

Models for Financial Economics  
Economics 765

Take-home exam due on or before Saturday June 24th 2017.

This exam comprises 6 pages, including the cover page

1. Exercise 1.15 in Shreve. Let  $X$  be a random variable on a probability space  $(\Omega, \mathcal{F}, P)$ , and assume that  $X$  has a density function  $f(x)$  that is positive for all  $x \in \mathbb{R}$ . Let  $g$  be a strictly increasing differentiable function satisfying

$$\lim_{y \rightarrow -\infty} g(y) = -\infty, \quad \lim_{y \rightarrow \infty} g(y) = \infty,$$

and define the random variable  $Y = g(X)$ .

Let  $h(Y)$  be an arbitrary non-negative function satisfying  $\int_{-\infty}^{\infty} h(y) dy = 1$ . We want to change the probability measure so that  $h(y)$  is the density function for the random variable  $Y$ . To do this, we define

$$Z = \frac{h(g(X))g'(X)}{f(X)}.$$

(i) Show that  $Z$  is non-negative and that  $EZ = 1$ .

Now define  $\tilde{P}$  by

$$\tilde{P}(A) = \int_A Z dP \quad \text{for all } A \in \mathcal{F}.$$

(ii) Show that  $Y$  has density  $h$  under  $\tilde{P}$ .

2. In his Theorem 3.7.3, Shreve shows that the joint density of the pair  $(M(t), W(t))$  is

$$f_{M(t), W(t)}(m, w) = \frac{2(2m - w)}{t\sqrt{2\pi t}} \exp - \frac{(2m - w)^2}{2t}, \quad w \leq m, m > 0,$$

where  $W(t)$  is standard Brownian motion, and

$$M(t) = \max_{0 \leq s \leq t} W(s)$$

is the maximum to date of the Brownian motion.

Consider a Brownian motion with drift. Let

$$\hat{W}(t) = \alpha t + W(t),$$

and the corresponding maximum to date

$$\hat{M}(t) = \max_{0 \leq s \leq t} \hat{W}(s).$$

Use Girsanov's theorem to change the measure in such a way that  $\hat{W}(t)$  has no drift under the new measure. Change back to the original measure and show that, under it, the joint density of the pair  $(\hat{M}(t), \hat{W}(t))$  is

$$\hat{f}_{\hat{M}(t), \hat{W}(t)}(m, w) = \frac{2(2m - w)}{t\sqrt{2\pi t}} \exp \left[ \alpha w - \frac{1}{2} \alpha^2 t - \frac{(2m - w)^2}{2t} \right], \quad w \leq m, m > 0.$$

**3.** Let  $(\Omega, \mathcal{F}, P)$  be a probability space with  $\Omega$  the unit interval  $[0, 1]$ ,  $\mathcal{F}$  the Borel  $\sigma$ -algebra  $\mathcal{B}$ , and  $P$  Lebesgue measure. Let  $\mathcal{F}_n$  be the  $\sigma$ -algebra generated by the intervals of the form  $[(j-1)2^{-n}, j2^{-n}]$ ,  $j = 1, 2, \dots, 2^n$ , open to the left, closed to the right. Let  $X$  be a bounded continuous function on  $[0, 1]$ .

- (i) Give the explicit form of the conditional expectation  $E(X | \mathcal{F}_n)$ .
- (ii) Show that  $\mathcal{F}_n \subset \mathcal{F}_{n+1}$  for all  $n$ .
- (iii) Show that the sequence of random variables  $\{E(X | \mathcal{F}_n)\}$ ,  $n = 1, 2, \dots$ , converges for almost all  $\omega \in [0, 1]$ . Characterise the limiting variable. **Hint:** Express  $\omega$  as the infinite sum

$$\omega = \sum_{j=1}^{\infty} \omega_j 2^{-j}$$

where  $\omega_j = 0$  or  $1$  for all  $j$ . Each sequence  $\{\omega_j\}$  defines a unique real number  $\omega$ , and each real  $\omega \in [0, 1]$  defines a unique sequence, provided that one excludes sequences such that there exists  $J > 0$  finite with  $\omega_j = 1$  and  $\omega_j = 0$  for all  $j > J$ .

- (iv) Show directly that  $E(E(X | \mathcal{F}_n)) = E(X)$ .

**4.** Let  $X$  be a real-valued random variable defined on the probability space  $(\Omega, \mathcal{F}, P)$ .

- (i) Suppose that  $E|X|^p$  exists, where  $p > 0$ . Prove the *Markov inequality*, which states that, for all  $\varepsilon > 0$ ,

$$P(|X| > \varepsilon) \leq \frac{E|X|^p}{\varepsilon^p}. \quad (1)$$

- (ii) Let  $g : \mathbb{R} \rightarrow \mathbb{R}_+$  be an increasing function. Show that, for all real  $a$ ,

