

## Assignment 2.

Given September 12, due September 19.

Last revised, September 15.

**Objective:** Conditioning and Markov chains.

1. Suppose that  $\mathcal{F}$  and  $\mathcal{G}$  are two algebras of sets and that  $\mathcal{G}$  adds information to  $\mathcal{F}$  in the sense that any  $\mathcal{F}$  measurable event is also  $\mathcal{G}$  measurable. Since  $\mathcal{F}$  and  $\mathcal{G}$  are collections of events, this may be written  $\mathcal{F} \subset \mathcal{G}$ . Suppose that the probability space  $\Omega$  is finite and that  $X(\omega)$  is a variable defined on  $\Omega$  (that is, a function of the random variable  $\omega$ ). The conditional expectations (in the modern sense) of  $X$  with respect to  $\mathcal{F}$  and  $\mathcal{G}$  are  $Y = E[X | \mathcal{F}]$  and  $Z = E[X | \mathcal{G}]$ . In each case below, state whether the statement is true or false and explain your answer with a proof or a counterexample.
  - a.  $Z \in \mathcal{F}$ .
  - b.  $Y \in \mathcal{G}$ .
  - c.  $Z = E[Y | \mathcal{G}]$ .
  - d.  $Y = E[Z | \mathcal{F}]$ .
2. For any event  $A \subseteq \Omega$  we can define the “indicator function” (also called the “characteristic function”, particularly by people who learned probability late in life),  $\mathbf{1}_A(\omega) = 1$  if  $\omega \in A$  and  $\mathbf{1}_A(\omega) = 0$  if  $\omega \notin A$ . People who call this the “characteristic function” (usually people who learned probability late in life) write  $\chi_A(\omega)$  for  $\mathbf{1}_A(\omega)$ .
  - a. Show that  $E[\mathbf{1}_A] = P(A)$ .
  - b. For any event,  $B$ , show that the classical  $E[\mathbf{1}_A | B]$  is the same as the Bayes’ rule conditional probability of  $A$ .

The “modern” definition of conditional probability, conditioning on an algebra of sets rather than a single set, is  $P(A | \mathcal{F}) = E[\mathbf{1}_A | \mathcal{F}]$ . The relationship between the classical and modern definition of conditional probability is more or less the same as the relation between classical and modern expected value.

3. Let  $\mathcal{S}$  be a finite state space and  $\Omega$  be the set of paths of length  $T$  from  $\mathcal{S}$ . Let  $P(X)$  be the probability of a path  $X \in \Omega$ . For any  $t$  in the range  $1 < t < T$ , let  $\mathcal{F}_t$  be the algebra of events in  $\Omega$  generated by the values of  $X_s$  for  $1 \leq s \leq t$ . Let  $\mathcal{G}_t$  be the smaller algebra generated only by  $X_t$ . Finally, let  $\mathcal{H}_t$  be the “complementary” algebra (based on complementary information) generated by the values  $X_t, \dots, X_T$ . An event in  $\mathcal{H}_t$  is a statement about the path from time  $t$  on without saying anything about the beginning values  $X_1, \dots, X_{t-1}$ . Show that the Markov property is equivalent to either of the following,

a.

$$E[\mathbf{1}_A | \mathcal{F}_t] = E[\mathbf{1}_A | \mathcal{G}_t] \quad \text{for any } A \in \mathcal{H}_{t+1}.$$

b.

$$E[F(X) | \mathcal{F}_t] = E[F(X) | \mathcal{G}_t] \quad \text{for any } F \in \mathcal{H}_{t+1}.$$

**Notes:**(i) Part a is a special case of part b (why?). (ii) Part b implies that  $E[F(X) | \mathcal{F}_t]$  is a function of  $X_t$  only. This is supposed to be intuitively clear as a consequence of the Markov property. (iii) Together with question 1, part b is a justification for the backward equation for expected values of final payouts.

4. Suppose we have a 3 state Markov chain with transition matrix

$$P = \begin{pmatrix} .6 & .2 & .2 \\ .3 & .5 & .2 \\ .1 & .2 & .7 \end{pmatrix}$$

and suppose that  $X_1 = 1$ .

- a. Show that the probability distribution of the first  $t$  steps conditioned on  $\mathcal{G}_{t+1}$  is the same as that conditioned on  $\mathcal{H}_{t+1}$ . This is a kind of backwards Markov property: a forward Markov chain is a backward Markov chain also.
- b. Calculate  $P(X_2 = 2 | \mathcal{G}_3)$ . This consists of 3 numbers.