

## BROWNIAN MOTION AND STOCHASTIC CALCULUS HW 3

TOMMENIX YU  
ID: 12370130  
STAT 38500

Discussed with classmates.

### Exercise 1.

*Proof.*

By Taylor we can write

$$MV(f; 0, \varepsilon) = \int_{B_\varepsilon} \left( f(0) + \sum \partial_i f(0) x_i + \frac{1}{2} \sum \partial_i^2 f(0) x_i^2 + \frac{1}{1!1!} \sum_{i \neq j} \partial_i \partial_j f(0) x_i x_j \right) ds + o(|x|^2)$$

and we note that the second and fourth term above is actually 0:

$$\int_{B_\varepsilon} \sum \partial_i f(0) x_i ds = \sum \partial_i f(0) \int_{B_\varepsilon} x_i ds = 0$$

by symmetry, and note that the operator

$$F_j : (\mathbb{R}^d)^* \rightarrow (\mathbb{R}^d)^*$$

where

$$F_j(f(x_1, x_2, \dots, x_n)) = f(x_1, \dots, -x_j, \dots, x_n)$$

is just flipping the space, thus is invariant under integral over  $\varepsilon$ -sphere for  $f = \int_{B_\varepsilon} ds$ , i.e.

$$\begin{aligned} \int_{B_\varepsilon} \sum_{i \neq j} \partial_i \partial_j f(0) x_i x_j ds &= \sum_{i \neq j} \partial_i \partial_j f(0) \int_{B_\varepsilon} ds = \sum_{i \neq j} \partial_i \partial_j f(0) F \left( \int_{B_\varepsilon} ds \right) \\ &= - \sum_{i \neq j} \partial_i \partial_j f(0) \int_{B_\varepsilon} ds \end{aligned}$$

and hence the value is 0. Now putting everything together we have (in definition of  $MV$ , the surface integral is normalized)

$$\begin{aligned} \lim_{\varepsilon \downarrow 0} \frac{MV(f; 0, \varepsilon) - f(0)}{\varepsilon^2} &= \lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon^2} \int_{B_\varepsilon} \frac{1}{2} \sum \partial_i^2 f(0) x_i^2 ds + o(\varepsilon^2) \\ &= \lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon^2} \frac{1}{2d} \Delta f(0) = \frac{1}{2d} \Delta f(0) \end{aligned}$$

□

**Exercise 2.***Proof.*

(1): First, WLOG let's shift the whole process down by  $\pi/2$  so that we start now at 0. Note that  $B_T \in \{-\pi/2, \pi/2\}$ , and by Doob's stopping time we know

$$\mathbb{P}(B_T = \pi/2) = \mathbb{P}(B_T = -\pi/2) = \frac{1}{2}.$$

For  $a \in \mathbb{R}$  we can have

$$\mathbb{P}(T < a, B_T = \pi/2) = \mathbb{P}(T < a, B_T = -\pi/2)$$

since  $X \sim -X$ . But note that

$$\mathbb{P}(\{T < a, B_T = \pi/2\} \cup \{T < a, B_T = -\pi/2\}) = \mathbb{P}(T < a)$$

by definition of  $B_T$  and thus

$$\mathbb{P}(T < a, B_T = \pi/2) = \frac{1}{2}\mathbb{P}(T < a) = \mathbb{P}(B_T = \pi/2)\mathbb{P}(T < a)$$

now for any  $b \notin [-\pi/2, \pi/2)$  the relation

$$\mathbb{P}(T < a, B_T > b) = \mathbb{P}(B_T > b)\mathbb{P}(T < a)$$

trivially holds, and for any  $b \in [-\pi/2, \pi/2)$  we know  $\{B_T > b\} = \{B_T = \pi/2\}$  and thus

$$\mathbb{P}(T < a, B_T > b) = \mathbb{P}(B_T > b)\mathbb{P}(T < a)$$

also holds. So we know  $T$  and  $B_T$  are independent.

(2): WLOG we shift again to a Brownian motion starting at 0 and the stopping time  $T$  becomes  $\min\{t : B_t = a \vee B_t = -b\}$  where  $a + b = \pi$  and WLOG let  $a < b$ .

By Doob's stopping time we know  $B_T$  is a Martingale (same argument as in last time) and  $\mathbb{E}[B_T] = \mathbb{E}[B_0] = 0$ . Thus

$$\text{Cov}(T, B_T) = \mathbb{E}[TB_T]$$

it suffices to show it nonzero.

