$. Prove that $\mathbb{P}[\exists t \geq 0 : |B_t^x| = r] = 1$ and deduce that

$$\mathbb{P}[\exists 0 < t_1 < t_2 < \dots : t_k \rightarrow \infty \text{ and } |B_{t_k}^x| = r] = 1$$

for any $x \in \mathbb{R}^d$ and $r > 0$ (in other words, the Brownian motion in 2D is *recurrent*).

- (d) Let $d \geq 3$. Prove that $B_t \rightarrow \infty$ as $t \rightarrow \infty$ almost surely (in other words, the Brownian motion in 3+D is *transient*). *Hint*: consider events $A_n := \{\exists s < t : B_s \geq n^3, B_t \leq n\}$.
(e) (*) Let $d \geq 2$, $U \in \mathbb{R}^d$ is a bounded open set, $x \in \partial U$ and there exist an open cone C (with the vertex at x) such that $B(x, r) \cap C \subset \mathbb{R}^d \setminus U$ for some $r > 0$. Prove that x is a *regular boundary point* of U , i.e. $\inf\{t > 0 : B_t^x \notin U\} = 0$ almost surely.

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15. N-DIMENSIONAL BROWNIAN MOTION: ITÔ'S CALCULUS, LOCAL MARTINGALES AND BESSEL PROCESSES

- (a) Let $\alpha \in \mathbb{R}$ and B_t be a N -dimensional Brownian motion started at $x \neq 0$. Consider (local) semi-martingale $X_t := |B_t|^\alpha$. Compute dX_t using Itô's calculus.
- (b) Note that X_t is a local martingale if $\alpha = 2 - N$, which agrees with the fact that the function $H(x) = |x|^{2-N}$ is harmonic in \mathbb{R}^N (one should consider $\log |B_t|$ for $N = 2$).
Let $N = 3$, $\alpha = -1$ and $X_t = |B_t|^{-1}$ so that X_t is a local martingale. Note that, almost surely, the process $(X_t)_{t \geq 0}$ is well-defined for all $t \geq 0$ and $\lim_{t \rightarrow \infty} X_t = 0$.
- (c) Give an example of a sequence $(\tau_n)_{n \in \mathbb{N}}$ of localizing stopping times for $(X_t)_{t \geq 0}$.
- (d) Check that $\mathbb{E}[X_t^2] \leq \text{const} < +\infty$ for some constant independent of $t \geq 0$.
- (e) What is the law of the random variable $\max_{t \in [0, +\infty)} X_t$? Does it belong to $L^1(\Omega)$?
- (f) Using the explicit formula for the Gaussian density, prove that $\mathbb{E}[X_T] < X_0 = x^{-1}$.

Hint: Use the identity $|S_r|^{-1} \int_{S_r} |y+x|^{-1} d\lambda_{S_r}(x) = \min\{x^{-1}, r^{-1}\}$, where λ_{S_r} denotes the Lebesgue measure on the sphere $S_r = \{y \in \mathbb{R}^3 : |y|=r\}$ and $|S_r|$ is the area of S_r .

Remark. Thus, X_t is a local martingale but not a (true) martingale.

Let $(B_t)_{t \geq 0}$ be a N -dimensional Brownian motion started at $x \neq 0$. Denote

$$\beta_t := \sum_{k=1}^N \int_0^t \frac{B_t^k}{|B_t|} dB_t^k,$$

(we set $B_t^k/|B_t| = 0$ if $|B_t| = 0$). Check that this stochastic integral is well-defined.

- (g) Prove that the process $(\beta_t)_{t \geq 0}$ is a 1-dimensional Brownian motion started at 0.
Hint: Check that $(\beta_t)_{t \geq 0}$ and $(\beta_t^2 - t)_{t \geq 0}$ are local martingales, and use Lévy's theorem.
- (h) Using (a), show that

$$\begin{aligned} |B_t| &= |x| + \beta_t + \frac{N-1}{2} \int_0^t \frac{ds}{|B_s|}, \\ |B_t|^2 &= |x|^2 + 2 \int_0^t |B_s| d\beta_s + Nt. \end{aligned}$$

Definition. Let $(\beta_t)_{t \geq 0}$ be a standard 1-dimensional Brownian motion started at 0 and $m \geq 0$. A process $(X_t)_{t \geq 0}$ satisfying the stochastic differential equation

$$dX_t = 2\sqrt{X_t} d\beta_t + mdt$$

is called a squared Bessel process of dimension m . The process $Y_t := \sqrt{X_t}$ (if $m < 2$, there is an issue with signs of Y_t) is called an m -dimensional Bessel process.

