

## Solution of homework assignment : On the Brownian bridge.

### Exercise 1 — Absolute continuity.

You have shown in the second exercise session that

$$\text{Law}(B_{|[0,1]}|B_1 \in dx) = \text{Law}(x\text{Id} + \beta),$$

In other words, for every bounded measurable  $H$ ,

$$\mathbb{E}[H(B_{|[0,1]}, B_1)] = \int_{\mathbb{R}} \mathbb{E}[H(x\text{Id} + \beta, x)] \mathbb{P}_{B_1}(dx).$$

- (1) For  $\varepsilon > 0$ , let  $\nu_\varepsilon = \text{Law}(B_{|[0,1]} | -\varepsilon \leq B_1 \leq \varepsilon)$  be the (deterministic!) probability measure such that for every bounded measurable  $H$ ,

$$\int_{\mathcal{C}([0,1])} H(\varphi) \nu_\varepsilon(d\varphi) = \frac{\mathbb{E}[H(B_{|[0,1]}) \mathbf{1}_{|B_1| \leq \varepsilon}]}{\mathbb{P}(|B_1| \leq \varepsilon)}.$$

Show that it converges (in the weak topology of measures), as  $\varepsilon \rightarrow 0$ , to  $\text{Law}(\beta)$ .

Let  $H$  be a bounded continuous functional.

$$\begin{aligned} \int_{\mathcal{C}([0,1])} H(\varphi) \nu_\varepsilon(d\varphi) &= \frac{\mathbb{E}[H(B_{|[0,1]}) \mathbf{1}_{|B_1| \leq \varepsilon}]}{\mathbb{P}(|B_1| \leq \varepsilon)} = \frac{\int \mathbb{P}(B_1 \in dx) \mathbf{1}_{x \leq \varepsilon} \mathbb{E}[H(x\text{Id} + \beta)]}{\int \mathbb{P}(B_1 \in dx) \mathbf{1}_{x \leq \varepsilon}} \\ &= \frac{\int_{-\varepsilon}^{\varepsilon} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \mathbb{E}[H(x\text{Id} + \beta)] dx}{\int \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx} \\ &\underset{\varepsilon \rightarrow 0}{\sim} \frac{2\varepsilon \frac{1}{\sqrt{2\pi}} \mathbb{E}[H(\beta)]}{2\varepsilon \frac{1}{\sqrt{2\pi}}} = \mathbb{E}[H(\beta)] \end{aligned}$$

because both  $x \mapsto e^{-x^2/2} \mathbb{E}[H(x\text{Id} + \beta)]$  and  $x \mapsto e^{-x^2/2}$  are continuous at  $x = 0$  (the first one thanks to the dominated convergence theorem). Hence the question.

- (2) For  $0 < a < 1$ , what does the Markov property say about the joint distribution of  $(B_{|[0,a]}, B_1)$ ? Deduce that, for  $H$  positive bounded continuous  $\mathcal{C}([0, a]) \rightarrow \mathbb{R}$ , the following quantity:

$$\mathbb{E}[H(B_{|[0,a]}) | |B_1| < \varepsilon] = \frac{\mathbb{E}[H(B_{|[0,a]}) \mathbf{1}_{|B_1| \leq \varepsilon}]}{\mathbb{P}(|B_1| \leq \varepsilon)} = \int_{\mathcal{C}([0,1])} H(\varphi_{|[0,a]}) \nu_\varepsilon(d\varphi).$$

converges, as  $\varepsilon \rightarrow 0$ , to

$$\int_{\mathcal{C}([0,a])} H(\phi) \frac{1}{\sqrt{1-a}} \exp\left(-\frac{\phi(a)^2}{2(1-a)}\right) \mathbb{P}_{B_{|[0,a]}}(d\phi).$$

Markov's property says the distribution of  $B_1 - B_a$  is a centered Gaussian of variance  $1 - a$ , independent of  $B_{|[0,a]}$ . Hence

$$\begin{aligned}
& \mathbb{E}[H(B_{|[0,a]}) \mathbb{1}_{|B_1| < \varepsilon}] \\
&= \mathbb{E}[H(B_{|[0,a]}) \mathbb{1}_{|B_1 - B_a + B_a| < \varepsilon}] \\
&= \int_{\mathcal{C}([0,a])} \mathbb{P}(B_{|[0,a]} \in d\phi) \int_{\mathbb{R}} \frac{dx}{\sqrt{2\pi(1-a)}} \exp\left(-\frac{x^2}{2(1-a)}\right) H(\phi) \mathbb{1}_{|x+\phi(a)| < \varepsilon} \\
&= \int_{\mathcal{C}([0,a])} \mathbb{P}(B_{|[0,a]} \in d\phi) H(\phi) \int_{\phi(a)-\varepsilon}^{\phi(a)+\varepsilon} \frac{dx}{\sqrt{2\pi(1-a)}} \exp\left(-\frac{x^2}{2(1-a)}\right) \\
&\sim 2\varepsilon \int_{\mathcal{C}([0,a])} \mathbb{P}(B_{|[0,a]} \in d\phi) H(\phi) \frac{1}{\sqrt{2\pi(1-a)}} \exp\left(-\frac{\phi(a)^2}{2(1-a)}\right),
\end{aligned}$$

by dominated convergence. As  $\mathbb{P}(|B_1| \leq \varepsilon) \sim \frac{2\varepsilon}{\sqrt{2\pi}}$ , dividing these two asymptotics yields the result.

