## AFM Brief Solution to Assignment 3

1. A stochastic process is generated by the equation

$$dX_t = \mu dt + \sigma dW_t.$$

with the initial condition  $X_0 = 1$ . Which equation governs the process  $Y(t, X_t) = (1 + t)^2 e^{aX_t}$ ? Here a is a constant.

**Solution:** According to Ito's lemma, we have

$$dY_t = \left[ \frac{\partial Y}{\partial t} + \mu \frac{\partial Y}{\partial X_t} + \frac{1}{2} \sigma^2 \frac{\partial^2 Y}{\partial X_t^2} \right] dt + \sigma \frac{\partial Y}{\partial X_t} dW_t.$$

As

$$\frac{\partial Y}{\partial t} = \frac{2Y}{1+t}, \ \frac{\partial Y}{\partial X_t} = a(1+t)^2 e^{aX_t} = aY, \ \frac{\partial^2 Y}{\partial X_t^2} = a^2(1+t)^2 e^{aX_t} = a^2Y,$$

we have

$$dY_t = \left\lceil \frac{2}{1+t} + \mu a + \frac{1}{2}\sigma^2 a^2 \right\rceil Y_t dt + \sigma a Y_t dW_t.$$

with the initial condition  $Y_0 = e^a$ 

2. Consider a stochastic process  $X_t$  governed by the equation

$$dX_t = X_t(\mu dt + \sigma dW_t),$$

with unknown constant  $\mu$  and  $\sigma$ . It is known that the process  $Y_t = e^{-3t}X_t^2$  is governed by the equation

$$dY_t = Y_t(dt + dW_t).$$

Determine the value of  $\mu$  and  $\sigma$ .

Solution: According to Ito's lemma, we have

$$dY_t = \left[\frac{\partial Y}{\partial t} + \mu X_t \frac{\partial Y}{\partial X_t} + \frac{1}{2}\sigma^2 X_t^2 \frac{\partial^2 Y}{\partial X_t^2}\right] dt + \sigma X_t \frac{\partial Y}{\partial X_t} dW_t.$$

As

$$\frac{\partial Y}{\partial t} = -3e^{-3t}X_t^2, \ \frac{\partial Y}{\partial X_t} = 2e^{-3t}X_t, \ \frac{\partial^2 Y}{\partial X_t^2} = 2e^{-3t},$$

we have

$$\begin{split} dY_t &= \left[ -3e^{-3t}X_t^2 + 2\mu e^{-3t}X_t^2 + \sigma^2 e^{-3t}X_t^2 \right] dt + 2\sigma X_t^2 e^{-3t} dW_t \\ &= \left[ -3 + 2\mu + \sigma^2 \right] Y_t dt + 2\sigma Y_t dW_t \\ &= Y_t \left[ (-3 + 2\mu + \sigma^2) dt + 2\sigma dW_t \right]. \end{split}$$

Hence we get

$$\left\{ \begin{array}{l} -3 + 2\mu + \sigma^2 = 1 \\ 2\sigma = 1 \end{array} \right. \Rightarrow \left\{ \begin{array}{l} \mu = \frac{15}{8} \\ \sigma = \frac{1}{2} \end{array} \right.$$

3. Find u(X,t) and v(X,t) where

$$dX_t = udt + vdW_t$$

and

(a) 
$$X_t = W_t^3$$
,

(b) 
$$X_t = -7t^2 + W_t^3$$
,

(c) 
$$X_t = g(t)e^{9W_t}$$
,

where f is a bounded, differentiable function.

## Solution:

(a) Take  $f(t, W_t) = W_t^4$ . Then,

$$\frac{\partial f}{\partial t} = 0, \ \frac{\partial f}{\partial W_t} = 3W_t^2, \ \frac{\partial^2 f}{\partial W_t^2} = 6W_t.$$

As

$$df(t,W_t) = \left[\frac{\partial f}{\partial t} + \frac{1}{2}\frac{\partial^2 f}{\partial W_t^2}\right]dt + \frac{\partial f}{\partial W_t}dW_t,$$

we have

$$dX_t = 3W_t dt + 3W_t^2 dW_t = 3X_t^{\frac{1}{3}} dt + 3X_t^{\frac{2}{3}} dW_t,$$

hence,  $u(X,t) = 3X_t^{\frac{1}{3}}$  and  $v(X,t) = 3X_t^{\frac{2}{3}}$ .

(b) Take  $f(t, W_t) = -7t^2 + W_t^3$ . Then,

$$\frac{\partial f}{\partial t} = -14t, \ \frac{\partial f}{\partial W_t} = 3W_t^2, \ \frac{\partial^2 f}{\partial W_t^2} = 6W_t.$$

As

$$df(t, W_t) = \left[\frac{\partial f}{\partial t} + \frac{1}{2} \frac{\partial^2 f}{\partial W_t^2}\right] dt + \frac{\partial f}{\partial W_t} dW_t,$$

we have

$$dX_t = (-14t + 3W_t)dt + 3W_t^2 dW_t = [-14t + 3(X_t + 7t^2)^{\frac{1}{3}}]dt + 3(X_t + 7t^2)^{\frac{2}{3}} dW_t,$$

hence,  $u(X,t) = -14t + 3(X_t + 7t^2)^{\frac{1}{3}}$  and  $v(X,t) = 3(X_t + 7t^2)^{\frac{2}{3}}$ .

