

Introduction to Stochastic Processes

Assignment 8
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8.1 Exercise

Assuming $B^1(t)$ and $B^2(t)$ are two independent Brownian motions, show that the following are also Brownian motion:

- (1) $X(t) = B^1(A + t) - B^1(A)$ for $A > 0$.
- (2) $X(t) = \alpha B^1(t) + \sqrt{1 - \alpha^2} B^2(t)$ for $0 < \alpha < 1$.

- (1) $X(0) = 0$ and it is also trivially almost surely continuous. And

$$X(t + h) - X(t) = B^1(A + t + h) - B^1(A + t) \sim \mathcal{N}(0, h)$$

And in general $X(t_n) - X(t_{n-1}) = B(t_n) - B(t_{n-1})$ so differences are independent.

- (2) We will show in general that if $\sum_{j=1}^k \alpha_j^2 = 1$ and B^1, \dots, B^k are independent, then $X(t) = \sum_{j=1}^k \alpha_j B^j$ is Brownian motion. $X(0) \stackrel{as}{=} 0$ and $X(t)$ is also almost surely continuous as the sum of almost surely continuous functions.

$$X(t + h) - X(t) = \sum_{j=1}^k \alpha_j (B^j(t + h) - B^j(t)) \sim \sum_{j=1}^k \alpha_j \mathcal{N}(0, h) = \mathcal{N}\left(0, h \sum_{j=1}^k \alpha_j^2\right) = \mathcal{N}(0, h)$$

And since

$$\begin{aligned} \text{Cov}(X(a) - X(b), X(c) - X(d)) &= \sum_{j=1}^k \sum_{i=1}^k \alpha_j \alpha_i \text{Cov}(B^j(a) - B^j(b), B^i(c) - B^i(d)) \\ &= \sum_{j=1}^k \alpha_j^2 \text{Cov}(B^j(a) - B^j(b), B^j(c) - B^j(d)) = 0 \end{aligned}$$

Since for $i \neq j$, B^j and B^i are independent. Thus $X(t_n) - X(t_{n-1}), \dots, X(t_1) - X(t_0)$ are independent (since if coefficients of a Gaussian vector are pairwise uncorrelated, the coefficients are independent).

8.2 Exercise

Set $0 \leq a < b$, show that Brownian motion is almost surely not monotonic on $[a, b]$. Then show that Brownian motion is almost surely not monotonic on any $[a, b]$.

Set a sequence $t_1 < \dots < t_n < \dots$ where $a \leq t_i \leq b$. Then $\mathbb{P}(B(t_i) - B(t_{i-1}) > 0) = \mathbb{P}(\mathcal{N}(0, t_i - t_{i-1}) > 0) = \frac{1}{2}$ and similarly $\mathbb{P}(B(t_i) - B(t_{i-1}) < 0) = \frac{1}{2}$. And we know that $\{B(t_i) - B(t_{i-1})\}$ is independent, and since

$$\sum_{i=1}^{\infty} \mathbb{P}(B(t_i) - B(t_{i-1}) > 0) = \sum_{i=1}^{\infty} \mathbb{P}(B(t_i) - B(t_{i-1}) < 0) = \infty$$

Thus by Borel-Cantelli, $\mathbb{P}(B(t_i) > B(t_{i-1}) \text{ i.o.}) = \mathbb{P}(B(t_i) < B(t_{i-1}) \text{ i.o.}) = 1$. So almost surely, there is a subsequence of t_n on which B is strictly increasing and another on which B is strictly decreasing. So $B(t)$ is almost surely not monotonic on t_n and therefore almost surely not monotonic on $[a, b]$.

Now, $B(t)$ is monotonic on some $[a, b]$ if and only if it is monotonic on some $[p, q]$ for $p, q \in \mathbb{Q}$ by the density of the rationals. Thus

$$\begin{aligned} \mathbb{P}(B(t) \text{ is monotonic on some } [a, b]) &= \mathbb{P}\left(\bigcup_{p < q \in \mathbb{Q}} B(t) \text{ is monotonic on } [p, q]\right) \\ &\leq \sum_{p < q \in \mathbb{Q}} \mathbb{P}(B(t) \text{ is monotonic on } [p, q]) = 0 \end{aligned}$$

The last equality is since $B(t)$ is almost surely not monotonic on any set $[a, b]$. Thus $B(t)$ is almost surely not monotonic on any $[a, b]$.

8.3 Exercise

Suppose f is a continuous function on $[0, 1]$ such that $f(0) = 0$. Let $\varepsilon > 0$, show that

$$\mathbb{P}\left(\sup_{0 \leq t \leq 1} |B(t) - f(t)| < \varepsilon\right) > 0$$

Using Lèvy's construction, we have that $G_n \rightrightarrows B$ in $[0, 1]$ where G_n are functions which are linearly interpolated through points in D_n . Since G_n converges to B uniformly, there exists an N such that for every $n \geq N$, $\|B - G_n\|_\infty < \frac{\varepsilon}{3}$. And since $[0, 1]$ is compact and f is continuous, it is uniformly continuous. So let us define $f_n(d) = f(d)$ for $d \in D_n$ and interpolate f_n linearly through the points in D_n . Since f is uniformly continuous on $[0, 1]$ there exists a $\delta > 0$ such that if $|x - y| < \delta$ then $|f(x) - f(y)| < \frac{\varepsilon}{6}$. There must exist an n such that the difference in points in D_n ($= \frac{1}{2^n}$) is less than δ , and so if $x \in [d_1, d_2]$ for $d_1, d_2 \in D_n$,

$$|f(x) - f_n(x)| \leq |f(x) - f(d_1)| + |f(d_1) - f_n(d_1)| + |f_n(d_1) - f_n(x)| < \frac{\varepsilon}{3}$$

The final inequality is due to $f(d_1) = f_n(d_1)$ and $|f_n(d_1) - f_n(x)| \leq |f_n(d_1) - f_n(d_2)|$ since f_n is linear in $[d_1, d_2]$. So there exists an N such that for every $n \geq N$ and $0 \leq t \leq 1$, $|B(t) - G_n(t)|, |f(t) - f_n(t)| < \frac{\varepsilon}{3}$. Now,

$$|B(t) - f(t)| \leq |B(t) - G_n(t)| + |G_n(t) - f_n(t)| + |f_n(t) - f(t)| < \frac{2\varepsilon}{3} + |G_n(t) - f_n(t)|$$

So we must show there is a non-zero probability that $|G_n(t) - f_n(t)| < \frac{\varepsilon}{3}$. Since f_n and G_n are both linearly interpolated through points on D_n , their maximum distance will be taken on D_n . For every $d \in D_n \setminus D_{n-1}$, by definition

$$G_n(t) = \frac{G_{n-1}(d - 2^{-n}) + G_{n-1}(d + 2^{-n})}{2} + \frac{Z_d}{\sqrt{2^{n+1}}}$$

And so

$$\mathbb{P}\left(|G_n(t) - f_n(t)| < \frac{\varepsilon}{3}\right) = \mathbb{P}\left((\forall d \in D_n) |G_n(d) - f_n(d)| < \frac{\varepsilon}{3}\right)$$

Now since the values of $G_n(d)$ are determined by a finite number of independent normal distributions (which have full range), this probability must be nonzero.