

EXAM - ÉPOCA NORMAL - 15/06/2023 (1)  
 EN-2023 - STOCHASTIC CALCULUS

$$1(a) \quad Y_t = f\left(t, \underbrace{B_t^{(1)}}_{x_1}, \underbrace{B_t^{(2)}}_{x_2}, \underbrace{U_t}_{x_3}\right) = B_t^{(1)} B_t^{(2)} + 4(B_t^{(2)})^2 + 6 \underbrace{\int_0^t s B_s^{(1)} ds}_{= U_t} -$$

$$-kt - ct^2 B_t^{(1)}$$

$dU_t = 6t B_t^{(1)} dt$ , by Ito's formula, since  $f \in C^{1,2}$ ,

$$dY_t = \underbrace{(-k - 2ct B_t^{(1)})}_{\frac{\partial f}{\partial t}} dt + \underbrace{(B_t^{(2)} - ct^2)}_{\frac{\partial f}{\partial x_1}} dB_t^{(1)} + \underbrace{(B_t^{(1)} + 8B_t^{(2)})}_{\frac{\partial f}{\partial x_2}} dB_t^{(2)}$$

$$+ \frac{1}{2} \underbrace{(8)}_{\frac{\partial^2 f}{\partial x_2^2}} dt + \underbrace{6t B_t^{(1)}}_{\frac{\partial^2 f}{\partial x_1 \partial x_2}} dt = (-k - 2ct B_t^{(1)} + 4 + 6t B_t^{(1)}) dt +$$

$$+ (B_t^{(2)} - ct^2) dB_t^{(1)} + (B_t^{(1)} + 8B_t^{(2)}) dB_t^{(2)}$$

$$\text{So, } Y_t = \int_0^t (-k - 2cs B_s^{(1)} + 4 + 6s B_s^{(1)}) ds +$$

$$+ \int_0^t (B_s^{(2)} - cs^2) dB_s^{(1)} + \int_0^t (B_s^{(1)} + 8B_s^{(2)}) dB_s^{(2)}$$

stochastic integral is a martingale      stochastic integral is a martingale

So,  $Y_t$  is a martingale if and only if

$$-k - 2cs B_s^{(1)} + 4 + 6s B_s^{(1)} = 0 \Leftrightarrow \begin{cases} k=4 \\ 2c=6 \end{cases} \Leftrightarrow \begin{cases} k=4 \\ c=3 \end{cases}$$

Note that the processes  $B_t^{(2)} - ct^2 \in L^2_{a,r}$  and  $B_t^{(1)} + 8B_t^{(2)} \in L^2_{a,r}$  (are polynomial functions of  $B_t$  and  $t$ ) (1)

(b)  $X_t = B_{2t}^{(1)} - B_t^{(1)} \sim N(0, 2t-t) \sim N(0, t)$  (2)

It is a Gaussian process

$$E[X_t] = 0$$

$$\begin{aligned} \text{COV}[X_t, X_s] &= E[X_t X_s] = E[(B_{2t}^{(1)} - B_t^{(1)})(B_{2s}^{(1)} - B_s^{(1)})] \\ &= E[B_{2t}^{(1)} B_{2s}^{(1)}] - E[B_{2t}^{(1)} B_s^{(1)}] - E[B_t^{(1)} B_{2s}^{(1)}] + E[B_t^{(1)} B_s^{(1)}] \\ &= \min\{2t, 2s\} - \min\{2t, s\} - \min\{t, 2s\} + \min\{t, s\} \end{aligned}$$

By the covariance function of  $B^{(1)}$

$$= \begin{cases} 2t - 2t - t + t = 0 \\ 2t - s - t + t = 2t - s \\ 2s - s - 2t + s = 0 \\ 2s - s - t + s = 2s - t \end{cases}$$

$$\begin{cases} \text{i) } 2t < s \\ \text{ii) } t < s \text{ and } 2t > s \\ \text{iii) } 2s < t \\ \text{iv) } s < t \text{ and } 2s > t \end{cases}$$

Since the covariance function of  $X$  is different from the covariance function of the SBM,  $X$  is not a standard Brownian motion.

