

MATH4210: Financial Mathematics Tutorial 7

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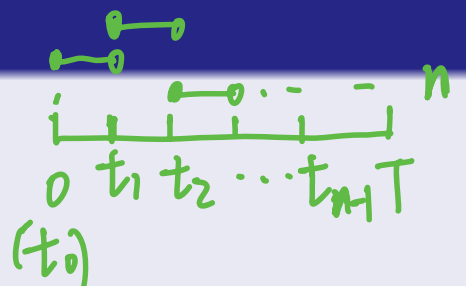
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Content Review

Question

- Let θ_t be a simple process.


$$\int_0^T \theta_t dB_t = \sum_{i=0}^{n-1} \alpha_i (B_{t_{i+1}} - B_{t_i}) = \alpha_0 (B_{t_1} - 0) + \cdots + \alpha_{n-1} (B_{t_n} - B_{t_{n-1}}).$$

- For each $\theta_t \in \mathbb{H}^2([0, T])$, there exists a sequence of simple processes $(\theta_t^n)_{n \geq 1}$ such that

$$\lim_{n \rightarrow \infty} \mathbb{E} \left[\int_0^T (\theta_t - \theta_t^n)^2 dt \right] = 0.$$

Question

- If $(\theta_t^n)_{n \geq 1}$ is a sequence of simple processes such that

$$\lim_{n \rightarrow \infty} \mathbb{E} \left[\int_0^T (\theta_t - \theta_t^n)^2 dt \right] = 0,$$

then the sequence of stochastic integrals $\int_0^T \theta_t^n dB_t$ has a unique limit in $\mathbb{L}^2(\Omega)$ as $n \rightarrow \infty$. And we define

$$\int_0^T \theta_t dB_t := \lim_{n \rightarrow \infty} \int_0^T \theta_t^n dB_t$$

as this limitation.

Question

- Let $\theta_t \in \mathbb{H}^2([0, T])$, then

$$\mathbb{E}\left[\int_0^T \theta_t dB_t\right] = 0, \quad \mathbb{E}\left[\left(\int_0^T \theta_t dB_t\right)^2\right] = \mathbb{E}\left[\int_0^T \theta_t^2 dt\right].$$

- Let H_t be a adapted process and $\Delta B_{k+1}^n := B_{t_{k+1}^n} - B_{t_k^n}$, then

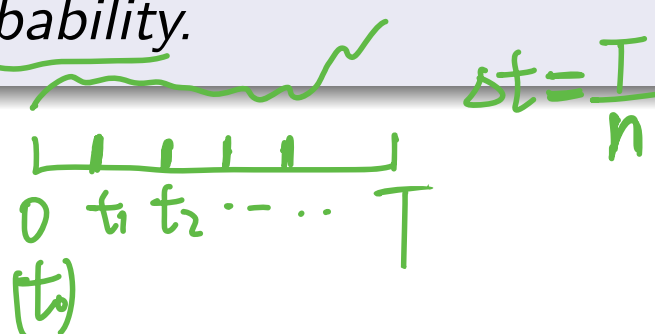
$$\sum_{k=0}^{n-1} H_{t_k^n} ((\Delta B_{k+1}^n)^2 - \Delta t) \rightarrow 0, \text{ in probability.}$$

$$\sum_{k=0}^{n-1} ((\Delta B_{k+1}^n)^2 - \Delta t) \rightarrow 0$$

$$\sum_{k=0}^{n-1} (\Delta B_{k+1}^n)^2 - \sum_{k=0}^{n-1} \Delta t \rightarrow 0$$

$$\frac{1}{n} \cdot n$$

$$\Rightarrow \sum_{k=0}^{n-1} (\Delta B_{k+1}^n)^2 \rightarrow T \text{ in probability}$$



Stochastic Integration

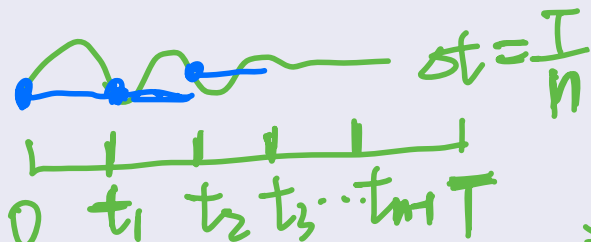
$$\xRightarrow{(n_j)} \sum_{k=0}^{n_j-1} (\Delta B_{k+1}^{n_j})^2 \rightarrow T, \text{ a.s.}$$

$X_n \rightarrow X$ in prop.

\Rightarrow subsequence $X_{n_j} \rightarrow X$ a.s.

Question

For fixed $T > 0$, prove that



$$\int_0^T B_t dB_t = \frac{1}{2} B_T^2 - \frac{1}{2} T$$

from sketch. $(B_t)_{t \geq 0}$ is a standard Brownian motion.

$$\theta_t^n := B_{t_k}, t \in [t_k, t_{k+1})$$

$$\underline{B_t = \theta_t}$$

Stochastic Integration

$$\textcircled{1} \lim_{n \rightarrow \infty} E \left[\int_0^T (B_t - \theta_t^n)^2 dt \right] = 0$$

$$E \left[\int_0^T (B_t - \theta_t^n)^2 dt \right]$$

$$= E \left[\sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} (B_t - \theta_t^n)^2 dt \right]$$

$$= E \left[\sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} (B_t - B_{t_k})^2 dt \right]$$

$$= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} E(B_t - B_{t_k})^2 dt$$

