

Assignment 2

1. Let B_t be a standard Brownian Motion. In class we showed $dB_t dB_t = dt$. Using similar notation, show that $dB_t dt = 0$.

→

As

$$a) \quad dB_t dt = \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^{n-1} (B_{t_i} - B_{t_{i-1}})(t_i - t_{i-1})$$

$$E[dB_t dt] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^{n-1} (t_i - t_{i-1}) E[B_{t_i} - B_{t_{i-1}}]$$

$$\text{but } E[B_{t_i} - B_{t_{i-1}}] = 0 \quad \forall i = 0 \dots n$$

$$\therefore E[dB_t dt] = 0 \quad \text{--- (1)}$$

$$b) \quad \text{Var}[dB_t dt] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^{n-1} \text{Var}[(B_{t_i} - B_{t_{i-1}})(t_i - t_{i-1})]$$

$$\text{as } B_{t_i} - B_{t_{i-1}} \quad \forall i = 0 \dots n \text{ are i.i.d.}$$

$$\begin{aligned} \text{Var}[(B_{t_i} - B_{t_{i-1}})(t_i - t_{i-1})] &= (t_i - t_{i-1})^2 \text{Var}[B_{t_i} - B_{t_{i-1}}] \\ &= (t_i - t_{i-1})^2 \cdot (t_i - t_{i-1}) \\ &= (t_i - t_{i-1})^3 \end{aligned}$$

$$\begin{aligned} \therefore \text{Var}[dB_t dt] &= \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^{n-1} (t_i - t_{i-1})^3 \\ &= \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^{n-1} (t_i - t_{i-1})^2 \cdot (t_i - t_{i-1}) \end{aligned}$$

$$\begin{aligned} \therefore \text{Var}[dB_t dt] &\leq \lim_{\|\pi\| \rightarrow 0} \sum_{i=0}^n \|\pi\|^2 (t_i - t_{i-1}) \\ &\leq \lim_{\|\pi\| \rightarrow 0} \|\pi\| \cdot T \end{aligned}$$

$$\text{Var}[dB_t dt] \leq 0$$

but Variance is always ≥ 0

$$\therefore \text{Var}[dB_t dt] = 0 \quad - (2)$$

From (1) & (2)

$$E[dB_t dt] = 0 \quad \& \quad \text{Var}[dB_t dt] = 0$$

\therefore by defⁿ of L^2 convergence

$$\underline{\underline{dB_t dt = 0}}$$

2. Show that the brownian motion B_t has unbounded first variation, i.e.,

$$\lim_{\|\Pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| = \infty,$$

where Π is the partition as defined in class.

→ let $x \sim N(0, \sigma^2)$

$$E[x | x > 0] = \int_0^{\infty} x \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}x^2} dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_0^{\infty} 2x \cdot e^{-\frac{1}{2\sigma^2}x^2} dx$$

let $x^2 = u \quad \therefore 2x dx = du$

As $x \rightarrow 0, u \rightarrow 0, x \rightarrow \infty, u \rightarrow \infty$

$$\therefore E[x | x > 0] = \frac{1}{\sqrt{2\pi}\sigma} \int_0^{\infty} e^{-\frac{1}{2\sigma^2}u} du$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \left[\frac{e^{-\frac{1}{2\sigma^2}u}}{-\frac{1}{2\sigma^2}} \right]_0^{\infty}$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \left[2\sigma^2 \right] = \sqrt{\frac{2}{\pi}} \cdot \sigma$$

by symmetry

$$\therefore E[x | x < 0] = -\sqrt{\frac{2}{\pi}} \sigma$$

$$\therefore E[-x | x < 0] = \sqrt{\frac{2}{\pi}} \sigma$$

$$\begin{aligned}
 \therefore E[|X|] &= E[E[|X| | X]] \\
 &= \frac{1}{2} E[|X| | X > 0] + \frac{1}{2} E[|X| | X < 0] \\
 &= \frac{1}{2} E[X | X > 0] + \frac{1}{2} E[-X | X < 0] \\
 &= \frac{1}{2} \sqrt{\frac{2}{\pi}} \sigma + \frac{1}{2} \sqrt{\frac{2}{\pi}} \sigma = \sqrt{\frac{2}{\pi}} \sigma
 \end{aligned}$$

$$\therefore E[|B_{t_{j+1}} - B_{t_j}|] = \sqrt{\frac{2}{\pi}} (t_{j+1} - t_j)$$

$$\begin{aligned}
 \therefore E \left[\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| \right] &= \lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} E[|B_{t_{j+1}} - B_{t_j}|] \\
 &= \lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} \sqrt{\frac{2}{\pi}} (t_{j+1} - t_j) \\
 &= \lim_{\|\pi\| \rightarrow 0} \sqrt{\frac{2}{\pi}} \sum_{j=0}^{n-1} t_{j+1} - t_j \\
 &= \sqrt{\frac{2}{\pi}} \cdot T \quad [\because \text{telescopic sum}] \quad \text{--- (1)}
 \end{aligned}$$

