

4.1 a

$$Y_t = B_t^2$$

$$dY_t = 2B_t dB_t + 2 dt.$$

b

$$Y_t = 2 + t + e^{B_t}.$$

$$dY_t = e^{B_t} dB_t + \left(\frac{1}{2} e^{B_t} + 1 \right)$$

c

$$dY_t = 2 \left[B_t^{(1)} dB_t^{(1)} + B_t^{(2)} dB_t^{(2)} + \right.$$

d

$$\left. \frac{t dt}{dB_t} \right].$$

e

$$dY_t = \left[\begin{array}{l} dB_t^{(1)} + dB_t^{(2)} + dB_t^{(3)} \\ 2B_t^{(2)} dB_t^{(2)} + 2dt - B_t^{(1)} dB_t^{(3)} - B_t^{(3)} dB_t^{(1)} \end{array} \right].$$

4.2

$$d(B_t^3) = 3B_t^2 dB_t + 3B_t dt.$$

$$\Rightarrow B_t^3 = \int_0^t 3B_t^2 dB_t + \int_0^t 3B_t dt \Rightarrow \text{Rearrange.}$$

4.3

For $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, by THEOREM 4.2.1,

$$df(X_t, Y_t) = \frac{\partial f}{\partial x} dX_t + \frac{\partial f}{\partial y} dY_t + \frac{1}{2} \frac{\partial^2 f}{\partial y^2} d[Y]_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} d[X]_t + \frac{\partial^2 f}{\partial x \partial y} d[X, Y]_t.$$

When $f(x, y) = xy$.

$$\frac{\partial f}{\partial x} = y, \quad \frac{\partial f}{\partial y} = x, \quad \frac{\partial f}{\partial x \partial y} = 1, \quad \text{and} \quad \frac{\partial^2 f}{\partial x^2} = \frac{\partial^2 f}{\partial y^2} = 0.$$

Thus

$$d(X_t Y_t) = X_t dY_t + Y_t dX_t + dX_t \cdot dY_t.$$

$$\begin{aligned} \underline{4.4} \quad dZ_t &= Z_t \left[\theta(t, \omega) dB_t - \frac{1}{2} \theta^2(t, \omega) dt + \frac{1}{2} (\theta(t, \omega) dB_t)^2 \right] \\ &= Z_t \theta(t, \omega) dB_t \end{aligned}$$

* If you are uneasy about this bit,
write $X_t = \int_0^t \theta(s, \omega) dB_s$. Then $dX_t = \theta dB_t + \frac{1}{2} \theta^2 dB_t^2$.

b Clearly holds by Proof of Theorem 3.2.5, which states

$$\int_0^t \lambda_s dB_s$$

is always a martingale.

Discussion about the remark.

In order for Z_t to be a martingale, we need to show $Z_t \theta_t^{(k)} \in \mathcal{V}[0, T]$ for each dimension $k=1, \dots, n$.

Recall, $\mathcal{V}(S, T) = \left\{ \begin{array}{l} \text{I } (t, \omega) \rightarrow \phi(t, \omega) \text{ } \mathcal{B} \times \mathcal{F} \text{ measurable} \\ \text{II } \phi(t, \omega) \text{ } \mathcal{F}_t\text{-adapted} \\ \text{III } \mathbb{E} \left[\int_S^T \phi(t, \omega)^2 dt \right] < \infty. \end{array} \right\}.$

In the context of 19.4, the function $\phi^{(k)}(t, \omega) = Z(t, \omega) \theta(t, \omega)$ clearly satisfies I and II, since:

By the claim we proved in 19.7,

$$\theta_t \in \mathcal{V}(0, T) \Rightarrow \int_0^t \theta_s dB_s \in \mathcal{V}(0, T).$$

Thus

$$\phi^{(k)}(t, \omega) = \exp \left(\underbrace{\int_0^t \theta_s dB_s}_{\in \mathcal{V}(0, T)} - \frac{1}{2} \underbrace{\int_0^t \theta_s^2 ds}_{\in \mathcal{V}(0, T)} \right) \underbrace{\theta_t}_{\in \mathcal{V}(0, T)}.$$

By comment immediately above.

So ϕ_t is a Borel function of things in $\mathcal{V}(0, T)$ at time t .

So I and II follow by properties of measurable functions.

Left to verify is whether III holds. The remark states that

$$\text{NOVIKOV} \Rightarrow \text{KAZAMAKI} \Rightarrow \text{III}.$$

(P.T.O)

4.5

$$d(B_t^k) = k B_t^{k-1} dB_t + \frac{1}{2} k(k-1) B_t^{k-2} (dB_t)^2$$

$$B_t^k = k \int_0^t B_s^{k-1} dB_s + \frac{1}{2} k(k-1) \int_0^t B_s^{k-2} ds.$$

Hence, *martingale starting at 0.*

$$\mathbb{E}[B_t^k] = \frac{1}{2} k(k-1) \int_0^t \mathbb{E}[B_s^{k-2}] ds. \quad (*)$$

$$\begin{aligned} \underline{a} \quad \mathbb{E}[B_t^4] &= \frac{1}{2} 4 \cdot 3 \int_0^t \mathbb{E}[B_s^2] ds \\ &= 6 \int_0^t s ds = 3t^2. \end{aligned}$$

$$\underline{b} \quad \mathbb{E}[B(t)^{2k+1}] = 0 \text{ since } \mathbb{E}[B_t^1] = 0,$$

and by (*), this holds for all odd numbers.

