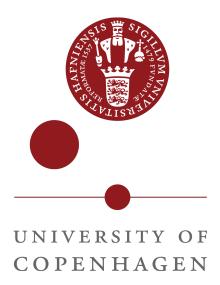
University of Copenhagen

CONTINUOUS TIME FINANCE (FINKONT)

Exam Papers

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Exam sets

In progress

Problem 2

Consider a standard Black-Scholes model, that is, a model consisting of a bank account B(t) with P-dynamics given by

$$dB(t) = rB(t) dt$$

with B(0) = 1 and a stock S(t) with P-dynamics given by

$$dS(t) = \alpha S(t) dt + \sigma S(t) d\overline{W}(t),$$

with S(0) = s > 0 where $r, \alpha \in \mathbb{R}$ and $\sigma > 0$ are constants and $\overline{W}(t)$ is a P-Brownian motion. Let T > 0 be a given and fixed (expiry) date.

Consider the derivative that at time T pays $X = \min \left[\max \left[S(T), K_1 \right], K_2 \right]$ where $0 < K_1 < K_2$ are constants. Let F(t,s) be the pricing function of the derivative.

- a. i. Determine the equations satisfied by the pricing function F(t,s).
 - ii. Find a hedging portfolio for the derivative X. (Hint: Draw a picture of the payoff function).

Let $h(t) = (h_0(t), h_1(t))$ be a self-financing portfolio given by

$$h_0(t) = (1-u)\frac{V^h(t)}{B(t)}, \ h_1(t) = u\frac{V^h(t)}{S(t)}$$

where u is a constant and set $V^h(0) = 1$. Note that $h_0(t)$ is the number of units of the bank account at time t, and $h_1(t)$ is the number of shares in the stock at time t, and $V^h(t)$ denotes the associated value process. Consider the derivative that at time T pays $Y = \sqrt{V^h(T)}$.

- b. Determine the arbitrage free price of derivative Y at time t=0.
- Solution (a).
- Solution (b).

Problem 3

Solution (a).

Solution (b).

Solution (c).

Solution (d).

Solution (e).

Exam 2017/18

Problem 1

Let W_t denote a Brownian motion and let

$$\mathcal{F}_t = \mathcal{F}_t^W = \sigma(\{W_s \mid 0 \le s \le t\}).$$

Let T > 0 be a given and fixed time.

Let f(t) be a bounded deterministic continuous function. Define the two processes

$$\begin{cases} X_t = \int_0^t f(u) \ dW_u, \\ M_t^{(\lambda)} = \exp\left\{\lambda X_t - \frac{\lambda^2}{2} \int_0^t f^2(u) \ du\right\}, \end{cases}$$

where $\lambda \in \mathbb{R}$ is a constant.

a. Show that $M^{(\lambda)}$ is a martingale with $E[M_t^{(\lambda)}]=1.$

Let 0 < s < t and $\lambda_1, \lambda_2 \in \mathbb{R}$ be given and fixed.

b. i. Show that

$$M_s^{(\lambda_1)} = E\left[\frac{M_s^{(\lambda_1)} M_t^{(\lambda_2)}}{M_s^{(\lambda_2)}} \middle| \mathcal{F}_s\right]$$

$$= E\left[\exp\left\{\lambda_1 X_s + \lambda_2 (X_t - X_s) - \frac{1}{2}\lambda_1^2 \int_0^s f^2(u) \ du - \frac{1}{2}\lambda_2^2 \int_s^t f^2(u) \ du\right\} \middle| \mathcal{F}_s\right]$$

ii. Show that X_s and $X_t - X_s$ are normally distributed and independent.

c. Compute the mean value of $M_T^{(\lambda)} \log(M_T^{(\lambda)})$.

Solution (a).

First, we see that since X_t is on integral form we know that

$$\begin{cases} dX_t = f(t) \ dW_t \\ X_0 = 0. \end{cases}$$

Hence we may represent M as $M_t^{(\lambda)} = g(t, X_t, Y_t)$ given by

$$g(t, x, y) = \exp\left\{\lambda x - \frac{\lambda^2}{2}y\right\},$$

where $Y_t = \int_0^t f^2(u) \ du$ with dynamics

$$\begin{cases} dY_t = f^2(t) \ dt \\ Y_0 = 0. \end{cases}$$

Hence by the multidimensional Ito's formula we have the dynamics of M given by

$$\begin{split} dM_t^{(\lambda)} &= g_t \ dt + g_x \ dX_t + g_y \ dY_t + \frac{1}{2} g_{yy} \ (dY_t)^2 + \frac{1}{2} g_{xx} \ (dX_t)^2 + f_{xy} (dX_t) (dY_t) \\ &= 0 + \lambda g \ dX_t - \frac{\lambda^2}{2} g \ dY_t + 0 + \frac{1}{2} \lambda^2 g \ (dX_t)^2 + 0 \\ &= \lambda M_t^{(\lambda)} f(t) \ dW_t - \frac{1}{2} \lambda^2 M_t^{(\lambda)} f^2(t) \ dt + \frac{1}{2} \lambda M_t^{(\lambda)} f^2(t) \ dt \\ &= \lambda f(t) M_t^{(\lambda)} \ dW_t, \end{split}$$

And so we see that M is a martingale as it only has dynamics wrt. the Brownian motion W (assuming $\lambda f_t M_t^{(\lambda)} \in \mathcal{L}^2$). Furthermore we have that

$$M_0^{(\lambda)} = g(0, X_0, Y_0) = \exp\left\{\lambda X_0 - \frac{1}{2}\lambda^2 Y_0\right\} = e^0 = 1$$

and so we have $E[M_t^{(\lambda)}] = M_0^{(\lambda)} = 1$ as desired. \square

Solution (b).

"(i)" We have from the previous exercise

$$\begin{split} &\frac{M_s^{(\lambda_1)} M_t^{(\lambda_2)}}{M_s^{(\lambda_2)}} \\ &= \exp\left\{\lambda_1 X_s - \frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du\right\} \exp\left\{\lambda_2 X_t - \frac{1}{2} \lambda_2^2 \int_0^t f^2(u) \ du\right\} \exp\left\{\frac{1}{2} \lambda_2^2 \int_0^s f^2(u) \ du - \lambda_2 X_s\right\} \\ &= \exp\left\{\lambda_1 X_s - \frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du + \lambda_2 X_t - \frac{1}{2} \lambda_2^2 \int_0^t f^2(u) \ du + \frac{1}{2} \lambda_2^2 \int_0^s f^2(u) \ du - \lambda_2 X_s\right\} \\ &= \exp\left\{\lambda_1 X_s + \lambda_2 (X_t - X_s) - \frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du - \frac{1}{2} \lambda_2^2 \int_s^t f^2(u) \ du\right\} \end{split}$$

and so the conclusion follows. \Box

"(ii)" We have that from lemma 4.18 that

$$X_s = \int_0^s f(u) \ dW_u \sim \mathcal{N}\left(0, \int_0^s f^2(u) \ dW_u\right)$$

furthermore we have that

$$X_t - X_s = \int_s^t f(u) \ dW_u \sim \mathcal{N}\left(0, \int_s^t f^2(u) \ dW_u\right).$$

In regard to the independence claim we could check identity below

$$E[e^{t_1X}e^{t_2Y}] = E[e^{t_1X}]E[e^{t_2Y}]$$

where X, Y are independent random variables. The above identity holds if and only if X and Y are independent. From above we have that

$$M_s^{(\lambda_1)} = E[e^{\lambda_1 X_s} e^{\lambda_2 (X_t - X_s)} \mid \mathcal{F}_s] e^{-\frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du - \frac{1}{2} \lambda_2^2 \int_s^t f^2(u) \ du}$$

and so taking expectation we have

$$1 = E[e^{\lambda_1 X_s} e^{\lambda_2 (X_t - X_s)}] e^{-\frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du - \frac{1}{2} \lambda_2^2 \int_s^t f^2(u) \ du$$

Which the gives

$$E[e^{\lambda_1 X_s} e^{\lambda_2 (X_t - X_s)}] = e^{\frac{1}{2} \lambda_1^2 \int_0^s f^2(u) \ du + \frac{1}{2} \lambda_2^2 \int_s^t f^2(u) \ du} = E[e^{\lambda_1 X_s}] E[e^{\lambda_2 (X_t - X_s)}]$$

and so the conclusion is that X_s and $X_t - X_s$ are independent. \square Solution (c).

