## SOLUTIONS **FINANCIAL DERIVATIVES AND STOCHASTIC ANALYSIS** (CTH[*tma*285], GU[*MMA*710])

December 13, 2011, morning, V No aids. Questions on the exam: Christer Borell, 0705 292322 Each problem is worth 3 points.

1. Let  $W = (W_1(t), W_2(t))_{0 \le t \le T}$  be a two-dimensional standard Brownian motion and  $(\mathcal{F}(t))_{0 \le t \le T}$  a filtration for W. Find an adapted process  $(\Gamma(t))_{0 \le t \le T}$  such that

$$W_1^2(t) - W_2^2(t) = \int_0^t \Gamma(u) \cdot dW(u), \ 0 \le t \le T.$$

Solution. We have

$$d(W_1^2(t) - W_2^2(t)) = dW_1^2(t) - dW_2^2(t) = 2W_1(t)dW_1(t) + \frac{1}{2}2(dW_1(t))^2$$
$$-2W_2(t)dW_2(t) - \frac{1}{2}2(dW_2(t))^2 = 2W_1(t)dW_1(t) + dt - 2W_2(t)dW_2(t) - dt$$
$$= 2W_1(t)dW_1(t) - 2W_2(t)dW_2(t)$$

and

$$W_1^2(t) - W_2^2(t) = \int_0^t (2W_1(u), -2W_2(u)) \cdot dW(u).$$

Thus  $\Gamma(t) = (2W_1(t), -2W_2(t)), 0 \le t \le T$ , is a solution of the problem.

2. (Black-Scholes model for two stocks with non-singular volatility matrice  $\sigma = (\sigma_{ij})_{1 \le i,j \le 2}$  and interest rate r > 0) Suppose K, T > 0 and consider a financial derivative of European type with the payoff

$$Y = (\frac{S_1(T)}{S_2(T)} - K)^+$$

at time of maturity T. (a) Find the price  $\Pi_Y(t)$  of the derivative at time  $t \in [0, T[$ . (b) Find adapted processes  $(\Delta_i(t))_{0 \le t < T}$ , i = 1, 2, such that

$$d\Pi_Y(t) = \sum_{i=1}^2 \Delta_i(t) dS_i(t) + r \left\{ \Pi_Y(t) - \sum_{i=1}^2 \Delta_i(t) S_i(t) \right\} dt$$

for  $0 \le t < T$ .

Solution. (a) With  $\sigma_i = (\sigma_{i1}, \sigma_{i2})$  and standard notation, we have

$$S_i(t) = S_i(0)e^{(r - \frac{|\sigma_i|^2}{2})t + \sigma_i \cdot \tilde{W}(t)}, \ i = 1, 2.$$

Hence, if  $\tau = T - t > 0$ ,

$$\Pi_{Y}(t) = e^{-r\tau} \tilde{E} \left[ \left( \frac{S_{1}(T)}{S_{2}(T)} - K \right)^{+} \mid \mathcal{F}(t) \right]$$
  
$$= e^{-r\tau} \tilde{E} \left[ \left( \frac{S_{1}(t)}{S_{2}(t)} e^{\left(\frac{|\sigma_{2}|^{2}}{2} - \frac{|\sigma_{1}|^{2}}{2}\right)\tau + (\sigma_{1} - \sigma_{2}) \cdot (\tilde{W}(T) - \tilde{W}(t))} - K \right)^{+} \mid \mathcal{F}(t) \right]$$
  
$$= e^{-r\tau} \tilde{E} \left[ \left( se^{\left(\frac{|\sigma_{2}|^{2}}{2} - \frac{|\sigma_{1}|^{2}}{2}\right)\tau + \sqrt{\tau}(\sigma_{1} - \sigma_{2}) \cdot \tilde{W}(1)} - K \right)^{+} \right]_{s = \frac{S_{1}(t)}{S_{2}(t)}}$$
  
$$= e^{-r\tau} E \left[ \left( se^{\left(\frac{|\sigma_{2}|^{2}}{2} - \frac{|\sigma_{1}|^{2}}{2}\right)\tau + \sqrt{\tau}|\sigma_{1} - \sigma_{2}|G} - K \right)^{+} \right]_{s = \frac{S_{1}(t)}{S_{2}(t)}}$$

where  $G \in N(0, 1)$ . Thus, if

$$\hat{r} - \frac{|\sigma_1 - \sigma_2|^2}{2} = \frac{|\sigma_2|^2}{2} - \frac{|\sigma_1|^2}{2}$$

we have

$$\hat{r} = \mid \sigma_2 \mid^2 - \sigma_1 \cdot \sigma_2$$

and

$$\Pi_Y(t) = a e^{-\hat{r}\tau} E\left[ \left( s e^{\left(\hat{r} - \frac{|\sigma_1 - \sigma_2|^2}{2}\right)\tau + \sqrt{\tau}|\sigma_1 - \sigma_2|G} - K \right)^+ \right]_{s = \frac{S_1(t)}{S_2(t)}}$$

where

$$a = e^{(\hat{r} - r)\tau}.$$

Now by the Black-Scholes call price formula

$$\Pi_Y(t) = v(t, S_1(t), S_2(t)) = a\left(\frac{S_1(t)}{S_2(t)}\Phi(d_+(\tau, \frac{S_1(t)}{S_2(t)})) - Ke^{-\hat{r}\tau}\Phi(d_-(\tau, \frac{S_1(t)}{S_2(t)}))\right)$$