$$\underline{\underline{To prove}} \quad \mathbb{E}[B(t)^{2k}] = \frac{(2k)! t^k}{2^k k!} \quad k=1, 2, \dots$$

we proceed by induction. Note it is true when $k=1$, and

$$\begin{aligned} \mathbb{E}[B(t)^{2(k+1)}] &= \frac{1}{2} (2k+2)(2k+1) \int_0^t \frac{(2k)!}{k!} \frac{s^k}{2^k} ds \\ &= \frac{1}{2} \frac{(2k+2)!}{k!} \frac{1}{2^k} \int_0^t s^k ds \\ &= \frac{(2(k+1))!}{(k+1)!} \frac{1}{2^{k+1}} t^{k+1}. \end{aligned}$$

4.6 11a

$$X_t = e^{ct + \alpha B_t}.$$

$$dX_t = \left[c dt + \alpha dB_t + \frac{1}{2} \alpha^2 dt \right] X_t. \text{ Rearrange}$$

$$\stackrel{11b}{=} X_t = \exp \left(ct + \sum_{j=1}^n \alpha_j B_j(t) \right).$$

$$dX_t = \left[c dt + \sum \left(\alpha_j dB_t + \frac{1}{2} \alpha_j^2 dt \right) \right] X_t \text{ Rearrange.}$$

11c

Note that if $c = -\frac{1}{2} \sum \alpha_j^2$, then X_t is a martingale.

4.7 a $v \equiv 1$. Then $X_t = B_t$, and X_t^2 is not a martingale.

b Prove if v bounded

$$\text{Then } \Rightarrow M_t = X_t^2 - \int_0^t |v_s|^2 ds$$

is a martingale.

I M_t clearly adapted. ✓

$$\text{II } \mathbb{E}[|M_t|] = \mathbb{E}\left[X_t^2 - \int_0^t |v_s|^2 ds\right]$$

$$\leq \mathbb{E}\left[\int_0^t |v_s| dB_s\right] + \mathbb{E}\left[\int_0^t |v_s|^2 ds\right]$$

$$\leq M \mathbb{E}[|B_t|] + t M^2$$

where $|v| \leq M$.

$$\leq M \sqrt{\mathbb{E}[B_t^2]} + t M^2$$

$$\leq M\sqrt{t} + t M^2 < \infty \quad \checkmark$$

III Let $s < t$.

$$\mathbb{E}[M_t | \mathcal{F}_s] = \mathbb{E}\left[X_t^2 - \int_0^t |v_s|^2 ds \mid \mathcal{F}_s\right]$$

$$= \mathbb{E}\left[X_t^2 - X_s^2 - \int_s^t |v_s|^2 ds \mid \mathcal{F}_s\right] + X_s^2$$

$$= \mathbb{E}\left[\underbrace{2X_s(X_t - X_s)}_{=0} + \underbrace{(X_t - X_s)^2}_{\text{by Ito ISOMETRY}} - \int_s^t |v_s|^2 ds \mid \mathcal{F}_s\right] + X_s^2$$

since X martingale. $= \mathbb{E}\left[\int_s^t |v_s|^2 ds \mid \mathcal{F}_s\right]$
by Ito ISOMETRY.

$$= X_s^2$$

4.8a (in THEOREM 4.2.1)

By the Multidimensional Itô Formula, if

$$g: [0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}^p.$$

Then $\forall k = 1, \dots, p$

$$d(g(t, X)) = dY_t^{(k)} = \frac{\partial g^{(k)}}{\partial t} dt + \sum_{i=1}^n \frac{\partial g^{(k)}}{\partial x_i} dX_t^{(i)} + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 g^{(k)}}{\partial x_i \partial x_j} dX_t^{(i)} dX_t^{(j)}$$

With $p=1$, \dots

$$\begin{aligned} df(B_t) &= \sum_{i=1}^n \frac{\partial f^{(i)}}{\partial x^{(i)}} dB_t^{(i)} + \frac{1}{2} \sum_i \sum_j \frac{\partial^2 f^{(i)}}{\partial x^{(i)} \partial x^{(j)}} dB_t^{(i)} dB_t^{(j)} \\ &= \underbrace{(\nabla f) \cdot dB_t}_{\nabla f = \left(\frac{\partial f}{\partial x^{(1)}}, \dots, \frac{\partial f}{\partial x^{(n)}} \right)} + \frac{1}{2} \underbrace{\Delta f}_{\Delta f = \sum_{i=1}^n \frac{\partial^2 f}{\partial x^{(i)2}}} dt \\ &\quad \text{where } \frac{\partial^2 f^{(i)}}{\partial x^{(i)} \partial x^{(j)}} = \mathbb{1}\{i=j\} \frac{\partial^2 f}{\partial x^{(i)2}} \end{aligned}$$

Result follows.

4.8b Let $g: \mathbb{R} \rightarrow \mathbb{R} \in C^2(\mathbb{R}) \cap C^2(\mathbb{R} - \{z_1, \dots, z_N\})$.

Now, $\exists f_k: \mathbb{R} \rightarrow \mathbb{R} : f_k \Rightarrow g$

$$\bullet f_k' \Rightarrow g'$$

$$\bullet \|f_k''\| \leq M.$$

$$\bullet f_k'' \rightarrow g'' \text{ on } \mathbb{R} - \{z_1, \dots, z_N\}.$$

We need to establish ①, ②, ③, ④.

$$f_k(B_t) = f_k(B_0) + \int_0^t f'_k(B_s) ds + \frac{1}{2} \int_0^t f''_k(B_s) ds$$

$$g(B_t) = g(B_0) + \int_0^t g'(B_s) ds + \frac{1}{2} \int_0^t g''(B_s) ds$$

↓ ①
↓ ②
↓ ③
↓ ④

Well, ① and ② follow by pointwise convergence of $f_k \rightarrow g$.