**Lemma 0.1.**  $M_t := B_t^3 - 3tB_t$  is a Martingale.

*Proof.* We omit the proof of  $L^1$  and adapted since it's the same as we did in last homework, question 3. For Martingale property, we note that for  $s < t$

$$B_t^3 = (B_s + (B_t - B_s))^3 = B_s^3 + 3B_s^2(B_t - B_s) + 3B_s(B_t - B_s)^2 + (B_t - B_s)^3$$

and taking conditional expectation we have

$$\begin{aligned} \mathbb{E}[B_t^3 | \mathcal{F}_s] &= B_s^3 + 3B_s^2\mathbb{E}[B_t - B_s | \mathcal{F}_s] + 3B_s\mathbb{E}[(B_t - B_s)^2 | \mathcal{F}_s] + \mathbb{E}[(B_t - B_s)^3 | \mathcal{F}_s] \\ &= B_s^3 + 3B_s^2 \cdot 0 + 3B_s(t - s) + 0 = B_s^3 + 3(t - s)B_s \end{aligned}$$

and

$$\begin{aligned}\mathbb{E}[3tB_t|\mathcal{F}_s] &= 3sB_s + 3s\mathbb{E}[B_t - B_s|\mathcal{F}_s] + 3(t-s)\mathbb{E}[B_s|\mathcal{F}_s] \\ &= 3sB_s + 3s \cdot 0 + 3B_s(t-s)\end{aligned}$$

and thus

$$\mathbb{E}[B_t^3 - 3tB_t|\mathcal{F}_s] = B_s^3 - 3sB_s = \mathbb{E}[B_s^3 - 3sB_s|\mathcal{F}_s]$$

so it is indeed a martingale. □

(A corollary of the above is that we can find Martingales of any degree using the coefficients of  $(x - y)^k$  for even degrees on  $y$ , but we'll not need that here.)

Now Lemma + Doob's stopping time theorem says  $B_T^3 - 3TB_T$  is a Martingale and thus we can compute

$$3\mathbb{E}[TB_T] = \mathbb{E}[B_T^3] = a^3 \frac{b}{a+b} - b^3 \frac{a}{a+b} = ab(a-b) \neq 0$$

thus

$$\text{Cov}(T, B_T) = \mathbb{E}[TB_T] \neq 0$$

which concludes the proof. □

**Exercise 3.***Proof.*

(1):

To solve:

$$\begin{cases} \partial_t p_t(y) = \frac{1}{2} \Delta_y p_t(y) \\ P_0(y) = \delta_x \end{cases}$$

Let's assume the solution is of the form

$$p_t(y) = e^{-\lambda t} \phi(y)$$

then the equation gives

$$\phi''(y) = -2\lambda \phi(y)$$

which yields only solution

$$\phi(y) = c_1 \sin(\sqrt{2\lambda}y) + c_2 \cos(\sqrt{2\lambda}y)$$

where imposing the boundary condition  $p_t(0) = p_t(\pi)$  really gives us that  $\lambda = \frac{k^2}{2}$  for  $k \in \mathbb{Z}^*$  (since we want it to decay) and  $p_t(y) = \sin(ky)$ . Now linearity of solution yields formally

$$p_t(y) = \sum_{k=1}^{\infty} a_k e^{-k^2 t/2} \sin(ky)$$

where it's only defined if it converges. But plugging this in initial condition gives  $a_k$  to be the Fourier coefficients of  $\delta_x(y)$  which in turn gives

$$a_k = \frac{2}{\pi} \int_0^{\pi} \delta_x(y) \sin(ky) dy = \frac{2}{\pi} \sin(kx)$$

and thus simply plugging in to our assumed function we get

$$p_t(y) = \frac{2}{\pi} \sum_{k=1}^{\infty} e^{-k^2 t/2} \sin(kx) \sin(ky)$$

which is indeed convergent since  $e^{-k^2} \ll o(k^{-2})$  as  $k \rightarrow \infty$  (this is even a computable geometric series).