We recall the definition of $M_t^{(\lambda)}$ and observe that

$$\log M_t^{(\lambda)} = \lambda X_t - \frac{1}{2} \lambda^2 \int_0^t f^2(u) \ du.$$

Furthermore we have the dynamics of $M^{(\lambda)}$ given by the differential form

$$dM_t^{(\lambda)} = \lambda f(t) M_t^{(\lambda)} dW_t.$$

with $M_0^{(\lambda)}=1.$ Since we know that $M_t^{(\lambda)}$ is a martingale we have

$$E^{P}[M_{T}^{(\lambda)}] = E^{P}[M_{0}^{(\lambda)}] = 1,$$

and so we may define a new probability measure as

$$d\tilde{P} = M_T^{(\lambda)} dP$$

on \mathcal{F}_T . We then have a new Brownian motion \tilde{W} such that

$$dW_t = \lambda f(t) dt + d\tilde{W}_t.$$

We can then see

$$\begin{split} E^P[M_T^{(\lambda)}\log M_T^{(\lambda)}] &= \int M_T^{(\lambda)}\log M_T^{(\lambda)} \ dP = \int M_T^{(\lambda)}\log M_T^{(\lambda)} \frac{1}{M_T^{(\lambda)}} \ d\tilde{P} \\ &= \int \log M_T^{(\lambda)} \ d\tilde{P} = E^{\tilde{P}}[\log M_T^{(\lambda)}]. \end{split}$$

Then we can evaluate the mean value by seeing the X has representation wrt. \tilde{P} by

$$X_t = \int_0^t f(u) (\lambda f(u) du + d\tilde{W}_u) = \lambda \int_0^t f^2(u) du + \int_0^t f(u) d\tilde{W}_u.$$

Giving that

$$\begin{split} E^P[M_T^{(\lambda)}\log M_T^{(\lambda)}] &= E^{\tilde{P}}[\log M_T^{(\lambda)}] \\ &= E^{\tilde{P}}\left[\lambda X_T - \frac{1}{2}\lambda^2 \int_0^T f^2(u) \ du\right] \\ &= E^{\tilde{P}}\left[\lambda^2 \int_0^T f^2(u) \ du + \lambda \int_0^T f(u) \ d\tilde{W}_u - \frac{1}{2}\lambda^2 \int_0^T f^2(u) \ du\right] \\ &= \lambda E^{\tilde{P}}\left[\frac{1}{2}\lambda \int_0^T f^2(u) \ du + \int_0^T f(u) \ d\tilde{W}_u\right] \\ &= \frac{1}{2}\lambda^2 \int_0^T f^2(u) \ du + \lambda E^{\tilde{P}}\left[\int_0^T f(u) \ d\tilde{W}_u\right] \\ &= \frac{1}{2}\lambda^2 \int_0^T f^2(u) \ du \end{split}$$

Since

$$\tilde{X}_T = \int_0^T f(u) \ d\tilde{W}_u,$$

is a \tilde{P} -martingale. \square

Problem 2

Consider a standard Black-Scholes model, that is, a model consisting of a bank account B_t with P-dynamics given by

$$dB_t = rB_t \ dt, \ B_0 = 1$$

and a stock S_t with P-dynamics given by

$$dS_t = \alpha S_t dt + \sigma S_t d\overline{W}_t, S_0 = s > 0$$

where $r, \alpha \in \mathbb{R}$ and $\sigma > 0$ are constants and \overline{W}_t is a P-Brownian motion. Let T > 0 be a given and fixed date.

Consider the derivative that at time T pays

$$X = \max \left\{ \min \left\{ S_T, K_2 \right\}, K_1 \right\},\,$$

where $0 < K_1 < K_2$ are constants.

a. Determine the arbitrage free price of derivative X at time t < T.

Consider a new derivative that at time T pays

$$Y = (S_T^2 - K^2)^+ - (K^2 - S_T^2)^+.$$

- b. i. Determine the arbitrage free price of derivative Y at time t < T.
 - ii. Find a hedging portfolio for derivative Y.

Let $h(t) = (h_0(t), h_1(t))$ be a portfolio where

$$h_0(t) = -e^{r(T-2t)+\sigma^2(T-t)}S^2(t)$$

is the number of units in the bank account at time t and

$$h_1(t) = 2e^{(r+\sigma^2)(T-t)}S(t)$$

is the number of shares in the stock at time t. Let $V^h(t)$ denote the associated value process.

- c. Determine whether the portfolio h is self-financing or not.
- d. Compute $V^h(T)$.

Solution (a).

We see that the derivative is the bull spread given by the payout function

$$X = \begin{cases} K_2 & \text{if } S_T > K_2, \\ S_T & \text{if } K_1 \le S_T \le K_2, \\ K_1 & \text{if } S_T < K_1. \end{cases}$$

We know from exercise 10.3 that this can be replicated by holding K_1 bonds, one call option with strike K_1 and a short on a call with strike K_2 . The arbitrage free price of the derivative is then the value process of the mentioned portfolio i.e.

$$\Pi_t[X] = K_1 e^{-r(T-t)} + c(K_1; t, T) - c(K_2; t, T),$$

where c denotes the pricing function for a European call option (non-instructive parameters supressed). \square Solution (b).

(i): We start by seeing that the derivative pays out

$$Y = \begin{cases} S_T^2 - K^2 & \text{if } S_T^2 \ge K^2, \\ -(K^2 - S_T^2) & \text{if } S_T^2 < K^2. \end{cases}$$

hence the payout is $Y = S_T^2 - K^2 = \Phi(S_T)$ where $\Phi(s) = s^2 - K^2$. That is Y is in fact a simple claim. We have from the risk neutral valueation formula 7.11 that

$$\Pi_t[Y] = e^{-r(T-t)} E_{t,s}^Q [S_T^2 - K^2]$$
$$= e^{-r(T-t)} E_{t,s}^Q [S_T^2] - e^{-r(T-t)} K^2.$$

Recall that under the martingale measure Q we have that S_t is a GBM hence

$$S_t = s \cdot \exp\left\{ \left(r - \frac{1}{2}\sigma^2 \right) (T - t) + \sigma \left(W_T^Q - W_t^Q \right) \right\}$$

then

$$S_T^2 = s^2 \cdot \exp \left\{ 2 \left(r - \frac{1}{2} \sigma^2 \right) (T - t) + 2 \sigma \left(W_T^Q - W_t^Q \right) \right\}.$$

Inserting this into the risk neutral valuation formula we get

$$\begin{split} \Pi_t[Y] &= e^{-r(T-t)} E_{t,s}^Q[S_T^2] - e^{-r(T-t)} K^2 \\ &= e^{-r(T-t)} s^2 e^{2\left(r - \frac{1}{2}\sigma^2\right)(T-t)} E^Q \left[\exp\left\{2\sigma\left(W_T^Q - W_t^Q\right)\right\} \right] - e^{-r(T-t)} K^2 \\ &= e^{-r(T-t)} s^2 e^{2\left(r - \frac{1}{2}\sigma^2\right)(T-t)} e^{\frac{1}{2}4\sigma^2(T-t)} - e^{-r(T-t)} K^2 \\ &= e^{-r(T-t)} \left(s^2 e^{(2r - \sigma^2)(T-t) + \frac{1}{2}4\sigma^2(T-t)} - K^2 \right) \\ &= e^{-r(T-t)} \left(s^2 e^{(2r + \sigma^2)(T-t)} - K^2 \right). \end{split}$$

The arbitrage free price of the derivative is then given above. \square

(ii): From theorem 8.5 we can determine a hedging portfolio with weightings

$$\begin{split} w_t^B &= \frac{\Pi_t - S_t \frac{\partial \Pi}{\partial s}}{\Pi_t} \\ &= 1 - \frac{S_t 2 S_t e^{-r(T-t)} e^{(2r+\sigma^2)(T-t)}}{e^{-r(T-t)} \left(S_t^2 e^{(2r+\sigma^2)(T-t)} - K^2 \right)} \\ &= 1 - \frac{2 S_t^2 e^{(2r+\sigma^2)(T-t)}}{S_t^2 e^{(2r+\sigma^2)(T-t)} - K^2} \\ &= 1 - \frac{2}{1 - K^2 S_t^{-2} e^{(2r+\sigma^2)(t-T)}} \\ w_t^S &= \frac{2}{1 - K^2 S_t^{-2} e^{(2r+\sigma^2)(t-T)}}. \end{split}$$

In absolute terms we will hold the portfolio

$$h_t^S = 2S_t e^{-r(T-t)} e^{(2r+\sigma^2)(T-t)}$$

$$h_t^B = \frac{e^{-r(T-t)} \left(s^2 e^{(2r+\sigma^2)(T-t)} - K^2 \right) - S_t h_t^S}{B_t}$$

$$= \frac{e^{-r(T-t)} \left(s^2 e^{(2r+\sigma^2)(T-t)} - K^2 \right) - S_t h_t^S}{e^{rt}}$$

$$= e^{-rT} s^2 e^{(2r+\sigma^2)(T-t)} - e^{-rT} K^2 - e^{-rt} S_t h_t^S.$$

The portfolio above will hedge Y with probability one. \square

Solution (c).

