

### Assignment 3 ITô Integral

1. Let  $\sigma(t)$  be deterministic function of time,  $\beta$  be constant and define

$$X(T) = \int_0^T \sigma(t) e^{-\beta(T-t)} dB_t.$$

Find the expectation and Variance of  $X(T)$ .  
What is the distribution of  $X(T)$ ?

→ By definition of itô integral

$$a) \quad X(T) = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^n \underbrace{\sigma(t_i) e^{-\beta(T-t_i)}}_{\text{deterministic function}} (B_{t_{i+1}} - B_{t_i})$$

$$E[X(T)] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^n \sigma(t_i) e^{-\beta(T-t_i)} E[B_{t_{i+1}} - B_{t_i}]$$

$$\text{As } E[B_{t_{i+1}} - B_{t_i}] = 0 \quad \left[ \begin{array}{l} \text{Expected value of increment} \\ \text{of brownian motion} \end{array} \right]$$

$$\therefore E[X(T)] = \lim_{\|\pi\| \rightarrow 0} \sum_{i=1}^n \sigma(t_i) e^{-\beta(T-t_i)} \cdot 0$$

$$E[X(T)] = 0$$

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

$$E[(X(T))^2] = E\left[\left(\int_0^T \sigma(t) e^{-\beta(T-t)} dB_t\right)^2\right]$$

by Itô's isometry,

$$= E\left[\int_0^T \left(\sigma(t) e^{-\beta(T-t)}\right)^2 dt\right]$$

$$E[(X(T))^2] = E\left[\int_0^T \sigma^2(t) e^{-2\beta(T-t)} dt\right]$$

$$\therefore \text{Var}(X(T)) = E\left[\int_0^T \sigma^2(t) e^{-2\beta(T-t)} dt\right]$$

$$\text{Var}(X(T)) = \int_0^T \sigma^2(t) e^{-2\beta(T-t)} dt \quad [\text{As everything is deterministic}]$$

c) As

$$X(T) = \lim_{\|n\| \rightarrow 0} \sum_{i=1}^n \sigma(t_i) e^{-\beta(T-t_i)} (B_{t_{i+1}} - B_{t_i}) \quad (1)$$

$\Delta B_{t_{i+1}} - B_{t_i}$  is normally distributed.

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

as they are increments of Brownian motion independent

Hence the RHS of (1) is sum of scaled independent increments of brownian motion,

with scaling factor =  $\sigma(t_i) e^{-\beta(T-t_i)}$   
i.e. sum of scaled i.i.d. normally distributed random variables.

Hence  $X(T)$  also follows normal distribution with mean 0 & Variance

$$E \left[ \int_0^T \sigma(t)^2 e^{-2\beta(T-t)} dt \right]$$

2. For given functions  $\mu(t, \omega)$  and  $\sigma(t, \omega)$  Itô process  $X_t$  is defined as

$$X_t = X_0 + \int_0^t \mu(s, \omega) ds + \int_0^t \sigma(s, \omega) dB_s(\omega)$$

What is the quadratic variation of process  $X_t$  on a fixed interval  $[0, T]$ ?

In your derivation you can assume that functions  $\mu$  and  $\sigma$  are simple functions, that is piecewise constants.

$t$



As  $\mu$  &  $\sigma$  are simple functions

$$\int_0^t \mu(s, \omega) ds = \sum_i \mu_i (s_{i+1} - s_i)$$

$$\int_0^t \sigma(s, \omega) dB_s = \sum_j \sigma_j (B_{j+1} - B_j)$$

$$\therefore \langle X, X \rangle = \lim_{\|\pi\| \rightarrow 0} \sum_k \left[ X_0 + \sum_{i=0}^k \mu_i (s_{i+1} - s_i) + \sum_{j=0}^k \sigma_j (B_{j+1} - B_j) - X_0 - \sum_{i=0}^{k-1} \mu_i (s_{i+1} - s_i) - \sum_{j=0}^{k-1} \sigma_j (B_{j+1} - B_j) \right]^2$$

$$= \lim_{\|\pi\| \rightarrow 0} \sum_k \left( \mu_k (s_{k+1} - s_k) + \sigma_k (B_{k+1} - B_k) \right)^2$$

$$= \lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k^2 (s_{k+1} - s_k)^2 + 2 \sum_k \mu_k \sigma_k (s_{k+1} - s_k) (B_{k+1} - B_k)$$

$$+ \sum_k \sigma_k^2 (B_{k+1} - B_k)^2 - (0)$$

a) As  $\pi = \max_k (s_{k+1} - s_k)$

$$\lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k^2 (s_{k+1} - s_k)^2 \leq \lim_{\|\pi\| \rightarrow 0} \pi \sum_k \mu_k^2 (s_{k+1} - s_k)$$

$$\leq \lim_{\|\pi\| \rightarrow 0} \pi \int_0^T \mu^2 ds \leq 0$$

$$= 0 \quad - (1)$$

As sum of +ves less than equal to 0 should be zero.

