

230Q: Problem Set #3

Professor Johan Walden

Group 20

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Solution

Part A

$$X(t) = e^{at}$$
$$dX = \alpha e^{\alpha t} dt$$

Part B

$$X(t) = \int_0^t g(s)dW(s)$$

$$dX = g(t)dW(t)$$

Part C

$$X(t) = e^{\alpha W(t)}$$

$$dX = \alpha e^{\alpha W(t)} dW + \frac{1}{2} \alpha^2 e^{\alpha W(t)} dt$$

Part D

$$\begin{split} X(t) &= e^{\alpha Y(t)} \\ dX &= \alpha e^{\alpha Y(t)} dY + \frac{1}{2} \alpha^2 e^{\alpha Y(t)} (dY)^2 \\ &= \alpha e^{\alpha Y(t)} (\mu dt + \sigma dW) + \frac{1}{2} \alpha^2 e^{\alpha Y(t)} \cdot \sigma^2 dt \\ &= \alpha e^{\alpha Y(t)} (\mu + \frac{1}{2} \alpha \sigma^2) dt + \alpha e^{\alpha Y(t)} \sigma dW \end{split}$$

Part E

$$x(t) = y^{2}(t)$$

$$dx = 2ydy + (dy)^{2}$$

$$= 2y(\alpha ydt + \sigma ydW) + \sigma^{2}y^{2}dt$$

$$= y^{2}(2\alpha + \sigma^{2})dt + 2\sigma y^{2}dW$$

Part F

$$\begin{split} x(t) &= \frac{1}{y(t)} \\ dx &= -\frac{1}{y^2} dy + 2\frac{1}{y^3} (dy)^2 \\ &= -\frac{1}{y^2} (\alpha y dt + \sigma y dW) + \frac{2}{y^3} \cdot \sigma^2 y^2 dt \\ &= \frac{1}{y} \left(-\alpha + 2\sigma^2 \right) dt - \frac{\sigma}{y} dW \end{split}$$

Solution

First, we calculate the derivatives of X as follows

$$X(t) = \int_0^t \sigma_s dw_s \Rightarrow \begin{cases} dx = \sigma_t dw \\ (dx)^2 = \sigma_t^2 dt \end{cases}$$

Then, Let $Y(t) = e^{iuX(t)}$. We know Y(0) = 1.

Applying Ito lemma, we get

$$dY = iue^{iux(t)}dx + \frac{1}{2}(iu)^2 e^{iux(t)}(dx)^2$$
$$= iue^{iux(t)}\sigma_t dw - \frac{1}{2}u^2 e^{iux(t)}\sigma_t^2 dt$$

Take an integral on both sides

$$Y(t) - 1 = \int_0^t iue^{iux(t)} \sigma_e dW - \frac{1}{2} \int_0^t u^2 e^{iux(t)} \sigma_t^2 dt$$

Take the expectation of the formula. By the martingale properties, we know $\int_0^t iue^{iux(t)}\sigma_t dw = 0$. So, we have

$$E[Y(t)] = 1 - \frac{1}{2}u^2 \int_0^t \sigma^2(s) E\left[e^{iux(t)}\right] dt$$

Let $m_t = E[Y(t)] = E[e^{iux(t)}]$. We know that m(0) = 1

Then, solve the ODE question as below

$$\begin{split} m_t &= 1 - \frac{1}{2}u^2 \int_0^t \sigma^2(s) m_s ds \\ m_t' &= -\frac{1}{2}u^2 \sigma^2(t) m_t \\ m_t &= e^{-\frac{1}{2}u^2 \int_0^t \sigma^2(s) ds} \end{split}$$

So, we have

$$E\left[e^{iux(t)}\right] = e^{-\frac{u^2}{2}\int_0^t \sigma^2(s)ds}$$

Solution

$$dX = \alpha X dt + \sigma_t dW$$

$$\int_0^t dX = \int_0^t \alpha X dt + \int_0^t \sigma_t dw$$

$$E[X_t - X_0] = E\left[\int_0^t \alpha X_s ds\right] + E\left[\int_0^t \sigma_s dW_s\right]$$

$$E[X_t] = E[X_0] + E\left[\int_0^t \alpha X_s ds\right]$$

$$E[X_t] = E[X_0] + \int_0^t \alpha E[X_s] ds$$

Let $E[x_t] = m_t$, $m_0 = E[x_0]$. We solve the following ODE problem

$$m_{t} = E[X_{0}] + \int_{0}^{t} \alpha m_{s} dS$$
$$m'_{t} = \alpha m_{t} \Rightarrow m_{t} = E[X_{0}] \cdot e^{\alpha t}$$

