# Part IA — Markov Chains Example Sheet 2

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Suppose there exists some state  $i \in S, i \neq s$ , st. i is recurrent. Then

$$\mathbb{P}_i(T_i < \infty) = 1$$
 where  $T_i := \min\{n \ge 1 : X_n = i\}$ 

But  $i \to s$ , so by definition there exists some r > 0 st.  $p_{i,s}(r) > 0$ , and as s is absorbing,  $p_{s,i}(n) = 0 \,\,\forall\,\, n \geq 0$ . Hence there is a non-zero probability of being 'trapped' in s, ie. a non-zero probability that  $T_i = \infty$ . Hence, for all  $i \in S$ ,  $\mathbb{P}_i(T_i < \infty) < 1$ , ie. i is transient.

Let P denote the transition matrix in question. The characteristic equation  $\det(P - \kappa \iota) = 0$  gives roots  $\kappa_1 = 1$ ,  $\kappa_2 = 1 - 2p$ ,  $\kappa_3 = 1 - 4p$ . Thus

$$P^{n} = U^{-1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & (1 - 2p)^{n} & 0 \\ 0 & 0 & (1 - 4p)^{n} \end{pmatrix} U$$

for some invertible matrix U, and so

$$p_{1,1}(n) = A + B(1-2p)^n + C(1-4p)^n$$
 Using  $p_{1,1}(0) = 1, p_{1,1}(1) = 1 - 2p, p_{1,1}(2) = (1-2p)^2 + 2p^2$  gives

$$A + B + C = 1$$

$$A + B(1 - 2p) + C(1 - 4p) = 1 - 2p$$

$$A + B(1 - 2p)^{2} + C(1 - 4p)^{2} = 1 - 4p + 6p^{2}$$

Setting  $p = \frac{1}{2}$  in the second equation,  $p = \frac{1}{4}$  in the third yields

$$A + B + C = 1$$
$$A - C = 0$$
$$A + B/4 = 3/8$$

which gives A = C = 1/4, B = 1/2, thus

$$p_{1,1}(n) = \frac{1}{4} + \frac{1}{2}(1 - 2p)^n + \frac{1}{4}(1 - 4p)^n$$

We have  $1\leftrightarrow 2\leftrightarrow 3$  so the chain is irreducible (all states recurrent of all states transient). Thus as  $n\to\infty,\ p_{1,1}\to 1/4$ , so state 1 is recurrent. hence all states are recurrent.

Using the symmetry of the problem, the invariant probabilities are  $\pi_i = 1/2^3$  for all vertices x. For a finite irreducible MC, the mean recurrent time  $\mu_i$  to state i is

$$\mu_i = \frac{1}{\pi_i}$$

Thus the expected number of steps until the particle first returns to v is  $2^3 = 8$ .

Should be a way to do the next two using invariant distributions but I can't see it. Let  $e_i$  be the expected number of steps to reach w given we are i steps away from w. Hence

$$e_0 = 0$$

$$e_1 = 1 + \frac{1}{4}e_1 + \frac{1}{2}e_2$$

$$e_2 = 1 + \frac{1}{2}e_1 + \frac{1}{4}e_2 + \frac{1}{4}e_3$$

$$e_3 = 1 + \frac{3}{4}e_2 + \frac{1}{4}e_3$$

Solving gives  $e_3 = 40/3$ , the expected number of steps to reach v starting in v.

For the last part: as the chain is irreducible and positive recurrent, the mean number of visits to state w between two consecutive visits to state v equals  $\pi_w/\pi_v = 8/8 = 1$ .

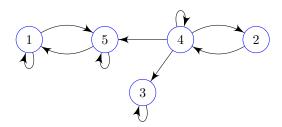
Suppose after one step the particle is in state j,  $a_j > 0$ . Now the particle can only travel to state j or j-1; once it is in state j-1, it can only travel to state j-1 or j-2, and so on... eventually it returns to state 0, where it is then sent to some other state k st.  $a_k > 0$ . In other words, we can only travel to a higher state by doing so from the origin. Motivated by this, we define

$$J = \sup\{j \mid a_j > 0\}$$

It is now easy to see that all states  $\geq J$  are transient; after we leave them (which we will do eventually), we can never return there.

Not sure how to find the mean recurrent times.

Possible transitions of the chain are illustrated below:



The communicating classes are  $C_1 = \{1,5\}, C_2 = \{3\}$  and  $C_3 = \{2,4\}$ . The classes  $C_1$  and  $C_3$  are not closed, but  $C_2$  is closed. We know that inside communicating classes, every state is recurrent or every state is transient. It is easy to see from the diagram that

 $C_1$  recurrent ,  $C_2$  recurrent ,  $C_3$  transient

- There is a non zero probability that a MC beginning in a transient state will never return to that state
- There is a guarantee that a process beginning in a recurrent state will return to that state.

