

4.1 Straightforward applications
of Itô's formula.

$$4.2 \begin{pmatrix} U(t) \\ V(t) \end{pmatrix} = \begin{pmatrix} \cos B(t) \\ \sin B(t) \end{pmatrix}$$

by applying Itô's formula one
gets

$$d \begin{pmatrix} U(t) \\ V(t) \end{pmatrix} = -\frac{1}{2} \begin{pmatrix} U(t) \\ V(t) \end{pmatrix} dt + \begin{pmatrix} -V(t) \\ U(t) \end{pmatrix} dB(t).$$

4.4

$$a) X(t) = X(0) e^{-\gamma t} + a \int_0^t e^{-\gamma(t-s)} dB(s)$$

$$b) E(X|t) = x_0 e^{-\gamma t}$$

let $0 \leq s \leq t < \infty$:

$$\text{Cov}(X(t), X(s)) =$$

$$a^2 E \left(\int_0^t e^{-\gamma(t-u)} dB(u) \cdot \int_0^s e^{-\gamma(s-v)} dB(v) \right) =$$

$$a^2 \int_0^s e^{-\gamma(t+s-2u)} du =$$

$$\frac{a^2}{2\gamma} \left(e^{-\gamma|t-s|} - e^{-\gamma(t+s)} \right)$$

$$\textcircled{c} \quad E\left(Y_{k+1}^{(n)} - Y_k^{(n)} \mid Y_k^{(n)} = y\right) = -\frac{2}{n} \left(y - \frac{n}{2}\right)$$

$$\text{Var}\left(Y_{k+1}^{(n)} - Y_k^{(n)} \mid Y_k^{(n)} = y\right) = 1 - \frac{4}{n^2} \left(y - \frac{n}{2}\right)^2$$

May think about ω :

$$Y_{k+1}^{(n)} - Y_k^{(n)} = -\frac{2}{n} \left(Y_k^{(n)} - \frac{n}{2}\right) + \underbrace{\epsilon_{k+1}^{(n)}}_{\uparrow}$$

+ Small error ~~$\epsilon_{k+1}^{(n)}$~~ \rightarrow random variable independent of the past

$$E\left(\underbrace{\epsilon_{k+1}^{(n)}}_{\epsilon_{k+1}^{(n)}}\right) = 0$$

$$\text{Var}\left(\underbrace{\epsilon_{k+1}^{(n)}}_{\epsilon_{k+1}^{(n)}}\right) = 1$$

Now, let $X^{(n)}(t) := \frac{1}{\sqrt{n}} \left(Y_{\lfloor nt \rfloor}^{(n)} - \frac{t}{2} \right)$

$$dt \approx \frac{1}{N}$$

to get

$$dX^{(n)}(t) \approx -2X^{(n)}(t) dt + dB(t) \dots$$

4.5

$$a) Af(x) = \beta f'(x) + \frac{\alpha^2}{2} x^2 f''(x)$$

$$b) f: \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R} \quad f = f(u, x)$$

$$Af(u, x) = \frac{\partial f}{\partial u}(u, x) - \gamma x \frac{\partial f}{\partial x}(u, x) + \frac{\alpha^2}{2} \frac{\partial^2 f}{\partial x^2}(u, x)$$

$$c) f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$Af(x_1, x_2) = \frac{\partial f}{\partial x_1}(x_1, x_2) + x_2 \frac{\partial f}{\partial x_2}(x_1, x_2) + \frac{1}{2} e^{2x_1} \frac{\partial^2 f}{\partial x_2^2}(x_1, x_2)$$

$$d) f: \mathbb{R}^2 \rightarrow \mathbb{R}, \quad f = f(x_1, x_2)$$

$$Af(x_1, x_2) = \frac{\partial f}{\partial x_1}(x_1, x_2) + \frac{1}{2} \frac{\partial^2 f}{\partial x_1^2}(x_1, x_2) + \frac{\Delta^2}{2} \frac{\partial^2 f}{\partial x_2^2}(x_1, x_2)$$

