

Q1:

a): $F = \int_0^T B^2(s) ds$.

define $f(t) = \int_0^t B^2(s) ds$

$df(t) = B^2(t) dt$

$E[F] = E\left[\int_0^T B^2(s) ds\right] = \int_0^T E[B^2(s)] ds$
 $= \int_0^T s ds = \frac{T^2}{2}$.

~~$d\left(\frac{B^3(t)}{3}\right) = \frac{B^3(t)}{3} dt - \frac{2}{3} B^2(t) dB(t) - \frac{1}{3} B(t) dt$~~

(Continued later)

b) $Y(t) = \log(X(t))$.

~~$X(t) = e^{Y(t)}$~~

$\Rightarrow dX(t) = e^{Y(t)} dY(t) + \frac{1}{2} e^{Y(t)} dY^2(t)$

$\Rightarrow e^{Y(t)} dY(t) + \frac{1}{2} e^{Y(t)} dY^2(t) = 2(3 - Y(t)) e^{Y(t)} dt + 4e^{Y(t)} dB(t)$

$\Rightarrow dY + \frac{1}{2} dY^2 = 2(3 - Y) dt + 4 dB(t)$

$e^{2t} Y = Z$

$Y = Ze^{-2t}$

$dY = e^{-2t} dZ - 2Ze^{-2t} dt + 0 \cdot \frac{1}{2} dt$

$\Rightarrow e^{-2t} dZ = 2Ze^{-2t} dt + \frac{1}{2} dZ^2 = 2dt - 2Ze^{-2t} dt + 4dB(t)$ (1)

$\Rightarrow Z(t) = Z(0) + \int_0^t 4e^{2s} dB(s)$

$\Rightarrow Y(t) = e^{-2t} Y(0) + (1 - e^{-2t}) + e^{-2t} \int_0^t 4e^{2s} dB(s)$

$\Rightarrow X(t) = \exp\left(\underbrace{\log(X(0))}_{-1} (1 - e^{-2t}) + e^{-2t} \int_0^t 4e^{2s} dB(s)\right)$

the factor of -1 comes from $\frac{dZ}{dZ}$ in dX since d only $4dB(t)$ contributes to it

$$Y(t) = \log(X(t))$$

$$= \frac{1}{4} - (1 - e^{-2t}) + e^{-2t} \int_0^t 4e^{2s} dB(s)$$

Ito integral \Rightarrow expectation =

$$\int_0^t 4e^{2s} dB(s) = 0$$

So $E[Y(t)] = \frac{1}{4} - (1 - e^{-2t})$

For distribution, note that

The Ito integral is integral of ~~continuous~~ deterministic function, so it is normally distributed

mean = 0

$$\text{variance} = \int_0^t (4e^{2s})^2 ds = \int_0^t 16e^{4s} ds = \frac{e^{4t} - 1}{4} \times 16 = 4(e^{4t} - 1)$$

So ~~mean of~~ $Y(t)$ is normally distributed with mean $\frac{1}{4} - (1 - e^{-2t})$ & variance $e^{-4t} \cdot (4(e^{4t} - 1)) = 4(1 - e^{-4t})$

$$\Rightarrow Y(t) \sim N\left(\frac{1}{4} - (1 - e^{-2t}), 4(1 - e^{-4t})\right)$$

Note: ~~we got Z~~ We got Z by guessing the form of $dZ = \alpha dt + \beta(t) dB(t)$ & solving for α & β since $dZ \cdot dZ = \beta^2(t) \cdot dt$ makes it easier

It is possible to ~~also~~ multiply eqn ① by dZ to get to the same conclusion.

Q2 (i) we can write it as (omitting the (t))

$$\begin{pmatrix} dY_1 \\ dY_2 \end{pmatrix} = \begin{pmatrix} 2 \\ \beta \end{pmatrix} dt + \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} dB_1 \\ dB_2 \\ dB_3 \end{pmatrix}$$

We find u_1, u_2, u_3 such that

$$\begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} = \begin{pmatrix} 2 \\ \beta \end{pmatrix}$$

$$\Rightarrow u_1 + 2u_2 + 3u_3 = 2$$

$$u_1 + 2u_2 + 2u_3 = \beta$$

$$\Rightarrow u_3 = 2 - \beta$$

$$\text{Now let } u_2 = \alpha$$

$$\text{So } u_1 + 2u_2 = 2 - 3(2 - \beta) = 3\beta - 4$$

$$\Rightarrow u_1 = 3\beta - 2\alpha - 4$$

Now consider

$$Z(t) = \exp \left(- \int_0^t ((3\beta - 2\alpha - 4) dB_1 + \alpha dB_2 + (2 - \beta) dB_3) - \int_0^t ((3\beta - 2\alpha - 4)^2 + \alpha^2 + (2 - \beta)^2) ds \right)$$

Varying α over \mathbb{R} generates an infinite family of stochastic processes Z .

Now by corollary to Girsanov's theorem, the ~~process~~ Q

$$dQ = Z dP$$

~~Q~~ satisfies

$$dY = \underbrace{\left(\begin{pmatrix} 2 \\ \beta \end{pmatrix} - \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} \right)}_{0} dt + \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} dB_1 \\ dB_2 \\ dB_3 \end{pmatrix}$$

$\Rightarrow Y$ is a martingale (since second term integrates to martingale as $\int dQ$ to integral it is)

$\Rightarrow Q$ is a martingale measure for $Y(t)$

& there are ∞ -ely many such Q , each generated by a different Y .

(ii)

$$dX = \frac{1}{3} X^{1/3} dt + X^{2/3} dB$$

$$Y = X^{1/3}$$

$$dY = \frac{1}{3} X^{-2/3} dX + \frac{1}{3} \cdot \left(-\frac{2}{3}\right) \times \frac{1}{2} \times \cancel{X^{1/3}} X^{-1/3} dB$$

$$= \frac{1}{3} X^{-2/3} \left(\frac{1}{3} X^{1/3} dt + X^{2/3} dB \right) + \cancel{\left(-\frac{1}{9}\right) X^{-1/3} dt}$$

$$= \frac{1}{3} dB$$



$$Y(0) = 3^{1/3}$$

$$Y(t) - Y(0) = \frac{B(t) - B(0)}{3} = \frac{B(t)}{3}$$

$$\Rightarrow Y(t) = 3^{1/3} + \frac{B(t)}{3}$$

$$\Rightarrow X(t) = \left(3^{1/3} + \frac{B(t)}{3} \right)^3$$

Conditions ^{in the} existence & uniqueness theorem for SDE:

given: $a(t, x) = \frac{1}{3} x^{1/3}$ $\sigma(t, x) = x^{2/3}$

Conditions: $\exists C, D$ such that

$$\left| \frac{1}{3} x^{1/3} \right| + |x^{2/3}| \leq C(1 + |x|)$$

$$\left| \frac{1}{3} (x^{1/3} - y^{1/3}) \right| + |x^{2/3} - y^{2/3}| \leq D|x - y|$$

Note that for arbitrarily small x, y , this inequality is false. For example, $x = 8\varepsilon^3$, $y = \varepsilon^3$

$$\left| \frac{1}{3} \varepsilon \right| + |3\varepsilon^2| \leq 7D\varepsilon^3 \rightarrow \text{false for small } \varepsilon$$

So, we can't use the existence & uniqueness theorem