(2):

When  $t \rightarrow \infty$  the leading term is just  $k = 1$  and thus

$$p_t(x, y) \sim \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y)$$

in particular, to fit the form of question we have

$$\lambda = -\frac{1}{2}, \quad c(x) = \frac{2}{\pi} \sin(x), \quad \tilde{c}(y) = \sin(x).$$

(3):

We can of course give a full characterization of things using the formula

$$p_t(y) = \frac{2}{\pi} \sum_{k=1}^{\infty} e^{-k^2 t/2} \sin(kx) \sin(ky)$$

and integrate against  $y$ , but since we are eventually taking limit  $t \rightarrow \infty$  and  $t$  and  $y$  are separable so eventually only the leading term matters. But to be less sloppy let's say that for any  $\varepsilon$  there exists large enough  $t$  such that (by (2))

$$\frac{1 - \varepsilon \int_a^b \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy}{1 + \varepsilon \int_0^\pi \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy} \leq \frac{\int_a^b p_t(y) dy}{\int_0^\pi p_t(y) dy} \leq \frac{1 + \varepsilon \int_a^b \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy}{1 - \varepsilon \int_0^\pi \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy}$$

and since  $\varepsilon$  is arbitrary we see that after limit the terms are really the same, thus plugging in we get

$$\begin{aligned} \lim_{t \rightarrow \infty} \mathbb{P}^x \{B_t \in I | t < T\} &= \lim_{t \rightarrow \infty} \frac{\int_a^b p_t(y) dy}{\int_0^\pi p_t(y) dy} = \lim_{t \rightarrow \infty} \frac{\int_a^b \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy}{\int_0^\pi \frac{2}{\pi} e^{-t/2} \sin(x) \sin(y) dy} \\ &= \frac{\cos a - \cos b}{2}. \end{aligned}$$

(4):

To not stop at  $2t$  is to arrive at a certain point  $z$  at time  $t$  then go to  $y$  after another  $t$ .

To be explicit, we have that as  $t \rightarrow \infty$  the leading term behaves like

$$\begin{aligned} p_{2t}(x, y) &= \int_0^\pi p_t(x, z) p_t(z, y) dz = \int_0^\pi \frac{2}{\pi} e^{-t/2} \sin(x) \sin(z) \frac{2}{\pi} e^{-t/2} \sin(z) \sin(y) dz \\ &= \frac{4}{\pi^2} e^{-t} \sin(x) \sin(y) \int_0^\pi \sin^2(z) dz = \frac{2}{\pi} e^{-t} \sin(x) \sin(y) \end{aligned}$$

and to compute the nominator we use a similar integral which states: density such that  $B_t \in L$  conditioned on  $T > 2t$ :

$$\begin{aligned} p_{2t,a,b}(x, y) &= \int_a^b p_t(x, z) p_t(z, y) dz = \int_a^b \frac{2}{\pi} e^{-t/2} \sin(x) \sin(z) \frac{2}{\pi} e^{-t/2} \sin(z) \sin(y) dz \\ &= \frac{4}{\pi^2} e^{-t} \sin(x) \sin(y) \int_a^b \sin^2(z) dz = \frac{(2b - 2a + \sin(2a) - \sin(2b))}{4} \frac{4}{\pi^2} e^{-t} \sin(x) \sin(y) \end{aligned}$$

and thus by putting those together (skipping the exact same approximating argument as above):

$$\begin{aligned} \lim_{t \rightarrow \infty} \mathbb{P}^x \{B_t \in I | 2t < T\} &= \lim_{t \rightarrow \infty} \frac{\int_0^\pi p_{2t,a,b}(x, y) dy}{\int_0^\pi p_{2t}(x, y) dy} = \lim_{t \rightarrow \infty} \frac{\frac{(2b - 2a + \sin(2a) - \sin(2b))}{4} \frac{4}{\pi^2} e^{-t} \sin(x) \sin(y)}{\frac{2}{\pi} e^{-t} \sin(x) \sin(y)} \\ &= \frac{(2b - 2a + \sin(2a) - \sin(2b))}{2\pi}. \end{aligned}$$

□

**Exercise 4.***Proof.*

(1): By Markov property, a Brownian motion starting at a new time is just a shifted Brownian motion, hence by independent increment

$$\sup_{x \in D} \mathbb{P}^x \{T > s + t\} = \sup_{x \in D} \mathbb{P}^x \{T > s\} \mathbb{P}^{B_s^x} \{T > t\} \leq \sup_{x \in D} \mathbb{P}^x \{T > s\} \sup_{x \in D} \mathbb{P}^x \{T > t\}$$

in other words

$$q_{s+t} \leq q_s q_t.$$

(2):

Notice that  $f(t) := \log q_t$  is subadditive since  $\log q_{t+s} \leq \log q_t q_s = \log q_t + \log q_s$ . Denote  $s := \inf_{t>0} \frac{f(t)}{t}$  (note this can be  $-\infty$ ). We want to show that

$$\lim_{t \rightarrow \infty} \frac{f(t)}{t} = \inf_{t>0} \frac{f(t)}{t}.$$

Suppose contrary, then there exists a sequence  $t_n$  such that  $\frac{f(t_n)}{t_n} \geq s + \varepsilon$  for some  $\varepsilon$ .