(c) Take  $f(t, W_t) = g(t)e^{9W_t}$ . Then,

$$\frac{\partial f}{\partial t} = g'(t)e^{9W_t}, \ \frac{\partial f}{\partial W_t} = 9g(t)e^{9W_t}, \ \frac{\partial^2 f}{\partial W_t^2} = 81g(t)e^{9W_t}.$$

As

$$df(t, W_t) = \left[ \frac{\partial f}{\partial t} + \frac{1}{2} \frac{\partial^2 f}{\partial W_t^2} \right] dt + \frac{\partial f}{\partial W_t} dW_t,$$

we have

$$dX_t = (g'(t) + \frac{81}{2}g(t))e^{9W_t}dt + 9g(t)e^{9W_t}dW_t = \left[\frac{g'(t)}{g(t)} + \frac{81}{2}\right]X_tdt + 9X_tdW_t,$$

hence, 
$$u(X,t) = (\frac{g'(t)}{g(t)} + \frac{81}{2})X_t$$
 and  $v(X,t) = 9X_t$ .

4.  $S_t$  is generated by the stochastic differential equation

$$dS_t = S_t(\mu dt + \sigma dW_t)$$
, with  $S_0 = a$ ,

where  $\mu$ ,  $\sigma$  and a are constants. Evaluate the probability of  $\lambda K \leq S_T \leq 2\lambda K$ , where T > 0,  $\lambda > 0$ , and K is a positive constant.

## **Solution:**

By solving the stochastic differential equation, we can get that

$$S_t = ae^{(\mu - \frac{1}{2}\sigma^2)t + \sigma W_t}.$$

Thus, we have

$$Prob(\lambda K \leq S_T \leq 2\lambda K) = Prob\left(ae^{(\mu - \frac{1}{2}\sigma^2)T + \sigma W_T} \leq 2\lambda K\right)$$
$$- Prob\left(ae^{(\mu - \frac{1}{2}\sigma^2)T + \sigma W_T} \leq \lambda K\right)$$
$$= Prob\left(\frac{W_T}{\sqrt{T}} \leq \frac{\ln\left(\frac{2\lambda K}{a}\right) - (\mu - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}\right)$$
$$- Prob\left(\frac{W_T}{\sqrt{T}} \leq \frac{\ln\left(\frac{\lambda K}{a}\right) - (\mu - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}\right).$$

As  $W_T$  follows the normal distribution with mean 0 and variance T,  $\frac{W_T}{\sqrt{T}}$  follows the standard normal distribution, we get that

$$Prob(\lambda K \leq S_T \leq 2\lambda K) = N \left( \frac{\ln\left(\frac{2\lambda K}{a}\right) - (\mu - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}} \right) - N \left( \frac{\ln\left(\frac{\lambda K}{a}\right) - (\mu - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}} \right)$$

where  $N(\cdot)$  is the cumulative distribution function of the standard normal distribution.

5. A stochastic process is generated by the equation

$$dX_t = \mu dt + \sigma X_t dW_t$$

with the initial condition of  $X_0 = 1$ . Which equation governs the process  $Y(t, X_t) = e^{(1+t)X_t + t^2}$ ?

**Solution:** According to Ito's lemma, we have

$$dY_t = \left[ \frac{\partial Y}{\partial t} + \mu \frac{\partial Y}{\partial X_t} + \frac{1}{2} \sigma^2 X_t^2 \frac{\partial^2 Y}{\partial X_t^2} \right] dt + \sigma X_t \frac{\partial Y}{\partial X_t} dW_t.$$

Since

$$\frac{\partial Y}{\partial t} = (X_t + 2t)Y, \ \frac{\partial Y}{\partial X_t} = (1+t)Y, \ \frac{\partial^2 Y}{\partial X_t^2} = (1+t)^2Y, X_t = \frac{\ln(Y_t) - t^2}{1+t}$$

we have

$$dY_t = \left[ (X_t + 2t) + \mu(1+t) + \frac{1}{2}\sigma^2 X_t^2 (1+t)^2 \right] Y_t dt + \sigma X_t (1+t) Y_t dW_t.$$

with the initial condition of  $Y_0 = e$ .

6. Consider a stochastic process  $X_t$  governed by the equation

$$dX_t = X_t(a(t)dt + b(t)dW_t),$$

where a(t) and b(t) are unknown functions of t. It is known that the process  $Y_t = f(t)X_t^2$  is governed by the equation

$$dY_t = Y_t t(dt + 9t^2 dW_t).$$

Determine the functions a(t) and b(t).

Solution: According to Ito's lemma, we have

$$dY_t = \left[ \frac{\partial Y}{\partial t} + a(t)X_t \frac{\partial Y}{\partial X_t} + \frac{1}{2}b(t)^2 X_t^2 \frac{\partial^2 Y}{\partial X_t^2} \right] dt + b(t)X_t \frac{\partial Y}{\partial X_t} dW_t.$$

As

$$\frac{\partial Y}{\partial t} = f'(t)X_t^2, \ \frac{\partial Y}{\partial X_t} = 2f(t)X_t, \ \frac{\partial^2 Y}{\partial X_t^2} = 2f(t),$$

we have

$$\begin{split} dY_t &= \left[ f'(t) + f(t)(2a(t) + b(t)^2) \right] X_t^2 dt + 2f(t)b(t) X_t^2 dW_t \\ &= Y_t \left[ \left( \frac{f'(t)}{f(t)} + 2a(t) + b(t)^2 \right) dt + 2b(t) dW_t \right]. \end{split}$$

Hence we get

$$\left\{ \begin{array}{l} 2b(t) = 9t^3 \\ \frac{f'(t)}{f(t)} + 2a(t) + b(t)^2 = t \end{array} \right. \Rightarrow \left\{ \begin{array}{l} b(t) = \frac{9}{2}t^3 \\ a(t) = \frac{1}{2}(t - \frac{81}{4}t^6 - \frac{d \ln f(t)}{dt}) \end{array} \right.$$