2)  $\exp(2B_t) = ?$  The process  $V_t \in L^2_{\text{loc}}$  exists by the Itô representation theorem. Let  $X_t = \exp(2B_t - 2t) = f(t, B_t)$ . By Itô's formula,

$$dX_t = \underbrace{-2 \exp(2B_t - 2t)}_{\frac{\partial f}{\partial t}} dt + \underbrace{2 \exp(2B_t - 2t)}_{\frac{\partial f}{\partial x}} dB_t + \frac{1}{2} \times \underbrace{4 \exp(2B_t - 2t)}_{\frac{\partial^2 f}{\partial x^2}} dt$$

(2)

Therefore,

$$dX_t = 2 \exp(2B_t - 2t) dB_t$$

$$\Leftrightarrow X_T - \underbrace{X_0}_{\exp(0)=1} = \int_0^T 2 \exp(2B_t - 2t) dB_t \quad \Leftrightarrow \underbrace{X_T}_{\exp(2B_T - 2T)} = 1 + \int_0^T 2 \exp(2B_t - 2t) dB_t$$

$$\Leftrightarrow \exp(2B_T) \exp(-2T) = 1 + \int_0^T 2 \exp(2B_t - 2t) dB_t$$

$$\Leftrightarrow \exp(2B_T) = e^{2T} + \int_0^T 2 e^{2T} \exp(2B_t - 2t) dB_t$$

For  $4B_t^2 + 5B_t$ , let us consider  $f(B_t) = 4B_t^2 + 5B_t = Y_t$  and apply the Ito formula to  $f$ :

$$dY_t = \underbrace{(8B_t + 5)}_{\frac{\partial f}{\partial x}} dB_t + \frac{1}{2} \underbrace{(8)}_{\frac{\partial^2 f}{\partial x^2}} dt \Rightarrow$$

$$\Rightarrow \underbrace{Y_T - Y_0}_{4B_T^2 + 5B_T - 0} = 4T + \int_0^T (8B_t + 5) dB_t$$

Therefore,  $\exp(2B_T) + 4B_T^2 + 5B_T = \underbrace{e^{2T} + 4T}_{f(T)} + \underbrace{\int_0^T (2e^{2t} \exp(2B_t - 2t) + 8B_t + 5) dt}_{N_t}$

$$(4) \quad 3a) \quad |f(t, x)| = |t^4 \ln(1+x^2)| \leq T^4 \ln(1+|x|^2)$$

$$= 2T^4 \ln(1+|x|) \leq 2T^4 (1+|x|)$$

Therefore  $f(t, x)$  satisfies the linear growth condition.   
 Like case  $\ln(1+|x|) \leq 1+|x| \Leftrightarrow 1+|x| \leq e^{1+|x|} \leq 1+1+|x| + \frac{(1+|x|)^2}{2!} + \dots$

$$|g(t, x)| = |t^2 \exp(\sin(x+1))| \leq T^2 e \leq T^2 L \leq T^2 L (1+|x|)$$

satisfies the linear growth condition

$$|\frac{\partial f}{\partial x}(t, x)| = |t^4 \frac{2x}{1+x^2}| \leq \begin{cases} T^4 \cdot 2|x| \leq T^4 \cdot 2 & \text{if } |x| \leq 1 \\ T^4 \frac{2|x|}{|x|^2} = T^4 \frac{2}{|x|} \leq T^4 \cdot 2 & \text{if } |x| > 1 \end{cases}$$

So  $|\frac{\partial f}{\partial x}(t, x)| \leq T^4 \cdot 2$  is bounded and  $f$  is differentiable with respect to  $x \Rightarrow f$  satisfies the Lipschitz property with respect to  $x$ .

$$|\frac{\partial g}{\partial x}(t, x)| = |t^2 e^{\sin(x+1)} \times \cos(x+1)| \leq T^2 L \Rightarrow g \text{ satisfies the Lipschitz property with respect to } x.$$

By the existence and uniqueness theorem, exists one unique solution in  $[0, T]$ .