③ Now, since $f'_k \rightrightarrows g'$,

$$\forall \epsilon > 0 \exists N \in \mathbb{N} : \forall x \in \mathbb{R}, \forall k \geq N.$$

$$|g'(x) - f'_k(x)| \leq \epsilon.$$

Thus $\left| \int_0^t f'_k(B_s) ds - \int_0^t g'(B_s) ds \right| \leq \int_0^t |f'_k - g'| (B_s) ds \leq \epsilon t \rightarrow 0.$

④. This follows from bounded convergence theorem, on the measure space $(\text{Leb}, \mathcal{B}[0, t], [0, t])$

4.9 Let $T_n = \inf \{s > 0 : |X_s(\omega)| \geq n\}$. Prove

$$\begin{aligned} \int_0^t v \cdot \frac{\partial g_n}{\partial x}(s, X_s) \mathbb{1}_{\{s \leq T_n\}} dB_s &= \int_0^{t \wedge T_n} v \frac{\partial g_n}{\partial x}(s, X_s) dB_s \\ &= \int_0^{t \wedge T_n} v \cdot \frac{\partial g}{\partial x}(s, X_s) dB_s. \end{aligned}$$

Well, this follows since $\forall s \leq t \wedge T_n$, the integrands are identical, and when/if $T_n < s \leq t$, the first integrand is 0.

Then

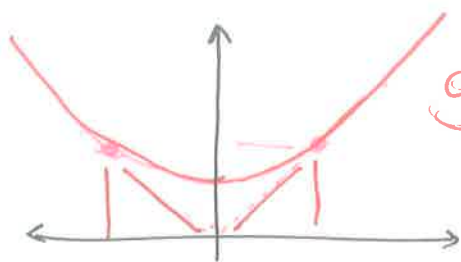
$$g(t \wedge T_n, X_{t \wedge T_n}) = g(0, X_0) + \int_0^{t \wedge T_n} \left(\frac{\partial g}{\partial s} + u \frac{\partial g}{\partial x} + \frac{1}{2} v^2 \frac{\partial^2 g}{\partial x^2} \right) ds + \int_0^{t \wedge T_n} v \frac{\partial g}{\partial x} dB_s.$$

Finally, $\forall t$ fixed, X_t is a finite random variable.

$$\text{Thus } 1 = \mathbb{P}(X_t \in \mathbb{R}) = \mathbb{P}\left(\limsup_{n \rightarrow \infty} \{|X_t| \leq n\}\right)$$

$$\Rightarrow \mathbb{P}(|X_t| \leq n) \uparrow 1 \text{ as } n \rightarrow \infty.$$

Thus taking $n \rightarrow \infty$ adds a.s. on both sides.

4.10 TANAKA'S FORMULA & LOCAL TIME

$$g_\epsilon(x) = \begin{cases} |x| & |x| \geq \epsilon \\ \frac{1}{2}(\epsilon^2 + \frac{x^2}{\epsilon}) & |x| < \epsilon \end{cases}$$

a By 4.8 b), we note that g_ϵ is $C^2(\mathbb{R} - \{-1, 1\})$.

$$g_\epsilon''(x) = \frac{1}{\epsilon} \mathbb{1}\{|x| < \epsilon\}.$$

Thus,

$$\begin{aligned} g_\epsilon(B_t) &= g_\epsilon(B_0) + \int_0^t g_\epsilon'(B_s) dB_s + \frac{1}{2} \int_0^t g_\epsilon''(B_s) ds \\ &= g_\epsilon(B_0) + \int_0^t g_\epsilon'(B_s) dB_s + \frac{1}{2\epsilon} \int_0^t \mathbb{1}\{B_s \in (-\epsilon, \epsilon)\} ds \\ &= g_\epsilon(B_t) + \int_0^t g_\epsilon'(B_s) dB_s + \frac{1}{2\epsilon} \text{Leb}\{s : B_s \in (-\epsilon, \epsilon)\}. \end{aligned}$$

$$\text{b} \quad \mathbb{E} \left[\left(\int_0^t \frac{B_s}{\epsilon} \mathbb{1}\{B_s \in (-\epsilon, \epsilon)\} dB_s \right)^2 \right]$$

$$= \mathbb{E} \left[\int_0^t \frac{B_s^2}{\epsilon^2} \mathbb{1}\{B_s \in (-\epsilon, \epsilon)\} ds \right]$$

$$\leq \mathbb{E} \left[\int_0^t \mathbb{1}\{B_s \in (-\epsilon, \epsilon)\} ds \right]$$

$$\left[\text{Now } P(|B_s| < \epsilon) = \int_{-\epsilon}^{\epsilon} \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2}x^2} dx \leq \frac{2\epsilon}{\sqrt{2\pi t}} \right]$$

$$\leq \int_0^t \frac{2\epsilon}{\sqrt{2\pi s}} ds = \sqrt{\frac{2}{\pi}} \epsilon \int_0^t s^{-1/2} ds = \frac{1}{\sqrt{2\pi}} \left[s^{1/2} \right]_0^t \epsilon = \frac{t^{1/2}}{\sqrt{2\pi}} \epsilon \rightarrow 0$$

