

## Solutions 2

1.

$$\begin{aligned} E[dS] &= E[\mu dt] + E[\sigma dW] \\ &= \mu dt + \sigma E[dW] \\ &= \mu dt + \sigma \times 0 \end{aligned}$$

Recall  $E[dW] = 0$ ,  $E[dW^2] \rightarrow dt$  as  $dt \rightarrow 0$ . Then

$$\begin{aligned} \text{Var}[dS] &= E[dS^2] - E[dS]^2 \\ &= E[\mu^2 dt^2 + 2\mu\sigma dt dW + \sigma^2 dW^2] - (E[\mu dt + \sigma dW])^2 \\ &= \mu^2 dt^2 + \sigma^2 dt - \mu^2 dt \\ &= \sigma^2 dt \end{aligned}$$

$$\begin{aligned} E[dS^2] &= \mu^2 dt^2 + 2\mu\sigma dt E[dW] + \sigma^2 E[dW^2] \\ &= \mu^2 dt^2 + 0 + \sigma^2 dt \end{aligned}$$

Thus,

$$\text{Var}[dS] = \mu^2 dt^2 + \sigma^2 dt - \mu^2 dt^2 = \sigma^2 dt$$

as required.

2. Use Itô's Lemma.

(a)

$$\frac{\partial f}{\partial S} = A, \quad \frac{\partial^2 f}{\partial S^2} = 0 \quad \frac{\partial f}{\partial t} = 0$$

and so, from Itô's lemma

$$df = A\mu S dt + A\sigma S dW.$$

(b)

$$\frac{\partial f}{\partial S} = nS^{n-1}, \quad \frac{\partial^2 f}{\partial S^2} = n(n-1)S^{n-2} \quad \frac{\partial f}{\partial t} = 0$$

and so, from Itô's lemma

$$df = nS^n \left( \mu + \frac{1}{2}(n-1)\sigma^2 \right) dt + \sigma nS^n dW$$

and

$$df = n f \left[ \left( \mu + \frac{1}{2}(n-1)\sigma^2 \right) dt + \sigma dW \right]$$

3. In this case  $f(S) = \log S$ . In which case

$$\frac{\partial f}{\partial S} = \frac{1}{S}, \quad \frac{\partial^2 f}{\partial S^2} = -\frac{1}{S^2}, \quad \frac{\partial f}{\partial t} = 0$$

and so by Itô's lemma

$$df = \left(\mu - \frac{1}{2}\sigma^2\right)dt + \sigma dW.$$

This is just a generalised Brownian motion as in question 2 and as such it is Normally distributed with

$$E[df](= E[d(\log S)]) = \left(\mu - \frac{1}{2}\sigma^2\right)dt$$

and

$$\text{Var}[df](= \text{Var}[d(\log S)]) = \sigma^2 dt$$

Thus the log of  $S$  is Normally distributed, or  $S$  is *lognormally* distributed, with mean  $(\mu - \frac{1}{2}\sigma^2)t$  and variance  $\sigma^2 t$ .

4. Performing a 'Taylor' expansion in both  $S$  and  $t$  we have

$$\begin{aligned} f(S + dS, t + dt) - f(S, t) &= df = \frac{\partial f}{\partial S}dS + \frac{\partial f}{\partial t}dt + \frac{1}{2}\frac{\partial^2 f}{\partial S^2}dS^2 + \frac{1}{2}\frac{\partial^2 f}{\partial t^2}dt^2 + \frac{\partial^2 f}{\partial S\partial t}dSdt \\ &\quad + O(dS^3) + O(dt^3) \end{aligned}$$

substituting in the expression for  $dS$  (writing  $A = A(S, t)$  and  $B = B(S, t)$ ) we obtain

$$\begin{aligned} df &= \frac{\partial f}{\partial S}[Adt + BdW] + \frac{\partial f}{\partial t}dt + \frac{1}{2}\frac{\partial^2 f}{\partial S^2}[A^2dt^2 + 2ABdtdW + B^2dW^2] \\ &\quad + \frac{1}{2}\frac{\partial^2 f}{\partial t^2}dt^2 + \frac{\partial^2 f}{\partial S\partial t}[Adt^2 + BdWdt] + O(dt^3) + O(dS^3) \end{aligned}$$

