

## IEOR 6711, HMWK 8, Professor Sigman

Throughout, we consider the binomial lattice model,  $S_n = S_0 \times Y_1 \cdots \times Y_n$ ,  $n \geq 0$ , with given parameters  $0 < d < 1 + r < u$ , and  $0 < p < 1$ .

1. Consider the case  $u = 1.1$ ,  $d = 0.9$ ,  $p = 0.5$ ;  $P(Y = 1.1) = 0.5$  and  $P(Y = 0.9) = 0.5$ ; thus  $E(Y) = 1$ . (Assume that  $S_0 > 0$ .) Prove (using straightforward basic probability theory) that  $S_n \rightarrow 0$ , wp1, even though  $E(S_n) = S_0$ ,  $n \geq 1$ . (Hint: Take logarithms.)
2. *Continuation:* Change  $u$  to 1.11 leaving all else the same. Prove that now,  $S_n \rightarrow 0$ , wp1, while  $E(S_n) \rightarrow \infty$ : On average you will become infinitely rich even though with certainty you will go broke!!
3. Let  $p = p^*$ , the risk-neutral probability;

$$p^* = \frac{1 + r - d}{u - d}.$$

Define  $M_n = S_n / (1 + r)^n$ ,  $n \geq 0$ .

- (a) Argue that  $\{M_n\}$  is a Markov chain.
  - (b) Prove that  $\{M_n\}$  is a martingale;  $E(M_{n+1} | M_n, \dots, M_0) = M_n$ ,  $n \geq 0$ .
  - (c) Prove that  $p^*$  is the unique probability making  $\{M_n\}$  into a martingale. (e.g., no other  $0 < p < 1$  will work.)
4. Consider a process with state space  $(0, 1)$  defined via

$$X_{n+1} = \begin{cases} a + bX_n, & \text{with prob. } X_n, \\ bX_n & \text{with prob. } 1 - X_n. \end{cases}$$

where  $0 < a < 1$ ,  $0 < b < 1$  and  $a + b = 1$ . (And assume that  $0 < X_0 < 1$ .)

- (a) Show that  $\{X_n\}$  forms a MG.
  - (b) Explain why  $E(X_\tau) = E(X_0)$  holds here for ANY stopping time  $\tau$ .
5. Consider a process  $\{X_n : n \geq 0\}$  such that  $E|X_n| < \infty$ ,  $n \geq 0$ , and

$$E(X_{n+1} | \mathcal{F}_n) = aX_n + bX_{n-1}, \quad n \geq 1$$

where  $0 < a < 1$ ,  $0 < b < 1$  and  $a + b = 1$ . (Here, as usual,  $\mathcal{F}_n = \sigma\{X_0, \dots, X_n\}$ .)

Find a value of  $c$  such that the process  $M_n = cX_n + X_{n-1}$  forms a MG with respect to  $\mathcal{F}_n$ :  $E(M_{n+1} | \mathcal{F}_n) = M_n$ ,  $n \geq 0$ .

6. Let  $\{X_n\}$  be a MG, and define the increments  $\Delta_n = X_n - X_{n-1}$ ,  $n \geq 1$ . We thus can re-write  $X_n = \Delta_1 + \cdots + \Delta_n$ ,  $n \geq 1$ . Prove that these increments are uncorrelated:  $E(\Delta_n \Delta_m) = 0$  for  $m < n$ . (Hint: Condition on  $\mathcal{F}_n$ .)
7. Consider the Markov chain  $\{X_n\}$  on  $\{0, 1, 2, 3, 4\}$  with transition matrix

$$\mathbf{P} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 2/3 & 0 & 0 & 1/3 & 0 \\ 1/5 & 1/5 & 1/5 & 1/5 & 1/5 \\ 0 & & 1/2 & 0 & 1/2 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

Suppose that  $X_0 = i$  with  $0 < i < 4$  and let  $\tau = \min\{n \geq 0 : X_n \in \{0, 4\}\}$ . By using Martingale theory, find  $p(i) =$  the probability that the chain hits 4 before hitting 0.

8. Suppose that  $\{X_n : n \geq 0\}$  is a SUBMG such that  $E(X_n) = c$ ,  $n \geq 0$  (e.g., the means are all identical). Prove that in fact,  $\{X_n : n \geq 0\}$  must be a martingale.
9. Consider a random walk  $R_n = \sum_{j=1}^n \Delta_j$ ,  $R_0 = 0$ , with iid  $\Delta_j$  such that  $P(\Delta \neq 0) > 0$  and such that

$$m(\theta) = E(e^{\theta\Delta}) < \infty,$$

for all  $\theta \in (-\delta, \delta)$  some  $\delta > 0$ . That is,  $\Delta$  has a finite moment generating function in a neighborhood of the origin.

- (a) For  $m(\theta) < \infty$ , Prove that  $X_n = m(\theta)^{-n} e^{\theta R_n}$ ,  $n \geq 0$  forms a martingale.
- (b) Assume that there exists a  $\theta^* \in (-\delta, \delta)$  such that  $m(\theta^*) = 1$ , so that then  $X_n = e^{\theta^* R_n}$  forms a MG. Let  $a, b > 0$  and define a stopping time (easily shown to be proper)

$$\tau = \min\{n \geq 1 : R_n \notin (-b, a)\},$$

the first time that the random walk either goes  $\geq a$  or  $\leq -b$ . Since we are not assuming that this random walk is simple, there typically will be an *overshoot* when the random walk crosses  $a$  or  $-b$ . If  $a$  is crossed first, then  $X_\tau = e^{a\theta^*} e^{H_a\theta^*}$ , where  $H_a = R_\tau - a$ , the overshoot above  $a$ . Similarly, if  $-b$  is crossed first, then  $X_\tau = e^{-b\theta^*} e^{-H_b\theta^*}$ , where  $H_b = |R_\tau - (-b)|$ , the overshoot below  $-b$ . Letting  $p(a) = P(R_\tau \geq a)$  and  $p(b) = 1 - p(a) = P(R_\tau \leq -b)$ . Assuming the appropriate UI condition holds, use the MG optional stopping theorem to obtain an expression for  $p(a)$ .

- (c) Assuming that both  $H_a$  and  $H_b$  are very very small (in comparison to  $a, b$ ), use your expression in (b) to obtain the following explicit approximation to  $p(a)$ :

$$p(a) \approx \frac{1 - e^{-b\theta^*}}{e^{a\theta^*} - e^{-b\theta^*}}.$$

- (d) (Assume here that  $E(\Delta) \neq 0$ .) Using a similar method as in (b) on  $R_n$  itself together with the above approximation for  $p(a)$ , derive the following approximation for  $E(\tau)$ :

$$E(\tau) \approx \frac{a(1 - e^{-b\theta^*}) - b(e^{a\theta^*} - 1)}{(e^{a\theta^*} - e^{-b\theta^*})E(\Delta)}.$$

- (e) Suppose specifically that  $\Delta = S - T$ , where  $S$  and  $T$  are independent and  $S \sim \exp(\mu)$ ,  $T \sim \exp(\lambda)$  and  $\lambda < \mu$ .

In this case, find  $\theta^*$  and then use it (via the expression in (b)) to compute  $p(a)$  exactly.

Also, compute  $E(\tau)$  exactly.