4.6

$$a) dX(t) = dt + \sqrt{2} dB(t)$$

$$b) dY(t) = \begin{pmatrix} dt \\ (cX(t))dt + \alpha X(t)dB(t) \end{pmatrix}$$

4.8

$$x > 0$$

$$Af(x) = \frac{\delta-1}{2x} f'(x) + \frac{1}{2} f''(x)$$

$Af(x) = 0$ general soln:

$$f(x) = \begin{cases} C_1 + C_2 x^{2-\delta} & \delta \neq 2 \\ C_1 + C_2 \ln x & \delta = 2 \end{cases}$$

(b) $0 < r < x < R < \infty$

$$P_x(\tau_r < \tau_R) =$$

$$\left\{ \frac{x^{2-\delta} - R^{2-\delta}}{r^{2-\delta} - R^{2-\delta}} \right. \quad \text{if } \delta \neq 2$$

$$\left. \frac{\ln x - \ln R}{\ln r - \ln R} \right\} \quad \text{if } \delta = 2$$

c) $\delta > 2$

$$\lim_{R \rightarrow \infty} P(\tau_r < \tau_R) = \left(\frac{r}{x}\right)^{\delta-2} < 1$$

$$\lim_{r \rightarrow 0} \lim_{R \rightarrow \infty} P(\tau_r < \tau_R) = 0$$



in this order!!!

d) $\delta = 2$

$$\lim_{R \rightarrow \infty} P_x(\tau_r < \tau_R) = 1$$

$$\lim_{r \rightarrow 0} P_x(\tau_r < \tau_R) = 0$$

e) $\delta < 2$

$$\lim_{r \rightarrow 0} P_x(\tau_r < \tau_R) = 1 - \left(\frac{x}{R}\right)^{2-\delta}$$

$$\lim_{R \rightarrow \infty} \lim_{r \rightarrow 0} P_x(\tau_r < \tau_R) = 1.$$

4.9

The process in question is:

$$dX(t) = \alpha X(t) dt + \beta X(t) dB(t)$$

$$X(0) = x$$

i.e. geometric Brownian Motion:

$$X(t) = x \exp \left\{ \beta B(t) + \left(\alpha - \frac{\beta^2}{2} \right) t \right\}$$

then apply Ito +
distribution of BM.

4.10 Assume first that g is discrete
with partition elements $(G_k)_{k=1,2,\dots}$

Then

$$E_Q(X|G_k) = \frac{\int_{G_k} X(\omega) dQ(\omega)}{\int_{G_k} dQ(\omega)} =$$

$$= \frac{\int_{G_k} X(\omega) g(\omega) dP(\omega)}{\int_{G_k} g(\omega) dP(\omega)} \cdot \frac{P(G_k)}{P(G_k)}$$

$$= \frac{E_P(X \cdot g | G_k)}{E_P(g | G_k)}$$

General case (Conditional expectation a la
Kolmogorov)

let $Z \in \mathcal{L}^\infty(\Omega, \mathcal{G}, \mathbb{P})$ (bdd,
 \mathcal{G} -measurable)

$$\mathbb{E}_Q(Z \cdot \mathbb{E}_Q(X|\mathcal{G})) = \mathbb{E}_Q(ZX)$$

by def. of conditional expectation

$$= \mathbb{E}_P(gZ \cdot X)$$

$$\mathbb{E}_Q\left(Z \cdot \frac{\mathbb{E}_P(gX|\mathcal{G})}{\mathbb{E}_P(g|\mathcal{G})}\right) = \mathbb{E}_P\left(gZ \frac{\mathbb{E}_P(gX|\mathcal{G})}{\mathbb{E}_P(g|\mathcal{G})}\right)$$

