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## TD8 : Construction of the Brownian Motion

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**Exercise 1** — *Transformations.*

Let  $(B_t)_{t \geq 0}$  be a Brownian motion.

- (1) Show that for any  $\lambda \in \mathbb{R}_+^*$ , the process  $(\lambda^{-1/2} B_{\lambda t})_{t \geq 0}$  is a Brownian motion.
- (2) Show that  $B_1 - B_{1-t}$  is a Brownian motion on  $[0, 1]$ .

We first consider the finite-dimensional marginals of the new process  $(X_t)_t$  in these two cases. Remark at first that they still form centered Gaussian vectors, since they are each obtained by a very simple linear transform of some f.d.m. of  $B$ . Now we only need to compute covariances.

- (1)  $\text{Cov}(X_s, X_t) = \text{Cov}(\lambda^{-1/2} B_{\lambda s}, \lambda^{-1/2} B_{\lambda t}) = \lambda^{-1} \text{Cov}(B_{\lambda s}, B_{\lambda t}) = \lambda^{-1} (\lambda s \wedge \lambda t) = s \wedge t.$
- (2) For  $0 \leq s, t \leq 1$ ,  $\text{Cov}(X_s, X_t) = \text{Cov}(B_1 - B_{1-s}, B_1 - B_{1-t}) = \text{Cov}(B_1, B_1) - \text{Cov}(B_1, B_{1-t}) - \text{Cov}(B_{1-s}, B_1) + \text{Cov}(B_{1-s}, B_{1-t}) = 1 - (1-t) - (1-s) + (1-s) \wedge (1-t) = 1 + (t-1) \wedge (s-1) = t \wedge s.$

Now since those processes are continuous on their domain of definition, they are Brownian motions.

**Exercise 2** — *Constructing a Brownian motion indexed by  $\mathbb{R}_+$ .*

Let  $(B^{(n)})_n$  be a sequence of independent Brownian motions defined on  $[0, 1]$ . For every  $t \geq 0$ , define

$$B_t = B_{t - \lfloor t \rfloor}^{(\lfloor t \rfloor)} + \sum_{i=0}^{\lfloor t \rfloor - 1} B_1^{(i)}.$$

Show that  $(B_t)_{t \geq 0}$  is a Brownian motion.

We can check continuity for all  $\omega$  manually. Now a f.d.m.  $B_{t_1}, \dots, B_{t_k}$  is a very simple linear transform of (some f.d.m. of  $B^{(1)}$ , some f.d.m. of  $B^{(2)}$ , ..., some f.d.m. of  $B^{(\lfloor t_k \rfloor)}$ ). Because of the independence assumption, this is a big Gaussian vector. Now we compute covariances. Let  $s \leq t$ .

$$\begin{aligned} \text{Cov}(B_s, B_t) &= \text{Cov} \left( B_{s - \lfloor s \rfloor}^{(\lfloor s \rfloor)} + \sum_{i=0}^{\lfloor s \rfloor - 1} B_1^{(i)}, B_{t - \lfloor t \rfloor}^{(\lfloor t \rfloor)} + \sum_{i=0}^{\lfloor t \rfloor - 1} B_1^{(i)} \right) \\ &= \begin{cases} \sum_{i=0}^{\lfloor s \rfloor - 1} \text{Var}(B_1^{(i)}) + \text{Cov}(B_{s - \lfloor s \rfloor}^{(\lfloor s \rfloor)}, B_{t - \lfloor t \rfloor}^{(\lfloor t \rfloor)}) & \text{if } \lfloor t \rfloor = \lfloor s \rfloor \\ \sum_{i=0}^{\lfloor s \rfloor - 1} \text{Var}(B_1^{(i)}) + \text{Cov}(B_{s - \lfloor s \rfloor}^{(\lfloor s \rfloor)}, B_1^{(\lfloor s \rfloor)}) & \text{if } \lfloor t \rfloor > \lfloor s \rfloor \end{cases} \\ &= s. \end{aligned}$$

This completes the proof.

**Exercise 3** — *Lévy's construction of the Brownian motion.*

Let  $H = L^2([0, 1])$  with the usual inner product. For  $t \geq 0$  let  $I_t = \mathbb{1}_{[0, t]} \in H$ . We also set  $(e_i)_{i \in \mathbb{N}}$  to be an orthonormal basis of  $H$ .

