SOLUTIONS: FINANCIAL DERIVATIVES AND STOCHASTIC ANALYSIS (CTH[tma285]&GU[MMA710])

December 16, 2008, morning (4 hours), v Each problem is worth 3 points. No aids. Examiner: Christer Borell, telephone number 0705292322

If not otherwise stated W denotes a one-dimensional standard Brownian motion.

1. (Black-Scholes model) A portfolio consists of a shares of the stock and b shares of the bond, where a, b > 0 are constants. The value process of the portfolio is denoted by $N = (N(t))_{0 \le t \le T}$.

(a) Determine an adapted process $(\nu(t))_{0 \le t \le T}$ such that

$$dN(t) = N(t)(rdt + \nu(t)d\tilde{W}(t)), 0 \le t \le T.$$

(b) The process $\tilde{W}^{(N)}(t) = \tilde{W}(t) - \int_0^t \nu(u) du$, $0 \le t \le T$, is a standard Brownian motion with respect to the probability measure $\tilde{P}^{(N)}$. Determine adapted processes $(\gamma_S(t))_{0\le t\le T}$ and $(\gamma_B(t))_{0\le t\le T}$ such that

$$d\frac{S(t)}{N(t)} = \frac{S(t)}{N(t)}\gamma_S(t)d\tilde{W}^{(N)}(t)$$

and

$$d\frac{B(t)}{N(t)} = \frac{B(t)}{N(t)}\gamma_B(t)d\tilde{W}^{(N)}(t),$$

respectively.

Solution. (a) We have that

$$dN(t) = adS(t) + bdB(t)$$

= $aS(t)(rdt + \sigma d\tilde{W}(t)) + brB(t)dt$
= $r(aS(t) + bB(t))dt + \sigma aS(t)d\tilde{W}(t)$
= $N(t)(rdt + \frac{\sigma aS(t)}{aS(t) + bB(t)}d\tilde{W}(t)).$

Thus

$$\nu(t) = \frac{\sigma a S(t)}{a S(t) + b B(t)}.$$

(b) It is known that

$$\gamma_S(t) = \sigma - \nu(t)$$
$$= \sigma - \frac{\sigma a S(t)}{a S(t) + b B(t)} = \frac{\sigma b B(t)}{a S(t) + b B(t)}$$

and

$$\begin{split} \gamma_B(t) &= 0 - \nu(t) \\ &= -\frac{\sigma a S(t)}{a S(t) + b B(t)}. \end{split}$$

2. (Black-Scholes model for two stocks) Suppose t < T. Find the price at time t of a derivative paying the amount $Y = \max(0, \ln \frac{S_1(T)}{S_2(T)})$ at time of maturity T.

Solution. Suppose

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

If $\sigma = \sigma_1 - \sigma_2$, $\tau = T - t$, and $x = \frac{S_1(t)}{S_2(t)}$, then

$$\frac{S_1(T)}{S_2(T)} = x e^{(\frac{|\sigma_2|^2}{2} - \frac{|\sigma_1|^2}{2})\tau + \sigma \cdot (\tilde{W}(T) - \tilde{W}(t))}.$$

Now set $a = \frac{|\sigma_2|^2}{2} - \frac{|\sigma_1|^2}{2}$ and suppose $G \in N^{\tilde{P}}(0,1)$ so that

$$\Pi_{Y}(t) = e^{-r\tau} \tilde{E} \left[\left(\ln \frac{S_{1}(T)}{S_{2}(T)} \right)^{+} \mid \mathcal{F}_{t} \right]$$
$$= e^{-r\tau} \tilde{E} \left[\left(\ln (x e^{a\tau + \sigma \cdot (\tilde{W}(T) - \tilde{W}(t))}) \right)^{+} \mid \mathcal{F}_{t} \right]$$
$$= e^{-r\tau} \tilde{E} \left[\left(\ln (x e^{a\tau + |\sigma| \sqrt{\tau}G}) \right)^{+} \right]$$
$$= e^{-r\tau} \tilde{E} \left[\left(a\tau + \ln x - \mid \sigma \mid \sqrt{\tau}G \right)^{+} \right]$$
$$= e^{-r\tau} \int_{-\infty}^{\frac{a\tau + \ln x}{|\sigma| \sqrt{\tau}}} (a\tau + \ln x - \mid \sigma \mid \sqrt{\tau}y) e^{-\frac{y^{2}}{2}} \frac{dy}{\sqrt{2\pi}}$$

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$$= e^{-r\tau} \left\{ (a\tau + \ln x)N(\frac{\ln x + a\tau}{|\sigma|\sqrt{\tau}}) + |\sigma| \sqrt{\frac{\tau}{2\pi}} e^{-\frac{1}{2} \left\{ \frac{a\tau + \ln x}{|\sigma|\sqrt{\tau}} \right\}^2} \right\}.$$

3. Suppose T is a positive constant and $\sigma(t)$, $0 \leq t \leq T$, a nonrandom continuous function not identically equal to zero.