$$= \mathbb{E}_P\left(g \frac{\mathbb{E}_P(gZX|\mathcal{G})}{\mathbb{E}_P(g|\mathcal{G})}\right) =$$

$$= \mathbb{E}_P\left(\mathbb{E}_P(g|\mathcal{G}) \cdot \frac{\mathbb{E}_P(gZX|\mathcal{G})}{\mathbb{E}_P(g|\mathcal{G})}\right) =$$

$$= \mathbb{E}_P(\mathbb{E}_P(gZX|\mathcal{G})) = \mathbb{E}_P(gZX) \checkmark$$

$$4.11 \quad \Omega_n = \{0, 1\}^n \quad T=0, H=1$$

$$P_n(\omega_1 \omega_2 \dots \omega_n) = \left(\frac{2}{3}\right)^{\sum_{k=1}^n \omega_k} \left(\frac{1}{3}\right)^{\sum_{k=1}^n (1-\omega_k)}$$

$$Q_n(\omega_1 \omega_2 \dots \omega_n) = \left(\frac{1}{2}\right)^n$$

$$\frac{dQ_n}{dP_n}(\underline{\omega}) = Z_n(\underline{\omega}) = \left(\frac{3}{2}\right)^n \cdot \left(2\right)^{-\sum_{k=1}^n \omega_k}$$

$$Z_{m+1} = Z_m \cdot \frac{3}{2} \cdot 2^{-\omega_m}$$

(a) given (Z_1, Z_2, \dots, Z_m) , $Z_{m+1} = \begin{cases} Z_m \cdot \frac{3}{2} & \text{with prob. } \frac{1}{3} \\ Z_m \cdot \frac{3}{4} & \text{with prob. } \frac{2}{3} \end{cases}$
 under P

(b) $E_Q(X | \mathcal{F}_2) = \omega_1 + \omega_2 + \frac{1}{2}$

~~$$E_P(X | \mathcal{F}_2) = E_P(\omega_1 + \omega_2 + \frac{1}{2} | \mathcal{F}_2)$$~~

$$E_P\left(X \cdot \frac{Z_3}{Z_2} \mid \mathcal{F}_2\right) = E_P\left((\omega_1 + \omega_2 + \omega_3) \cdot \frac{3}{2} \cdot 2^{-\omega_3} \mid \mathcal{F}_2\right)$$

$$= \frac{3}{2} (\omega_1 + \omega_2) \underbrace{E_p(2^{-\omega_3} | \mathcal{F}_2)}_{= 2/3} +$$

$$\frac{3}{2} \underbrace{E_p(\omega_3 2^{-\omega_3} | \mathcal{F}_2)}_{= 1/3}$$

$$= \omega_1 + \omega_2 + \frac{1}{2} \quad \checkmark$$

© ... interpretation ...

4.12 (a) Follows from Girsanov

$$X(t) := F(t) + B(t)$$

$$E\left(\Phi(X(u): 0 \leq u \leq T)\right) =$$

$$E\left(\Phi(B(u): 0 \leq u \leq T) e^{\int_0^T f(u) dB(u) - \frac{1}{2} \int_0^T |f(u)|^2 du}\right)$$

(b) if $\int_0^T |f(u)|^2 = \infty$ then the R.N derivative doesn't exist

$$4.14 \quad Y(t) = B(t) + t$$

$$M(t) = \exp \left\{ -B(t) - \frac{t}{2} \right\}$$

$$d\mathbb{Q}_t = M(t) d\mathbb{P}_t$$

$t \mapsto Y(t)$ - is drifted BM
under \mathbb{P}

- is standard BM
under \mathbb{Q}

\mathbb{Q}_∞ and \mathbb{P}_∞ are mutually singular

$$4.15 \quad \mathbb{P}(X(t) > M) > 0$$



$$\mathbb{Q}(X(t) > M) > 0$$

$$\parallel$$

$$\mathbb{P}(B(t) > M)$$