

Theorem: If $\Delta_n := \{a = t_0 < t_1 < \dots < t_{n-1} < t_n = b\}$ is a partition of $[a, b]$, then

$$\sum_{i=1}^n |B_{t_i} - B_{t_{i-1}}|^2 \rightarrow b - a \quad \text{in } L^2$$

as

$$\|\Delta_n\| = \max_{1 \leq i \leq n} (t_i - t_{i-1}) \rightarrow 0.$$

That is, $Q_2(B; a, b)$, the quadratic variation of Brownian motion on the interval $[a, b]$, exists and equals $b - a$.

Proof: To begin, notice that

$$\sum_{i=1}^n (t_i - t_{i-1}) = b - a.$$

Let

$$Y_n = \sum_{i=1}^n |B_{t_i} - B_{t_{i-1}}|^2 - (b - a) = \sum_{i=1}^n \left[|B_{t_i} - B_{t_{i-1}}|^2 - (t_i - t_{i-1}) \right] = \sum_{i=1}^n X_i$$

where

$$X_i = |B_{t_i} - B_{t_{i-1}}|^2 - (t_i - t_{i-1}),$$

and note that

$$Y_n^2 = \sum_{i=1}^n \sum_{j=1}^n X_i X_j = \sum_{i=1}^n X_i^2 + 2 \sum_{i < j} X_i X_j.$$

The independence of the Brownian increments implies that $\mathbb{E}(X_i X_j) = 0$ for $i \neq j$; hence,

$$\mathbb{E}(Y_n^2) = \sum_{i=1}^n \mathbb{E}(X_i^2).$$

But

$$\begin{aligned} \mathbb{E}(X_i^2) &= \mathbb{E} \left((B_{t_i} - B_{t_{i-1}})^4 - 2(t_i - t_{i-1}) \mathbb{E} \left((B_{t_i} - B_{t_{i-1}})^2 \right) + (t_i - t_{i-1})^2 \right) \\ &= 3(t_i - t_{i-1})^2 - 2(t_i - t_{i-1})^2 + (t_i - t_{i-1})^2 \\ &= 2(t_i - t_{i-1})^2 \end{aligned}$$

since the fourth moment of a normal random variable with mean 0 and variance $t_i - t_{i-1}$ is $3(t_i - t_{i-1})^2$. Therefore,

$$\mathbb{E}(Y_n^2) = \sum_{i=1}^n \mathbb{E}(X_i^2) = 2 \sum_{i=1}^n (t_i - t_{i-1})^2 \leq 2 \|\Delta_n\| \sum_{i=1}^n (t_i - t_{i-1}) = 2(b - a) \|\Delta_n\| \rightarrow 0$$

as $\|\Delta_n\| \rightarrow 0$ from which we conclude that $Y_n \rightarrow 0$ in L^2 ; that is, $Q_2(B; a, b) = b - a$. \square