- (3) Deduce that the distribution of  $\beta_{|[0,a]}$  is absolutely continuous with regard to that of  $B_{|[0,a]}$  when  $a < 1$ . Is it the case when  $a = 1$  ?

By question 1, since the restriction  $B_{|[0,1]} \mapsto B_{|[0,a]}$  is a continuous map, we have as  $\varepsilon \rightarrow 0$

$$\mathbb{E}[H(B_{|[0,a]}) \mid |B_1| \leq \varepsilon] \rightarrow \mathbb{E}[H(\beta_{|[0,a]})]$$

Equating this with question 2 gets

$$\mathbb{E}[H(\beta_{|[0,a]})] = \int_{\mathcal{C}([0,a])} \mathbb{P}(B_{|[0,a]} \in d\phi) H(\phi) \frac{1}{\sqrt{1-a}} \exp\left(-\frac{\phi(a)^2}{2(1-a)}\right)$$

Hence (because equality against all bounded continuous functions characterizes equality of measures) the law of  $\beta_{|[0,a]}$  has density  $\phi \mapsto \frac{1}{\sqrt{1-a}} \exp\left(-\frac{\phi(a)^2}{2(1-a)}\right)$  w.r.t. the law of  $B_{|[0,a]}$ .

When  $a = 1$  this is not the case anymore: look at the set  $\{\phi, \phi(1) = 0\}$ .  $\beta$  belongs to it with probability 1, and  $B$  with probability 0.

### Exercise 2 — Location of the minimum.

We want to compute the distribution of  $T = \inf\{t \geq 0, \beta_t = \min_{[0,1]} \beta\}$ .

- (1) Show that the global minimum of  $\beta$  is almost surely reached exactly once. You may use the fact that for every  $a < b < c < d \in \mathbb{Q} \cap (0, 1)$ , the global minimum of  $B$  on  $[a, b]$  and  $[c, d]$  are almost surely different (4th exercise session)

For every  $a < b < c < d < 1$  this almost sure property of  $B_{|[0,d]}$  is also true for  $\beta_{|[0,d]}$  by absolute continuity. Hence by countable union, almost surely for every  $a < b < c < d \in \mathbb{Q} \cap (0, 1)$ , the global minimum of  $\beta$  on  $[a, b]$  and  $[c, d]$  are different.

Moreover we know that almost surely the global minimum of  $\beta$  on  $[0, 1]$  is not 0 (otherwise since  $B = \beta + B_1 \text{Id}$  then  $\liminf_{t \rightarrow 0} B_t/t > -\infty$  which is almost surely not the case.) These two properties imply that almost surely the global minimum of  $\beta$  is reached only once.

- (2) Show that the Brownian bridge is cyclically exchangeable, i.e. that for every  $x \in [0, 1)$ , the process  $t \mapsto \beta_{(x+t) \bmod 1} - \beta_x$  is still distributed like  $\beta$ . (You may start by reasoning on the Brownian motion.)

A first idea is to define  $X_t = B_{(x+t) \bmod 1} - B_x$ . However this has a jump discontinuity at  $t = 1 - x$ . Hence we rather consider the continuous process

$$X_t = B_{(x+t) \bmod 1} - B_x + B_1 \mathbf{1}_{x+t > 1}.$$

This process is continuous and we may rewrite it as such:

$$\begin{aligned} X_t &= (B_{x+t} - B_x) \mathbf{1}_{x+t \leq 1} + (B_{x+t-1} + B_1 - B_x) \mathbf{1}_{x+t > 1} \\ &= \tilde{B}_t \mathbf{1}_{t \leq 1-x} + (B_{t-(1-x)} + \tilde{B}_{1-x}) \mathbf{1}_{t > 1-x} \end{aligned}$$

after setting  $\tilde{B}_u = B_{x+u} - B_x$ . By Markov's property  $\tilde{B}$  is independent from  $B_{|[0,x]}$ . Hence we have a process that follows a Brownian motion up to time  $1 - x$ , then follows an independent Brownian motion afterwards. This is exactly the description of a Brownian motion given by Markov's property at  $t = 1 - x$ .

Now we look at the Brownian bridge derived from  $X$ . Remark that we have  $B_1 = X_1$ . Hence

$$\begin{aligned} X_t - tX_1 &= B_{(x+t) \bmod 1} - B_x + B_1 \mathbf{1}_{x+t > 1} - tB_1 \\ &= B_{(x+t) \bmod 1} - B_x - B_1(t - \mathbf{1}_{x+t > 1}) \\ &= B_{(x+t) \bmod 1} - (B_x - xB_1) - B_1(t + x - \mathbf{1}_{x+t > 1}) \\ &= B_{(x+t) \bmod 1} - B_1((x+t) \bmod 1) - (B_x - xB_1) = \beta_{(x+t) \bmod 1} - \beta_x. \end{aligned}$$

As a result  $t \mapsto \beta_{(x+t) \bmod 1} - \beta_x$  is a Brownian bridge.