\subseteq Thus as $\epsilon \rightarrow 0$,

$$\begin{aligned} \lim_{\epsilon \downarrow 0} g_\epsilon(B_t) &= \lim_{\epsilon \downarrow 0} \left[g_\epsilon(B_0) + \int_0^t g'_\epsilon(B_s) dB_s + \frac{1}{2\epsilon} \text{Leb}\{s: |B_s| < \epsilon\} \right] \\ &= |B_0| + \int_0^t g'_\epsilon(B_s) \mathbb{1}_{\{|B_s| \geq \epsilon\}} dB_s + \underbrace{\frac{1}{2} \int_0^t g'_\epsilon(B_s) \mathbb{1}_{\{|B_s| < \epsilon\}} ds}_{\rightarrow 0 \text{ in } L^2} + L_t \\ &= |B_0| + \int_0^t \text{sign}(B_s) \mathbb{1}_{\{|B_s| \geq \epsilon\}} dB_s + L_t. \end{aligned}$$

Now note

$$\mathbb{E} \left[\left(\int_0^t \text{sign}(B_s) \mathbb{1}_{\{|B_s| < \epsilon\}} dB_s \right)^2 \right] = \mathbb{E} \left[\int_0^t \mathbb{1}_{\{|B_s| < \epsilon\}} ds \right] \rightarrow 0$$

by argument in part b.

$$\text{Thus} \quad = |B_0| + \int_0^t \text{sign}(B_s) dB_s + L_t.$$

4.11

a

$$dX_t = \frac{1}{2} e^{\frac{1}{2}t} \cos(B_t) dt - e^{\frac{1}{2}t} \sin(B_t) dB_t - \frac{1}{2} e^{\frac{1}{2}t} \cos(B_t) dt$$

$$= -e^{\frac{1}{2}t} \sin(B_t) dB_t$$

This martingale.

b

$$dX_t = e^{\frac{1}{2}t} \cos(B_t) dB_t. \quad \text{Same.}$$

c

$$f(x, t) = (x+t) \exp(-x - \frac{1}{2}t)$$

$$\frac{\partial f}{\partial t} = \exp(-x - \frac{1}{2}t) - \frac{1}{2}(x+t) \exp(-x - \frac{1}{2}t)$$

$$= [1 - \frac{1}{2}x - \frac{1}{2}t] \exp(-x - \frac{1}{2}t)$$

$$\frac{\partial f}{\partial x} = [1 - x - t] \exp(-x - \frac{1}{2}t)$$

$$\frac{\partial^2 f}{\partial x^2} = [-1 - (1 - x - t)] \exp(-x - \frac{1}{2}t)$$

$$dX_t = (1 - B_t - t) \exp(-B_t - \frac{1}{2}t) dB_t$$

$$+ \left[\underbrace{(1 - \frac{1}{2}B_t - \frac{1}{2}t)}_{=0} + \frac{1}{2}(-2 + x + t) \right] dt$$

$$= [1 - B_t - t] e^{-B_t - \frac{1}{2}t} dB_t$$

$$\begin{aligned} \underline{4.12} \quad 0 &= \mathbb{E}[X_s - X_t \mid \mathcal{F}_t^{(n)}] \\ &= \mathbb{E}\left[\int_t^s u(r, \omega) dr \mid \mathcal{F}_t^{(n)}\right] \end{aligned}$$

Then

$$\begin{aligned} 0 &= \frac{d}{ds} \mathbb{E}\left[\int_t^s u(r, \omega) dr \mid \mathcal{F}_t^{(n)}\right] \\ &= \mathbb{E}\left[\frac{d}{ds} \int_t^s u(r, \omega) dr \mid \mathcal{F}_t^{(n)}\right] \quad \forall \text{ a.a } s > t. \\ &= \mathbb{E}[u(s, \omega) \mid \mathcal{F}_t^{(n)}] \end{aligned}$$

Now, as $t \uparrow s$, by COROLLARY C-9,

$$0 = \mathbb{E}[u(s, \omega) \mid \mathcal{F}_t^{(n)}] \rightarrow \mathbb{E}[u(s, \omega) \mid \mathcal{F}_s^{(n)}] = u(s, \omega).$$

Thus $u(s, \omega) \quad \forall \text{ a.a } (s, \omega).$

4.13

$$dX_t = u dt + dB_t.$$

Show $Y_t = X_t M_t$ is a martingale, where

$$M_t = \exp \left(- \int_0^t u dB_r - \frac{1}{2} \int_0^t u^2 dr \right).$$

(is also a martingale).

$$dY_t = M_t dX_t + X_t dM_t + dX_t dM_t.$$

$$\left(\begin{aligned} \text{Now } dM_t &= M_t \left(-u dB_t - \frac{1}{2} u^2 dt + \frac{1}{2} u^2 (dB_t)^2 \right) \\ &= -u M_t dB_t. \end{aligned} \right)$$

$$= M_t \left[u dt + dB_t + X_t (-u dB_t) + (-u dB_t)(u dt + dB_t) \right]$$

$$= M_t \left[u dt + dB_t + u X_t dB_t - u^2 dB_t dt - u (dB_t)^2 \right]$$

$$= M_t \left[1 - u X_t \right] dB_t.$$

4.13 (cont.)Remarks.a With $X_t = 1 dt + 1 dB_t$ (in 4.11 c),Clearly by 4.13.

$$X_t \exp\left(-\int_0^t 1 dB_s - \frac{1}{2} \int_0^t 1^2 ds\right)$$

is a martingale.