Collapsing the tree into a random walk on  $\mathbb{Z}^+$ , where the root of the tree R is represented by 0, with two branches extending to 1, four extending to 2 etc.

The walker moves rightwards with probability 2/3 and leftwards with probability 1/3, at at R moves rightwards with probability 1.

It is now intuitively obvious that this random walk is transient. For a concrete proof, we will argue that the state 0 is transient, and as the MC is irreducible, all states are transient. By the usual arguments, the probability of return in 2n steps is given by

$$p_0(2n) = \binom{2n}{n} \left(\frac{2}{3}\right)^n \left(\frac{1}{3}\right)^n$$

Using Stirling's formula:  $n! \sim (n/e)^n \sqrt{2\pi n}$  as  $n \to \infty$ , we have

$$p_0(2n) = \frac{(2n)!}{(n!)^2} \left(\frac{2}{3}\right)^n \left(\frac{1}{3}\right)^n$$
$$\sim \frac{2^{3n}}{3^{2n}\sqrt{n\pi}}$$
$$= \left(\frac{8}{9}\right)^n \frac{1}{\sqrt{n\pi}}$$

We can conclude  $\sum_{n=0}^{\infty} p_0(2n) < \infty$  as this series can be shown to converge using the ratio test.

The walk is at the origin  $\mathbf{0} = (0, 0, 0, 0)$  at time 2n if and only if it has taken equal number of steps negative and positive in each dimension (ie. in 1D, equal number right and left). Therefore,

$$p_{\mathbf{0},\mathbf{0}}(2n) = \left(\frac{1}{8}\right)^{2n} \sum_{i_1,\dots,i_4} \frac{(2n)!}{(i_1!i_2!i_3!i_4!)^2}$$

where  $i_1 + i_2 + i_3 + i_4 = n$ Thus

$$p_{\mathbf{0},\mathbf{0}}(2n) = \left(\frac{1}{2}\right)^{2n} \binom{2n}{n} \sum_{i_1,\dots,i_4} \left(\frac{n!}{4^n i_1! i_2! i_3! i_4!}\right)^2$$

$$\leq \left(\frac{1}{2}\right)^{2n} \binom{2n}{n} M \sum_{i_1,\dots,i_4} \frac{n!}{4^n i_1! i_2! i_3! i_4!} \tag{*}$$

where

$$M = \max \left\{ \frac{n!}{4^n i_1! i_2! i_3! i_4!} : i_1 \cdots, i_4 \ge 0, \sum_{r=1}^4 i_r = n \right\}$$

It is not difficult to see that the maximum is attained when i, j and k are all closest to  $\frac{1}{3}n$ , so that

$$M \le \frac{n!}{4^n(\lfloor \frac{1}{4}n \rfloor!)^4}$$

Furthermore, the final summation in (\*) equals 1, since the summand is the probability that, in allocating n balls randomly to four urns, the urns contain  $i_1, \dots, i_4$  balls respectively. It follows that

$$p_{\mathbf{0},\mathbf{0}}(2n) \le \frac{(2n)!}{16^n n! (|\frac{1}{4}n|!)^4}$$

which, by Stirling's formula, is strictly no bigger than  $Cn^{-2}$ , for some constant C. Therefore:

$$\sum_{n=0}^{\infty} p_{\mathbf{0},\mathbf{0}}(2n) < \infty$$

implying that the origin  ${\bf 0}$  is transient.

Shorter way: project onto  $\mathbb{Z}^3$  by discarding all coordinates except the first 3. Now we have a new possibility of the random walk  $X_n^{\text{proj}}$  staying where it is with probability  $\frac{4-3}{4} = \frac{1}{4}$  (when the original walk jumps in one of the discarded directions), and when it jumps,

$$\mathbb{P}(X_n^{\mathrm{proj}} = \mathbf{i} + \mathbf{e}_\alpha \mid X_n^{\mathrm{proj}} = \mathbf{i}) = \frac{1/8}{1 - 1/4} = \frac{1}{6} \quad \alpha = 1, 2, 3$$

where  $\mathbf{e}_1 = (1, 0, 0), \, \mathbf{e}_2 = (0, 1, 0), \, \mathbf{e}_3 = (0, 0, 1).$ 

Clearly, if the original walk is recurrent then the projected walk is too. But we see when the chain jumps, it behaves like the simple symmetric random walk on  $\mathbb{Z}^3$ , which we know is transient. Hence the projected walk is transient, and so must the original chain be.