By definition of  $\inf$  there exist  $T \in (0, \infty)$  such that  $\frac{f(T)}{T} < s + \frac{\varepsilon}{2}$ . Since  $q_t$  is a probability, we have  $f(t) \leq 0$ , and thus if we denote  $t_n \in (mT, (m+1)T)$  then

$$\frac{f(t_n)}{t_n} \leq \frac{f(mT) + f(t_n - mT)}{t_n} \leq \frac{f(T)}{t_n/m} + \frac{f(t_n - mT)}{t_n}$$

where as  $t_n \rightarrow \infty$  we know  $\frac{t_n}{m} \leq T + \frac{T}{m} \rightarrow T$  and hence

$$\lim_{n \rightarrow \infty} \frac{f(t_n)}{t_n} \leq \frac{f(T)}{T} < s + \varepsilon/2$$

is a contradiction. Hence

$$\lim_{t \rightarrow \infty} \frac{\log q_t}{t} = \inf_{t>0} \frac{\log q_t}{t}$$

and we only need to show

$$\inf_{t>0} \frac{\log q_t}{t} \in (-\infty, 0).$$

The  $< 0$  direction is clear, so we only need to convince ourselves that  $\lim_{t \rightarrow \infty} \log q_t = O(t)$  for which we use the later bound.

Now, let  $r := r_x := \text{dist}(x, D)$  and  $R := R_x := \max_{t \in \partial D} \text{dist}(x, t)$  which are just the smallest and largest distance of  $x$  to the boundary. Now define  $T_R^x := \min\{t : |B_t - x| = R\}$  and  $T_r^x := \min\{t : |B_t - x| = r\}$ , also, denote

$$T_R^{x,1} = \min\{t : B_t^1 - x_1 \geq R\}, \quad T_r^{x,1} = \min\{t : B_t^1 - x_1 \geq r/\sqrt{d}\}$$

then

$$\mathbb{P}(T_R^{x,1} > t) \leq \mathbb{P}^x\{T_R^x > t\} \leq \mathbb{P}^x\{T > t\} \leq \mathbb{P}\{T_r^x > t\} \leq 2\mathbb{P}(T_r^{x,1} > t)$$

since to reach outside of  $B_R$  contains (as an event) going above  $+R$  in one dimension, and to reach  $B_r$  at least one dimension has to go beyond  $r/\sqrt{d}$ , which again we use union bound to say that (probability of either reaching from below and from above)  $\leq$  (probability of reaching from below) + (probability of reaching from above).

So we can get our result if we can find a lower bound exponential in  $t$  of  $\mathbb{P}(T_r^{x,1} > t)$ . To be exponential in  $t$  roughly suggests that we should cut time in  $\lceil t \rceil$  pieces, so we only need to show that there's positive probability that  $\mathbb{P}(T_r^{x,1} > 1, \text{ start and ends in a same situation })$ .

From discussion in class, we can find a neighborhood  $N_x$  of  $x$ , contained in ball of radius  $r$ , such that  $\forall y \in N_x, \mathbb{P}(T_r^{y,1} > 1, B_1 \in N_x) > c_1$  for some constant  $c_1$ . In particular  $x \in N_x$  so we get a bound. Thus

$$\mathbb{P}(T_r^{x,1} > t) \geq c_1^{\lceil t \rceil \frac{1}{|t|}} \geq c_1^{2\lceil t \rceil}$$

hence taking log

$$\log \mathbb{P}(T_r^{x,1} > t) \gtrsim O(t)$$

hence we are done for approximation of one point. But since all inequalities holds for all point (well, we can bound  $r_x \leq \text{diam}(D)$ ), we get the bound that

$$\lim_{t \rightarrow \infty} \frac{\log q_t}{t} = C > -\infty$$

hence  $\lambda \in (0, \infty)$ .