- (3) Deduce the law of  $T$ .

Since the minimum is reached only once, if  $T$  is the argmin of  $\beta$ , then  $U = T - x \bmod 1$  is the argmin of  $t \mapsto \beta_{(x+t) \bmod 1} - \beta_x$ . Hence  $T - x \bmod 1$  is distributed like  $U$ . An immediate consequence is that  $\mathbb{E}[e^{2i\pi n T}] = 0$  for  $n \neq 0$ , hence the bounded random variable  $e^{2i\pi T}$  has the same moments as  $e^{2i\pi U}$  where  $U$  is uniform. So  $e^{2i\pi T} \stackrel{d}{=} e^{2i\pi U}$  and  $T \stackrel{d}{=} U$ .

**Exercise 3** — *Maximum of  $|\beta|$ .*

(1) (a) Show that

$$\mathbb{P}(S > a, |B_1| < \varepsilon) = 2\mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon)$$

In what follows,  $\tilde{B}$  will be the Brownian motion  $B$  reflected at  $T_a$ , and  $\tilde{T}$  the associated hitting times.

$$\begin{aligned} \mathbb{P}(S > a, |B_1| < \varepsilon) &= \mathbb{P}(T_a < 1 \wedge T_{-a}, |B_1| < \varepsilon) + \mathbb{P}(T_{-a} < 1 \wedge T_a, |B_1| < \varepsilon) \\ &= 2\mathbb{P}(T_a < 1 \wedge T_{-a}, |B_1| < \varepsilon) \\ &= 2\mathbb{P}(\tilde{T}_a < \tilde{T}_{-a}, |\tilde{B}_1 - 2a| < \varepsilon) \end{aligned}$$

Hence the result since  $B$  and  $\tilde{B}$  are likewise distributed.

(b) Show that

$$\mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon) = \mathbb{P}(|B_1 - 2a| < \varepsilon) - \mathbb{P}(T_a < T_{-a}, |B_1 - 4a| < \varepsilon)$$

We have

$$\begin{aligned} \mathbb{P}(|B_1 - 2a| < \varepsilon) &= \mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon) + \mathbb{P}(T_{-a} < T_a, |B_1 - 2a| < \varepsilon) \\ &= \mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon) + \mathbb{P}(T_a < T_{-a}, |B_1 + 2a| < \varepsilon) \\ &= \mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon) + \mathbb{P}(\tilde{T}_a < \tilde{T}_{-a}, |\tilde{B}_1 - 4a| < \varepsilon) \end{aligned}$$

Hence the result.

(c) Keep working and deduce an explicit series which equals  $\mathbb{P}(S > a, |B_1| < \varepsilon)$ .

We may show exactly as above the equalities, for  $k \geq 1$

$$\mathbb{P}(T_a < T_{-a}, |B_1 - 2ka| < \varepsilon) = \mathbb{P}(|B_1 - 2ka| < \varepsilon) - \mathbb{P}(T_a < T_{-a}, |B_1 - (2k+1)a| < \varepsilon)$$

Hence, combining all this gives

$$\begin{aligned} \mathbb{P}(T_a < T_{-a}, |B_1 - 2a| < \varepsilon) &= \sum_{k=1}^n (-1)^{k+1} \mathbb{P}(|B_1 - 2ka| < \varepsilon) \\ &\quad + (-1)^n \mathbb{P}(T_a < T_{-a}, |B_1 - (2n+1)a| < \varepsilon). \end{aligned}$$

As the remainder goes to 0 (at the speed  $e^{-n^2/2}$ ...), we have an equality:

$$\mathbb{P}(S > a, |B_1| < \varepsilon) = 2 \sum_{k=1}^{\infty} (-1)^{k+1} \mathbb{P}(|B_1 - 2ka| < \varepsilon)$$

(2) Deduce the cumulative distribution function of  $K$ .

We deduce the expression of the conditional probability

$$\mathbb{P}(S > a | |B_1| < \varepsilon) = 2 \sum_{k=1}^{\infty} (-1)^{k+1} \frac{\mathbb{P}(|B_1 - 2ka| < \varepsilon)}{\mathbb{P}(|B_1| < \varepsilon)}$$

The summands converge each to  $(-1)^{k+1} e^{-2k^2 a^2}$ , and assuming  $\varepsilon < 1$ ,  $k \mapsto e^{-2(k-1)^2 a^2} e^{1/2}$  is a summable dominating function. With the dominated convergence theorem we get

$$\begin{aligned} \mathbb{P}(S > a | |B_1| < \varepsilon) &\xrightarrow{\varepsilon \rightarrow 0} 2 \sum_{k=1}^{\infty} (-1)^{k+1} e^{-2k^2 a^2} \\ &\xrightarrow{\varepsilon \rightarrow 0} \mathbb{P}(K > a) \text{ if } a \text{ is a point of continuity for } K \end{aligned}$$

As we have an equality between a decreasing continuous function and a decreasing function outside of a countable number of points, we have equality everywhere and we are done.