

Homework 11 (Sol)

Math 622

May 4, 2016

1.

The solution follows the method in class for deriving the Black-Scholes-Merton formula with random interest rate (see Theorem 9.4.2 in Shreve).

According to Theorem 9.2 (here the role of $\nu(t)$ is played by $3-t$),

$$\widetilde{W}^{T_3}(t) = \widetilde{W}(t) - \int_0^t (3-s) ds = \widetilde{W}(t) + (1/2)[(3-t)^2 - 9]$$

is a Brownian motion under $\tilde{\mathbf{P}}^{(3)}$, for $t \leq 3$, and

$$d_t \text{For}_S(t) = \text{For}_S(t)[\sigma - (3-t)] d\widetilde{W}^{T_3}(t) = \text{For}_S(t)[t+\sigma-3] d\widetilde{W}^{T_3}(t).$$

Then $\text{For}_S(3) = \frac{s_0}{B(0,3)} e^{X-\nu^2/2}$, where

$$X = \int_0^3 [s+\sigma-3] d\widetilde{W}^{T_3}(s),$$

is, under the measure $\tilde{\mathbf{P}}^{(3)}$, a Gaussian random variable with mean 0 and variance $\nu^2 = \int_0^3 [s+\sigma-3]^2 ds = (1/3)[\sigma^3 - (\sigma+3)^3]$. Therefore, using Theorem 2 in Lecture Notes 12, the forward price of the call at $t = 0$ is

$$\begin{aligned} \frac{V(0)}{B(0,3)} &= \tilde{E}^{T_3}[(S(3) - K)^+] \\ &= \frac{s_0}{B(0,3)} N\left(\frac{\ln(s_0/B(0,3)K) + \nu^2/2}{\nu}\right) - KN\left(\frac{\ln(s_0/B(0,3)K) - \nu^2/2}{\nu}\right) \end{aligned}$$

As a result,

$$V(0) = s_0 N\left(\frac{\ln(s_0/B(0,3)K) + \nu^2/2}{\nu}\right) - KB(0,3)N\left(\frac{\ln(s_0/B(0,3)K) - \nu^2/2}{\nu}\right).$$

2. (Shreve, 10.9). By equation (5) in Lecture Notes 12,

$$f(t, T) = f(0, T) + \int_0^t \alpha(u, T) du + \int_0^t \sum_{j=1}^d \sigma_j(u, T) dW_j(u). \quad (1)$$

Following the derivation of section 10.3.2, show that

$$d_t \left(- \int_t^T f(t, v) dv \right) = R(t) dt - \alpha^*(t, T) dt - \sum_{j=1}^d \sigma_j^*(t, T) dW_j(t)$$

By applying Itô's rule to $B(t, T) = e^{-\int_t^T f(t, v) dv}$, it follows that

$$\begin{aligned} dB(t, T) &= B(t, T) \left[R(t) - \alpha^*(t, T) + \frac{1}{2} \sum_{j=1}^d (\sigma_j^*(t, T))^2 \right] dt \\ &\quad - B(t, T) \sum_{j=1}^d \sigma_j^*(t, T) dW_j(t) \end{aligned}$$

Write

$$\begin{aligned} dB(t, T) &= B(t, T) \left[R(t) - \alpha^*(t, T) + \frac{1}{2} \sum_{j=1}^d (\sigma_j^*(t, T))^2 + \sum_{j=1}^d \sigma_j^*(t, T) \Theta_j(t) \right] dt \\ &\quad - B(t, T) \sum_{j=1}^d \sigma_j^*(t, T) d\widetilde{W}_j(t) \end{aligned}$$

where $\widetilde{W}_j(t) = W_j(t) + \int_0^t \Theta_j(s) ds$. When we change measure to make $\widetilde{W}(t) = (\widetilde{W}_1(t), \dots, \widetilde{W}_d(t))$ a Brownian motion, the model will be risk-neutral if

$$\alpha^*(t, T) = \frac{1}{2} \sum_{j=1}^d (\sigma_j^*(t, T))^2 + \sum_{j=1}^d \sigma_j^*(t, T) \Theta_j(t).$$

Take derivatives on both sides with respect to T to obtain,

$$\alpha(t, T) = \sum_{j=1}^d \sigma_j^*(t, T) \sigma_j(t, T) + \sum_{j=1}^d \sigma_j(t, T) \Theta_j(t).$$

(ii) Suppose there is a solution $\Theta(t) = (\Theta_1(t), \dots, \Theta_d(t))$. If T_1, \dots, T_d is a set of distinct times then

$$\alpha(t, T_i) = \sum_{j=1}^d \sigma_j^*(t, T_i) \sigma_j(t, T_i) + \sum_{j=1}^d \sigma_j(t, T_i) \Theta_j(t)$$

for $1 \leq i \leq d$. If the matrix $[\sigma_j(t, T_i)]_{1 \leq i, j \leq d}$ is invertible, this system has a unique solution $\Theta(t)$ for all $t \leq \min_k T_k$.

3. a) In this case $\sigma^*(t, T) = \int_t^T tv \, dv = (1/2)t(T^2 - t^2)$. In order that the model be arbitrage-free, there must exist a solution $\theta(t)$ to $\alpha(t, T) = \sigma(t, T)[\sigma^*(t, T) + \theta(t)]$, or

$$\frac{T^3 t^2}{2} + 5Tt - \frac{Tt^4}{2} = tT[(1/2)t(T^2 - t^2) + \theta(t)].$$

Subtracting $\frac{T^3 t^2}{2} - \frac{Tt^4}{2}$ from both sides leaves $5Tt = tT\theta(t)$. Therefore there is a solution with $\theta(t) \equiv 5$, and so the model is arbitrage-free.

b) Let $\sigma_1(t, T) = 1$ and $\sigma_2(t, T) = 2T$. Then $\sigma_1^*(t, T) = T - t$ and $\sigma_2^*(t, T) = \int_t^T 2u \, du = T^2 - t^2$.