(3):

First, we show that the sup is taken at  $x = 0$ . Thus, now shifting  $x$  to  $a/2$  the boundary condition implies  $\lambda$  (from Q3 (1)) is  $\frac{k^2 \pi^2}{8a^2}$  which shows

$$p_t(y) = \sum_{k=1}^{\infty} a_k e^{-k^2 \pi^2 t / 8a^2} \sin\left(\frac{k\pi}{2a} y\right)$$

and by letting  $t \rightarrow \infty$  we get

$$p_t(x, y) \sim \frac{2}{\pi} e^{-\pi^2 t / 8a^2} \sin\left(\frac{\pi}{2a} x\right) \sin\left(\frac{\pi}{2a} y\right)$$

and hence

$$\begin{aligned} \lim_{t \rightarrow \infty} \frac{\log q_t}{t} &= \lim_{t \rightarrow \infty} \frac{\log \mathbb{P}(T > t)}{t} = \lim_{t \rightarrow \infty} \frac{\log \int_0^{2a} \frac{2}{\pi} e^{-\pi^2 t / 8a^2} \sin\left(\frac{\pi}{2a} x\right) \sin\left(\frac{\pi}{2a} y\right) dy}{t} \\ &= \lim_{t \rightarrow \infty} \frac{\log \frac{4}{\pi} - \frac{t\pi^2}{8a^2}}{t} \stackrel{x=a}{=} -\frac{\pi^2}{8a^2} + \lim_{t \rightarrow \infty} O(t^{-1}) = -\frac{\pi^2}{8a^2} = -\lambda_D. \end{aligned}$$

□

**Exercise 5.***Proof.*

(1): By DCT (compact support) and simple function approximation we can exchange the limit

$$\mathbb{E}^x \left[ \int_0^\infty f(B_t) dt \right] = \int_0^\infty \mathbb{E}^x[f(B_t)] dt = \int_{\mathbb{R}^d} f(y) P_t(x, y) dy$$

for  $P_t(x, y) = \frac{1}{(2\pi t)^{d/2}} e^{-|y-x|^2/2t}$  as defined in class. Plugging in we have

$$\begin{aligned} \mathbb{E}^x \left[ \int_0^\infty f(B_t) dt \right] &= \int_0^\infty \mathbb{E}^x[f(B_t)] dt = \int_0^\infty \int_{\mathbb{R}^d} f(y) \frac{1}{(2\pi t)^{d/2}} e^{-|y-x|^2/2t} dy dt \\ &= \int_{\mathbb{R}^d} \int_0^\infty \frac{1}{(2\pi t)^{d/2}} e^{-|y-x|^2/2t} dt f(y) dy = \int_{\mathbb{R}^d} G(x, y) f(y) dy \end{aligned}$$

where we've changed integral again by  $f$  compact support.

(2):

(Idea by Zihao He)

For  $\varepsilon > 0$  define stopping time  $T_\varepsilon := \inf\{t \geq 0 : |B_t - x| \geq \varepsilon\}$ . And we separate the integral to get

$$\phi(x) = \mathbb{E}^x \left[ \int_0^{T_\varepsilon} f(B_t) dt \right] + \mathbb{E}^x \left[ \int_{T_\varepsilon}^\infty f(B_t) dt \right]$$

and by Strong Markov property we just shift by  $B_{T_\varepsilon}$  which is the same thing as uniformly shift to the  $\varepsilon$  ball:

$$\mathbb{E}^x \left[ \int_{T_\varepsilon}^\infty f(B_t) dt \right] = \mathbb{E}^x \left[ \mathbb{E}^{B_{T_\varepsilon}} \left[ \int_0^\infty f(B_{t+T_\varepsilon}) dt \right] \right] = \mathbb{E}^x[\phi(B_{T_\varepsilon})] = MV(\phi; x, \varepsilon)$$

but now miraculously we plug in question 1 to get

$$\frac{1}{2} \Delta \phi(x) = d \lim_{\varepsilon \rightarrow 0} \frac{MV(\phi; x, \varepsilon) - \phi(x)}{\varepsilon^2} = d \lim_{\varepsilon \rightarrow 0} \frac{-\mathbb{E}^x \left[ \int_0^{T_\varepsilon} f(B_t) dt \right]}{\varepsilon^2}$$

and as  $\varepsilon \rightarrow 0$  by Taylor (compact so  $\|f'\| < \infty$ ) on  $f$  we have  $\int_0^{T_\varepsilon} f(B_t) dt = f(x) \cdot T_\varepsilon + o(\varepsilon)$  and from last homework we know the expectation of stopping time is  $\mathbb{E}^x[T_\varepsilon] = \frac{\varepsilon^2}{d}$  and thus

$$\frac{-\mathbb{E}^x \left[ \int_0^{T_\varepsilon} f(B_t) dt \right]}{\varepsilon^2} = \frac{-d \frac{\varepsilon^2}{d} f(x) + o(\varepsilon^3)}{\varepsilon^2}$$

taking the limit it goes to  $-f(x)$ .

□