$$3(h) \quad dX_t = \frac{1}{4} X_t dt + \frac{1}{2} X_t dB_t \quad \Leftarrow \alpha = 1 \quad (5)$$

let  $Z_t = \ln(X_t)$ . by Ito's formula,

$$dZ_t = \frac{1}{X_t} dX_t + \frac{1}{2} \left( -\frac{1}{X_t^2} \right) (dX_t)^2 = \frac{1}{X_t} \left( \frac{1}{4} X_t dt + \frac{1}{2} X_t dB_t \right)$$

$$+ \frac{1}{2} \left( -\frac{1}{X_t} \right) \left( \frac{1}{4} X_t^2 dt \right) \quad \Leftrightarrow dZ_t = \left( \frac{1}{4} - \frac{1}{8} \right) dt + \frac{1}{2} dB_t$$

$$Z_t = Z_0 + \frac{1}{8} t + \frac{1}{2} B_t \quad \text{and} \quad X_t = e^{Z_t} = \frac{x}{X_0} e^{\frac{1}{8} t + \frac{1}{2} B_t}$$

If  $\alpha = -2$ :  $\begin{cases} dX_t = \frac{1}{4X_t^2} dt + \frac{1}{2} X_t dB_t \\ X_0 = x > 0 \end{cases}$

Interpreting factor:  $F_t = \exp\left(\int_0^t \frac{1}{2} dB_s - \frac{1}{2} \int_0^t (\frac{1}{2})^2 ds\right) = \exp\left(\frac{1}{2} B_t - \frac{1}{8} t\right) \Rightarrow F_t^{-1} = e^{\frac{1}{8} t - \frac{1}{2} B_t}$

$$X_t = F_t Y_t \Leftrightarrow Y_t = F_t^{-1} X_t = e^{\frac{1}{8} t - \frac{1}{2} B_t} X_t$$

$$f(t, x) = \frac{1}{4} x^{-2} \Rightarrow dY_t = (F_t)^{-1} f(t, F_t Y_t) = e^{\frac{1}{8} t - \frac{1}{2} B_t} \frac{1}{4} F_t^{-2} Y_t^{-2} dt$$

$$\Leftrightarrow \begin{cases} dY_t = \frac{1}{4} e^{\frac{3}{8} t - \frac{3}{2} B_t} Y_t^{-2} dt \\ Y_0 = x \end{cases} \Leftrightarrow \begin{cases} Y_t^2 dY_t = \frac{1}{4} e^{\frac{3}{8} t - \frac{3}{2} B_t} dt \\ Y_0 = x \end{cases}$$

$$\Leftrightarrow \left[ \frac{Y_s^3}{3} \right]_0^t = \int_0^t \frac{1}{4} e^{\frac{3}{8} s - \frac{3}{2} B_s} ds \Leftrightarrow$$

$$\Leftrightarrow \frac{1}{3} Y_t^3 - \frac{1}{3} x^3 = \frac{1}{4} \int_0^t e^{\frac{3}{8} s - \frac{3}{2} B_s} ds \Leftrightarrow Y_t = \left( x^3 + \frac{3}{4} \int_0^t e^{\frac{3}{8} s - \frac{3}{2} B_s} ds \right)^{\frac{1}{3}}$$

$$\Rightarrow X_t = F_t Y_t = e^{\frac{1}{2} B_t - \frac{1}{8} t} \left( x^3 + \frac{3}{4} \int_0^t e^{\frac{3}{8} s - \frac{3}{2} B_s} ds \right)^{\frac{1}{3}}$$