4.14
a

$$B_T(\omega) = \int_0^T 1 dB_t.$$

b $d(tB_t) = B_t dt + t dB_t.$

$$\Rightarrow TB_T = \int_0^T B_t dt + \int_0^T t dB_t.$$

$$\int_0^T B_t dt = \int_0^T (T-t) dB_t.$$

c $d(B_t^2) = 2B_t dB_t + dt.$

$$B_T^2(\omega) = T + \int_0^T 2B_t dB_t.$$

d $d(B_t^3) = 3B_t^2 dB_t + 3B_t dt.$
 $= 3B_t^2 dB_t + 3d(tB_t) - 3t dB_t.$

$$B_T^3(\omega) = \int_0^T 3B_t^2 + 3T - 3t dB_t.$$

e $d(e^{B_t}) = e^{B_t} dB_t + \frac{1}{2} e^{B_t} dt.$

$$d(te^{B_t}) = t d(e^{B_t}) + e^{B_t} dt$$

So $d(e^{B_t}) = e^{B_t} dB_t + \frac{1}{2} [d(te^{B_t}) - t d(e^{B_t})]$

$$(1 + \frac{1}{2}t) d(e^{B_t}) = e^{B_t} dB_t + \frac{1}{2} d(te^{B_t}).$$

e

This one is a bit trickier to spot:

$$\begin{aligned} \text{Note } d(e^{B_t - \frac{1}{2}t}) &= e^{B_t - \frac{1}{2}t} dB_t + \frac{1}{2} e^{B_t - \frac{1}{2}t} dB_t^2 - \frac{1}{2} e^{B_t - \frac{1}{2}t} dt \\ &= e^{B_t - \frac{1}{2}t} dB_t. \end{aligned}$$

Thus

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

$$\Rightarrow \underline{e^{B_T} = e^{\frac{1}{2}T} + \int_0^T e^{\frac{1}{2}T} (e^{B_t - \frac{1}{2}t}) dB_t.}$$

f

Again, like in e, let's resort to finding a martingale:

$$\begin{aligned} d(e^{\frac{1}{2}t} \sin B_t) &= e^{\frac{1}{2}t} \cos(B_t) dB_t - \frac{1}{2} e^{\frac{1}{2}t} \sin B_t dB_t^2 + \frac{1}{2} e^{\frac{1}{2}t} \sin B_t dt \\ &= e^{\frac{1}{2}t} \cos(B_t) dB_t. \end{aligned}$$

Thus

$$e^{\frac{1}{2}T} \sin(B_T) = \int_0^T e^{\frac{1}{2}t} \cos(B_t) dB_t.$$

$$\Rightarrow \sin(B_T) = \int_0^T e^{-\frac{1}{2}T} (e^{\frac{1}{2}t} \cos(B_t)) dB_t$$

4.15

$$X_t = (x^{1/3} + \frac{1}{3} B_t)^3$$

$$= f(t, B_t) = f(B_t), \quad f(u) = (x^{1/3} + \frac{1}{3} u)^3$$

$$f'(u) = (x^{1/3} + \frac{1}{3} u)^2 = f(u)^{2/3}$$

$$f''(u) = \frac{2}{3} (x^{1/3} + \frac{1}{3} u) = \frac{2}{3} f(u)^{1/3}$$

$$\text{Itô: } dX_t = f'(B_t) dB_t + \frac{1}{2} f''(B_t) d\langle B \rangle_t$$

$$dX_t = X_t^{2/3} dB_t + \frac{1}{3} X_t^{1/3} dt$$

□

4.16aLet $Y \in L^2(\mathcal{F}_T)$. $M_t = \mathbb{E}[Y | \mathcal{F}_t]$.a Show $\mathbb{E}[M_t^2] < \infty \quad \forall t \in [0, T]$.By 3.16, $\mathbb{E}[\mathbb{E}[Y | \mathcal{F}_t]^2] \leq \mathbb{E}[Y^2]$ since $\mathcal{F}_t \subseteq \mathcal{F}_T$.b Find g :

$$\mathbb{E}[B_T^2 | \mathcal{F}_t] = \mathbb{E}[B_T^2] + \int_0^t g(s, \omega) dB_s.$$

Now,

$$d(B_t^2 - t) = 2B_t dB_t.$$

$$\text{ie } B_T^2 - T = \int_0^T 2B_s dB_s, \quad \text{so}$$

$$\mathbb{E}[B_T^2 | \mathcal{F}_t] = T + \int_0^t 2B_s dB_s.$$

Checking:

$$\begin{aligned} \mathbb{E}[B_T^2 | \mathcal{F}_t] &= \mathbb{E}[(B_T - B_t) + B_t]^2 | \mathcal{F}_t \\ &= (T-t) + B_t^2 \\ &= T + (B_t^2 - t) \\ &= T + \int_0^t 2B_s dB_s \end{aligned}$$

4.16 b u

We've seen before that

$$\mathbb{E}[B_T^3 | \mathcal{F}_t] = B_t^3 + 3B_t(T-t)$$

And in 4.14 d,

$$B_t^3 = \int_0^t \underbrace{3B_s^2 + 3t - 3s}_{\text{These integrands shouldn't depend on } t \text{ in the final answer...}} dB_s.$$

Also,

$$3B_t(T-t) = \int_0^t \underbrace{3(T-t)}_{\text{These integrands shouldn't depend on } t \text{ in the final answer...}} dB_s.$$