$\sim N(0, t - t_k)$

$$(\text{Var}(X) = E(X^2) - [E(X)]^2)$$

$\leftarrow \frac{1}{n} = t - t_k$

$$= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} (t - t_k) dt$$

$$\leftarrow \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} \frac{1}{n} dt = \left(\sum_{k=0}^{n-1} \frac{1}{n} (t_{k+1} - t_k) \right) \frac{1}{n} \rightarrow \frac{1}{2} T + \frac{1}{2} B_T^2$$

$\Delta B_{t_k}^2 \left(\sum_{k=0}^{n-1} \Delta B_{t_k}^2 \rightarrow T, a.s \right)$

$$\textcircled{2} \int_0^T B_t dB_t = \lim_{n \rightarrow \infty} \int_0^T \theta_t^n dB_t$$

$$\int_0^T \theta_t^n dB_t \rightarrow \frac{1}{2} T + \frac{1}{2} B_T^2$$

$$= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} \theta_t^n dB_t$$

$$= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} B_{t_k} dB_t$$

$$= \sum_{k=0}^{n-1} B_{t_k} (B_{t_{k+1}} - B_{t_k})$$

$$= \sum_{k=0}^{n-1} \frac{1}{2} (-2 B_{t_k} B_{t_{k+1}} + \underbrace{B_{t_k}^2})$$

$$= \sum_{k=0}^{n-1} \frac{1}{2} [(B_{t_k} - B_{t_{k+1}})^2 - B_{t_{k+1}}^2 + B_{t_k}^2]$$

$$= \sum_{k=0}^{n-1} \frac{1}{2} [(B_{t_k} - B_{t_{k+1}})^2] - \sum_{k=0}^{n-1} \frac{1}{2} [B_{t_{k+1}}^2 - B_{t_k}^2]$$

$$\Delta B_{t_k}^2 \left(\sum_{k=0}^{n-1} \Delta B_{t_k}^2 \rightarrow T, a.s \right)$$

$$\sum_{k=0}^{n-1} [B_{t_{k+1}}^2 - B_{t_k}^2]$$

Ito Formula

$$= \frac{T^2}{n^2} \cdot n = \frac{T^2}{n} \rightarrow 0$$

(subsequence, a.s)

$$\begin{aligned} & B_{t_1}^2 - B_{t_0}^2 + B_{t_2}^2 - B_{t_1}^2 \\ & \dots \dots B_{t_n}^2 - B_{t_{n-1}}^2 \\ & = B_{t_n}^2 - B_{t_0}^2 \\ & = B_T^2 - 0 \end{aligned}$$

$$Y_t = f(B_t)$$

$$dB_t \sim O(t^{\frac{1}{2}}) \quad dt \sim O(t)$$

$$\begin{aligned} dY_t &= f'(B_t) dB_t + \frac{f''(B_t)}{2} (dB_t)^2 \\ &= f'(B_t) dB_t + \frac{f''(B_t)}{2} dt \end{aligned}$$

Question

Consider a standard Brownian motion $(B_t)_{t \geq 0}$. Let $T > 0$, compute

- (a) $\int_0^T B_t dB_t$ $\int_0^T d(B_t^2)$
 (b) $\int_0^T \exp(B_t - \frac{1}{2}t) dt$

~~using Ito formula.~~

(a) $f(x) := x^2 \quad f' = 2x \quad f'' = 2$

$$df(B_t) = f'(B_t) dB_t + \frac{f''(B_t)}{2} dt$$

$$dB_t^2 = 2B_t dB_t + dt$$

$$\Rightarrow B_T^2 = 2 \int_0^T B_t dB_t + T$$

$$\Rightarrow \int_0^T B_t dB_t = \frac{B_T^2 - T}{2}$$

$$\xrightarrow{0 \rightarrow T} \int_0^T B_t^2 - B_0^2 = 2 \cdot \int_0^T B_t dB_t + T - 0$$

Stochastic Integration

$$f(t, B_t) := Y_t$$

$$dY_t = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dB_t + \frac{\partial^2 f}{\partial^2 x} \cdot \frac{1}{2} (dB_t)^2 \approx dt$$

$$\boxed{\cancel{dB_t dt} \approx dt^{1.5}}$$

$$= \left(\frac{\partial f}{\partial t} + \frac{\partial^2 f}{\partial^2 x} \cdot \frac{1}{2} \right) dt + \frac{\partial f}{\partial x} dB_t$$

$$f(t, B_t) = e^{x - \frac{1}{2}t} \quad \frac{\partial f}{\partial t} = e^{x - \frac{1}{2}t} \cdot \left(-\frac{1}{2}\right) \cdot \frac{\partial f}{\partial x} = e^{x - \frac{1}{2}t} \cdot \frac{\partial^2 f}{\partial x^2} = e^{x - \frac{1}{2}t}$$

$$df(t, B_t) = \left(\left(-\frac{1}{2}\right) e^{B_t - \frac{1}{2}t} + e^{B_t - \frac{1}{2}t} \cdot \frac{1}{2} \right) dt + e^{B_t - \frac{1}{2}t} dB_t = 0$$

$$\int_0^T \Rightarrow f(T, B_T) - f(0, B_0) = \int_0^T e^{B_t - \frac{1}{2}t} dB_t$$

$$e^{B_T - \frac{1}{2}T} - e^0 = \int_0^T e^{B_t - \frac{1}{2}t} dB_t$$

$$e^{B_T - \frac{1}{2}T} - 1 = \int_0^T e^{B_t - \frac{1}{2}t} dB_t$$