Now, as  $dt \rightarrow 0$  then  $dW^2 \rightarrow dt$  and  $dWdt = o(dt)$  leaving

$$df = \left[ A(S, t)\frac{\partial f}{\partial S} + \frac{\partial f}{\partial t} + \frac{1}{2}B^2(S, t)\frac{\partial^2 f}{\partial S^2} \right] dt + B(S, t)\frac{\partial f}{\partial S}dW$$

as required.

5. Performing a 'Taylor' expansion in  $S_1, \dots, S_n$  gives

$$\begin{aligned} df &= \frac{\partial f}{\partial S_1}dS_1 + \dots + \frac{\partial f}{\partial S_n}dS_n + \frac{1}{2}\frac{\partial^2 f}{\partial S_1^2}dS_1^2 + \dots + \frac{1}{2}\frac{\partial^2 f}{\partial S_n^2}dS_n^2 + \frac{\partial^2 f}{\partial S_1\partial S_2}dS_1dS_2 \\ &\quad + \frac{\partial^2 f}{\partial S_1\partial S_3}dS_1dS_3 + \dots + \frac{\partial^2 f}{\partial S_{n-1}\partial S_n}dS_{n-1}dS_n + O(S_1^3) + \dots + O(S_n^3) \end{aligned}$$

Now using the rules given and the expressions for  $dS_i$  this reduces, eventually, to

$$df = \left[ \sum_i \mu_i S_i \frac{\partial f}{\partial S_i} + \frac{1}{2} \sum_i \sigma_i^2 S_i^2 \frac{\partial^2 f}{\partial S_i^2} + \frac{1}{2} \sum_i \sum_{j \neq i} \frac{\partial^2 f}{\partial S_i \partial S_j} \sigma_i \sigma_j S_i S_j \rho_{ij} \right] dt + \sum_i \sigma_i S_i \frac{\partial f}{\partial S_i} dW_i$$

6. (a) The process  $Y$  is given by

$$f(X, t) = e^{\kappa t} X$$

then

$$\frac{\partial f}{\partial t} = \kappa Y$$

$$\frac{\partial f}{\partial X} = e^{\kappa t}$$

$$\frac{\partial^2 f}{\partial X^2} = 0$$

so

$$df = [\kappa Y + \kappa(\theta - X)e^{\kappa t}]dt + \sigma e^{\kappa t} dW$$

(b)

$$df = \kappa\theta e^{\kappa t} dt + \sigma e^{\kappa t} dW$$

(c)

$$\int_0^T df = \int_0^T \kappa\theta e^{\kappa t} dt + \int_0^T \sigma e^{\kappa t} dW$$

we get

$$f_T - f_0 = [\theta(e^{\kappa T} - 1)] + \sigma \int_0^T e^{\kappa t} dW$$

and

$$e^{\kappa T} X_T - X_0 = [\theta(e^{\kappa T} - 1)] + \int_0^T \sigma e^{\kappa t} dW$$

$$X_T = X_0 e^{-\kappa T} + [\theta(1 - e^{-\kappa T})] + \int_0^T \sigma e^{-\kappa(T-t)} dW_t$$

(d)

$$E[X_T] = X_0 e^{-\kappa T} + [\theta(1 - e^{-\kappa T})]$$

$$\text{var}[X_T] = \text{var} \left[ \int_0^T \sigma e^{-\kappa(T-t)} dW_t \right]$$

$$= E \left[ \left( \int_0^T \sigma e^{-\kappa(T-t)} dW_t \right)^2 \right]$$

$$= \int_0^T \sigma^2 e^{-2\kappa(T-t)} dt$$

$$= \frac{\sigma^2}{2\kappa} [1 - e^{-2\kappa T}]$$