To conclude, $E[X_t] = E[X_0] \cdot e^{\alpha t}$

Solution

By applying Taylor expansion, we have

$$\begin{split} dR &= 2XdX + 2YdY + \frac{1}{2} \left[2(dX)^2 + 2(dY)^2 + 0(dX)(dY) \right] \\ &= 2XdX + 2YdY + (dX)^2 + (dY)^2 \\ &= 2X(\alpha Xdt - YdW) + 2Y(\alpha Ydt + XdW) + Y^2dt + X^2dt \\ &= \left(2\alpha X^2 + 2\alpha Y^2 + X^2 + Y^2 \right) dt + (2XY - 2XY)dw \\ &= \left(X^2 + Y^2 \right) (2\alpha + 1)dt \end{split}$$

To conclude, we have $dR = (2\alpha + 1)Rdt$

Solution

Part A

First, calculate the derivatives of Y

$$Y = e^{-\alpha t} X_t, Y_t = -\alpha Y, Y_x = e^{-\alpha t}, Y_{xx} = 0$$

Then, apply Ito lemma

$$dY = -\alpha Y dt + e^{-\alpha t} dX$$

$$= -\alpha Y dt + e^{-\alpha t} (\alpha x dt + \sigma dw)$$

$$= (-\alpha Y + \alpha X e^{-\alpha t}) dt + \sigma e^{-\alpha t} dW$$

$$= \sigma e^{-\alpha t} dW$$

Part B

$$\int_0^t dY = \int_0^t \sigma e^{-\alpha s} dW_s$$

$$Y_t - Y_0 = \int_0^t \sigma e^{-\alpha s} dW_s$$

$$e^{-\alpha t} X_t - X_0 = \int_0^t \sigma e^{-\alpha s} dW_s$$

$$e^{-\alpha t} X_t = X_0 + \int_0^t \sigma e^{-\alpha s} dW_s$$

$$X_t = e^{\alpha t} x_0 + \sigma \int_0^t e^{\alpha (t-s)} dW_s$$

Part C

 $X(t) \mid \tilde{F}_0$ is normally distributed since each increment is independent and normally distributed. The increment comprises $\alpha X dt$, which is a constant. The increment also comprises σdW , which is independent normally distributed. So, $X(t) \mid \tilde{F}_0$ is normally distributed. For the expected value,

$$E[x(T) \mid x(0)] = E\left[e^{\alpha T}x_0\right] + E\left[\sigma \int_0^T e^{\alpha(t-s)}dW_s\right]$$
$$= e^{\alpha T}x_0 + \sigma e^{\alpha T}E\left[\int_0^T e^{-\alpha s}dW_s\right]$$
$$= e^{\alpha T}X_0$$

For the variance,

$$\operatorname{Var}[X(T) \mid X(0)] = \operatorname{Var}\left(e^{\alpha T} X_0 + \sigma \int_0^T e^{\alpha (T-s)} dW_s\right)$$

$$= \sigma^2 \operatorname{Var}\left(\int_0^T e^{\alpha (T-s)} dW_s\right)$$

$$= \sigma^2 \left(E\left[\left(\int_0^T e^{\alpha (T-s)} dW_s\right)^2\right] - E\left[\int_0^T e^{\alpha (T-s)} dW_s\right]^2\right)$$

$$= \sigma^2 \int_0^T E\left[e^{2\alpha (T-s)}\right] ds$$

$$= \sigma^2 \left(-\frac{1}{2\alpha}\right) \cdot e^{2\alpha (T-s)}\Big|_0^T$$

$$= -\frac{\sigma^2}{2\alpha} \left(1 - e^{2\alpha T}\right)$$

$$= \frac{\sigma^2}{2\alpha} \left(e^{2\alpha T} - 1\right)$$