We assume no dividends and no consumption that is $c_t = 0$ and $dD_t^i = 0$ for i = 0, 1. Then the portfolio is self-financing if and only if the value process has dynamics.

$$h_0(t) dB_t + h_1(t) dS_t = 0$$

This is given in lemma 6.12.

THE BELOW IS IN WORKS AND NOT CORRECT!

Now we have that the value process is given by

$$V_t^h = h_0(t)B_t + h_1(t)S_t.$$

Using the representation $V_t^h = f(h_0(t), B_t) + f(h_1(t), S_t)$ given by f(x, y) = xy we have

$$dV_t^h = df(h_0(t), B_t) + df(h_1(t), S_t).$$

Using Ito's formula on each term we have

$$df(h_0(t), B_t) = B_t dh_0(t) + h_0(t) dB_t + (dB_t)(dh_0(t)),$$

$$df(h_1(t), S_t) = S_t dh_1(t) + h_1(t) dS_t + (dS_t)(dh_1(t)),$$

since of cause $f_{xx} = f_{yy} = 0$. We can the determine the dynamics of the portfolio by

$$\begin{split} dh_0(t) &= -(-2t - \sigma^2) S_t^2 e^{r(T - 2t) + \sigma^2(T - t)} \ dt \\ &- 2S_t e^{r(T - 2t) + \sigma^2(T - t)} \ dS_t \\ &- \frac{1}{2} 2 e^{r(T - 2t) + \sigma^2(T - t)} \ (dS_t)^2 \\ &= (-2t - \sigma^2) h_0(t) \ dt + \frac{2}{S_t} h_0(t) \ (\mu S_t \ dt + \sigma S_t \ dW_t) + \frac{1}{S_t^2} h_0(t) \sigma^2 S_t^2 \ dt \\ &= (\mu - 1) 2h_0(t) \ dt + 2\sigma h_0(t) \ dW_t \end{split}$$

and

$$dh_1(t) = (-r - \sigma^2) 2e^{(r+\sigma^2)(T-t)} S_t dt$$

$$+ 2e^{(r+\sigma^2)(T-t)} dS_t + 0$$

$$= (-r - \sigma^2) h_1(t) dt + \frac{1}{S_t} h_1(t) (\mu S_t dt + \sigma S_t dW_t)$$

$$= (-r - \sigma^2 + \mu) h_1(t) dt + h_1(t) \sigma dW_t$$

And so in total

$$\begin{split} dV_t^h(t) &= df(h_0(t), B_t) + df(h_1(t), S_t) \\ &= B_t \ dh_0(t) + h_0(t) \ dB_t + (dB_t)(dh_0(t)) \\ &+ S_t \ dh_1(t) + h_1(t) \ dS_t + (dS_t)(dh_1(t)) \\ &= B_t \ ((\mu - 1)2h_0(t) \ dt + 2\sigma h_0(t) \ dW_t) + h_0(t) \ rB_t \ dt + 0 \\ &+ S_t \ ((-r - \sigma^2 + \mu)h_1(t) \ dt + h_1(t)\sigma \ dW_t) + h_1(t) \ (\mu S_t \ dt + \sigma S_t \ dW_t) + \sigma^2 S_t h_1(t) \ dt \\ &= \left[B_t(\mu - 1)2h_0(t) + h_0(t)rB_t + S_t(-r - \sigma^2 + \mu)h_1(t) + h_1\mu S_t + \sigma^2 S_t h_1(t) \right] \ dt \\ &+ \left[B_t 2\sigma h_0(t) + S_t h_1\sigma + h_1\sigma S_t \right] \ dW_t \\ &= \left[(2\mu - 2 + r)B_t h_0(t) + (-r + 2\mu)S_t h_1(t) \right] \ dt \\ &+ \left[B_t h_0(t) + h_1 S_t \right] 2\sigma \ dW_t \\ &= V_t^{+} 2\mu \ dt + V_t^{+} \ dW_t \end{split}$$

Solution (d).

We compute V_T^h easily by inserting h_0 and h_1 below

$$V_T^h = B_T h_0(T) + S_T h_1(T)$$

$$= B_T \left(-e^{r(T-2T) + \sigma^2(T-T)} S_T^2 \right) + S_T \left(2e^{(r+\sigma^2)(T-T)} S_T \right)$$

$$= -S_T^2 + 2S_T^2 = S_T^2.$$

and so h hedge the payout $\Phi(S_T) = S_T^2$. \square

Problem 3

Consider a two-dimensional model. The market model consist of three assets: A bank account B_t and two stocks S_1 and S_2 . The P-dynamics of B_t is

$$dB_t = rB_t \ dt, \ B_0 = 1,$$

where $r \in \mathbb{R}$ is a constant interest rate. The P-dynamics of S_1 and S_2 are given by

$$dS_1(t) = \alpha_1 S_1(t) \ dt + \sigma_1 S_1(t) \ d\overline{W}_1(t), \qquad S_1(0) = s_1 > 0,$$

$$dS_2(t) = \alpha_2 S_2(t) \ dt + \sigma_2 S_2(t) \ d\overline{W}_2(t), \qquad S_2(0) = s_2 > 0,$$

where $\alpha_1, \alpha_2 \in \mathbb{R}$ are constants. Moreover, $\sigma_1 > 0$ is a constant and $\sigma_2(t) = \sigma_0 e^{-\gamma t}$ where $\sigma_0 > 0$ and $\gamma > 0$ are constants and $\overline{W}_1(t)$ and $\overline{W}_2(t)$ are two independent *P*-Brownian motions. The filtration is the one generated by the two Brownian motions, that is, $\mathcal{F}_t = \sigma(\overline{W}_1(s), \overline{W}_2(s) \mid 0 \le s \le t)$. Let T > 0 be a given and fixed (expiry) date.

- a. i. Is the model arbitrage free?
 - ii. Is the model complete?

Consider the derivative that at time T pays $X = S_1(T)S_2(T)$ and let $F(t, s_1, s_2)$ be the pricing function of the derivative.

- b. i. Determine the arbitrage free price of derivative X at time t = 0.
 - ii. Determine the equation satisfied by the pricing function $F(t, s_1, s_2)$.

Consider a new derivative that at time T pays $Y = \log(S_2(T))$.

c. Determine the arbitrage free price of derivative Y at time t < T.

Solution (a).

(i): We know that the model is arbitrage free if and only if there exist a martingale measure Q. This is equivalent with finding a likelihood process L with Radon-Nikodym derivative φ given by the solution to the equation

$$\sigma_t \varphi_t = r_t - \mu_t.$$

We see that

$$\sigma_t = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_0 e^{-\gamma t} \end{bmatrix} \Rightarrow \sigma_t^{-1} = \begin{bmatrix} 1/\sigma_1 & 0 \\ 0 & e^{\gamma t}/\sigma_0 \end{bmatrix}.$$

Hence we trivially have a solution given by

$$\varphi_t = \begin{bmatrix} 1/\sigma_1 & 0 \\ 0 & e^{\gamma t}/\sigma_0 \end{bmatrix} \begin{bmatrix} r - \alpha_1 \\ r - \alpha_2 \end{bmatrix} = \begin{bmatrix} \frac{r - \alpha_1}{\sigma_1} \\ \frac{r - \alpha_2}{\sigma_0} e^{\gamma t} \end{bmatrix}.$$

Proposition 14.1 gives now that if L, given by

$$dL_t = \varphi_t^\top L_t \ dW_t, \ L_0 = 1,$$

is a martingale then the market is arbitrage free. This is true if the Novikov condition is satisfied. We have

$$E^{P}\left[e^{\frac{1}{2}\int_{0}^{T}\|\varphi_{t}\|^{2} dt}\right] = e^{\frac{1}{2}\int_{0}^{T}\|\varphi_{t}\|^{2} dt} = e^{\frac{1}{2}\int_{0}^{T}\left(\frac{r-\alpha_{1}}{\sigma_{1}}\right)^{2} + \left(\frac{r-\alpha_{2}}{\sigma_{0}}e^{\gamma t}\right)^{2} dt} < \infty$$

since of cause

$$\int_0^T (\frac{r-\alpha_1}{\sigma_1})^2 + (\frac{r-\alpha_2}{\sigma_0}e^{\gamma t})^2 dt = T(\frac{r-\alpha_1}{\sigma_1})^2 + (\frac{r-\alpha_2}{\sigma_0})^2 \int_0^T e^{2\gamma t} dt < \infty$$

for all $T \geq 0$. Then the Novikov condition is satisfied and L is martingale with $E[L_T] = 1$. \square

(ii): The model is complete if the martingale measure is unique. This is equivalent with $Ker[\sigma_t] = \{0\}$ and since σ_t is invertible (diagonal) we have that the model is complete. \square

Solution (b).