⑥ 4(a)  $Z_s = F(s, X_s)$ . By the Ito formula,

$$dZ_s = \frac{\partial F}{\partial s} ds + \frac{\partial F}{\partial x} dX_s + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (dX_s)^2 =$$

$$= \frac{\partial F}{\partial s} ds + \frac{\partial F}{\partial x} (2X_s ds + 6X_s dB_s) + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (36X_s^2 ds) =$$

$$= \left( \frac{\partial F}{\partial s} + 2X_s \frac{\partial F}{\partial x} + 18X_s^2 \frac{\partial^2 F}{\partial x^2} \right) ds + 6X_s \left( \frac{\partial F}{\partial x} \right) dB_s =$$

by the PDE =  $-X_s$

$$= -X_s ds + 6X_s \frac{\partial F(s, X_s)}{\partial x} dB_s \quad \text{Interpreting between } t \text{ and } T:$$

$$\int_t^T dF(s, X_s) = \int_t^T \cancel{-X_s} ds + \int_t^T 6X_s \frac{\partial F(s, X_s)}{\partial x} dB_s \Leftrightarrow$$

$$\Leftrightarrow F(T, X_T) - F(t, X_t) = - \int_t^T X_s ds + 6 \int_t^T X_s \frac{\partial F(s, X_s)}{\partial x} dB_s$$

$$\Rightarrow \underbrace{\mathbb{E}_{t,x} [F(T, X_T)] - \mathbb{E}_{t,x} [F(t, X_t)]}_{= F(t,x)} = - \int_t^T \underbrace{\mathbb{E}_{t,x} [X_s]}_{0} ds + 6 \underbrace{\mathbb{E}_{t,x} \left[ \int_t^T X_s \frac{\partial F(s, X_s)}{\partial x} dB_s \right]}_0$$

$$\Leftrightarrow F(t,x) = \underbrace{\mathbb{E}_{t,x} [F(T, X_T)]}_{h_1(X_T^4)} + \int_t^T \mathbb{E}_{t,x} [X_s] ds$$

$$\Leftrightarrow F(t,x) = \mathbb{E}_{t,x} [h_1(X_T^4)] + \int_t^T \mathbb{E}_{t,x} [X_s] ds //$$

4(h) In order to apply the formula let us calculate  $\mathbb{E}_t$   
 $X_s$ :  $\begin{cases} dX_s = 2X_s ds + 6X_s dB_s \\ X_t = x \end{cases} \Rightarrow$  solution is the G.B.M.

$$X_s = x e^{(2 - \frac{1}{2}6^2)(s-t) + 6(B_s - B_t)} \Leftrightarrow$$

$$\Leftrightarrow X_s = x e^{-16(s-t) + 6(B_s - B_t)}$$

By the formula,  $F(t, x) = \mathbb{E}_{t, x} \left[ \ln(x^4 e^{-64(\tau-t) + 24(B_\tau - B_t)}) \right] +$

$$+ \int_t^\tau \mathbb{E}_{t, x} \left[ x e^{-16(s-t) + 6(B_s - B_t)} \right] ds =$$

$$= \mathbb{E}_{t, x} \left[ \ln(x^4) + (-64(\tau-t) + 24(B_\tau - B_t)) \right] + \int_t^\tau x e^{-16(s-t)} e^{18(s-t)} ds$$

$$= \ln(x^4) - 64(\tau-t) + x \left[ \frac{e^{2(s-t)}}{2} \right]_t^\tau =$$

$$= \ln(x^4) - 64(\tau-t) + \frac{x}{2} (e^{2(\tau-t)} - 1) //$$

5) Under  $\mathbb{Q}$ , we have:

$$\begin{cases} ds_t = \Omega S_t dt + \sigma S_t dW_t \\ S_0 = S_0 \end{cases}, \text{ where } W \text{ is a S.B.M.}$$

Under  $\mathbb{Q}$ ,  
 and  $\Omega$  is the risk-free  
 interest rate

Solution is the G.B.M.:

$$S_t = S_0 e^{(\Omega - \frac{1}{2}\sigma^2)t + \sigma W_t}$$

⑧ The price at time 0 is

$$\begin{aligned}
 F(0, S_0) &= e^{-rT} \mathbb{E}_Q [\Phi(S_T)] = \\
 &= e^{-rT} \mathbb{E}_Q \left[ \frac{1}{K} S_T^4 + \ln(S_T) + K \mathbb{1}_{M_1 < S_T < M_2} \right] = \\
 &= e^{-rT} \left\{ \frac{1}{K} \mathbb{E}_Q \left[ S_0^4 e^{4(r - \frac{1}{2}\sigma^2)T + 4\sigma W_T} \right] + \mathbb{E}_Q \left[ \ln(S_0) + (r - \frac{1}{2}\sigma^2)T + \sigma W_T \right] \right. \\
 &\quad \left. + K \mathbb{P}_Q [M_1 < S_T < M_2] \right\} = e^{-rT} \left\{ \frac{1}{K} S_0^4 e^{(4r - \frac{1}{2}\sigma^2)4T} e^{\frac{1}{2}4\sigma^2 T} + \right. \\
 &\quad \left. + \ln(S_0) + (r - \frac{1}{2}\sigma^2)T + K \mathbb{P}_Q (\ln(M_1) < \ln(S_T) < \ln(M_2)) \right\} \\
 &= e^{-rT} \left\{ \frac{S_0^4}{K} e^{(4r + 6\sigma^2)T} + \ln(S_0) + (r - \frac{1}{2}\sigma^2)T + \right. \\
 &\quad \left. + K \mathbb{P}_Q \left[ \underbrace{\frac{\ln(M_1) - \ln(S_0) - (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}}_{\equiv d_1} < Z < \underbrace{\frac{\ln(M_2) - \ln(S_0) - (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}}_{\equiv d_2} \right] \right\} \\
 &= e^{-rT} \left\{ \frac{S_0^4}{K} e^{(4r + 6\sigma^2)T} + \ln(S_0) + (r - \frac{1}{2}\sigma^2)T + \right. \\
 &\quad \left. + K (N[d_2] - N[d_1]) \right\}, \text{ where } N \text{ is the} \\
 &\quad \text{cumulative distribution function} \\
 &\quad \text{of the } N(0,1) \text{ distribution}
 \end{aligned}$$



6. Let  $m > m$  and consider

(9)!

$$\|z_t^{(m)} - z_t^{(n)}\|_{L^2(\mathbb{P})} = \left\| \sum_{k=m}^{n-1} (z_t^{(k+1)} - z_t^{(k)}) \right\|_{L^2(\mathbb{P})} \leq$$

$$\leq \sum_{k=m}^{n-1} \|z_t^{(k+1)} - z_t^{(k)}\|_{L^2(\mathbb{P})} = \sum_{k=m}^{n-1} \left( \mathbb{E} \|z_t^{(k+1)} - z_t^{(k)}\|^2 \right)^{\frac{1}{2}}$$

By the norm triangle inequality property

$$\leq \sum_{k=m}^{\infty} \left( \frac{c^{k+1} t^{k+1}}{(k+1)!} \right)^{\frac{1}{2}} \rightarrow \text{This series is convergent because?}$$

$$\frac{a_{k+1}}{a_k} = \frac{\left( \frac{c^{k+2} t^{k+2}}{(k+2)!} \right)^{\frac{1}{2}}}{\left( \frac{c^{k+1} t^{k+1}}{(k+1)!} \right)^{\frac{1}{2}}} = \left( \frac{c t}{k+2} \right)^{\frac{1}{2}} \xrightarrow{k \rightarrow \infty} 0$$

Therefore  $\sum_{k=m}^{\infty} \left( \frac{c^{k+1} t^{k+1}}{(k+1)!} \right)^{\frac{1}{2}} \xrightarrow{m \rightarrow \infty} 0$ .

So,  $\|z_t^{(m)} - z_t^{(n)}\|_{L^2(\mathbb{P})} \longrightarrow 0$  as  $m, n \rightarrow \infty$  and

the sequence is a Cauchy sequence in  $L^2(\mathbb{P})$ . Since this space is a complete space, a Cauchy sequence is a convergent sequence. Therefore  $z_t^{(m)}$  converges to a limit as  $m \rightarrow \infty$ .

So,  $z_t^{(m)} \xrightarrow{m \rightarrow \infty} z_t$  in  $L^2(\mathbb{P})$ .

Every sequence that converges in  $L^2(\mathbb{P})$  has a subsequence that converges a.s. Therefore, exists a subsequence

$z_t^{(m_k)} \longrightarrow z_t$  a.s.

(9)