Fortunately, the t -dependence cancels:

$$B_t^3 + 3B_t(T-t) = \int_0^t 3B_s^2 + 3(T-s) dB_s.$$

So

$$\mathbb{E}[B_T^3 | \mathcal{F}_t] = \underbrace{0}_{\mathbb{E}[B_T^3] = 0} + \int_0^t 3B_s^2 + 3(T-s) dB_s$$

4.16 b iii

$$\begin{aligned}
\mathbb{E}[Y | \mathcal{F}_t] &= \mathbb{E}[e^{\sigma B_T} | \mathcal{F}_t] \\
&= e^{\frac{1}{2}\sigma^2 T} \mathbb{E}[e^{\sigma B_T - \frac{1}{2}\sigma^2 T} | \mathcal{F}_t] \\
&= e^{\frac{1}{2}\sigma^2 T} e^{\sigma B_t - \frac{1}{2}\sigma^2 t}
\end{aligned}$$

With $X_t = e^{\sigma B_t - \frac{1}{2}\sigma^2 t}$

$$dX_t = \sigma(e^{\sigma B_t - \frac{1}{2}\sigma^2 t}) dB_t$$

Thus $X_t = 1 + \int_0^t \sigma e^{\sigma B_s - \frac{1}{2}\sigma^2 s} dB_s$.

It follows that

$$\mathbb{E}[Y | \mathcal{F}_t] = e^{\frac{1}{2}\sigma^2 T} + \int_0^t \sigma e^{\sigma B_s + \frac{1}{2}\sigma^2(T-s)} dB_s$$

4.17LEMMA 4.3.1

$$\forall \epsilon > 0, \exists n, \exists t_1, \dots, t_n, \exists \phi \in \mathcal{C}_0^\infty(\mathbb{R}^n) :$$

$$\|Y - \phi(B_{t_1}, \dots, B_{t_n})\|_{L^2} = \mathbb{E}[|Y - \phi(B_{t_1}, \dots, B_{t_n})|^2]^{\frac{1}{2}} < \epsilon$$

Assuming this Lemma, we want to show,

$\forall Y \in L^2(\mathcal{F}_T)$, there exists $f(t, \omega)$ such that

Y has representation

$$Y(\omega) = \mathbb{E}[F] + \int_0^T f(t, \omega) dB_t \quad (4.3.14).$$

Now the question says

"In view of this Lemma it is enough to prove the representation holds for Y of the 'form' .

(★)

$$Y = \phi(B_{t_1}, \dots, B_{t_n}) \quad \phi \in \mathcal{C}_0^\infty(\mathbb{R}^n) "$$

Before we answer a, b, c, let us check

(★) is true.

claim

- IF
1. Every $Y \in \mathcal{L}^2(\mathcal{F}_T)$ can be approximated in \mathcal{L}^2 by function of form $\phi(B_{t_1}, \dots, B_{t_n})$ (+)
 2. Every function of form $\phi(B_{t_1}, \dots, B_{t_n})$ can be represented

$$\phi(B_{t_1}, \dots, B_{t_n}) = \mathbb{E}[\phi(B_{t_1}, \dots, B_{t_n})] + \int_0^T f(t, \omega) dB_t.$$
- Lemma 4.3.1
- What we have to show in [4.17]

Then. Every $Y \in \mathcal{L}^2(\mathcal{F}_T)$ can be represented.

$$Y = \mathbb{E}[Y] + \int_0^T f(t, \omega) dB_t.$$

Proof.

- Let $Y \in \mathcal{L}^2(\mathcal{F}_T)$. Then $\exists \phi_k$ of the form (+) such that $\phi_k \rightarrow Y$ in \mathcal{L}^2 .
- Without loss of generality, assume $\mathbb{E}[\phi_k] = \mathbb{E}[Y]$.
 (otherwise replace $\hat{\phi}_k = \phi_k + a$, $a = \mathbb{E}[Y] - \mathbb{E}[\phi_k] \in \mathbb{R}$).
 Then $\hat{\phi}_k$ clearly also has the form (+).
- Since $\phi_k \rightarrow Y$ in \mathcal{L}^2 , by [2.19], the ϕ_k are Cauchy.
- Furthermore,
$$\phi_k = \mathbb{E}[Y] + \int_0^T f_k(t, \omega) dB_t \quad \forall k.$$

- [We now show that $f_k \rightarrow f^*$ in $\mathcal{L}^2[0, T]$, and that
$$Y = \mathbb{E}[Y] + \int_0^T f^* dB_t$$
]
- KEY IDEA

- Well, (ϕ_k) Cauchy $\Rightarrow \forall \epsilon > 0 \exists N: \forall k, j \geq N$
 $\|\phi_k - \phi_j\|_2^2 < \epsilon.$

- By ITO'S ISOMETRY

$$\mathbb{E}[(\phi_k - \phi_j)^2] = \mathbb{E}\left[\int_0^T (f_k - f_j)^2 dt\right]$$

- This implies (f_k) Cauchy, and $\mathcal{L}^2(0, T)$ is complete,
 Thus $f_k \rightarrow f^* \in \mathcal{L}^2(0, T).$

... continued.