(i): We may determine the price of the derivative using the risk neutral valueation formula

$$\Pi_t[X] = E^Q \left[e^{-\int_t^T r(u) \ du} X \ \middle| \ \mathcal{F}_t \right]$$

Hence we have for t = 0 and $S_1(0) = s_1$ and $S_2(0) = s_2$ that

$$\Pi_0[X] = E^Q \left[e^{-\int_0^T r(u) \ du} X \ \middle| \ \mathcal{F}_0 \right] = e^{-rT} E^Q \left[S_1(T) S_2(T) \ \middle| \ \mathcal{F}_0 \right],$$

Since we have that S_1 and S_2 have dynamics wrt. two independent Brownian motions we know that the price processes are independent. If we multiply by B(T)/B(T) we obtain two martingale processes under the measure Q:

$$\Pi_0[X] = e^{-rT} B(T)^2 E^Q \left[\frac{S_1(T)}{B(T)} \middle| \mathcal{F}_0 \right] E^Q \left[\frac{S_2(T)}{B(T)} \middle| \mathcal{F}_0 \right]
= e^{-rT} e^{2rT} s_1(0) s_2(0) = e^{rT} s_1(0) s_2(0),$$

and so the arbitrage free price is given above. \square

(ii): We have from Bjork (14.31) that Π satisfies the PDE below

$$\begin{cases} F_t + \sum_{i=1}^2 r s_i F_{s_i} + \frac{1}{2} \text{tr}[\sigma_t^\top D(S) F_{ss} D(S) \sigma_t] - rF = 0 \\ F(T, s_1, s_2) = \Phi(s_1, s_2) \end{cases}$$

The PDE is in detail

$$\begin{split} 0 + rS_1(t)S_2(t) + rS_2S_1 + \frac{1}{2}\mathrm{tr} \begin{bmatrix} S_1\sigma_1 & 0 \\ 0 & S_2\sigma_0e^{-\gamma t} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} S_1\sigma_1 & 0 \\ 0 & S_2\sigma_0e^{-\gamma t} \end{bmatrix} - r\Pi_t \\ &= 2rS_1(t)S_2(t) + \frac{1}{2}\mathrm{tr} \begin{bmatrix} S_1\sigma_1 & 0 \\ 0 & S_2\sigma_0e^{-\gamma t} \end{bmatrix} \begin{bmatrix} 0 & S_2\sigma_0e^{-\gamma t} \\ S_1\sigma_1 & 0 \end{bmatrix} \\ &= 2rS_1(t)S_2(t) + \frac{1}{2}\mathrm{tr} \begin{bmatrix} 0 & S_1S_2\sigma_0\sigma_1e^{-\gamma t} \\ S_1S_2\sigma_0\sigma_1e^{-\gamma t} & 0 \end{bmatrix} - r\Pi_t \\ &= 2rS_1(t)S_2(t) - r\Pi_t = 0. \end{split}$$

or

$$F(t, s_1, s_2) = 2s_1s_2, F(T, s_1, s_2) = s_1s_2$$

this ends the question. \square

Solution (c).

We have the derivative $Y = \log(S_2(T))$. By the risk neutral valuation formula we have that the arbitrage free price is given by

$$\Pi_t[Y] = E^Q \left[e^{-\int_t^T r(u) \ du} Y \ \middle| \ \mathcal{F}_t \right] = e^{-r(T-t)} E^Q \left[\log(S_2(T)) \ \middle| \ \mathcal{F}_t \right].$$

Under the measure Q the dynamics of S_2 is that of a GBM hence

$$d\log(S_2(t)) = \left(r - \frac{1}{2}\sigma_0^2 e^{-2\gamma t}\right) dt + \sigma_0^2 e^{-2\gamma t} dW_t^Q,$$

and so with the knowledge that $S_2(t) = s_2$ we have

$$\begin{split} \Pi_t[Y] &= e^{-r(T-t)} E^Q \left[\log(s_2) + \int_t^T \left(r - \frac{1}{2} \sigma_0^2 e^{-2\gamma s} \right) \, ds + \int_t^T \sigma_0^2 e^{-2\gamma t} \, dW_t^Q \, \middle| \, \mathcal{F}_t \right] \\ &= e^{-r(T-t)} \left[\log(s_2) + \int_t^T \left(r - \frac{1}{2} \sigma_0^2 e^{-2\gamma s} \right) \, ds \right] \\ &= e^{-r(T-t)} \left[\log(s_2) + r(T-t) - \frac{1}{2} \sigma_0^2 \int_t^T e^{-2\gamma s} \, ds \right] \\ &= e^{-r(T-t)} \left[\log(s_2) + r(T-t) + \frac{1}{4\gamma} \sigma_0 \left[e^{-2\gamma s} \right]_t^T \right] \\ &= e^{-r(T-t)} \left[\log(s_2) + r(T-t) + \frac{1}{4\gamma} \sigma_0 (e^{-2\gamma T} - e^{-2\gamma t}) \right]. \end{split}$$

The arbitrage free price of the derivative is then given above. \Box

Exam 2018/19

Problem 1

Let W(t) denote a Brownian motion and let $\mathcal{F}_t = \mathcal{F}_t^W$. Let T > 0 be a given and fixed time.

Consider the stochastic differential equation

$$dX(t) = \alpha \ dt + \sqrt{X(t)} \ dW(t),$$

and X(0) = x > 0 where $\alpha \in \mathbb{R}$.

- a. i. Compute the mean value of X(T).
 - ii. Compute the variance of X(T).
- b. Find the solution of the partial differential equation

$$4F_t(t,x) + 8x^2F_{xx}(t,x) + 3xF_x(t,x) = 5F(t,x)$$
 for $t < T$ and $x > 0$.
 $F(T,x) = x^3$.

Let $\widetilde{W}(t)$ be another Brownian motion such that W(t) and $\widetilde{W}(t)$ are two independent Brownian motions. Let Y(t) and Z(t) be two martingales given by the following dynamics

$$\begin{split} dY(t) &= W(t) \ dW(t) + \widetilde{W}(t) \ d\widetilde{W}(t), \\ dZ(t) &= \widetilde{W}(t) \ dW(t) - W(t) \ d\widetilde{W}(t). \end{split}$$

with Y(0) = Z(0) = 0.

c. Show that M(t) = Y(t)Z(t) is a martingale.

Solution (a).

(i): We start by writing X on integral form given as

$$X(t) = x + \int_0^t \alpha \ dt + \int_0^t \sqrt{X(s)} \ dW(s).$$

Taking expectation yields.

$$E[X(t)] = E\left[x + \alpha t + \int_0^t \sqrt{X(s)} \ dW(s)\right] = x + \alpha t,$$

since

$$E\left[\int_0^t \sqrt{X(s)} \ dW(s)\right] = E\left[\int_0^0 \sqrt{X(s)} \ dW(s)\right] = 0.$$

This result follows from lemma 4.10 as the process $M_t = \int_0^t \sqrt{X(s)} \ dW(s)$ is a martingale. From this we have that $E[X(T)] = x + \alpha T$. \square

(ii): We have that the variance is given by

$$Var(X(t)) = E(X^{2}(t)) - (E(X(t))^{2}.$$

and so we have

$$\begin{split} Var(X(t)) + E(X(t))^2 &= E\left[\left(x + t\alpha + \int_0^t \sqrt{X(s)} \ dW(s)\right)^2\right] \\ &= (x + \alpha t)^2 + E\left[\left(\int_0^t \sqrt{X(s)} \ dW(s)\right)^2\right] + 2(x + \alpha t)E\left[\int_0^t \sqrt{X(s)} \ dW(s)\right] \\ &= (x + \alpha t)^2 + E\left[\left(\int_0^t \sqrt{X(s)} \ dW(s)\right)^2\right]. \end{split}$$

Now by setting $Z(t) = \int_0^t \sqrt{X(s)} \ dW(s)$ we see that Z has dynamics $dZ(t) = \sqrt{X(t)} \ dW(t)$ with Z(0) = 0. By Ito's formula on the variable f(t, Z(t)) with $f(t, z) = z^2$ we have

$$df(t, Z(t)) = 0 dt + 2Z(t) dZ(t) + \frac{1}{2} 2 (dZ(t))^{2}$$
$$= 2Z(t)\sqrt{X(t)} dW(t) + X(t) dt.$$

Obviously when taking expectation on f(t, Z(t)) we see that the integral part related to the Brownian motion is a martingale with mean 0 and then

$$E[f(t, Z(t))] = E\left[\int_0^t X(s) \ ds\right].$$

In total we have

$$Var(X(t)) = (x + \alpha t)^2 + E\left[\int_0^t X(s) \ ds\right] - (x + \alpha t)^2 = E\left[\int_0^t X(s) \ ds\right].$$

Moving the expectation inside the integral then gives

$$Var(X(t)) = \int_0^t (x + \alpha s) \ ds = xt + \frac{1}{2}\alpha t^2.$$

Inserting t = T gives the desired result. \square

Solution (b).