$$P(X > a) \leq \frac{E(g(X))}{g(a)}. \quad (2)$$

- (iii) Suppose that a discrete-time filtration is defined on the measure space  $(\Omega, \mathcal{F})$  as a nested set of  $\sigma$ -algebras  $\mathcal{F}_t$ ,  $t = 0, 1, 2, \dots$ , with  $\mathcal{F}_s \subset \mathcal{F}_t$  for  $s < t$ . A martingale in this discrete-time context is a stochastic process  $X_t$ , defined for  $t = 0, 1, \dots$ , and such that  $X_t$  is  $\mathcal{F}_{t-}$ -measurable for all  $t$ , and

$$E(X_{t+1} | \mathcal{F}_t) = X_t. \quad (3)$$

Show that (3) implies that, for all  $t = 0, 1, \dots$ ,

$$E(X_{t+s} | \mathcal{F}_t) = X_t, \quad s = 1, 2, \dots \quad (4)$$

- (iv) A stopping time  $\tau$  relative to the filtration  $\mathcal{F}_t$  is a random variable that takes on the possible values  $0, 1, 2, \dots, \infty$ . It is such that the event  $\{\omega : \tau(\omega) = k\}$

belongs to  $\mathcal{F}_k$  for all  $k = 0, 1, \dots$ . The stopped process  $X_{t \wedge \tau}$  is defined by the equation

$$X_{t \wedge \tau}(\omega) = X_{\min(t, \tau(\omega))}(\omega).$$

Show that

$$X_{t \wedge \tau} = \sum_{s=1}^{t-1} \mathbf{I}(\tau = s) X_s + \mathbf{I}(\tau \geq t) X_t.$$

If  $X_t$  is a martingale, show that the process  $X_{t \wedge \tau}$  is also a martingale.

- (v) The  $\sigma$ -algebra  $\mathcal{F}_\tau$  defined by a stopping time  $\tau$  is such that  $A \in \mathcal{F}_\tau$  iff the event  $A \cap \{\tau = t\} \in \mathcal{F}_t$  for all  $t = 0, 1, 2, \dots$ . If  $\tau < n$  almost surely for some finite positive integer  $n$ , show that, for all  $t = 0, 1, \dots, n$ ,

$$\mathbb{E}(X_t | \mathcal{F}_\tau) = \sum_{s=0}^n \mathbb{E}(X_t \mathbf{I}(\tau = s) | \mathcal{F}_s).$$

**5.** Shreve's Exercise 6.7 on the Heston stochastic volatility model is very long. You should refer to it for details that may be helpful but are omitted here.

A stock price has the usual differential based on geometric Brownian motion:

$$dS(t) = rS(t) dt + \sqrt{V(t)}S(t) d\widetilde{W}_1(t),$$

where the interest rate  $r$  is constant, and the volatility  $\sqrt{V(t)}$  is itself a stochastic process governed by the equation

$$dV(t) = (a - bV(t)) dt + \sigma\sqrt{V(t)} d\widetilde{W}_2(t).$$

The parameters  $a$ ,  $b$ , and  $\sigma$  are positive constants, and  $\widetilde{W}_1(t)$  and  $\widetilde{W}_2(t)$  are correlated Brownian motions under a risk-neutral measure  $\widetilde{P}$ , with

$$d\widetilde{W}_1(t) d\widetilde{W}_2(t) = \rho dt.$$

The two-dimensional process  $(S(t), V(t))$  is a Markov process.

At time  $t$ , the price of a European call expiring at  $T > t$  is given by a function  $c(t, s, v)$  such that

$$c(t, S(t), V(t)) = \widetilde{\mathbb{E}}\left[e^{-r(T-t)}(S(T) - K)_+ \mid \mathcal{F}(t)\right], \quad 0 \leq t \leq T.$$

We are to show that  $c$  satisfies the partial differential equation

$$c_t + rsc_s + (a - bv)c_v + \frac{1}{2}s^2vc_{ss} + \rho\sigma sv c_{sv} + \frac{1}{2}\sigma^2vc_{vv} = rc \quad (5)$$

in the region  $0 \leq t < T$ ,  $s \geq 0$ ,  $v \geq 0$ , and that it satisfies the boundary condition

$$c(T, s, v) = (s - K)_+ \quad \text{for all } s \geq 0, v \geq 0. \quad (6)$$

Shreve suggests that we should proceed as follows.

- (i) Show that  $e^{-rt}c(t, S(t), V(t))$  is a martingale under  $\tilde{P}$ , and then derive the PDE (5) from this fact.
- (ii) Suppose that there are functions  $f(t, x, v)$  and  $g(t, x, v)$  satisfying

$$f_t + (r + \frac{1}{2}v)f_x + (a - bv + \rho\sigma v)f_v + \frac{1}{2}vf_{xx} + \rho\sigma vf_{xv} + \frac{1}{2}\sigma^2vf_{vv} = 0, \quad (7)$$

$$g_t + (r - \frac{1}{2}v)g_x + (a - bv)g_v + \frac{1}{2}vg_{xx} + \rho\sigma vg_{xv} + \frac{1}{2}\sigma^2vg_{vv} = 0, \quad (8)$$

in the region  $0 \leq t < T$ ,  $-\infty < x < \infty$ , and  $v \geq 0$ . Show that, if we define

$$c(t, s, v) = sf(t, \log s, v) - e^{-r(T-t)}Kg(t, \log s, v), \quad (9)$$

then  $c(t, s, v)$  satisfies (5).