$$b) \lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k \sigma_k (B_{k+1} - B_k) (s_{k+1} - s_k)$$

$$= \lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k \sigma_k dB ds$$

but by rules of stochastic calculus

$$dB ds = dB_t dt = 0$$

$$\therefore \lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k \sigma_k (B_{k+1} - B_k) (s_{k+1} - s_k)$$

$$= \lim_{\|\pi\| \rightarrow 0} \sum_k \mu_k \sigma_k \cdot 0 = 0 \quad - (2)$$

$$c) \lim_{\|\pi\| \rightarrow 0} \sum_k \sigma_k^2 (B_{k+1} - B_k)^2$$

$$\text{As } B_{k+1} - B_k = dB$$

$$(B_{k+1} - B_k)^2 = dB \cdot dB$$

$$\text{but } dB \cdot dB = dt = t_{k+1} - t_k \quad (\text{by rule of stochastic calculus})$$

$$\therefore \lim_{\|\pi\| \rightarrow 0} \sum_k \sigma_k^2 (B_{k+1} - B_k)^2$$

$$= \lim_{\|\pi\| \rightarrow 0} \sum_k \sigma_k^2 (t_{k+1} - t_k)$$

by def<sup>n</sup> of reiman integral

$$= \int_0^T \sigma^2 dt = \int_0^T \sigma^2(s, \omega) ds \quad - (3)$$

Substituting (1), (2), (3) in (0)

$$\langle X, X \rangle = 0 + 2 \cdot 0 + \int_0^T \sigma^2(s, \omega) ds$$

over  $[0, T]$

$\therefore \langle X, X \rangle =$  Quadratic variation of  $X_t$  over fixed interval  $[0, T]$

$$= \int_0^T \sigma^2(s, \omega) ds$$



3. Compute  $E e^{-\int_0^t \sqrt{s} dB_s}$

$$\text{let } x = -\int_0^t \sqrt{s} dB_s$$

$$E[x] = -E \int_0^t \sqrt{s} dB_s$$

$$= -E \left[ \lim_{\| \pi \| \rightarrow 0} \sum_k \sqrt{s_k} (B_{s_{k+1}} - B_{s_k}) \right]$$

$$= - \lim_{\| \pi \| \rightarrow 0} \sum_k \sqrt{s_k} \underbrace{E[B_{s_{k+1}} - B_{s_k}]}_0$$

$$E[x] = -1 \times 0 = 0$$

$$\text{Var}[x] = \text{Var} \left[ -\int_0^t \sqrt{s} dB_s \right]$$

$$= \text{Var} \left[ \int_0^t \sqrt{s} dB_s \right]$$

$$= E \left[ \left( \int_0^t \sqrt{s} dB_s \right)^2 \right]$$

by Itô's isometry

$$\text{Var}[x] = E \left[ \int_0^t s ds \right]$$

$$\text{Var}[x] = E \left[ \frac{t^2}{2} - \frac{0}{2} \right] = \frac{t^2}{2}$$



As

$$X = - \int_0^t \sqrt{s} dB_s$$

$X$  follows a normal distribution by the same argument as in question (1).

$\therefore e^X$  follows lognormal distribution

$$E e^X = E e^{- \int_0^t \sqrt{s} dB_s} = \text{mean of lognormal distribution}$$

As  $X \sim N(\mu, \sigma^2)$ ,  $Y = e^X$

$$Y \sim \text{lognormal}(\mu, \sigma^2)$$

$$E[Y] = E[e^X] = e^{\mu + \frac{\sigma^2}{2}}$$

$$\therefore E e^{- \int_0^t \sqrt{s} dB_s} = e^{0 + \frac{t^2}{4}}$$

$$\therefore E e^{- \int_0^t \sqrt{s} dB_s} = e^{\left(\frac{t^2}{4}\right)}$$