- (1) Check that  $\langle I_s, I_t \rangle = s \wedge t$ .

**Immediate.**

- (2) Assume that there exists a  $H$ -valued standard Gaussian random variable. That is, a random variable  $\xi \in H$ , such that for every  $x \in H$ ,  $\langle x, \xi \rangle \sim \mathcal{N}(0, |x|^2)$ .

- (a) Using the random variable  $\xi$  and the functions  $(I_t)_{t \geq 0}$ , build a Gaussian process  $(B_t)_{t \in [0, 1]}$  such that  $\text{Cov}(B_s, B_t) = s \wedge t$ .

**Setting  $B_t = \langle \xi, I_t \rangle$  yields a Gaussian process with the right covariance kernel. It can be checked by definition of  $\xi$  that every linear combination of coordinates of  $(B_{t_1}, \dots, B_{t_k})$  is Gaussian and that  $\mathbb{E}[(B_t - B_s)^2] = |t - s|$ .**

- (b) Let  $Z_i = \langle \xi, e_i \rangle$ , so that  $\xi = \sum_{i \in \mathbb{N}} Z_i e_i$ . Show that the  $(Z_i)$  are independent standard Gaussians (*Hint*: Compute the characteristic function of finite subvectors.). Deduce that the process of the previous question would satisfy,

$$(\dagger) \quad B_t = \sum_{n=0}^{\infty} Z_n \int_0^t e_i(s) ds.$$

We have,

$$\mathbb{E}[\exp(it_1 Z_{i_1} + \dots + it_p Z_{i_p})] = \mathbb{E}[\exp(i \langle t_1 e_1 + \dots + t_p e_p, \xi \rangle)] = \prod_{i=1}^p e^{-it_p^2/2}.$$

Hence the distribution is that of i.i.d. standard Gaussians.

$$(\dagger) \quad B_t = \langle \xi, I_t \rangle = \sum_{i=0}^{\infty} \langle \xi, e_i \rangle \langle I_t, e_i \rangle = \sum_{n=0}^{\infty} Z_n \int_0^t e_i(s) ds.$$

- (c) By computing  $|\xi|^2$ , show that  $\xi$  cannot exist.

$\|\xi\|^2 = \sum_{i=0}^{\infty} Z_i^2$  which is a.s. not convergent. Indeed,  $p = \mathbb{P}(Z_n \geq 1)$  is independent of  $n$  and  $> 0$  so  $\sum_n \mathbb{P}(Z_n \geq 1)$  diverges and by Borel-Cantelli

$$\mathbb{P}(\forall N \exists n \geq N, Z_n \geq 1) = 1.$$

This last event is contained in the event  $\{\sum_{i=0}^{\infty} Z_i^2 \text{ diverges}\}$ .

- (3) Define  $h_0 = 0$  and for  $n \geq 0$  and  $0 \leq k < 2^n$ ,

$$h_{k,n} := 2^{n/2} \left( \mathbb{1}_{\left[\frac{2k}{2^{n+1}}, \frac{2k+1}{2^{n+1}}\right]} - \mathbb{1}_{\left[\frac{2k+1}{2^{n+1}}, \frac{2k+2}{2^{n+1}}\right]} \right),$$

We admit (or recall) that  $(h_{k,n})_{k,n}$  is an orthonormal basis of  $H$  called the Haar wavelet basis. Let  $(Z_{n,k})_{n,k}$  be a family of independent standard Gaussian random variables. For every  $t \geq 0$  set

$$(\dagger\dagger) \quad B_t = tZ + \sum_{n=0}^{\infty} F_n(t),$$

where  $F_n(t) = \sum_{k=0}^{2^n-1} Z_{n,k} f_{n,k}(t)$  and  $f_{n,k}(t) = \int_0^t h_{n,k}(s) ds$ .

