

4.12)

$$(a) \quad M_t = tB_t - \frac{1}{3} B_t^3$$

$$E(M_t | F_s) = Z \quad \left\{ \begin{array}{l} \text{Expectation of } M_t \text{ given} \\ \text{filtration } F_s \end{array} \right\}$$

$t > s$

$$B_t = B_s + (B_t - B_s)$$

$$B_t^3 = (B_s)^3 + 3B_s^2(B_t - B_s) + 3(B_t - B_s)^2 B_s + (B_t - B_s)^3$$

- ①

Taking expectation of $B_t^3 | F_s$

$$E(B_s^3 | F_s) + 3E(B_s^2(B_t - B_s) | F_s) + 3E(B_s(B_t - B_s)^2 | F_s) + E((B_t - B_s)^3 | F_s)$$

$$\downarrow$$

$$B_s^3 + 3B_s^2 \cdot 0 + 3 \cdot B_s(t-s) + 0$$

$$\left\{ \begin{array}{l} (B_t - B_s) \rightarrow \text{odd moments} \\ \text{are zero} \sim N(0, t-s) \\ (B_t - B_s)^2 = t-s \end{array} \right.$$

$$E(t.B_t | F_s) = t.E(B_t | F_s) = t.B_s$$

$$\begin{aligned} Z &= t.B_s - \frac{1}{3} (B_s^3 + 3B_s(t-s)) \\ &= t.B_s - \frac{1}{3} B_s^3 - B_s.t + B_s.s \\ &= B_s.s - \frac{1}{3} B_s^3 = X_s \end{aligned}$$

$$E(X_t | F_s) = X_s \Rightarrow \text{So a martingale.}$$

(b) By stopping time.

$$\text{First} \rightarrow P(a) = \frac{b}{a+b} \quad \text{and} \quad P(b) = \frac{a}{a+b}.$$

$$\text{Now } X_t = t.B_t - \frac{1}{3} B_t^3$$

$$X_0 = 0.B_0 - \frac{1}{3} B_0^3 = 0$$

$$E\left(Z B_Z - \frac{1}{3} B_Z^3\right) = 0$$

$$E(Z B_Z) = E\left(\frac{1}{3} B_Z^3\right)$$

$$\rightarrow = \frac{b \cdot a^3}{3(a+b)} + \frac{a \cdot (-b)^3}{3(a+b)}$$

$$= \frac{ab}{3} \left[\frac{a^2 - b^2}{a+b} \right] = \frac{ab(a-b)}{3}$$

$$(c) E\left(e^{ab\tau - a^2\tau/2}\right)$$

$$= e^{-a^2\tau/2} E\left[e^{ab\tau}\right] \quad \text{--- (1)}$$

Since $e^{-a^2\tau/2}$ is constant for any τ ,

$$= E\left(e^{ab\tau}\right) \Rightarrow \text{M.G.F of B.M.}$$

$$B_\tau \sim N(0, \tau); E\left(e^{aB_\tau}\right) =$$

$$e^{\frac{a^2\tau}{2}} \quad [\text{M.G.F of a Normal}]$$

So ① becomes

$$e^{-\frac{a^2 z}{2} + \frac{a^2 z}{2}} = 1$$

$$\text{So } E\left(e^{+a\beta z - \frac{a^2 z}{2}}\right) = 1$$

$$(iv) e^{\alpha\beta z - \frac{1}{2}\alpha^2 z}$$

$$= e^{\alpha\beta z} e^{-\frac{1}{2}\alpha^2 z}$$

$$= \left(1 + \alpha\beta z + \frac{\alpha^2 \beta^2}{2} z^2 + \frac{\alpha^3 \beta^3}{6} z^3 + O(\alpha^4) \right) \cdot \left(1 - \frac{1}{2}\alpha^2 z + \frac{1}{4}\alpha^4 z^2 + O(\alpha^6) \right)$$

Collecting terms upto α^3

$$1 + \alpha\beta z + \frac{\alpha^2}{2}(\beta^2 - z) + \alpha^3\left(\frac{1}{6}\beta^3 - \frac{1}{2}z\beta\right) + O(\alpha^4)$$

$$E(\beta_z) = 0, E(\beta_z^2) = z$$

and $E(\text{the whole term}) = 1$

$$\begin{aligned} E(1) + \cancel{\alpha E(\beta_z)} + \frac{\alpha^2}{2} (\cancel{E(\beta_z^2)} - z) \\ + \alpha^3 (E(\frac{1}{6} \beta_z^3 - \frac{1}{2} z \beta_z)) \\ = 1 \end{aligned}$$