(a) Set

$$X(t) = \int_0^t \sigma(u) dW(u), \ 0 \le t \le T,$$

and

$$Y(t) = W(\int_0^t \sigma^2(u) du), \ 0 \le t \le T.$$

Prove that the processes $(X(t))_{0 \le t \le T}$ and $(Y(t))_{0 \le t \le T}$ have the same distribution.

(b) Consider an interest rate process $(R(t))_{0 \le t \le T}$ with R(0) = r and such that

$$dR(t) = (a - bR(t))dt + \sigma(t)dW(t), \ 0 \le t \le T,$$

where a and b are positive constants. Find the density of the random variable

$$Z = \max_{0 \le t \le T} \left\{ e^{bt} (R(t) - E[R(t)]) \right\}.$$

Solution. (a). Note that $(X(t))_{0 \le t \le T}$ and $(Y(t))_{0 \le t \le T}$ are mean-zero Gaussian processes. Therefore, if these processes have the same covariances they must have the same distribution.

If $0 \le s \le t \le T$,

$$E[X(s)X(t)] = E\left[\int_0^s \sigma(u)dW(u)\int_0^t \sigma(u)dW(u)\right]$$
$$= E\left[\int_0^t \mathbf{1}_{[0,s,]}(u)\sigma(u)dW(u)\int_0^t \sigma(u)dW(u)\right] = \int_0^t \sigma^2(u)\mathbf{1}_{[0,s]}(u)du$$
$$= \int_0^s \sigma^2(u)du$$

and, therefore, for all $0 \le s, t \le T$,

$$E[X(s)X(t)] = \int_0^{\min(s,t)} \sigma^2(u) du.$$

Moreover,

$$E[Y(s)Y(t)] = E\left[W(\int_0^s \sigma^2(u)du)W(\int_0^t \sigma^2(u)du)\right]$$
$$= \min(\int_0^s \sigma^2(u)du, \int_0^t \sigma^2(u)du) = \int_0^{\min(s,t)} \sigma^2(u)du$$

and Part (a) is solved.

(b) We have

$$d(e^{bt}R(t)) = ae^{bt} + \sigma(t)e^{bt}dW(t)$$

and, hence,

$$e^{bt}R(t) = r + \frac{a}{b}(e^{bt} - 1) + \int_0^t \sigma(u)e^{bu}dW(u).$$

Now $e^{bt}E[R(t)] = E[e^{bt}R(t)] = r + \frac{a}{b}(e^{bt} - 1)$ and we get

$$Z = \max_{0 \le t \le T} \int_0^t \sigma(u) e^{bu} dW(u).$$

Next let x be a positive number. By using Part (a) it follows that

$$P\left[Z \ge x\right] = P\left[\max_{0 \le t \le T} W\left(\int_0^t \sigma^2(s) e^{2bs} ds\right) \ge x\right]$$

and the Bachelier double law gives

$$P[Z \ge x] = 2P\left[W(\int_0^T \sigma^2(u)e^{2bu}ds) \ge x\right]$$
$$= 2(1 - N(\frac{x}{a}))$$

where

$$a = \sqrt{\int_0^T \sigma^2(u) e^{2bu} ds}.$$

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Hence, the density function of Z equals

$$f_Z(x) = \begin{cases} \frac{2}{a\sqrt{2\pi}} e^{-\frac{x^2}{2a^2}}, \ x > 0\\ 0, \text{ otherwise.} \end{cases}$$

4. Let $(\mathcal{F}(t))_{t\geq 0}$ be a filtration for W. Show that W is a Markov process.

5. Consider a model with m stocks and a discount process $(D(t))_{0 \le t \le T}$. The stock price processes are Itô processes defined by a d-dimensional Brownian motion W and a filtration \mathcal{F}_t , $0 \le t \le T$ for W and, in addition,

$$D(t) = e^{-\int_0^t R(u)du}, 0 \le t \le T,$$

where the process $(R(t))_{0 \le t \le T}$ is adapted.

(a) Explain the notion of risk-neutral measure.

(b) Let $X = (X(t))_{0 \le t \le T}$ be a portfolio value process and \tilde{P} a risk-neutral measure. Show that $(D(t)X(t))_{0 \le t \le T}$ is a martingale under \tilde{P} .