Left to verify is that this f^* actually does represent Y , i.e. that

$$Y(\omega) = \mathbb{E}[Y] + \int_0^T f(t, \omega) d\mathcal{B}_t(\omega).$$

Well

$$\begin{aligned} & \| Y - (\mathbb{E}[Y] + \int_0^T f^* d\mathcal{B}_t) \|_{\mathcal{L}^2(\mathcal{F}_T)} \\ & \leq \| Y - \phi_k \|_{\mathcal{L}^2(\mathcal{F}_T)} + \| \phi_k - (\mathbb{E}[Y] + \int_0^T f^* d\mathcal{B}_t) \|_{\mathcal{L}^2(\mathcal{F}_T)} \\ & = \| Y - \phi_k \|_{\mathcal{L}^2(\mathcal{F}_T)} + \| \int_0^T \phi_k - f^* d\mathcal{B}_t \|_{\mathcal{L}^2(\mathcal{F}_T)} \\ & \quad \quad \quad \downarrow \text{ITO ISOMETRY} \\ & = \| Y - \phi_k \|_{\mathcal{L}^2(\mathcal{F}_T)} + \| \phi_k - f^* \|_{\mathcal{L}^2(0, T)} \\ & \quad \quad \quad \xrightarrow{\text{as } k \rightarrow \infty} 0 \quad \quad \quad \xrightarrow{\text{as } k \rightarrow \infty} 0 \end{aligned}$$

a Let $v(t, B(t)) = w(t, B(t_1), \dots, B(t_{k-1}), B(t))$

By ITO's FORMULA, for $t \in [t_{k-1}, t_k]$

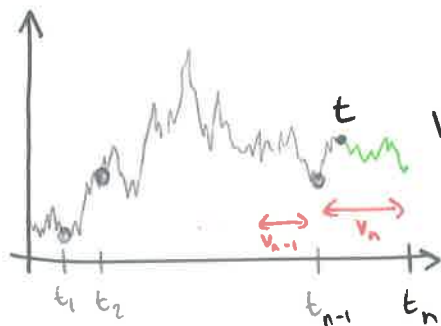
$$v(t, B(t)) = v(t, B(t_{k-1})) + \int_{t_{k-1}}^t \frac{\partial v}{\partial x}(s) dB_s + \int_{t_{k-1}}^t \left(\frac{\partial v}{\partial t}(s) + \frac{1}{2} \frac{\partial^2 v}{\partial x^2}(s) \right) ds.$$

That is,

$$w(t, B(t_1), \dots, B(t_{k-1}), B(t))$$

$$= w(t, B(t_1), \dots, B(t_k)) + \int_{t_{k-1}}^t \frac{\partial w}{\partial x_k}(s) dB_s + \int_{t_{k-1}}^t \frac{\partial w}{\partial t}(s) + \frac{1}{2} \frac{\partial^2 w}{\partial x_k^2}(s) ds.$$

b



$$V_n(t_n, x_1, \dots, x_n) = \phi(x_1, \dots, x_n)$$

For $t \in [t_{n-1}, t_n]$.

$$\frac{\partial V_n}{\partial t} + \frac{1}{2} \frac{\partial^2 V_n}{\partial x_n^2} = 0$$

(Preamble)

So think of $V_n(t, x_1, \dots, x_n) = \mathbb{E}[\phi(B_{t_1}, \dots, B_{t_n}) \mid B_{t_1} = x_1, \dots, B_{t_{n-1}} = x_{n-1} \text{ and } B_t = x_n \text{ currently}]$

And think of

$$t \in [t_{k-1}, t_k], V_k(t, x_1, \dots, x_k) = \mathbb{E}[V_{k+1}(t_k, x_1, \dots, x_k, x_k) \mid B_{t_1} = x_1, \dots, B_{t_k} = x_k]$$

etc.

$$= \mathbb{E}[\phi(B_{t_1}, \dots, B_{t_n}) \mid B_{t_1} = x_1, \dots, B_{t_k} = x_k].$$

So when $t \in [t_{k-1}, t_k]$, $V_k(t)$ is the "best guess" of $\phi(B_{t_1}, \dots, B_{t_n})$ given knowledge up to t .

4.17 b

Verify

$$V_k(\underline{t}, x_1, \dots, x_k) = \frac{1}{\sqrt{2\pi(t_k - t)}} \int_{\mathbb{R}} V_{k+1}(t_k, x_1, \dots, x_k, y) \exp\left[-\frac{(x_k - y)^2}{2(t_k - t)}\right] dy.$$

is the solution to

$$\left[\begin{array}{l} \text{I. } V_k(\underline{t}_k, x_1, \dots, x_k) = V_{k+1}(\underline{t}_k, x_1, \dots, x_k, x_k) \\ \text{II. } \frac{\partial V_k}{\partial t} + \frac{1}{2} \frac{\partial^2 V_k}{\partial x_k^2} \end{array} \right]$$

I.

Well, as $t \rightarrow t_k$, the RHS refers to integration over dirac at x_k : $\delta_{x_k}(y)$, hence I is satisfied.

II.

