

P1

Check axioms.

$$i) \tilde{B}_0 = B_1 - B_1 = 0$$

$$ii) \Delta \tilde{B}_{t_k} = \tilde{B}_{t_{k+1}} - \tilde{B}_{t_k} = B_{t_{k+1}+1} - B_{t_k+1} \\ = \Delta B_{t_k+1} \sim N(0, t_{k+1} - t_k)$$

$$iii) 0 \leq t_1 < t_2 < \dots < t_n$$

$$(\tilde{B}_{t_1}, \Delta \tilde{B}_{t_1}, \dots, \Delta \tilde{B}_{t_{n-1}})$$

$$= (B_{t_1+1}, \Delta B_{t_1+1}, \dots, \Delta B_{t_{n-1}+1})$$

indep. R.V.'s.

P2

$$a) \int_0^T \phi_N dB_s := \sum_{j=0}^N \phi_N(t_j, \omega) \Delta B_{t_j} = \sum_{j=0}^N B_{t_j}^2 \Delta B_{t_j}$$

$$\text{where } \Delta B_{t_j} = B_{t_{j+1}} - B_{t_j}$$

$$\int_0^T B_s^2 dB_s := \lim_{\substack{N \rightarrow \infty \\ \max_k |\Delta t_k| \rightarrow 0}} \int_0^t \phi_N dB_s \text{ in } L^2(\Omega)^P$$

(Need $E \int_0^t |\phi_N - B_s^2|^2 ds \xrightarrow{N \rightarrow \infty} 0$ which is ok here since B_t^2 is t -cont. and \mathcal{F}_t -adapted)

P2 b):

Compute exp and var of

$$(i) X_t = \int_0^t s dB_s$$

$$(ii) Y = \int_0^T B_s^2 dB_s$$

Ito-int. have exp. 0

$$\Rightarrow E(X_t) = 0 = E(Y)$$

Use Ito-isometry:

$$\text{Var}(X_t) = E|X_t|^2 = E\left|\int_0^t s dB_s\right|^2$$

$$\stackrel{\text{Ito-isom}}{=} \int_0^t E(s^2) ds = \underline{\underline{\frac{1}{3}t^3}}$$

$$\text{Var}(Y) = E|Y|^2 = E\left|\int_0^T B_s^2 dB_s\right|^2$$

$$\stackrel{\text{Ito-isom}}{=} \int_0^T \underbrace{E(B_s^4)}_{=3s^2} ds = \underline{\underline{\frac{1}{3}T^3}}$$

List of formulas

3)

P2 c)

Let

$$\varphi = \begin{cases} 0 \\ B_1^2(\omega) X_{[1, \infty)}(t) \end{cases}$$

Show by a direct argument that

$$X_t = \int_0^t \varphi(\omega, s) dB_s$$

is a mart. w. r. t. filtr. $\{\mathcal{F}_t\}$ of B.M. B_t .

$$X_t = \begin{cases} 0 & 0 \leq t \leq 1 \\ B_1^2(\omega)(B_t - B_1) & t \geq 1 \end{cases}$$

i) X_t \mathcal{F}_t -meas (since $\{\varphi, \mathcal{F}\} \subset \mathcal{F}_s \subset \mathcal{F}_t, s < t$)

$$ii) E|X_t| \leq E(B_1^2 | B_t - B_1|)$$

$$\stackrel{\text{indep.}}{=} E(B_1^2) E|B_t - B_1|$$

$$\stackrel{\text{incr.}}{\leq} E(B_1^2) \sqrt{E|B_t - B_1|^2} = 1 \cdot t < \infty$$

$$iii) E(X_t | \mathcal{F}_s) \stackrel{t > s}{=} E(X_s | \mathcal{F}_s) + E(X_t - X_s | \mathcal{F}_s)$$

$$= X_s + E(X_t - X_s | \mathcal{F}_s)$$

since X_s \mathcal{F}_s -meas. (i)

$$E[X_t - X_s] = \begin{cases} 0 & s, t < 1 \\ B_{1,t}^2 (B_t - B_1) & s < 1, t \geq 1 \\ B_1^2 (w) (B_t - B_s) & \text{otherwise} \end{cases}$$

(*)

$$E(0 | \mathcal{F}_s) = 0$$

$$E(B_{1,t}^2 (B_t - B_1) | \mathcal{F}_s)$$

$$\stackrel{s < 1}{=} \underbrace{E(B_{1,t}^2 (B_t - B_1) | \mathcal{F}_1)}_{\text{telescope}} | \mathcal{F}_s = 0$$

$$\begin{aligned} & \rightarrow B_{1,t}^2 \underbrace{E(B_t - B_1 | \mathcal{F}_1)}_{\mathcal{F}_1\text{-meas.}} \\ & = E(B_t - B_1) = 0 \\ & \quad \uparrow \\ & \quad \text{indep. of } \mathcal{F}_1 \end{aligned}$$

$$E(B_{1,t}^2 (B_t - B_s) | \mathcal{F}_s)$$

$$\begin{aligned} & B_{1,t}^2 \underbrace{E(B_t - B_s | \mathcal{F}_s)}_{\mathcal{F}_s\text{-meas.}} = 0 \\ & \stackrel{s \geq 1}{=} E(B_t - B_s) = 0 \\ & \quad \uparrow \\ & \quad \text{indep. of } \mathcal{F}_s \end{aligned}$$

Hence X_t is \mathcal{F}_t mart.

