

Ch. 4 (Markov chains, exercises:

4.14, 4.20, 4.23, 4.30

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4.14.) Specify the classes of the following Markov chains & determine whether transient or recurrent:

a) 
$$P_1 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix} \end{matrix}$$
 Note that  $1 \leftrightarrow 2$  and  $2 \leftrightarrow 3$ , so all states communicate.

Hence, there is only one class in the Markov chain, i.e., the Markov chain is irreducible.

Recurrence and transience are class properties (Cor. 4.2). Not all states in a finite Markov chain can be transient (see comment after Prop. 4.1). Hence,

Po. prob. of going from 1 to 2, 1 to 3, 2 to 1, 2 to 3, 3 to 1 and 3 to 2

since our Markov chain is irreducible & finite state, all states must be recurrent (see also Remark (ii) right before Ex. 4.17).

b) 
$$P_2 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$
 
$$\begin{matrix} 1 \rightarrow 4 \\ 2 \rightarrow 4 \\ 3 \rightarrow 1 \text{ or } 3 \rightarrow 2 \\ 4 \rightarrow 3 \end{matrix}$$

⇓

$1 \leftrightarrow 4, 1 \leftrightarrow 3, 2 \leftrightarrow 4, 2 \leftrightarrow 3 \Rightarrow$  all states communicate.

That is, we have a finite state, irreducible Markov chain. Hence, all states are recurrent (see ex. a).

$$c) P_3 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \end{matrix}$$

$$\left. \begin{array}{l} 1 \rightarrow 1, \quad 1 \rightarrow 3 \\ 2 \rightarrow 1, \quad 2 \rightarrow 2, \quad 2 \rightarrow 3 \\ 3 \rightarrow 1, \quad 3 \rightarrow 3 \end{array} \right\} \begin{array}{l} 3 \leftrightarrow 1 \\ 2 \end{array}$$

$$\left. \begin{array}{l} 4 \rightarrow 4, \quad 4 \rightarrow 5 \\ 5 \rightarrow 4, \quad 5 \rightarrow 5 \end{array} \right\} 4 \leftrightarrow 5$$

↓

The classes are:  $\{2\}$ ,  $\{1, 3\}$ ,  $\{4, 5\}$

Note that if starting in 2, we will not return (with probability 1). Hence,  $\{2\}$  is transient.

The classes  $\{1, 3\}$  and  $\{4, 5\}$  are both recurrent, since there is probability 1 of reentering any of these states at some point. F. ex: starting in 1, what is

$$f_1 = P(\text{reenter } 1 \mid \text{start in } 1) = 1 - P(\text{never reenter } 1 \mid \text{start in } 1)$$

$$= 1 - P(\text{go from } 1 \text{ to } 2 \text{ then always } 2 \mid \text{start in } 1)$$

$$= 1 - \left( \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdots \text{infinitely many times} \right)$$

$$\rightarrow 1 - 0 = 1$$



4.20.) A transition prob. matrix is doubly stochastic if the sum of each column is one;

$$\sum_i P_{ij} = 1 \quad \forall j$$

If such a chain is irreducible and consists of  $n+1$  states  $0, 1, \dots, M$ , show that the long run proportions are:

$$\pi_j = \frac{1}{M+1}, \quad j = 0, 1, \dots, M$$

The Markov chain is finite state (since # states =  $M+1$ ) and irreducible. Hence, it is positive recurrent (see Remark (ii) pg. 217).

Hence, from Thm. 4.1, the long run proportions are the unique solution of:

$$(*) \quad \begin{cases} \pi_j = \sum_i \pi_i P_{ij}, \quad j \geq 1 \\ \sum_j \pi_j = 1 \end{cases}$$

Note that for  $\pi_j = \frac{1}{M+1}, j = 0, \dots, M$ , we have

$$\sum_i \pi_i P_{ij} = \sum_i \frac{1}{M+1} P_{ij} = \frac{1}{M+1} \sum_i P_{ij} = \frac{1}{M+1} = \pi_j$$

doubly stochastic

Hence,  $\pi_j = \frac{1}{M+1}, j = 0, \dots, M$

is the unique solution of (\*) for our Markov chain, i.e. these are the long-run proportions. (4)

Otherwise, all long run proportions are 0: Not possible for finite # states

4.23) Good weather year: # storms  $\sim$  Poisson (1)