We see by dividing by 4 we have the PDE given by

$$F_t + 2x^2 F_{xx} + \frac{3}{4} x F_x = \frac{5}{4} F$$

hence by setting r = 5/4, $\mu = 3x/4$ and $\sigma^2 = 4x^2$ we have the boundary value problem

$$\begin{cases} F_t + \mu F_x + \frac{1}{2}\sigma^2 F_{xx} - rF = 0, \\ F(T, x) = x^3. \end{cases}$$

From Feynmann-Kac we know this has solution on $[0,T]\times\mathbb{R}$ given by the stochastic representation

$$F(t,x) = e^{-r(T-t)} E_{t,x}[X_T^3],$$

where X satisfies the SDE

$$dX_t = \frac{3}{4}X_t dt + 2X_t dW_t.$$

Giving that X(t) = x and X is a GBM we have

$$X_T = x \cdot e^{\left(r - \frac{1}{2}2^2\right)(T - t) + 2(W_T - W_t)} = x \cdot e^{\frac{-5}{4}(T - t) + 2(W_T - W_t)}.$$

The relevant mean value is then

$$F(t,x) = e^{-\frac{5}{4}(T-t)} E\left[x^3 \cdot e^{\frac{-15}{4}(T-t) + 6(W_T - W_t)}\right]$$

$$= e^{-\frac{5}{4}(T-t)} x^3 e^{\frac{-15}{4}(T-t)} E\left[e^{6(W_T - W_t)}\right]$$

$$= x^3 e^{\frac{-20}{4}(T-t)} e^{\frac{1}{2}6^2(T-t)} = x^3 e^{-5(T-t) + 18(T-t)}$$

$$= x^3 e^{13(T-t)}.$$

The solution is the given above. \Box

Solution (c).

We show that M has dynamics solely given in terms of Brownian motions. We have that M(t) = f(t, Y(t), Z(t)) for f(t, y, z) = yz the dynamics given by Ito's formula:

$$dM(t) = 0 dt + Z(t) dY(t) + Y(t) dZ(t) + (dY(t))(dZ(t))$$

since the only second derivative not zero is $f_{yz} = f_{zy} = 1$. The product (dY(t))(dZ(t)) is computed first

$$(dY(t))(dZ(t)) = (W(t) \ dW(t) + \widetilde{W}(t) \ d\widetilde{W}(t)) \cdot (\widetilde{W}(t) \ dW(t) - W(t) \ d\widetilde{W}(t))$$

$$= W(t)\widetilde{W}(t) \ dt - \widetilde{W}(t)W(t) \ dt = 0,$$

where we use that $dW(t)d\widetilde{W}(t) = dt$ is the only non-zero term. Then we obviously have

$$dM(t) = Z(t) \ dY(t) + Y(t) \ dZ(t)$$

= $Z(t)W(t) \ dW(t) + Z(t)\widetilde{W}(t) \ d\widetilde{W}(t) + Y(t)\widetilde{W}(t) \ dW(t) - Y(t)W(t) \ d\widetilde{W}(t).$

Giving that M(t) is a martingale. (lemma 4.11)

Problem 2

Consider a standard Black-Scholes model, that is, a model consisting of a bank account B(t) with P-dynamics given by

$$dB(t) = rB(t) dt$$

with B(0) = 1 and a stock S(t) with P-dynamics given by

$$dS(t) = \alpha S(t) dt + \sigma S(t) d\overline{W}(t),$$

with S(0) = s > 0 and where $r, \alpha \in \mathbb{R}$ and $\sigma > 0$ are constants and $\overline{W}(t)$ is a P-Brownian motion. Let T > 0 be a given fixed (expiry) date.

Let $h(t) = (h_0(t), h_1(t))$ be a portfolio where

$$h_0(t) = \exp\left(\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{\alpha - r}{2} - \frac{1}{8}\sigma^2\right)t\right)$$

is the number of units in the bank account at time t and

$$h_1(t) = \frac{1}{s} \exp\left(-\frac{1}{2}\sigma \overline{W}(t) + \left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2\right)t\right)$$

is the number of shares in the stock at time t. Let $V^h(t)$ denote the associated value process and let $u(t) = (u_0(t), u_1(t))$ denote the relative portfolio.

- a. i. Determine whether the portfolio h is self-financing or not.
 - ii. Compute $u_1(t)$.

Consider two derivatives that at time T pay $X_1 = \Phi_1(S(T))$ and $X_2 = \Phi_2(S(T))$. For i = 1, 2, the arbitrage free price of derivative X_i is given by $\pi_i(t) = F_i(t, S(t))$ where $F_i(t, s)$ is the pricing function of the derivative. Assume that $\pi_i(t) > 0$. The price process $\pi_i(t)$ has dynamics (under the P-measure) given by

$$d\pi_i(t) = \alpha_i(t)\pi_i(t) dt + \sigma_i(t)\pi_i(t) d\overline{W}(t).$$

- b. i. Determine $\alpha_i(t)$ and $\sigma_i(t)$ for i = 1, 2.
 - ii. Show that

$$\frac{r - \alpha_1(t)}{\sigma_1(t)} = \frac{r - \alpha_2(t)}{\sigma_2(t)}.$$

Let C(t, s; K, T) denote the Black-Scholes price at time t of an European call option with strike K and expiry date T when the current price of the underlying is s. Similarly, let P(t, s; K, T) denote the Black-Scholes price at time t of an European put option with strike K and expiry date T when the current price of the underlying is s. Consider a new derivative that at time T pays

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$$Y = \max \{C(T, S(T); K, T_1), P(T, S(T); K, T_1)\}$$

where $T < T_1$ is a fixed date.

c. Determine the arbitrage free price of derivative Y at time t < T. (Hint: recall $\max(x, y) = (x + y)^+ y$) Assume that the call option and the put option do not have the same strike prices, that is, a derivative that at time T pays

$$\widetilde{Y} = \max \{C(T, S(T); K_1, T_1), P(T, S(T); K_2, T_1)\}$$

where the strike prices $K_1 \neq K_2$. Let F(t,s) be the pricing function of the derivative.

d. Determine the equation satisfied by the pricing function F(t,s).

Solution (a).

We have that h is self-financing if and only if the equation

$$dV^{h}(t) = h_0(t) dB(t) + h_1(t) dS(t)$$

is satisfied. And so, we start by determining the dynamics of the number of assets denoted by h_0 and h_1 . From Ito's formula we can conclude that

$$\begin{split} dh_0(t) &= \left(\frac{\alpha - r}{2} - \frac{1}{8}\sigma^2\right)h_0(t)\ dt + \frac{1}{2}\sigma h_0(t)\ dW(t) + \frac{1}{2}\frac{1}{2}\sigma\frac{1}{2}\sigma h_0\ (dW(t))^2 \\ &= \left(\frac{\alpha - r}{2} - \frac{1}{8}\sigma^2\right)h_0(t)\ dt + \frac{1}{2}\sigma h_0(t)\ dW(t) + \frac{1}{2^3}\sigma^2 h_0\ dt \\ &= \left(\frac{\alpha - r}{2} - \frac{1}{8}\sigma^2 + \frac{1}{8}\sigma^2\right)h_0(t)\ dt + \frac{1}{2}\sigma h_0(t)\ dW(t) \\ &= \frac{\alpha - r}{2}h_0(t)\ dt + \frac{1}{2}\sigma h_0(t)\ dW(t). \end{split}$$

For the number of stocks we have

$$dh_1(t) = \left(\frac{r-\alpha}{2} + \frac{3}{8}\sigma^2\right)h_1(t) dt - \frac{1}{2}\sigma h_1(t) dW(t) + \frac{1}{2}\frac{1}{2}\sigma\frac{1}{2}\sigma h_1(t) (dW(t))^2$$
$$= \left(\frac{r-\alpha}{2} + \frac{1}{2}\sigma^2\right)h_1(t) dt - \frac{1}{2}\sigma h_1(t) dW(t).$$