5)

P3 a)

$$\ddot{X} + \dot{X} = W + \sin t$$

$$\Updownarrow$$

$$\begin{cases} \dot{X}_1 = X_2 \\ \dot{X}_2 = -X_2 + W + \sin t \end{cases}$$

↓ I to $\dot{X}_i = \frac{dX_i}{dt}$, " $W = \frac{dB_t}{dt}$ ", mult. by dt

$$\begin{cases} dX_1 = X_2 dt \\ dX_2 = -X_2 dt + dB_t + \sin t dt \end{cases}$$

or

$$\begin{cases} X_1(t) = x_1 + \int_0^t X_2(t) dt \\ X_2(t) = x_2 - \int_0^t X_2(t) dt + \int_0^t dB_s + \int_0^t \sin t dt \end{cases}$$

(e)

P3 b)

Brownian motion w. ^{outside} periodic forcing

$$m \ddot{X} + k\dot{X} = f_1 W + f_2 \sin t$$

$$dX_1 = X_2 dt$$

$$dX_2 = -X_2 dt + dB_t + \sin t$$

2nd eq'n uncoupled, solve it using int. factor.

Homogeneous eq'n,

$$dX_2 = -X_2 dt,$$

has int. factor e^t :

$$\begin{aligned} d(e^t X_2) &\stackrel{Ito}{=} e^t X_2 dt + e^t dX_2 + 0 \\ &= 0 \end{aligned}$$

Let $f(x,t) = x e^t$. Then, f solves 2nd eq'n

(here $f_t = a x e^t$, $f_x = e^t$, $f_{xx} = 0$)

($f \in C^2$) and find that

$$f(x, t) = f(x, 0) + \int_0^t f_t(x, s) ds$$

7.)

If X_2 sol'n (and hence Ito proc.), and since f is C^2 , we may use Ito's formula:

$$\begin{aligned} df(X_2, t) &= f_t dt + f_x dX_2 + \frac{1}{2} \varepsilon^2 f_{xx} \\ &= X_2 e^t dt + e^t (-X_2 dt + dB_t + \sin t dt) + 0 \\ &= e^t (dB_t + \sin t dt) \end{aligned}$$

$$\Rightarrow f(X_2(t), t) = f(X_2(0), 0) + \int_0^t e^s dB_s + \int_0^t e^s \sin s ds$$

$$\Rightarrow e^t X_2(t) = x_2 + \int_0^t e^s dB_s + \int_0^t e^s \sin s ds$$

$$X_2(t) = e^{-t} x_2 + \int_0^t e^{-(t-s)} dB_s + \int_0^t e^{-(t-s)} \sin s ds$$

p4

X_t str. sol'n if

$$i) X_t \text{ Ito proc. } \begin{cases} X_t \text{ } \mathcal{F}_t\text{-adapt., jointly meas.} \\ \int_0^t |b(s, X_s)| + |\sigma(s, X_s)|^2 ds < \infty \text{ a.s. } \forall t > 0 \end{cases}$$

ii) (1) holds a.s.

Sufficient cond'n for ex. and uniqueness =

$$(A1) |b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq L|x - y|$$

for all t, x, y and

$$(A2) z \in L^2(\Omega), z \text{ indep. of } \mathcal{F}_\infty = \mathcal{F}_{\{B_s: 0 \leq s < \infty\}}$$

P5

The eq'n may be written as

$$(*) \begin{cases} u_t = A u & , t > 0 \\ u = f & , t = 0 \end{cases}$$

where

$$A = \frac{1}{2} \partial_x^2 + \partial_x + y \partial_y$$

is the generator of the following Ito diffusion

$$\begin{cases} dX_t = dt + dB_t \\ dY_t = Y_t dt \end{cases} \Rightarrow \begin{cases} X_t^x = x + t + B_t \\ Y_t^y = y e^t \end{cases}$$

Since $f \in C_c^2$,

Eq'n (*) is then the "backward" Kolmogorov

eq'n satisfied by

$$u(t, x, y) = E(f(X_t^x, Y_t^y))$$

$$= \int_{\mathbb{R}} f(x+t+z, y e^t) \frac{1}{\sqrt{2\pi t}} e^{-\frac{z^2}{2t}} dz,$$

where we used the prob. density func'n of $B_t \sim N(0, t)$ given in formulas at the end of the exam.