From Exercise 10.9, for the model to be arbitrage-free there must be a solution $\theta_1(t), \theta_2(t)$, independent of T , to

$$\alpha(t, T) = T - t - 2Tt^2 = [T - t + \theta_1(t)] + 2T[T^2 - t^2 + \theta_2(t)].$$

By cancellation of terms common to both sides, $(\theta_1(t), \theta_2(t))$ must solve

$$0 = \theta_1(t) + 2T\theta_2(t) + 2T^3$$

for all $0 \leq t \leq T \leq \bar{T}$. If this were true, then taking partial derivatives with respect to T on both sides implies $2\theta_2(t) + 6T^2 = 0$. But this contradicts the condition that $\theta_2(t)$ is independent of T , and hence there can be no solution of the required form. Therefore, we conclude that the given model is not arbitrage-free.

4. (Shreve, Exercise 10.11) The value at $t = 0$ of a payment of δK at T is $\delta KB(0, T)$. The value at $t = 0$ of a series of payments of δK at time T_1, \dots, T_{n+1} is thus $\delta K \sum_{j=1}^{n+1} B(0, T_j)$. By Theorem 10.4.1, the value at $t = 0$ of a payment of amount $\delta L(T_{j-1}, T_{j-1})$ at T_j is $\delta B(0, T_j)L(0, T_{j-1})$ —see equation (10.4.5). The value of a contract at $t = 0$ promising fixed legs in return for paying floating legs is therefore

$$\delta K \sum_{j=1}^{n+1} B(0, T_j) - \delta \sum_{j=1}^{n+1} B(0, T_j)L(0, T_{j-1}) \quad (2)$$

5. For the one-factor Vasicek model, $dR(t) = (a - bR(t)) \, dt + \sigma R(t) \, d\widetilde{W}(t)$, the results of section 10.3.5 show that

$$d[D(t)B(t, T)] = -\sigma^*(t, T)[D(t)B(t, T)] \, d\widetilde{W}(t),$$

where $\sigma^*(t, T) = \frac{\sigma}{b}(1 - e^{-b(T-t)})$. By (10.4.9) and (10.4.15),

$$\begin{aligned} dL(t, T) &= \frac{1 + \delta L(t, T)}{\delta L(t, T)} [\sigma^*(t, T + \delta) - \sigma^*(t, T)] L(t, T) d\widetilde{W}^{T+\delta}(t) \\ &= [1 + \delta L(t, T)] \frac{\sigma e^{-b(T-t)}(1 - e^{-b\delta})}{\delta b} d\widetilde{W}^{T+\delta}(t). \end{aligned}$$

Let $Y(t) = 1 + \delta L(t, T)$. Then it follows that

$$dY(t) = \delta dL(t, T) = Y(t)\beta(t, T) d\widetilde{W}^{T+\delta}(t),$$

where $\beta(t, T) = \frac{\sigma e^{-b(T-t)}(1 - e^{-b\delta})}{b}$. This has the solution

$$Y(t) = [1 + \delta L(0, T)] \exp\left\{\int_0^t \beta(u, T) d\widetilde{W}^{T+\delta}(u) - \frac{1}{2} \int_0^t \beta^2(u, T) du\right\}.$$

Let $B = \int_0^T \beta(u, T) d\widetilde{W}^{T+\delta}(u)$. This is a normal random variable with mean 0 and variance $\int_0^T \beta^2(u, T) du$. Then

$$L(T, T) = \delta^{-1}[Y(T) - 1] = \delta^{-1}\left[(1 + \delta L(0, T))e^{B - (1/2)\int_0^T \beta^2(u, T) du} - 1\right].$$

Let $V(0)$ denote the price of a caplet at strike K for $[T, T + \delta]$. The $T + \delta$ -forward price is thus

$$\begin{aligned} \frac{V(0)}{B(0, T + \delta)} &= \tilde{E}^{T+\delta}\left[(L(t, T) - K)^+\right] \\ &= \tilde{E}^{T+\delta}\left[\left(\delta^{-1}(1 + \delta L(0, T))e^{B - (1/2)\int_0^T \beta^2(u, T) du} - \delta^{-1} - K\right)^+\right] \end{aligned}$$

The Black-Scholes formula tells us how to price this. It is the same as the price of a call at strike $\delta^{-1} + K$, when $\sigma^2 T = \int_0^T \beta^2(u, T) du$, $r = 0$ and the initial price is $\delta^{-1} + L(0, T)$. This is

$$(\delta^{-1} + L(0, T))N(\bar{d}_+) - (\delta^{-1} + K)N(\bar{d}_-),$$

where

$$d_{\pm} = \frac{1}{\sqrt{\int_0^T \beta^2(u, T) du}} \left[\log \frac{1 + \delta L(0, T)}{1 + \delta K} \pm \frac{1}{2} \int_0^T \beta^2(u, T) du \right].$$

We could compute $\int_0^T \beta^2(u, T) du$ explicitly, but have not done so here. Finally,

$$V(0) = B(0, T + \delta) \left[(\delta^{-1} + L(0, T))N(\bar{d}_+) - (\delta^{-1} + K)N(\bar{d}_-) \right].$$