- (iii) Suppose a pair of processes  $(X(t), V(t))$  is governed by the stochastic differential equations

$$\begin{aligned} dX(t) &= (r + \frac{1}{2}V(t)) dt + \sqrt{V(t)} dW_1(t), \\ dV(t) &= (a - bV(t) + \rho\sigma V(t)) dt + \sigma\sqrt{V(t)} dW_2(t), \end{aligned}$$

where  $W_1(t)$  and  $W_2(t)$  are Brownian motions under a measure  $P$  with  $dW_1(t) dW_2(t) = \rho dt$ . Define

$$f(t, x, v) = E^{t,x,v} \mathbf{I}(X(T) \geq \log K).$$

Show that  $f(t, x, v)$  satisfies (7) and the boundary condition

$$f(T, x, v) = \mathbf{I}(x \geq \log K) \quad \text{for all } x \in \mathbb{R}, v \geq 0.$$

- (iv) Suppose a pair of processes  $(X(t), V(t))$  is governed by the stochastic differential equations

$$\begin{aligned} dX(t) &= (r - \frac{1}{2}V(t)) dt + \sqrt{V(t)} dW_1(t), \\ dV(t) &= (a - bV(t)) dt + \sigma\sqrt{V(t)} dW_2(t), \end{aligned}$$

where  $W_1(t)$  and  $W_2(t)$  are Brownian motions under a measure  $P$  with  $dW_1(t) dW_2(t) = \rho dt$ . Define

$$g(t, x, v) = E^{t,x,v} \mathbf{I}(X(T) \geq \log K).$$

Show that  $g(t, x, v)$  satisfies (8) and the boundary condition

$$g(T, x, v) = \mathbf{I}(x \geq \log K) \quad \text{for all } x \in \mathbb{R}, v \geq 0.$$

- (v) Show that with  $f(t, x, v)$  and  $g(t, x, v)$  as in (iii) and (iv), the function  $c(t, x, v)$  of (9) satisfies the boundary condition (6).

6. This is Shreve's Exercise 8.5, about a perpetual American put that pays dividends. Consider a perpetual American put written on a geometric Brownian motion asset price paying dividends at a constant rate  $a > 0$ . The differential of this asset price is

$$dS(t) = (r - a)S(t) dt + \sigma S(t) d\widetilde{W}(t),$$

where  $\widetilde{W}(t)$  is a Brownian motion under the risk-neutral measure  $\widetilde{P}$ .

- (i) Suppose we adopt the strategy of exercising the put the first time the asset price is at or below  $L$ . What is the risk-neutral expected discounted payoff of this strategy? Write this as a function  $v_L(x)$  of the initial asset price  $x$ . (**Hint:** Define the positive constant

$$\gamma = \frac{1}{\sigma^2} \left( r - a - \frac{1}{2}\sigma^2 \right) + \frac{1}{\sigma} \sqrt{\frac{1}{\sigma^2} \left( r - a - \frac{1}{2}\sigma^2 \right)^2 + 2r}$$

and write  $v_L(x)$  using  $\gamma$ .)

- (ii) Determine  $L_*$ , the value of  $L$  that maximises the risk-neutral expected discounted payoff computed in (i).
- (iii) Show that, for any initial asset price  $S(0) = x$ , the process  $e^{-rt}v_{L_*}(S(t))$  is a supermartingale under  $\widetilde{P}$ . Show that if  $S(0) = x > L_*$  and  $e^{-rt}v_{L_*}(S(t))$  is stopped the first time the asset price reaches  $L_*$ , then the stopped supermartingale is a martingale. (**Hint:** Show that

$$r + (r - a)\gamma - \frac{1}{2}\sigma^2\gamma(\gamma + 1) = 0.)$$

- (iv) Show that, for any initial asset price  $S(0) = x$ ,

$$v_{L_*}(x) = \max_{\tau \in \mathcal{T}} \widetilde{\mathbb{E}}[e^{-r\tau}(K - S(\tau))],$$

where  $\tau$  is a stopping time realised in the non-negative real line.

7. This is taken from Shreve's Exercises 11.3 and 11.7. Let  $N(t)$  be a (simple) Poisson process with intensity  $\lambda$ , and let  $S(0) > 0$  and  $\sigma > -1$  be given. Then using just the independence of non-overlapping increments of the process  $N$  and the distribution of these increments, show that the Geometric Poisson process

$$S(t) = \exp\{N(t) \log(\sigma + 1) - \lambda\sigma t\} = (\sigma + 1)^{N(t)} e^{-\lambda\sigma t}$$

is a martingale.

Again using only the properties of Poisson increments, prove that a compound Poisson process is a Markov process. That is, show that, whenever we are given two times  $0 \leq t \leq T$  and a function  $h(x)$ , there is another function  $g(t, x)$  such that

$$\mathbb{E}[h(Q(T)) | \mathcal{F}(t)] = g(t, Q(t)).$$