Bad weather year: # storms  $\sim$  Poisson (3)

Suppose: Yearly weather only depend on last year.  $\therefore$  (1)

$$(2): P(X_n = G | X_{n-1} = G) = P(X_n = B | X_{n-1} = G) = \frac{1}{2}$$

$$(3): P(X_n = G | X_{n-1} = B) = \frac{1}{3}$$

$$(4): P(X_n = B | X_{n-1} = B) = \frac{2}{3} \quad \left( \begin{array}{l} B \text{ twice as likely to} \\ \text{be followed by B as G} \end{array} \right)$$

$$X_0 = G$$

a) Find  $E[\# \text{ storms in } X_1 \text{ \& } X_2 | X_0 = G]$ ?

$\{X_n\}_{n \geq 0}$  is a Markov chain (from (1)) with values

in  $S = \{G, B\}$  and index set  $\{0, 1, 2, \dots\}$ ,

which represents years.

Transition matrix (from (2) - (4)):

$$P = \begin{array}{c} G \\ B \\ G \quad B \end{array} \left[ \begin{array}{cc} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{3} & \frac{2}{3} \end{array} \right]$$

Define:

$N_i := \# \text{ storms in year } i$

We want to find:

$$E[N_1 + N_2 \mid X_0 = G] = E[N_1 \mid X_0 = G] + E[N_2 \mid X_0 = G]$$

expect. linear

M1

$$= E[E[N_1 \mid X_1] \mid X_0 = G]$$

double expectation rule

$$+ E[E[N_2 \mid X_2] \mid X_0 = G] = (\sim)$$

M2

First:

$$E[E[Y \mid X]] = E[Y]$$

so:

$$E[Y \mid Z] = E[E[Y \mid X] \mid Z]$$

M1:

$$E[E[N_1 \mid X_1] \mid X_0 = G]$$

$$= \sum_{i \in S} E[N_1 \mid X_1 = i] P(X_1 = i \mid X_0 = G)$$

def. expectation

def s

$$= E[N_1 \mid X_1 = G] P_{G,G} + E[N_1 \mid X_1 = B] P_{G,B}$$

$$= 1 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2} = \frac{1+3}{2} = \underline{2}$$

assumptions; # storms Poisson with means 1 & 3 resp.

+ transition matrix

M2:

def expectation

$$E[E[N_2 | X_2] | X_0 = G] = \sum_{i \in S} E[N_2 | X_2 = i] P(X_2 = i | X_0 = G)$$

$$= E[N_2 | X_2 = G] P(X_2 = G | X_0 = G)$$

def S

$$+ E[N_2 | X_2 = B] P(X_2 = B | X_0 = G)$$

$$= 1 \cdot P_{G,G}^{(2)} + 3 \cdot P_{G,B}^{(2)}$$

assumptions,

$\sim$  Poisson mean  
1 & 3 resp.

$$= \frac{5}{12} + \frac{3 \cdot 7}{12} = \frac{26}{12} = \frac{13}{6}$$

To find  $P_{G,G}^{(2)}$  &  $P_{G,B}^{(2)}$ :

$$P^2 = \begin{matrix} & \begin{matrix} G & B \end{matrix} \\ \begin{matrix} G \\ B \end{matrix} & \begin{bmatrix} \frac{5}{12} & \frac{7}{12} \\ \frac{7}{18} & \frac{11}{18} \end{bmatrix} \end{matrix}$$

Hence; from (2), M1 & M2:

$$E[N_1 + N_2 | X_0 = G]$$

$$= 2 + \frac{13}{6} = \frac{25}{6}$$

b) Find  $P(\text{no storms year 3} \mid X_0 = G)$ :

$$P(N_3 = 0 \mid X_0 = G) = \sum_{i \in S} P(N_3 = 0 \mid X_3 = i) P(X_3 = i \mid X_0 = G)$$

condition with  $X_3$

$$= P(N_3 = 0 \mid X_3 = G) P_{G,G}^{(3)} + P(N_3 = 0 \mid X_3 = B) P_{G,B}^{(3)}$$

$$P_{G,G}^{(3)} = \begin{bmatrix} 0,40 & 0,60 \\ 0,40 & 0,60 \end{bmatrix}$$

G                  B

Also:

$$P(N_3 = 0 \mid X_3 = G)$$

$\sim$  Poisson (1),

$$P(N_3 = 0 \mid X_3 = B) \sim \text{Poisson (3)}.$$

Hence:

$$P(N_3 = 0 \mid X_3 = G) = \frac{1^0 e^{-1}}{0!} = e^{-1}$$

$$P(N_3 = 0 \mid X_3 = B) = \frac{3^0 e^{-3}}{0!} = e^{-3}$$

$$= e^{-1} \cdot 0,4 + e^{-3} \cdot 0,6$$

$$\approx \underline{0,18}$$

c) Find the long-run average number of storms per year:

$$E[N_\infty \mid X_0 = G] = ?$$

Write  $N_\infty$  although it's a little imprecise.