where

$$d_{\pm}(\tau, \frac{S_1(t)}{S_2(t)}) = \frac{\ln \frac{S_1(t)}{KS_2(t)} + (\hat{r} \pm \frac{|\sigma_1 - \sigma_2|^2}{2})\tau}{|\sigma_2 - \sigma_2| \sqrt{\tau}}$$

that is

$$d_{+}(\tau, \frac{S_{1}(t)}{S_{2}(t)}) = \frac{\ln \frac{S_{1}(t)}{KS_{2}(t)} + \left(\frac{|\sigma_{1}|^{2}}{2} + \frac{3|\sigma_{2}|^{2}}{2} - 2\sigma_{1} \cdot \sigma_{2}\right)\tau}{|\sigma_{2} - \sigma_{1}|\sqrt{\tau}}$$

and

$$d_{-}(\tau, \frac{S_{1}(t)}{S_{2}(t)}) = \frac{\ln \frac{S_{1}(t)}{KS_{2}(t)} + \left(-\frac{|\sigma_{1}|^{2}}{2} + \frac{|\sigma_{2}|^{2}}{2}\right)\tau}{|\sigma_{2} - \sigma_{1}|\sqrt{\tau}}$$

(b) By martingale representation the derivative may be replicated and

$$d\Pi_Y(t) = v'_{s_1}(t, S_1(t), S_2(t))dS_1(t) + v'_{s_2}(t, s_1, S_2(t))dS_2(t) + (\dots)dt.$$

Thus remembering the delta of a standard call,

$$\Delta_1(t) = v'_{s_1}(t, S_1(t), S_2(t)) = \frac{a}{S_2(t)} \Phi(d_+(\tau, \frac{S_1(t)}{S_2(t)}))$$

and

$$\Delta_2(t) = v'_{s_1}(t, S_1(t), S_2(t)) = -\frac{aS_1(t)}{S_2(t)^2} \Phi(d_+(\tau, \frac{S_1(t)}{S_2(t)})).$$

3. An interest rate process  $(R(t))_{t\geq 0}$  is governed by the equation

$$dR(t) = a(b - R(t))dt + \sigma dW(t), \ t \ge 0$$

where  $W(t), t \ge 0$ , is a one-dimensional Brownian motion and a, b, and  $\sigma$  positive constants.

Suppose R(0) < b and let  $\tau = \min \{t > 0; R(t) = b\}$ . Find  $P[\tau \le T]$  if T > 0.

(Hint: Show that there is a strictly increasing function  $f(v), v \ge 0$ , such that the process

$$B(v) = \int_0^{f(v)} e^{at} dW(t), \ v \ge 0$$

is a standard Brownian motion.)

Solution. Set X(t) = R(t) - b so that

$$dX(t) = -aX(t)dt + \sigma dW(t), \ t \ge 0$$

and

$$X(t) = e^{-at} (X(0) + \sigma \int_0^t e^{au} dW(u)).$$

Thus

$$\tau = \min\left\{t > 0; \int_0^t e^{au} dW(u) = \frac{1}{\sigma}(b - R(0))\right\}.$$

But

$$Y(t) =_{def} \int_0^t e^{au} dW(u) \in N(0, f^{-1}(t))$$

with  $v = f^{-1}(t) = \frac{1}{2a}(e^{2at} - 1), 0 \leq t < \infty$ , and  $(Y(t))_{t\geq 0}$  is a centred Gaussian process with covariance

$$E[Y(s)Y(t)] = E\left[\int_{0}^{s} e^{au} dW(u) \int_{0}^{t} e^{au} dW(u)\right]$$
$$= E\left[\int_{0}^{\min(s,t)} e^{au} dW(u) \int_{0}^{\min(s,t)} e^{au} dW(u)\right] = \operatorname{Var}(Y(\min(s,t)))$$
$$= f^{-1}(\min(s,t)) = \min(f^{-1}(s), f^{-1}(t))$$

where the third equality follows from the Itô isometry. Now

$$B(v) = Y(f(v)) = Y(\frac{1}{2a}\ln(1+2av)), \ v \ge 0$$

is a centred Gaussian process with covariance

$$E\left[B(u)B(v)\right] = \min(u, v)$$

and we conclude that the process  ${\cal B}$  is a standard Brownian motion. Moreover, if

$$c = \frac{1}{\sigma}(b - R(0)),$$

the Bachelier double law yields

$$P[\tau \le T] = P\left[\max_{0 \le t \le T} Y(t) \ge c\right]$$
$$= P\left[\max_{0 \le v \le \frac{1}{2a}(e^{2aT} - 1)} B(v) \ge c\right] = 2(1 - \Phi(\frac{c}{\sqrt{\frac{1}{2a}(e^{2aT} - 1)}}))$$
$$= 2\Phi(\frac{R(0) - b}{\sigma\sqrt{\frac{1}{2a}(e^{2aT} - 1)}}).$$

4. Let W be a one-dimensional standard Brownian motion and set

$$Q = \sum_{i=1}^{n} (W(t_i) - W(t_{i-1}))^2$$

where  $0 = t_0 < t_1 < \dots t_{n-1} < t_n = T < \infty$ . Show that E[Q] = T and

$$\operatorname{Var}(Q) \le 2T \max_{1 \le i \le n} (t_i - t_{i-1}).$$

5. Let W be a one-dimensional standard Brownian motion and suppose m > 0 and  $\tau_m = \min \{t > 0; W(t) = m\}$ . Use the formula

$$P[\tau_m \le t, W(t) \le w] = P[W(t) \ge 2m - w], w \le m,$$

to prove that  $\tau_m$  has the density

$$f_m(t) = \frac{m}{t\sqrt{2\pi t}}e^{-\frac{m^2}{2t}}, \ t > 0.$$