Remark. If $m = N \geq 2$ is integer, then $(Y_t)_{t \geq 0} \stackrel{(d)}{=} (|B_t|)_{t \geq 0}$. Note that, almost surely, $|B_t| \neq 0$ for all $t \geq 0$. Thus in this case there exists a unique continuous choice of the square root of the squared process $(X_t)_{t \geq 0} \stackrel{(d)}{=} (|B_t|^2)_{t \geq 0}$.

BROWNIAN MOTION AND STOCHASTIC CALCULUS

Master class 2015-2016

16. LOCAL TIME AT ZERO FOR 1D BROWNIAN MOTION

Let $(B_t)_{t \geq 0}$ be a one-dimensional Brownian motion. For every $\varepsilon > 0$, we define a function

$$g_\varepsilon : \mathbb{R} \rightarrow \mathbb{R}_+, \quad g_\varepsilon(x) := \sqrt{x^2 + \varepsilon^2}.$$

Note that $g'_\varepsilon(x) \rightarrow \text{sign}(x)$ as $\varepsilon \rightarrow 0$ uniformly on each of the sets $\{x \in \mathbb{R} : |x| \geq \delta\}$, $\delta > 0$.

- (a) Apply Itô's formula to compute a decomposition $g_\varepsilon(B_t) = g_\varepsilon(B_0) + M_t^\varepsilon + A_t^\varepsilon$, where M_t^ε is a local martingale and A_t^ε is a bounded variation process. Observe that M_t^ε is a square integrable martingale and A_t^ε is increasing.
- (b) Show that, for each $T > 0$,

$$(M_t^\varepsilon)_{t \in [0, T]} \xrightarrow{\varepsilon \rightarrow 0} (\beta_t)_{t \in [0, T]} \quad \text{in } L^2(\Omega), \quad \text{where } \beta_t := \int_0^t \text{sign}(B_s) dB_s$$

and the convergence is understood in the metric of $C([0, T])$.

Remark. Recall that Lévy's theorem implies that the process $(\beta_t)_{t \geq 0}$ is another one-dimensional Brownian motion defined on the same filtration as $(B_t)_{t \geq 0}$.

- (c) Infer that there exists an increasing process $(L_t^0)_{t \geq 0}$ such that

$$|B_t| = \int_0^t \text{sign}(B_s) dB_s + L_t^0 \quad \text{for all } t \geq 0.$$

Observing that $(A_t^\varepsilon)_{t \in [0, T]} \rightarrow (L_t^0)_{t \in [0, T]}$ as $\varepsilon \rightarrow 0$ show that, for every choice of the segment $[u, v] \subset \mathbb{R}_+$ and $\delta > 0$, the following is true:

$$|B_t| \geq \delta \quad \text{for all } t \in [u, v] \quad \Rightarrow \quad L_u^0 = L_v^0.$$

Infer that L_t^0 is constant on each of the intervals $(u, v) \subset \mathbb{R}_+$ such that $B_t \neq 0$ on (u, v) .

Definition. The process L_t^0 is called a local time of the Brownian motion B_t at 0.

- (d) Given $\varepsilon > 0$, define two sequences of stopping times $(\sigma_n^\varepsilon)_{n \geq 1}$ and $(\tau_n^\varepsilon)_{n \geq 1}$ inductively by setting $\sigma_1^\varepsilon := 0$,

$$\tau_n^\varepsilon := \inf\{t > \sigma_n^\varepsilon : |B_t| = \varepsilon\} \quad \text{and} \quad \sigma_{n+1}^\varepsilon := \inf\{t > \tau_n^\varepsilon : B_t = 0\}.$$

Further, let $N_t^\varepsilon := \max\{n \geq 1 : \tau_n^\varepsilon \leq t\}$ with the convention $\max \emptyset := 0$. Show that

$$\varepsilon N_t^\varepsilon \xrightarrow{\varepsilon \rightarrow 0} L_t^0 \quad \text{in } L^2(\Omega).$$

Hint: Observe that

$$\left| L_t^0 + \int_0^t (\sum_{n \geq 1} \mathbf{1}_{[\sigma_n^\varepsilon, \tau_n^\varepsilon]}(s)) \text{sign}(B_s) dB_s - \varepsilon N_t^\varepsilon \right| \leq \varepsilon.$$

Remark. Note that L_t^0 can be reconstructed from the absolute value $|B_t|$ of the Brownian motion B_t . In particular, this means that the filtration generated by the Brownian motion $\beta_t = |B_t| - L_t^0$ is strictly smaller than the one generated by B_t itself.