

MATH4210: Financial Mathematics Tutorial 7

Xiangying Pang

The Chinese University of Hong Kong

xytang@math.cuhk.edu.hk

12 March, 2024

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

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

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

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

as this limitation.

Content Review

Question

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

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

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

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

$$\Delta B_1^n = B_{t_1^n} - B_{t_0^n}$$

$$\Delta t = \frac{T}{n}$$



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

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

Stochastic Integration

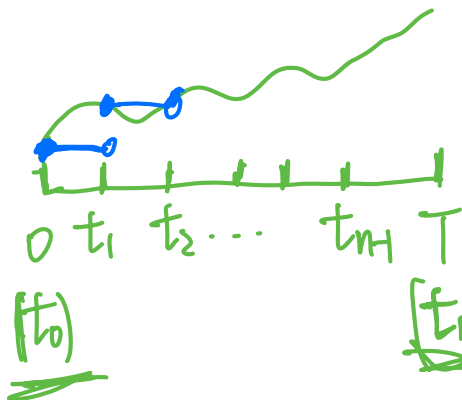
$$\begin{aligned} & \Rightarrow \sum_{k=0}^{n_j-1} (\Delta B_{k+1}^{n_j})^2 - T \rightarrow 0, \text{ a.s.} \\ & \left[(X_n)_{n \geq 0}, \mathbb{R}^d \right] X_n \rightarrow X, \text{ in prob} \\ & \Rightarrow X_{n_j} \rightarrow X, \text{ a.s.} \end{aligned}$$

Question

For fixed $T > 0$, prove that

$$\begin{aligned} & \textcircled{1} \text{ find a seq of simple processes } \theta_t^n \\ & \textcircled{2} \lim E \left[\int_0^T (\theta_t^n - B_t)^2 dt \right] = 0 \\ & \int_0^T B_t dB_t = \frac{1}{2} B_T^2 - \frac{1}{2} T \quad \textcircled{3} \int_0^T B_t dB_t = \lim \int_0^T \theta_t^n dB_t \end{aligned}$$

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



$$\Delta t = \frac{T}{n} \quad \textcircled{1} \quad \theta_t^n = B_{t_k}, t \in [t_k, t_{k+1})$$

Stochastic Integration

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

$$\begin{aligned} & E \left[\int_0^T (B_t - \theta_t^n)^2 dt \right] \\ &= E \left[\sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} (B_t - \underbrace{\theta_t^n}_{B_{t_k}})^2 dt \right] \\ &= E \left[\sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} \underbrace{(B_t - B_{t_k})^2}_{\sim N(0, t-t_k)} dt \right] \\ &= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} E(B_t - B_{t_k})^2 dt \\ &\quad (\text{Var}(X) = \underbrace{E(X^2)}_{\substack{\parallel \\ t-t_k}} - \underbrace{[E(X)]^2}_0) \end{aligned}$$

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

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

$$\textcircled{3} \int_0^T B_t dB_t = \lim_{n \rightarrow \infty} \int_0^T \theta_t^n dB_t$$

$$\begin{aligned} \int_0^T \theta_t^n dB_t &= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} \underbrace{\theta_t^n}_{B_{t_k}} dB_t \\ &= \sum_{k=0}^{n-1} \int_{t_k}^{t_{k+1}} \underbrace{B_{t_k}}_{\text{constant}} dB_t \\ &= \sum_{k=0}^{n-1} B_{t_k} (B_{t_{k+1}} - B_{t_k}) \\ &= \sum_{k=0}^{n-1} \underbrace{B_{t_k} B_{t_{k+1}} - B_{t_k}^2}_{\text{telescoping}} \\ &= \sum_{k=0}^{n-1} \frac{1}{2} (-2 B_{t_k} B_{t_{k+1}} + 2 B_{t_k}^2) \\ &= \sum_{k=0}^{n-1} \frac{1}{2} [(B_{t_k} - B_{t_{k+1}})^2 - B_{t_{k+1}}^2 + B_{t_k}^2] \\ &= \sum_{k=0}^{n-1} \frac{1}{2} \underbrace{(B_{t_k} - B_{t_{k+1}})^2}_{\Delta B_{t_k}^2} - \sum_{k=0}^{n-1} \frac{1}{2} \underbrace{(B_{t_{k+1}}^2 - B_{t_k}^2)}_{\text{telescoping}} \\ &= \sum_{k=0}^{n-1} \frac{1}{2} \Delta B_{t_k}^2 + \frac{1}{2} B_T^2 \end{aligned}$$

$$\left(\sum_{k=0}^{n-1} \Delta B_{t_k}^2 \rightarrow T \text{ as } n \rightarrow \infty \right)$$

Ito Formula

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

$$\xrightarrow{n \rightarrow \infty} -\frac{1}{2}T + \frac{1}{2}B_T^2$$

$$\begin{matrix} \parallel \\ B_{T_n}^2 - B_{T_0}^2 \\ \parallel \\ B_T^2 \\ \parallel \\ 0 \end{matrix}$$

$$(a) Y_t = f(B_t)$$

$$dY_t = f'(B_t)dB_t + \frac{f''(B_t)}{2} \underbrace{(dB_t)^2}_{\frac{dt}{dt}}$$

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

Question

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

$$(a) \int_0^T B_t dB_t \sim \int_0^T d(\frac{B_t^2}{2})$$

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

using Ito formula.

$$f(x) = x^2, f'(x) = 2x, f''(x) = 2$$

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

$$df(B_t) = f'(B_t)dB_t + \frac{f''(B_t)}{2}dt \Rightarrow \int_0^T B_t dB_t = \frac{1}{2}B_T^2 - \frac{1}{2}T$$

$$\Rightarrow dB_t^2 = 2 \cdot B_t \cdot dB_t + dt$$

Stochastic Integration

$$\frac{dB_t \cdot dt}{dt^{\frac{1}{2}}} = \underbrace{O(t^{\frac{1}{2}})}_{O(t')}$$

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

$$dY_t = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dB_t + \cancel{\frac{\partial^2 f}{\partial x \partial t} dB_t dt} + \frac{\partial^2 f}{\partial x^2} dt \cdot \underbrace{\frac{1}{2} (dB_t)^2}_{\sim O(t') \downarrow dt}$$

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

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

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

$$\begin{aligned} \int_0^T \Rightarrow f(T, B_T) - f(0, B_0) &= \int_0^T e^{B_t - \frac{1}{2}t} dB_t \\ e^{B_T - \frac{1}{2}T} - \underbrace{e^0}_1 &= \int_0^T e^{B_t - \frac{1}{2}t} dB_t \Rightarrow \int_0^T e^{B_t - \frac{1}{2}t} dB_t = e^{B_T - \frac{1}{2}T} - 1 \end{aligned}$$