P6

The generator of X_t^x is

$$A = \frac{1}{2} (2 + \sin x)^2 \partial_x^2$$

If $f \in C_c^2$ and τ st. f. w. $E(\tau) < \infty$,
then Dynkin's formula gives that

$$f(X_\tau^x) = f(x) + E \int_0^{\bar{\tau}} A f(X_s^x) ds$$

Take $f \in C_c^2$ s.t. $f_k = x^2$ for $|x| \leq R$

and let $\bar{\tau}_k = \min(\bar{\tau}, k)$.

Note that τ st. f. and $E(\tau) \leq k$.

By Dynkin

$$\begin{aligned} |X_{\bar{\tau}_k}^x|^2 &= |x|^2 + E \int_0^{\bar{\tau}_k} \underbrace{(2 + \sin X_s^x)^2}_{\geq 1} ds \\ &\geq |x|^2 + 1 \cdot E(\bar{\tau}_k) \end{aligned}$$

$$\Rightarrow E(\bar{\tau}_k) \leq |X_{\bar{\tau}_k}^x|^2 - |x|^2 \leq R^2 - |x|^2 < \infty$$

By monotone conv. then

$$E(\tau) = \lim E(\bar{\tau}_k) \leq R^2 - |x|^2 < \infty$$

P7

Let $f(x, y) = |x - y|^2$, then $f \in C^2$ and

$$f_x = (x - y), \quad f_y = -(x - y), \quad f_{xx} = 1$$

Then $f(X_t, X_t^\varepsilon)$ is an Ito proc. and

by the Ito formula

$$\begin{aligned} df(X_t, X_t^\varepsilon) &= f_x dX_t + f_y dX_t^\varepsilon \\ &\quad + \frac{1}{2} f_{xx} \varepsilon^2 dt + \frac{1}{2} f_{yy} 0 dt + f_{xy} 0 dt \end{aligned}$$

$$= (X_t - X_t^\varepsilon) (b(X_t) dt + \varepsilon dB_t)$$

$$- (X_t - X_t^\varepsilon) b(X_t^\varepsilon) + \frac{1}{2} \varepsilon^2 dt$$

$$= (X_t - X_t^\varepsilon) (b(X_t) - b(X_t^\varepsilon)) dt$$

$$+ \frac{1}{2} \varepsilon^2 dt + \varepsilon dB_t$$

Hence

$$|X_t - X_t^\varepsilon|^2 = \underbrace{|x_0 - x_0|^2}_{=0} + \int (\dots) dt + \frac{1}{2} \varepsilon^2 t + \varepsilon B_t$$

$$\leq \int_0^t L |X_s - X_s^\varepsilon|^2 ds + \frac{1}{2} \varepsilon^2 t + \varepsilon B_t$$

since $(X_t - X_t^\varepsilon) (b(X_t) - b(X_t^\varepsilon)) \leq |X_t - X_t^\varepsilon| \cdot L |X_t - X_t^\varepsilon|$

12)

Taking exp. and using Fubini,

$$\underbrace{E|X_t - X_t^\varepsilon|^2}_{v(t)} \leq L \int_0^t \underbrace{E|X_s - X_s^\varepsilon|^2}_{v(s)} ds + \frac{1}{2} \varepsilon^2 t$$

and then using Grönwall (list of formulas),

$$\underline{\underline{v(t) \leq \frac{1}{2} \varepsilon^2 t e^{Lt}}}$$

Alternative: Use SDE directly

$$\begin{aligned} \underbrace{E|X_t - X_t^\varepsilon|^2}_{v(t)} &\stackrel{(2),(3)}{=} E \left| x_0 + \int_0^t b(X_s) ds - x_0 - \int_0^t b(X_s^\varepsilon) ds + \varepsilon B_t \right|^2 \\ &\stackrel{(a+b)^2 \leq 2a^2 + 2b^2}{\leq} 2E \left| \int_0^t (b(X_s) - b(X_s^\varepsilon)) ds \right|^2 + 2E |\varepsilon B_t|^2 \end{aligned}$$

$$\leq 2E \left| \int_0^t L |X_s - X_s^\varepsilon| ds \right|^2 + 2\varepsilon^2 t$$

$$\stackrel{\text{Jensen}}{\leq} 2E t \int_0^t L^2 |X_s - X_s^\varepsilon|^2 ds + 2\varepsilon^2 t$$

$$= 2t L^2 \int_0^t v(s) ds + 2\varepsilon^2 t$$

$$\stackrel{\text{Grönwall}}{\Rightarrow} v(t) \leq 2\varepsilon^2 t e^{2L^2 t}$$