Actually, were interested in  $\lim_{n \rightarrow \infty} E[N_n \mid X_0 = G]$  but since

it's a pain to write over & over, we simplify by writing  $N_\infty$

$$E[N_\infty | X_0 = G] = E[N_\infty | X_\infty = G] P(X_\infty = G | X_0 = G) \\ + E[N_\infty | X_\infty = B] P(X_\infty = B | X_0 = G)$$

$$= E[N_\infty | X_\infty = G] \pi_G + E[N_\infty | X_\infty = B] \pi_B = (\dots)$$

def. of long run proportions

M: Need to find  $\vec{\pi} = (\pi_G, \pi_B)$ !

Solve  $\pi = \pi P = \pi \begin{bmatrix} \frac{1}{2} & \frac{1}{3} \\ \frac{1}{3} & \frac{2}{3} \end{bmatrix}$

$$\pi_G + \pi_B = 1$$

That is:

$$\pi_G = \frac{1}{2} \pi_G + \frac{1}{3} \pi_B$$

$\Downarrow$

$$\pi_B = \frac{3}{2} \pi_G = \frac{3}{2} (1 - \pi_B)$$

$$\pi_B + \frac{3}{2} \pi_B = \frac{3}{2}$$

$$\pi_B \left(1 + \frac{3}{2}\right) = \frac{3}{2}$$

$$\frac{5}{2} \pi_B = \frac{3}{2}$$

$$\pi_B = \frac{3}{5}, \text{ so } \pi_G = 1 - \frac{3}{5} = \frac{2}{5}$$

Our Markov chain is a finite state, and irreducible, hence positive recurrent.

From Thm. 4.1, the long-run proportions are the unique solution of  $\pi = \pi P$  &  $\sum_i \pi_i = 1$

Hence, from (v) and (M):

$$E[N_\infty | X_0 = G] = 1 \cdot \frac{2}{5} + 3 \cdot \frac{3}{5} = \frac{2+9}{5} = \frac{11}{5}$$

$X_\infty = G: N_\infty \sim \text{Poisson}(1)$

$X_\infty = B: N_\infty \sim \text{Poisson}(3)$

4.30) 3 out of 4 trucks are followed by a car.  
1 out of 5 cars is followed by a truck.  
What fraction of vehicles are trucks?

Make Markov chain of observed vehicles with transition matrix

$$P = \begin{matrix} & \begin{matrix} C & T \end{matrix} \\ \begin{matrix} C \\ T \end{matrix} & \begin{bmatrix} \frac{4}{5} & \frac{1}{5} \\ \frac{3}{4} & \frac{1}{4} \end{bmatrix} \end{matrix}, S = \{C, T\}$$

Want to find long-run proportion of trucks.

Our Markov chain is finite state and irreducible, hence positive recurrent. Thm. 4.1 tells us that the long-run proportions of cars and trucks,  $\pi := (\pi_C, \pi_T)$  is the unique

solution of

$$\begin{cases} \pi = \pi P \\ \pi_C + \pi_T = 1 \end{cases}$$

$$\text{So: } \pi_C = \frac{4}{5} \pi_C + \frac{3}{4} \pi_T$$

$$\frac{1}{5} \pi_C = \frac{3}{4} \pi_T$$

$$\pi_C = \frac{15}{4} \pi_T = \frac{15}{4} (1 - \pi_C)$$

$$\pi_C = \frac{15}{4} - \frac{15}{4} \pi_C$$

$$\pi_C \left( \frac{4+15}{4} \right) = \frac{15}{4}$$

$$\pi_C = \frac{15}{19}, \text{ so } \pi_T = 1 - \frac{15}{19} = \frac{4}{19}$$

The percentage of trucks on the road is  $\pi_T = \frac{4}{19}$ .

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