$$\text{Var} \left[\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| \right] = \lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} \text{Var}[B_{t_{j+1}} - B_{t_j}]$$

$$\begin{aligned}
 (1) - (2) \quad \text{Var}[B_{t_{j+1}} - B_{t_j}] &= E[(B_{t_{j+1}} - B_{t_j})^2] - (E[|B_{t_{j+1}} - B_{t_j}|])^2 \\
 &= (t_{j+1} - t_j) - \frac{2}{\pi} (t_{j+1} - t_j)^2
 \end{aligned}$$

$$\begin{aligned}
\therefore \text{var} \left[\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| \right] &= \lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} (t_{j+1} - t_j) - \frac{2}{\pi} (t_{j+1} - t_j)^2 \\
&= \lim_{\|\pi\| \rightarrow 0} T - \frac{2}{\pi} \sum_{j=0}^{n-1} (t_{j+1} - t_j)^2 \\
&\geq \lim_{\|\pi\| \rightarrow 0} T - \frac{2}{\pi} \|\pi\| \cdot \sum_{j=0}^{n-1} (t_{j+1} - t_j) \\
&\geq \lim_{\|\pi\| \rightarrow 0} T - \frac{2}{\pi} \|\pi\| \cdot T \\
&\geq \lim_{\|\pi\| \rightarrow 0} T \left(1 - \frac{2}{\pi} \|\pi\| \right) \\
&\geq T \quad \text{--- (2)}
\end{aligned}$$

\therefore From (1) & (2), For any time period $[0, T]$

$$E \left[\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| \right] = \sqrt{\frac{2}{\pi}} \cdot T$$

$$\text{var} \left[\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}| \right] \geq T$$

This implies $\lim_{\|\pi\| \rightarrow 0} \sum_{j=0}^{n-1} |B_{t_{j+1}} - B_{t_j}|$ can potentially
~~take~~ blow up i.e. Brownian motion has unbounded first variation

3. Show that $(dB_t)^n = 0$ for $n \geq 3$

$$\rightarrow (dB_t)^p = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^m (B_{t_i} - B_{t_{i-1}})^p \quad p=3,4,\dots$$

a)

$$\text{As } B_{t_i} - B_{t_{i-1}} \sim N(0, t_i - t_{i-1})$$

p^{th} moment of a normal distribution is

$$\hat{m}_p = \begin{cases} \sigma^p (p-1)! & \text{if } p \text{ is even} \\ 0 & \text{if } p \text{ is odd} \end{cases}$$

$$E[(dB_t)^p] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^m E(B_{t_i} - B_{t_{i-1}})^p$$

if p is even & $p \geq 3$

$$E[(dB_t)^p] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^m \sigma_i^p (p-1)!$$

$$\text{where } \sigma_i = t_i - t_{i-1}$$

$$E[(dB_t)^p] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^m (t_i - t_{i-1})^p (p-1)! \quad (1)$$

$$\leq \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^m \|\pi\|^p (p-1)!$$

but from (1) $E[(dB_t)^p]$ is always +ve $\therefore E[(dB_t)^p] = 0$

if p is odd & $p \geq 3$

$$E[(dB_t)^p] = \lim_{\| \Pi \| \rightarrow 0} \sum_{i=1}^m E[(B_{t_i} - B_{t_{i-1}})^p]$$

$$= \lim_{\| \Pi \| \rightarrow 0} \sum_{i=1}^m 0 = 0$$

$$E[(dB_t)^p] = 0 \quad \forall p \geq 3 \quad \text{--- (2)}$$

$$b) \text{Var}[(dB_t)^p] = \lim_{\| \Pi \| \rightarrow 0} \sum_{i=1}^m \text{Var}[(B_{t_i} - B_{t_{i-1}})^p]$$

$$\text{Var}[(B_{t_i} - B_{t_{i-1}})^p] = E[(B_{t_i} - B_{t_{i-1}})^{2p}] - \left(E[(B_{t_i} - B_{t_{i-1}})^p] \right)^2$$

but

$$E[(B_{t_i} - B_{t_{i-1}})^p] = 0 \quad \forall p \geq 3$$

Similarly

$$\& E[(B_{t_i} - B_{t_{i-1}})^{2p}] = 0$$

$$\therefore \text{Var}[(B_{t_i} - B_{t_{i-1}})^p] = 0$$

$$\therefore \text{Var}[(dB_t)^p] = \lim_{\| \Pi \| \rightarrow 0} \sum_{i=1}^m \text{Var}[(B_{t_i} - B_{t_{i-1}})^p]$$

$$= \sum_{i=1}^m 0 = 0 \quad \text{--- (3)}$$

From (2) & (3), $(dB_t)^p = 0$ for $p \geq 3$

replacing p by n $(dB_t)^n = 0$ for $n \geq 3$