Note $p(t, x, y) = \frac{1}{\sqrt{2\pi t}} \exp\left(-\frac{(x-y)^2}{2t}\right)$ satisfies the

Fokker-Planck equation:

$$\left[p_t = \frac{1}{2} p_{xx} = \frac{1}{2} p_{yy} \right] \quad (\text{This is easily verified})$$

Hence $q(t, x, y) = p(t_k - t, x, y)$ satisfies

$$q_t = -\frac{1}{2} p_{xx} = -\frac{1}{2} p_{yy}, \quad t < t_k.$$

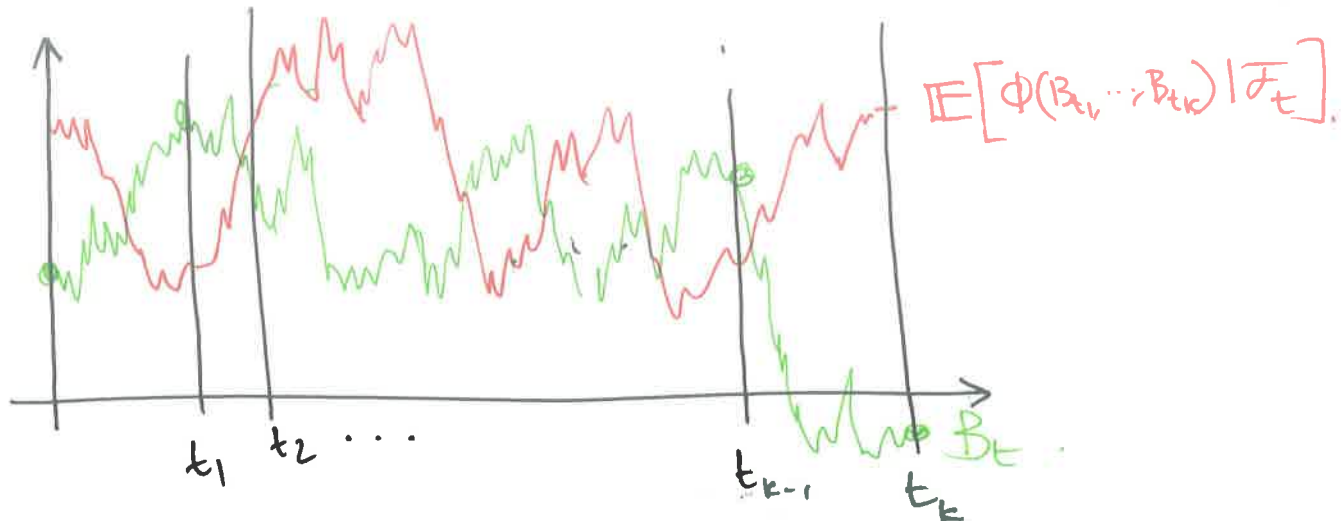
Now

$$f(t, u) := V_k(t, x_1, \dots, u) = \int_{\mathbb{R}} \psi(y) q(t, u, y) dy.$$

$$\text{where } \psi(y) := V_{k+1}(t_k, x_1, \dots, x_k, y)$$

Then

$$f_t = \int_{\mathbb{R}} \psi(y) q_t dy = \int_{\mathbb{R}} \psi(y) \left(-\frac{1}{2} q_{uu}\right) dy = -f_{uu}$$

4.17 \subseteq 

Let $f(t, \omega) = \frac{\partial v_k}{\partial x_k}(t, B(t_1), \dots, B(t_{k-1}), B(t))$.

for $t \in [t_{k-1}, t_k]$.

Show $f(t, \omega)$ satisfies the representation 4.3.15.

Well,

$$\begin{aligned}
 \Phi(B_{t_1}, \dots, B_{t_n}) &= v_n(t_n, B_{t_1}, \dots, B_{t_n}) \\
 &= v_n(t_{n-1}, B_{t_1}, \dots, B_{t_{n-1}}, B_{t_{n-1}}) \\
 &\quad + \int_{t_{n-1}}^{t_n} \frac{\partial v_n}{\partial x_n}(s, B_{t_1}, \dots, B_{t_{n-1}}, B_s) dB_s. \\
 &= v_{n-1}(t_{n-1}, B_{t_1}, \dots, B_{t_{n-1}}) \\
 &\quad + \int \dots dB_s \\
 &= v_{n-2}(t_{n-2}, B_{t_1}, \dots, B_{t_{n-2}}) + \int_{t_{n-1}}^{t_n} \frac{\partial v_n}{\partial x_n}(s, B_{t_1}, \dots, B_{t_{n-1}}, B_s) dB_s \\
 &\quad + \int_{t_{n-2}}^{t_{n-1}} \frac{\partial v_{n-1}}{\partial x_{n-1}}(s, B_{t_1}, \dots, B_{t_{n-1}}, B_s) dB_s.
 \end{aligned}$$

By induction.

$$\Phi(B_{t_1}, \dots, B_{t_n}) = V_0(0, x_0) + \sum_{j=1}^n \int_{t_{j-1}}^{t_j} \frac{\partial V}{\partial x_j}(s, B_{t_1}, \dots, B_{t_{j-1}}, B_s) dB_s,$$

where $V_0(0, x_0) = \mathbb{E}_{x_0}[\Phi(B_{t_1}, \dots, B_{t_n})]$ □

Now this question has a lot of Riddly notation in it, but the idea is quite simple, so let's see an example.

EXAMPLE

Let $\Phi(x, y) = x + y$. Let $0 < t_1 < t_2$.

Now for $t \in [0, t_1]$,

$$\begin{aligned} \mathbb{E}[\Phi(B_{t_1}, B_{t_2}) | \mathcal{F}_t] &= \mathbb{E}[B_{t_1} + B_{t_2} | \mathcal{F}_t] \\ &= 2B_t. \end{aligned}$$

For $t \in [t_1, t_2]$,

$$\mathbb{E}[\Phi(B_{t_1}, B_{t_2}) | \mathcal{F}_t] = B_{t_1} + B_t.$$

Indeed our function $f(t, \omega)$ expresses the rate of change of the expectation of $\Phi(B_{t_1}, B_{t_2})$ with changes in B_t .

Thus