We may derive the dynamics of the portfolio as

$$dV^{h}(t) = d(h_{0}(t)B(t) + h_{1}(t)S(t))$$

$$= B(t) dh_{0}(t) + h_{0}(t) dB(t) + (dh_{0}(t))(dB(t))$$

$$+ S(t) dh_{1}(t) + h_{1}(t) dS(t) + (dh_{1}(t))(dS(t))$$

and so we want that

$$(*) = B(t) dh_0(t) + (dh_0(t))(dB(t)) + S(t) dh_1(t) + (dh_1(t))(dS(t)) = 0.$$

Inserting the dynamics given and portfolio dynamics above we have

$$(*) = B(t)\frac{\alpha - r}{2}h_0(t) dt + B(t)\frac{1}{2}\sigma h_0(t) dW(t) + 0$$

$$+ S(t)\left(\frac{r - \alpha}{2} + \frac{1}{2}\sigma^2\right)h_1(t) dt - S(t)\frac{1}{2}\sigma h_1(t) dW(t)$$

$$- \frac{1}{2}\sigma h_1(t)\sigma S(t) dt$$

$$= \left(B(t)\frac{\alpha - r}{2}h_0(t) - \frac{\alpha - r}{2}S(t)h_1(t)\right) dt$$

$$+ \left(\frac{1}{2}B(t)\sigma h_0(t) - \frac{1}{2}S(t)\sigma h_1(t)\right) dW(t)$$

We see that this is zero if $h_0(t)B(t) = h_1(t)S(t)$. First we have

$$h_0(t)B(t) = \exp\left(\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{\alpha - r}{2} - \frac{1}{8}\sigma^2\right)t\right)B(t)$$

$$= \exp\left(\frac{1}{2}\left(\alpha - r - \frac{1}{4}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}(t)\right)B(t)$$

$$= \exp\left(\frac{1}{2}\left(\alpha - \frac{1}{2}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}(t)\right)\exp\left(\frac{1}{2}\left(-r + \frac{1}{4}\sigma^2\right)t\right)B(t)$$

$$= (S(t))^{1/2}\exp\left(\left(-\frac{r}{2} + \frac{1}{8}\sigma^2\right)t\right)B(t),$$

and

$$h_1(t)S(t) = \frac{1}{s} \exp\left(-\frac{1}{2}\sigma \overline{W}(t) + \left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2\right)t\right) s \cdot \exp\left(\left(\alpha - \frac{1}{2}\sigma^2\right)t + \sigma \overline{W}(t)\right)$$

$$= \exp\left(\frac{1}{2}\sigma \overline{W}(t) + \left(\frac{r+\alpha}{2} - \frac{7}{8}\sigma^2\right)t\right)$$

$$= \exp\left(\frac{1}{2}\sigma \overline{W}(t) + \frac{1}{2}\left(\alpha - \frac{2}{4}\sigma^2\right)t\right) \exp\left(\frac{1}{2}\left(r - \frac{5}{4}\sigma^2\right)t\right)$$

$$= (S(t))^{1/2} \exp\left(\left(-\frac{5}{8}\sigma^2\right)t\right) B(t)$$

Which does not hold. THIS EXERCISE SHOULD BE ABLE TO BE SOLVED.. \Box (ii): We have that

$$u_1(t) = \frac{h_1(t)S(t)}{V^h(t)}.$$

Using that S is a GBM and $B(t) = e^{rt}$ we have

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$$\begin{split} u_1(t) &= \frac{h_1(t)S(t)}{h_1(t)S(t) + h_0(t)B(t)} \\ &= \frac{e^{-\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2\right)t}e^{\left(\alpha - \frac{1}{2}\sigma^2\right)t + \sigma\overline{W}_t}}{e^{-\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2\right)t}e^{\left(\alpha - \frac{1}{2}\sigma^2\right)t + \sigma\overline{W}_t} + e^{\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{\alpha-r}{2} - \frac{1}{8}\sigma^2\right)t}e^{rt}} \\ &= \frac{se^{\left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2 + \alpha - \frac{1}{2}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}_t}}{e^{\left(\frac{r-\alpha}{2} - \frac{3}{8}\sigma^2 + \alpha - \frac{1}{2}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}_t} + e^{\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{\alpha-r}{2} - \frac{1}{8}\sigma^2 + r\right)t}} \\ &= \frac{e^{\left(\frac{r+\alpha}{2} - \frac{7}{8}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}_t}}{e^{\left(\frac{r+\alpha}{2} - \frac{7}{8}\sigma^2\right)t + \frac{1}{2}\sigma\overline{W}_t} + e^{\frac{1}{2}\sigma\overline{W}(t) + \left(\frac{\alpha+r}{2} - \frac{1}{8}\sigma^2\right)t}} \\ &= \frac{e^{-\frac{7}{8}\sigma^2t}}{e^{-\frac{7}{8}\sigma^2t} + e^{-\frac{1}{8}\sigma^2t}} = \frac{e^{-\frac{6}{8}\sigma^2t}}{e^{-\frac{6}{8}\sigma^2t} + 1}. \end{split}$$

OBVIOUSLY had the previous exercise been done correct we would have $h_1(t)S(t) = h_0(t)B(t)$ i.e. $u_1(t) = \frac{1}{2}$.

Solution (b).

(i): We know that $\pi_i(t) = F_i(t, S(t))$ and so from Ito's formula we have the dynamics (we suppress the argument (t, S(t)) in the derivatives):

$$d\pi_{i}(t) = \frac{\partial F_{i}}{\partial t} dt + \frac{\partial F_{i}}{\partial s} dS(t) + \frac{1}{2} \frac{\partial^{2} F_{i}}{\partial s^{2}} (dS(t))^{2}$$

$$= \frac{\partial F_{i}}{\partial t} dt + \frac{\partial F_{i}}{\partial s} (\alpha S(t) dt + \sigma S(t) d\overline{W}(t)) + \frac{1}{2} \frac{\partial^{2} F_{i}}{\partial s^{2}} \sigma^{2} S(t)^{2} dt$$

$$= \left(\frac{\partial F_{i}}{\partial t} + \frac{\partial F_{i}}{\partial s} \alpha S(t) + \frac{1}{2} \frac{\partial^{2} F_{i}}{\partial s^{2}} \sigma^{2} S(t)^{2} \right) dt + \frac{\partial F_{i}}{\partial s} \sigma S(t) d\overline{W}(t)$$

$$= \underbrace{\frac{\partial F_{i}}{\partial t} + \frac{\partial F_{i}}{\partial s} \alpha S(t) + \frac{1}{2} \frac{\partial^{2} F_{i}}{\partial s^{2}} \sigma^{2} S(t)^{2}}_{\pi_{i}(t)} \pi_{i}(t) dt + \underbrace{\frac{\partial F_{i}}{\partial s} \sigma S(t)}_{=\sigma_{i}(t)} \pi_{i}(t) d\overline{W}(t)$$

as desired. \square

(ii): We have

$$\begin{split} \frac{r - \alpha_i(t)}{\sigma_i(t)} &= \frac{r - \frac{\frac{\partial F_i}{\partial t} + \frac{\partial F_i}{\partial s} \alpha S(t) + \frac{1}{2} \frac{\partial^2 F_i}{\partial s^2} \ \sigma^2 S(t)^2}{\pi_i(t)}}{\frac{\frac{\partial F_i}{\partial s} \sigma S(t)}{\alpha_i(t)}} \\ &= \frac{r \pi_i(t) - \frac{\partial F_i}{\partial t} - \frac{\partial F_i}{\partial s} \alpha S(t) - \frac{1}{2} \frac{\partial^2 F_i}{\partial s^2} \ \sigma^2 S(t)^2}{\frac{\partial F_i}{\partial s} \sigma S(t)} \\ &= \frac{r \pi_i(t) - \frac{\partial F_i}{\partial t} - \frac{\partial F_i}{\partial s} \alpha S(t)}{\frac{\partial F_i}{\partial s} \sigma S(t)} \\ &= \frac{r \pi_i(t) - \frac{\partial F_i}{\partial t} - \frac{\partial F_i}{\partial s} r S(t) - \frac{1}{2} \frac{\partial^2 F_i}{\partial s^2} \ \sigma^2 S(t)^2 + \frac{\partial F_i}{\partial s} r S(t) - \frac{\partial F_i}{\partial s} \alpha S(t)}{\frac{\partial F_i}{\partial s} \sigma S(t)} \\ &= \frac{\frac{\partial F_i}{\partial s} r S(t) - \frac{\partial F_i}{\partial s} \alpha S(t)}{\frac{\partial F_i}{\partial s} \sigma S(t)} = \frac{r - \alpha}{\sigma}, \end{split}$$

where we used the Black-Scholes equation i.e.

