Stochastic Calculus, Fall 2002 (http://www.math.nyu.edu/faculty/goodman/teaching/StochCalc/)

## Assignment 3.

Given September 20, due September 26. Last revised, September 23.

**Objective:** Conditioning and Markov chains, II.

- 1. Suppose an  $n \times n$  matrix, A, has n linearly independent right eigenvectors (column vectors) with corresponding eigenvalues  $\lambda_j$ :  $Ar_j = \lambda_j r_j$ . Suppose the the corresponding left (row vector) eigenvectors are  $L_j$ :  $l_j A = \lambda_j l_j$ . Suppose that the  $r_j$  and  $l_k$  have been normalized to be biorthogonal:  $l_j r_k = \delta_{jk}$ , where  $\delta_{jk} = 1$  if j = k and  $\delta_{jk} = 0$  if  $j \neq k$  (this is the "Kronecker delta"; the  $\delta_{jk}$  are the entries in the identity matrix.).
  - **a.** Show that  $A = \sum_{j=1}^{n} \lambda_j r_j l_j$ . Note that each term in the sum is an  $n \times n$  matrix. In the other order,  $l_j r_j$  is a  $1 \times 1$  matrix.
  - **b.** Find a similar expression for  $A^t$ , the product of A with itself t times, not the transpose of A.
- 2. We have a three state Markov chain with transition matrix

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix}$$

For example, some of the transition probabilities are  $P(1 \to 1) = \frac{1}{2}$ ,  $P(3 \to 1) = \frac{1}{3}$ , and  $P(1 \to 2) = \frac{1}{4}$ . Let  $\tau = \min(t \mid X_t = 3)$ . Even though this  $\tau$  is not bounded (it could be arbitrarily large), we will see that  $P(\tau > t) \leq Ca^t$  for some a < 1 so that the probability of large  $\tau$  is very small. This is enough to prevent the stopping time paradox (take my word for it). Suppose that at time t = 1, all states are equally likely.

- **a.** Consider the quantities  $u_t(j) = P(X_t = j \text{ and } \tau > t)$ . Find a matrix evolution equation for a two component vector made from the  $u_t(j)$  and a submatrix of P.
- **b.** Solve this equation using the method of problem 1 to find a formula for  $m_t = P(\tau = t)$ . For this you will have to find the eigenvalues and the left and right eigenvectors of the 2 × 2 matrix A.
- c. Use the answer of part b to find  $E[\tau]$ . It might be helpful to use the formula

$$\sum_{t=1}^{\infty} t P(\tau = t) = \sum_{t=1}^{\infty} P(\tau \ge t) .$$

Verify the formula if you want to use it.

**d.** Consider the quantities  $f_t(j) = P(\tau \ge t \mid X_1 = j)$ . Find a matrix recurrence for them.

- e. Use the method of question 1 to solve this and find the  $f_t$ .
- **3.** Let P be the transition matrix for an n state Markov chain. Let v(k) be a function of the state  $k \in S$ . For this problem, suppose that paths in the Markov chain start at time t = 0 rather than t = 1, so that  $X = (X_0, X_1, \ldots)$ . For any complex number, z, with |z| < 1, consider the sum

$$E\left[\sum_{t=0}^{\infty} z^t v(X_t) \mid \mathcal{F}_0\right] . \tag{1}$$

Of course, this is a function of  $X_0 = k$ , which we call f(k). Find a linear matrix equation for the quantities f that involves P, z, and v. Hint: the sum

$$E\left[\sum_{t=1}^{\infty} z^t v(X_t) \mid \mathcal{F}_1\right]$$
.

may be related to (1) if we take out the common factor, z.

- 4. A particular random walk has states  $X_t$  that are integers, possibly negative. In general the transition probabilities are  $P(k \to k 1) = P(k \to k) = P(k \to k + 1) = \frac{1}{3}$  with all other transion probabilities being zero. Don't worry that the state space might be infinite or that stopping times might be infinite.
  - **a.** Show that  $F_t = X_t$  and  $G_t = X_t^2 \frac{2}{3}t$  are martingales.
  - **b.** Define the stopping time  $\tau_a = \min(t \mid |X_t| = a)$ . Use the result of part *a* and the Doob stopping time theorem to show that  $E(\tau_a \mid X_0 = 0) = \frac{3}{2}a^2$ .
  - c. Let f(k) be defined by  $f(X_0) = E[\tau_a | \mathcal{F}_0]$ . Find a system of linear equations satisfied by the numbers f(k) for -a < k < a. Express these equations in the form  $Af = \mathbf{1}$ , where A is a  $(2a-1) \times (2a-1)$  matrix and  $\mathbf{1}$  is a vector of all ones. To find the solution, calculate Aw where w(k) = 1, w(k) = k, and  $w(k) = k^2$ , then express the answer as a linear combination of these vectors.
  - **d.** Show that the martingale method of part b also applies to  $E[\tau_a \mid X_0 = k]$  for any |k| < a. Check that this is consistent with the answer to part c.