## Math 565a - Homework 5

1. In some problems the process itself is not a Markov process, but one can construct a closely related Markov process by "enlarging the state space." Then you can apply the theory of Markov processes to your problem. Here is a simple example.

We work on the lattice  $\mathbb{Z}^2$ . We generate a sample of the process as follows. Take  $X_0 = 0$ . Having generated  $X_1, X_2, \dots, X_n$ , consider the four lattice sites that are a distance 1 from  $X_n$ . One of them is  $X_{n-1}$ . Take  $X_{n+1}$  to be one of the other three with equal probability 1/3. The filtration is the natural one:  $\mathcal{F}_n = \sigma(X_1, X_2, \dots, X_n)$ . One can describe this process as a random walk that is not allowed to "backtrack."

(a) Show that  $X_n$  is not a Markov process.

(b) Let  $Y_n$  be the random vector  $(X_{n-1}, X_n)$  for  $n = 1, 2, \cdot$ . The filtration is the same as before, or equivalently  $\mathcal{F}_n = \sigma(Y_1, Y_2, \cdots, Y_n)$ . For the original process the state space is  $\mathbb{Z}^2$ . for the  $Y_n$  process it is  $\mathbb{Z}^2 \times \mathbb{Z}^2$ . Show that  $Y_n$ is a Markov process.

2. (Durrett 1.6, p. 284) Fix a positive integer N. Let  $\xi_1, \xi_2, \cdots$  be i.i.d. where each  $\xi_n$  is uniformly distributed on  $\{1, 2, \cdots, N\}$ . Consider the set  $\{\xi_1, \xi_2, \cdots, \xi_n\}$  of values that have occurred by time n. It is a subset of  $\{1, 2, \cdots, N\}$  and will eventually have repetitions. Let  $X_n$  be the number of distinct values that appear in this set. Show that this is a Markov process and find the transition probabilities.

3. (Durrett 1.7, p. 285) Let  $S_n$  be a simple random walk on  $\mathbb{Z}$  with  $S_0 = 0$ . Let  $M_n = \max\{S_k : 0 \le k \le n\}$ . Show that  $M_n$  is not a Markov process.

4. (Durrett 1.8, p. 285) Let  $\theta, U_1, U_2, \cdots$  be independent with each random variable uniformly distribution on [0, 1]. Define  $X_n$  by

$$X_n = \begin{cases} +1 & \text{if } U_n < \theta \\ -1 & \text{if } U_n \ge \theta \end{cases}$$

Then let  $S_n = \sum_{k=1}^n X_k$ . We can think of this as a sort of random walk. We first choose a "bias"  $\theta$ . We then flip a coin with probability  $\theta$  of heads and probability  $1 - \theta$  of tails to generate the random walk. (The bias does not change as we flip the coin.) Show that  $S_n$  is a Markov process which is not time homogeneous.

5. (Durrett 2.10, p. 290) Consider a Markov process with a countable state space S. Let  $A \subset S$  such that  $S \setminus A$  is finite. Assume that for each  $x \notin A$ ,

 $P_x(\tau_A < \infty) > 0$ . Let

$$\tau_A = \inf\{n \ge 0 : X_n \in A\}$$

and let  $g(x) = E_x \tau_A$ . Let p(x, y) be the transition probabilities. Show that

$$g(x) = 1 + \sum_{y \in S} p(x, y)g(y)$$

for  $x \notin A$ .

6. (the ballot problem) Let  $Y_j$  be i.i.d. with values in  $\{0, 1, 2, \dots\}$ . Let  $S_n = Y_1 + \dots + Y_n$ . Fix N. The goal of this problem is to show

$$P(S_j < j, 1 \le j \le N | S_N) = (1 - \frac{S_N}{N})^+$$

Note that if  $S_N \ge N$ , the equation is trivially true. So you can assume  $S_N < N$ . I showed in class that  $X_n = S_n/n$  is a backwards martingale with respect to the filtration  $\mathcal{F}_n = \sigma(X_k, k \ge n)$ . Define  $\tau = \max\{k : S_k \ge k, 1 \le k \le N\}$  if this set is non-empty. If the set is empty we define  $\tau = 1$ . We want to apply the optional sampling theorem to the backwards martingale and  $\tau$ . If we go backward in time  $X_n$  is a martingale.

(a) Show that  $\tau$  is a stopping time if we go backwards in time. (Part of the problem is to figure out what this means.)

(b) Show that  $X_{\tau} = 1_{G^c}$  where G is the event  $\{S_j < j, 1 \le j \le N\}$ .

(c) Apply the optional sampling theorem to get the result stated above.