- (a) Using the inequality  $\mathbb{P}(|Y| \geq \lambda) \leq \frac{\sqrt{2/\pi}}{\lambda} e^{-\lambda^2}$  for  $\lambda > 0$  and  $Y \sim \mathcal{N}(0, 1)$ , show that

$$\mathbb{P}\left(2^{-\frac{n+2}{2}} \max_{0 \leq k < 2^n} |Z_{n,k}| > \frac{1}{n^2}\right) = o\left(\frac{1}{n^2}\right).$$

$$\mathbb{P}\left(2^{-\frac{n+2}{2}} \max_{0 \leq k < 2^n} |Z_{n,k}| > \frac{1}{n^2}\right) \leq 2^n \mathbb{P}\left(2^{-\frac{n+2}{2}} |Z_{n,1}| > \frac{1}{n^2}\right) \leq \frac{n^2}{\sqrt{\pi}} 2^{\frac{n-3}{2}} \exp\left(-\frac{2^{n+1}}{n^4}\right) = o\left(\frac{1}{n^2}\right).$$

- (b) Show that  $\mathbb{P}\left(\|F_n\|_\infty \leq \frac{1}{n^2} \text{ for } n \text{ large enough}\right) = 1$ .

We have  $\|F_n\|_\infty \leq 2^{-\frac{n+2}{2}} \max_{0 \leq k < 2^n} |Z_{n,k}|$ , by the previous question  $\sum_n \mathbb{P}\left(\|F_n\|_\infty > \frac{1}{n^2}\right) \leq +\infty$  so by Borel-Cantelli, almost surely for  $n$  large enough  $\|F_n\|_\infty \leq 1/n^2$ .

- (c) Show that almost surely, the sum of functions in (††) converges uniformly on  $[0, 1]$  to a (random) continuous function.

The result follows directly from the previous question by looking at the difference between partial sums and using Cauchy's criterion.

- (4) (★) Prove the same result than in the previous question when we use the Fourier basis  $e_0 = 1$ , and  $e_m(t) = \sqrt{2} \cos(\pi mt)$  in (†) rather than the Haar wavelet basis.

#### Exercise 4 — Time inversion.

Let  $(B_t)_{t \geq 0}$  be a Brownian motion. Set  $X_t = tB_{1/t}$  for  $t > 0$  and  $X_0 = 0$ .

- (1) Show that  $X$  has the finite-dimensional marginals of a Brownian motion.

We first consider the finite-dimensional marginals of the new process  $(X_t)_t$  in this case. Remark at first that they still form centered Gaussian vectors, since they are each obtained by a very simple linear transform of some f.d.m. of  $B$ . Now we only need to compute covariances: If  $0 < s, t$ ,  $\text{Cov}(X_s, X_t) = \text{Cov}(sB_{1/s}, tB_{1/t}) = st(s^{-1} \wedge t^{-1}) = t \wedge s$ . If either  $t = 0$  or  $s = 0$ , then we get  $0 = s \wedge t$  for the covariance too.

- (2) Show that the set  $U = \{f \in \mathbb{R}^{\mathbb{Q}_+}, \lim_{t \rightarrow 0, t \in \mathbb{Q}} f_t = 0\} \subset \mathbb{R}^{\mathbb{Q}_+}$  is measurable.

Observe that,

$$U = \bigcap_{n \geq 1} \bigcup_{m \in \mathbb{N}} \bigcap_{\substack{q \in \mathbb{Q}_+ \\ q \leq 1/m}} \{A : |f_q| < 1/n\},$$

hence  $U$  belongs to the  $\sigma$ -algebra generated by finite-dimensional sets.

- (3) Deduce that  $(X_t)_t$  is continuous almost surely, hence may be modified on a negligible event to form a Brownian motion.

By the  $\pi$ - $\lambda$  (monotone class) theorem, two measures that coincide on a  $\pi$ -system  $\Pi$  (a family of sets stable by finite intersection), coincide on the generated  $\sigma$ -algebra  $\sigma(\Pi)$ . As a result, since  $B$  and  $X$  have the same finite-dimensional marginals, then

$$\mathbb{P}(X|_{\mathbb{Q}_+} \in U) = \mathbb{P}(B|_{\mathbb{Q}_+} \in U) = 1.$$

Hence we have with probability one that:

(a)  $t \mapsto X_t$  is continuous on  $(0, \infty)$ ,

(b)  $X_t \xrightarrow[t \rightarrow 0^+, t \in \mathbb{Q}]{} X_0$

wich together implies continuity on the whole of  $[0, \infty)$ . Now if we change the  $X$  to the constant zero function whenever  $X$  is not continuous, this makes  $X$  continuous for all  $\omega$  without changing the f.d.m.'s. So  $X$  is a Brownian motion.