Setting  $\pi = \pi P$  reveals

$$\frac{1}{2}\pi_1 + \frac{1}{2}\pi_5 = \pi_1$$

$$\frac{1}{2}\pi_2 + \frac{1}{4}\pi_4 = \pi_2$$

$$\pi_3 + \frac{1}{4}\pi_4 = \pi_3$$

$$\frac{1}{2}\pi_2 + \frac{1}{4}\pi_4 = \pi_4$$

$$\frac{1}{2}\pi_1 + \frac{1}{4}\pi_4 + \frac{1}{2}\pi_5 = \pi_5$$

which gives immediately  $\pi_4 = 0 = \pi_2$ ,  $\pi_1 = \pi_5$ . Thus any vector of the form  $\pi = (\pi_1, 0, \pi_3, 0, \pi_1)$  st.  $2\pi_1 + \pi_3 = 1$  is an invariant distribution

Let  $X_n$  be the number of molecules in A after n epoch of time.  $X = \{X_n \mid n \ge 0\}$  as a Markov chain, owing to the independence of the choice for the passing molecule from previous events. We have  $p_{i,i+1} = (N-i)/N$ ,  $p_{i,i-1} = i/N$ .

The detailed balance equations

$$\pi_i p_{i,i+1} = \pi_{i+1} p_{i+1,i}$$

are solved by  $\pi_{i+1} = \pi_i(N-i)/(i+1)$ , thus

$$\pi_N = \frac{1 \times 2 \times \dots \times (N-1) \times N}{N \times (N-1) \times \dots \times 2 \times 1} \pi_0 = \frac{N!}{N!} \pi_0 = \pi_0$$

So for some 0 < M < N,

$$\pi_{M} = \frac{N - (M - 1)}{M} \pi_{M-1} = \frac{(N - (M - 1)) \times (N - (M - 2)) \times \dots \times N}{M \times (M - 1) \times \dots \times 1} \pi_{0}$$
$$= \frac{N!}{(N - M)!M!} \pi_{0} = \binom{N}{M} \pi_{0}$$

As  $\pi_i$  is a distribution, must have  $\sum_i \pi_i = 1$ , ie.

$$\pi_0 \sum_{i=0}^{N} \binom{N}{i} = 1$$

Thus  $\pi_0 = 2^{-N}$ , and

$$\pi_i = 2^{-N} \binom{N}{i}, \qquad i = 0, 1, \cdots, N$$

ie  $\pi_i \sim \text{Bin } (N, 1/2)$ 

Equations yield

$$\pi_1 = p\pi_3$$

$$\pi_2 = \pi_1 + \frac{2}{3}\pi_2 + (1-p)\pi_3$$

$$\pi_3 = \frac{1}{3}\pi_2$$

Hence the invariant distribution is of the form C(p/3, 1, 1/3) for different normalisation constants C depending on the value of p. For p = 1/16, we obtain

$$\pi = \left(\frac{1}{65}, \frac{48}{65}, \frac{16}{65}\right)$$

For p = 1/6,

$$\pi = \left(\frac{1}{25}, \frac{18}{25}, \frac{6}{25}\right)$$

and for p = 1/12,

$$\pi = \left(\frac{1}{49}, \frac{36}{49}, \frac{12}{49}\right)$$

We can see that the first entry of the invariant distribution in each case is exactly what we would obtain letting  $n \to \infty$  in the calculation of  $p_{1,1}(n)$  in the previous example sheet.

(a) Detailed balance equations are

$$\pi_1(1-p) = \pi_2 q$$

Hence reversible.

(b) Detailed balance equations are

$$\pi_1 p = \pi_2 (1 - p)$$

$$\pi_2 p = \pi_3 (1 - p)$$

$$\pi_3 p = \pi_1 (1 - p)$$

Hence reversible, except when  $p = \frac{1}{2}$ 

(c) Have

$$\pi_i = \pi_{i+1}(1-p) \Rightarrow \pi_i = \frac{p^{i+1}}{(1-p)^i}\pi_0$$

Hence reversible.

(d) Detailed balance equations give  $\pi_i = \frac{1}{n}$  for each i. Thus reversible.

Detailed balance equations give

$$\pi_i \lambda_i = \pi_{i+1} \mu_{i+1} \Rightarrow \pi_i = \frac{\lambda_{i-1} \cdots \lambda_0}{\mu_i \cdots \mu_1} \pi_0$$

Hence reversible.