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$$r\pi_i(t) - \frac{\partial F_i}{\partial t} - \frac{\partial F_i}{\partial s} rS(t) - \frac{1}{2} \frac{\partial^2 F_i}{\partial s^2} \sigma^2 S(t)^2 = 0$$

for any derivative's arbitrage free pricing process. Since i is not included in the fraction above we have the desired result. \square

Solution (c).

We follow the hint and see that the payout is

$$Y = \max \left\{ C(T, S(T); K, T_1), P(T, S(T); K, T_1) \right\}$$

$$= \left(C(T, S(T); K, T_1) - P(T, S(T); K, T_1) \right)^+ + P(T, S(T); K, T_1)$$

$$= \left(C(T, S(T); K, T_1) - Ke^{-r(T_1 - T)} - C(T, S(T); K, T_1) + S(T) \right)^+ + P(T, S(T); K, T_1)$$

$$= \left(S(T) - Ke^{-r(T_1 - T)} \right)^+ + P(T, S(T); K, T_1)$$

$$= C(T, S(T); Ke^{-r(T_1 - T)}, T) + P(T, S(T); K, T_1)$$

Hence we can hedge this payout with a call option with strike $Ke^{-r(T_1-T)}$ at expiry T and a put with strike K at expiry T_1 , that is

$$\Pi_t[Y] = C(t, S(t); Ke^{-r(T_1 - T)}, T) + P(t, S(t); K, T_1)$$

as desired. \square

Solution (d).

We have that the arbitrage free pricing function F(t,s) has to satisfie the Black-Scholes equation 7.10 i.e.

$$F_t + rsF_s + \frac{1}{2}\sigma^2 s^2 F_{ss} - rF = 0$$

$$F(T,s) = \max \{C(T,s;K_1,T_1), P(T,s;K_2,T_1)\}.$$

which may be written differently in terms of call options, stock price s and zero-coupon bonds. \square

Problem 3

Consider a two-dimensional Black-Scholes model. The market model consist of three assets: A bank account B(t) and two stocks $S_1(t)$ and $S_2(t)$. The P-dynamics of B(t) is

$$dB(t) = rB(t) dt$$

with B(0) = 1 where $r \in \mathbb{R}$ is a constant interest rate. The P-dynamics of $S_1(t)$ and $S_2(t)$ are given by

$$dS_1(t) = \alpha_1 S_1(t) dt + \sigma S_1(t) d\overline{W}_1(t),$$

$$dS_2(t) = \alpha_2 S_2(t) dt + \sigma S_2(t) (d\overline{W}_1(t) + d\overline{W}_2(t)),$$

with $S_1(0) = s_1 > 0$ and $S_2(0) = s_2 > 0$ where $\alpha_1, \alpha_2 \in \mathbb{R}$ and $\sigma > 0$ are constants and $\overline{W}_1(t)$ and $\overline{W}_2(t)$ are two independent P-Brownian motions. The filtration is the one generated by the two Brownian motions. Let T > 0 be a given and fixed (expiry) date.

- a. i. Is the model arbitrage free?
 - ii. Is the model complete?
- b. Compute the covariance of $S_1(T)$ and $S_2(T)$. (Hint: recall cov(X,Y) = E[XY] E[X]E[Y]).

Consider the derivative that at time T pays $X = S_1(T_0) + S_2(T)$ where $0 < T_0 < T$ is a fixed date.

c. Find a hedge portfolio for derivative X.

Solution (a).

(i): The model is arbitrage free if and only if a martingale measure exists. That is if the equation

$$\sigma_t \varphi_t = r - \alpha$$

has at least one solution. We have the following market on matrix form

$$dS_t = D(S_t)\alpha_t dt + D(S_t)\sigma_t dW_t$$

or written out in total

$$\begin{bmatrix} dS_1(t) \\ dS_2(t) \end{bmatrix} = \begin{bmatrix} S_1(t) & 0 \\ 0 & S_2(t) \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix} \begin{bmatrix} dt \\ dt \end{bmatrix} + \begin{bmatrix} S_1(t) & 0 \\ 0 & S_2(t) \end{bmatrix} \begin{bmatrix} \sigma & 0 \\ \sigma & \sigma \end{bmatrix} \begin{bmatrix} d\overline{W}_1(t) \\ d\overline{W}_2(t) \end{bmatrix}.$$

Hence we want to solve

$$\begin{bmatrix} \sigma & 0 \\ \sigma & \sigma \end{bmatrix} \begin{bmatrix} \varphi_1(t) \\ \varphi_2(t) \end{bmatrix} = \begin{bmatrix} r - \alpha_1 \\ r - \alpha_2 \end{bmatrix}.$$

This is easy if σ is invertible. We see that we in fact have that the inverse of σ is

$$\sigma_t^{-1} = \begin{bmatrix} 1/\sigma & 0\\ -1/\sigma & 1/\sigma \end{bmatrix}$$

as we have

$$\begin{bmatrix} 1/\sigma & 0 \\ -1/\sigma & 1/\sigma \end{bmatrix} \begin{bmatrix} \sigma & 0 \\ \sigma & \sigma \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I.$$

Then we clearly have the solution

$$\varphi_t = \begin{bmatrix} 1/\sigma & 0 \\ -1/\sigma & 1/\sigma \end{bmatrix} \begin{bmatrix} r - \alpha_1 \\ r - \alpha_2 \end{bmatrix} = \begin{bmatrix} \frac{r - \alpha_1}{\sigma} \\ \frac{-r + \alpha_1 + r - \alpha_2}{\sigma} \end{bmatrix} = \begin{bmatrix} \frac{r - \alpha_1}{\sigma} \\ \frac{\alpha_1 - \alpha_2}{\sigma} \end{bmatrix}.$$

By defining the likelihood process L_t as

$$dL_t = \varphi_t^{\top} L_t \ d\overline{W}_t, \ L_0 = 1,$$

we know from the Novikov condition that if the integral $E^P[e^{1/2}\int_0^T\|\varphi_t\|^2]$ is finite then L is a martingale with $E^P[L_T]=1$. We see that

$$E^{P}\left[e^{\frac{1}{2}\int_{0}^{T}\|\varphi_{t}\|^{2}}dt\right] = E^{P}\left[e^{\frac{1}{2}\int_{0}^{T}\left(\frac{r-\alpha_{1}}{\sigma}\right)^{2} + \left(\frac{\alpha_{1}-\alpha_{2}}{\sigma}\right)^{2}}dt\right] = E^{P}\left[e^{\frac{1}{2}T\left(\frac{r-\alpha_{1}}{\sigma}\right)^{2} + \frac{1}{2}T\left(\frac{\alpha_{1}-\alpha_{2}}{\sigma}\right)^{2}}\right] < \infty.$$

Hence we have found a martingale measure defined by the likelihood process L above. We conclude that the market is arbitrage free. \square

(ii): The model is complete if the martingale measure is unique. This is equivalent with $Ker[\sigma_t] = \{0\}$ and since σ_t is invertible (diagonal) we have from theorem 14.7 that the model is complete. \square

Solution (b).

We have by definition:

$$cov(S_1(T), S_2(T)) = E[S_1(T)S_2(T)] - E[S_1(T)]E[S_2(T)].$$

Thus we set $Z(t) = S_1(t)S_2(t)$ and evaluate the mean value of Z. By Ito's formula on $f(s_1, s_2) = s_1s_2$ we have

$$\begin{split} dZ(t) &= df(S_1(t), S_2(t)) \\ &= S_2(t) \ dS_1(t) + S_1(t) \ dS_2(t) + \frac{1}{2}(dS_1(t))(dS_2(t)) \\ &= S_2(t)\alpha_1(t)S_1(t) \ dt + S_2(t)\sigma S_1(t) \ d\overline{W}_1(t) \\ &+ S_1(t)\alpha_2(t)S_2(t) \ dt + S_1(t)\sigma S_2(t) \ (d\overline{W}_1(t) + d\overline{W}_2(t)) \\ &+ \frac{1}{2}\sigma^2 S_1(t)S_2(t) \ d\overline{W}_1(t)(d\overline{W}_1(t) + d\overline{W}_2(t)) \\ &= (\alpha_1(t) + \alpha_2(t))S_1(t)S_2(t) \ dt + 2\sigma S_1(t)S_2(t) \ d\overline{W}_1(t) + \sigma S_1(t)S_2(t) \ d\overline{W}_2(t) \\ &+ \frac{1}{2}\sigma^2 S_1(t)S_2(t) \ dt \\ &= (\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)Z(t) \ dt + 2\sigma Z(t) \ d\overline{W}_1(t) + \sigma Z(t) \ d\overline{W}_2(t). \end{split}$$