$$E(\frac{1}{6} \beta_z^3 - \frac{1}{2} z \beta_z) = 0$$

$$E(z \beta_z) = E(\frac{1}{3} \beta_z^3)$$

Using that $E(\beta_z^3) = \frac{a^3 b}{a+b} - \frac{b^3 a}{a+b} = ab(a-b)$

$$E(z \beta_z) = \frac{ab(a-b)}{3}$$

$$4.11) \quad X_t = e^{\alpha N_t - \lambda t (e^{\alpha} - 1)}$$

for $t \geq s$

$$N_t = (N_t - N_s) + N_s$$

$$X_t = e^{\alpha (N_t - N_s) + \alpha N_s - \lambda t (e^{\alpha} - 1)}$$

$$E(X_t | \mathcal{F}_s) = E \left[e^{\alpha (N_t - N_s) + \alpha N_s - \lambda t (e^{\alpha} - 1)} \mid \mathcal{F}_s \right]$$

Now we know

$$N_t - N_s \approx \text{Poisson}(\lambda(t-s))$$

$N_t - N_s$ is independent of N_s

[Memorylessness of Poisson]

$$\Rightarrow e^{-\lambda t (e^{\alpha} - 1)} \cdot E[e^{\alpha (N_t - N_s)}] \cdot E[e^{\alpha N_s} \mid \mathcal{F}_s]$$

$\boxed{\hspace{10em}} \rightarrow \text{M.G.F of } N_t$

$$\begin{aligned} \text{MGF of } N &= e^{\lambda(e^t - 1)} \\ \Rightarrow e^{-\lambda t(e^t - 1)} e^{\lambda(t-s)(e^t - 1)} \dots e^{\lambda N_s} \\ \Rightarrow \boxed{e^{\lambda N_s} - \lambda s(e^t - 1)} &\Rightarrow E(s) \end{aligned}$$

Hence a martingale since $E(X_t | \mathcal{F}_s) = E(X_s)$
 and also it is integrable since it's a finite
 random variable with finite mean.

$$4.6) \quad C = \begin{bmatrix} 3/16 & 1/8 \\ 1/8 & 1/4 \end{bmatrix}$$

If it's positive semidefinite then it's not a degenerate random vector.

$$\text{Det}(C) = \frac{3}{16} \times \frac{1}{4} - \frac{1}{8} \times \frac{1}{8}$$

$$= \frac{2}{64} = \frac{1}{32} > 0$$

and all leading principal minors > 0 .

Hence it's not degenerate

$$(b) E(Y|X) = \left(\frac{8}{\text{Var}(X)} \right) X = \frac{1 \times 16}{8 \times 3} X$$

$$= \frac{2}{3} X$$

$$(c) Y = W + \frac{2}{3} X$$

$$\text{Cov}(Y, Y) = \text{Cov}(W, Y) + \frac{2}{3} \text{Cov}(Y, X)$$

$$\text{As } \text{Cov}(X, W) = 0$$

$$\frac{1}{4} = \text{Cov}(W, Y) + \frac{2}{3} \times \frac{1}{8}$$

$$\frac{1}{4} - \frac{1}{12} = \text{Cov}(W, Y)$$

$$\frac{1}{6} = \text{Cov}(W, Y)$$

$$E^0(Y) = E(W) + \frac{2}{3} E^0(X)$$

$$\boxed{E(W) = 0}$$

Now

$$\text{Cov}(W, Y) = \text{Cov}(W, W) + \text{Cov}(W, X) \rightarrow 0$$

$$W \sim N(0, \frac{1}{6})$$

$$E(Y|X) = E(W|X) + \frac{2}{3} E(X|X)$$

$$E(Y|X) = \underset{\substack{\downarrow \\ 0}}{E(W)} + \frac{2}{3} X$$

$$E(Y|X) = \frac{2}{3} X.$$

$$\text{Var}(Y|X) = \text{Var}(W) = \frac{1}{6}$$

$$\text{or } \text{PDF}(Y|X) \sim N\left(\frac{2}{3}X, \frac{1}{6}\right)$$

(d) Now

$$Y - \frac{2}{3}X = W \quad W \sim N(0, \frac{1}{6})$$

$$\sqrt{6} \left(Y - \frac{2}{3}X \right) = Z_1 \quad Z \sim N(0, 1) \\ = \sqrt{6}W$$

$$Z_2 = \frac{X}{\sqrt{\frac{3}{16}}} = \frac{4}{\sqrt{3}}X$$

Z_2, Z_1 are independent since
 W and X are independent,