Thus we have that the terms invovling the Brownian motions will vanish when takings expectation hence

$$\begin{split} E[Z(t)] &= Z(0) + E\left[\int_0^t (\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)Z(s) \ ds\right] \\ &= Z(0) + (\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)\int_0^t E\left[Z(s)\right] \ ds. \end{split}$$

Then we have the dynamics of E[Z(t)] is given as

$$dE[Z(t)] = (\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)E[Z(t)] dt.$$

We may then solve this using $E[Z(0)] = Z(0) = s_1 s_2$:

$$E[Z(T)] = Z(0)e^{(\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)T} = s_1 s_2 e^{(\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)T}.$$

Inserting in the formula for covariance we arrive at

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$$\begin{aligned} cov(S_1(T), S_2(T)) &= E\left[S_1(T)S_2(T)\right] - E[S_1(T)]E[S_2(T)] \\ &= s_1 s_2 e^{(\alpha_1(t) + \alpha_2(t) + \frac{1}{2}\sigma^2)T} - E[S_1(T)]E[S_2(T)] \\ &= s_1 s_2 e^{(\alpha_1 + \alpha_2 + \frac{1}{2}\sigma^2)T} - s_1 e^{\alpha_1 T} s_2^{\alpha_2 T} \\ &= s_1 s_2 e^{(\alpha_1 + \alpha_2)T} \left(e^{\frac{1}{2}\sigma^2 T} - 1\right). \end{aligned}$$

as desired. \square

Solution (c).

We may look at this problem on two subintervals: $[0, T_0]$ and $(T_0, T]$. On the latter we know that the portfolio should consist of $S_1(T_0)$ zero coupon bonds with expiry T and one position in the second stock. Hence on the interval $(T_0, T]$ the hedging portfolio is

$$h(t) = (h_0(t), h_1(t), h_2(t)) = (e^{-r(T-T_0)}S(T_0), 0, 1), t > T_0.$$

Hence we on the interval $[0, T_0]$ we want to replicate the derivative $\widetilde{X} = e^{-r(T-T_0)}S(T_0)$. This is obviously easy since we should hold $e^{-r(T-T_0)}$ of the first stock. Then we have

$$h(t) = \begin{cases} \left(0, e^{-r(T-T_0)}, 1\right) & \text{for } t \le T_0, \\ \left(e^{-r(T-T_0)}S(T_0), 0, 1\right) & \text{for } t > T_0. \end{cases}$$

This then give a self-financing portfolio with value process

$$V^{h}(t) = \begin{cases} S_{1}(t)e^{-r(T-T_{0})} + S_{2}(t) & \text{for } t \leq T_{0}, \\ e^{-r(T-t)}S(T_{0}) + S_{2}(t) & \text{for } t > T_{0}. \end{cases}$$

as desired. \square

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Problem 1

Let W(t) denote a Brownian motion and let $\mathcal{F}_t = \mathcal{F}_t^W$. Let T > 0 be a given and fixed time.

Consider the two dimensional stochastic differential equation

$$dX(t) = \frac{1}{2}X(t) dt + Y(t) dW(t),$$

$$dY(t) = \frac{1}{2}Y(t) dt + X(t) dW(t),$$

with X(0) = 0 and Y(1) = 1.

- a. Show that $(X(t), Y(t)) = (\sinh(W(t)), \cosh(W(t)))$ solves the two-dimensional stochastic differential equation. (Hint: Recall that $\sinh(x) = \frac{1}{2}(e^x e^{-x})$ and $\cosh(x) = \frac{1}{2}(e^x + e^{-x})$).
- b. i. Show that $M(t) = e^{-t/2} \cosh(W(t))$ is a martingale.
 - ii. Find a constant z and a process h(t) such that

$$\cosh(W(T)) = z + \int_0^T h(t) \ dW(t).$$

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Let L(t) be a Likelihood process and let dQ = L(T)dP be a new probability measure.

c. Determine the Likelihood process L(t) such that sinh(W(t)) is a martingale under the probability measure Q.

Solution (a).

Assume that $X(t) = \sinh(W(t))$ and $Y(t) = \cosh(W(t))$. The relevant derivatives is then

$$\frac{d}{dw}\sinh(w) = \frac{1}{2}e^w + \frac{1}{2}e^{-w} = \cosh(w), \ \frac{d^2}{dw^2}\sinh(w) = \frac{d}{dw}\cosh(w) = \frac{1}{2}e^w - \frac{1}{2}e^{-w} = \sinh(w).$$

That is sinus and cosinus hyperbolic are their each others derivative. Then by Ito's formula we have

$$dX(t) = \cosh(W(t)) \ dW(t) + \frac{1}{2} \sinh(W(t)) \ (dW(t))^{2}$$
$$= \frac{1}{2} \sinh(W(t)) \ dt + \cosh(W(t)) \ dW(t)$$
$$= \frac{1}{2} X(t) \ dt + Y(t) \ dW(t).$$

and

$$dY(t) = \sinh(W(t)) \ dW(t) + \frac{1}{2} \cosh(W(t)) \ (dW(t))^{2}$$
$$= \frac{1}{2} \cosh(W(t)) \ dt + \sinh(W(t)) \ dW(t)$$
$$= \frac{1}{2} Y(t) \ dt + X(t) \ dW(t).$$

And thus the result has been prooved. \square

Solution (b).

(i): Consider the function f(z,y) = zy. Then we have that M(t) = f(Z(t), Y(t)) for $Z(t) = e^{-t/2}$ hence M has dynamics given by Ito's formula:

$$\begin{split} dM(t) &= df(Z(t), Y(t)) \\ &= Y(t) \ dZ(t) + Z(t) \ dY(t) + (dZ(t))(dY(t)) \\ &= Y(t) \ (-\frac{1}{2}Z(t) \ dt) + Z(t) \ (\frac{1}{2}Y(t) \ dt + X(t) \ dW(t)) + (-\frac{1}{2}Z(t) \ dt)(\frac{1}{2}Y(t) \ dt + X(t) \ dW(t)) \\ &= X(t)Z(t) \ dW(t). \end{split}$$

and so we see that pr. lemma 4.11 M is a martingale. \Box

(ii): We have from above

$$M(T) = M(0) + \int_0^T X(t)Z(t) \ dW(t) = Z(T)\cosh(W(t))$$

Hence it follows that

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$$\cosh(W(T)) = \frac{M(0)}{Z(T)} + \int_0^T \frac{X(t)Z(t)}{Z(T)} dW(t).$$

Using that the martingale has initial value

$$M(0) = e^{-0/2} \cosh(0) = 1$$

we have

$$\cosh(W(T)) = e^{T/2} + \int_0^T \sinh(W(t))e^{(T-t)/2} \ dW(t).$$

In total we have $z = e^{T/2}$ and $h(t) = \sinh(W(t))e^{(T-t)/2}$ as desired. \square

Solution (c).

We have that under the measure Q the dynamics of W is given by the Girsanov Theorem

$$dW(t) = \varphi \ dt + dW_t^Q,$$

where φ is the Girsanov kernel associated with L. Then we know that X has dynamics under the Q measure:

$$dX(t) = \frac{1}{2}X(t) dt + Y(t) (\varphi dt + dW_t^Q)$$
$$= \left(\frac{1}{2}X(t) + \varphi Y(t)\right) dt + Y(t) dW_t^Q$$

and so we would have that X is a martingale under Q if

$$\varphi_t = -\frac{1}{2} \frac{X(t)}{Y(t)} = -\frac{1}{2} \frac{\sinh(W(t))}{\cosh(W(t))} = -\frac{1}{2} \tanh(W(t)).$$

Then we can define a Likelihood process with initial condition $L_0 = 1$ and dynamics $dL_t = \varphi_t L_t \ dW(t)$ i.e. L is the solution

$$L_t = \exp\left\{ \int_0^s -\frac{1}{2} \tanh(W(s)) \ dW(t) - \frac{1}{2} \int_0^t \left(-\frac{1}{2} \tanh(W(t)) \right)^2 \ dW(t) \right\} > 0.$$

We lastly show that the Novikov condition is satisfied i.e.

$$E^{P}\left[e^{\frac{1}{2}\int_{0}^{T}\left\|\varphi_{s}\right\|^{2}\ ds}\right] = E^{P}\left[e^{\frac{1}{8}\int_{0}^{T}\tanh^{2}(W(t))\ ds}\right] < \infty$$

and so L is a P-martingale and L is a Likelihood process. We thus have found a Likelihood process such that X is a martingale under the measure Q given by $dQ = L_T dP$. \square

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