Part IA — Markov Chains Example Sheet 1

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 $X = (X_n)$ is a Markov Chain, and so satisfies the Markov Property

$$\mathbb{P}(X_{n+1} = i_{n+1} \mid X_0 = i_0, \dots, X_n = i_n) = \mathbb{P}(X_{n+1} = i_{n+1} \mid X_n = i_n)$$

Given that $Z_k = X_{m+k}$, with $k \ge 0$, want to show that Z_k is a Markov Chain, ie.

$$\mathbb{P}(Z_{n+1} \mid Z_0, \cdots, Z_n) = \mathbb{P}(Z_{n+1} \mid Z_n) \qquad (*)$$

Substituting in our definition of Z_k ,

$$\mathbb{P}(Z_{n+1} \mid Z_0, \dots, Z_n) = \mathbb{P}(X_{m+n+1} | X_m, \dots, X_{m+n})$$

$$= \mathbb{P}(X_{m+n+1} | X_{m+n}) \quad \text{as } X \text{ is Markov}$$

$$= \mathbb{P}(Z_{n+1} \mid Z_n)$$

Hence Z_k satisfies the Markov property, and is thus a Markov Chain. $Z_0 = X_m = i$, so Z_k has starting state i.

Let X_1, \dots, X_n be independent random variables, so $\mathbb{P}(X_{n+1} = i_{n+1} \mid X_n = i_n) = \mathbb{P}(X_{n+1} = i_{n+1})$. But then

$$\mathbb{P}(X_{n+1} = i_{n+1} \mid X_0 = i_0, \dots, X_n = i_n) = \mathbb{P}(X_{n+1} = i_{n+1})$$

also by independence, so Markov property is satisfied.

Homogeneous if $\mathbb{P}(X_{n+1} = i_{n+1} \mid X_n = i_n)$ does not depend on the value of n, so must have X_n identically distributed.

As X_n is the maximum reading obtained after n throws, this depends only on the last maximum, so X_n is a Markov chain. In particular, the state transition matrix is

$$\begin{pmatrix} 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 0 & 2/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 0 & 0 & 3/6 & 1/6 & 1/6 & 1/6 \\ 0 & 0 & 0 & 4/6 & 1/6 & 1/6 \\ 0 & 0 & 0 & 0 & 5/6 & 1/6 \\ 0 & 0 & 0 & 0 & 0 & 6/6 \end{pmatrix}$$

We have

$$S_{n+1} = \begin{cases} S_n + 1 & \text{with probability } p \\ S_n - 1 & \text{with probability } q \end{cases}$$

and $S_0 = 0$. Let $X_n = |S_n|$

Clearly X_n satisfies the Markov property (where the walk is now only depends on where it was one step before), so X is a Markov Chain with initial state $X_0 = 0$. The transition probabilities are given by the transition matrix

$$\begin{pmatrix} 0 & 1 & 0 & 0 & \cdots \\ q & 0 & p & 0 & \cdots \\ 0 & q & 0 & p & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

Define $M_k := \max\{S_k \mid 0 \le k \le n\}$, and $Y_n = M_n - S_n$ Not sure about this part.

X is a Markov chain, thus

$$\mathbb{P}(X_{n+1} = i_{n+1} \mid X_0 = i_0, \dots, X_n = i_n) = \mathbb{P}(X_{n+1} = i_{n+1} \mid X_n = i_n)$$

And so

$$\mathbb{P}(Y_{r+1} \mid Y_0, \dots, Y_r) = \mathbb{P}(X_{n_{r+1}} \mid X_{n_0}, \dots, X_{n_r})$$

If $n_r = 2r$ and X is a simple random walk on Z st.

$$X_{n+1} = \begin{cases} X_n + 1 & \text{with probability } p \\ X_n - 1 & \text{with probability } q \end{cases}$$

the allowed states are $\{\cdots, -2, -1, 0, 1, 2, \cdots\}$. Of course, the probability of finding oneself in an odd state is always zero (we are always taking an even number of states).

It is difficult to visualise the state transition matrix as the state space is infinite, but a general row has the form

$$(\cdots \quad 0 \quad q^2 \quad 0 \quad 2pq \quad 0 \quad p^2 \quad \cdots)$$

With probability 2pq of returning to the current state, p^2 moving two to the right and q^2 moving two to the left. (Thus rows summing to $(p+q)^2=1$)

I think the answer should be no. X and Y are both Markov chains, so the value of $X_{n+1} + Y_{n+1}$ only depends on X_n , and Y_n . But we only have the value of $X_n + Y_n$ (loss of information).

For an explicit counterexample, try to smuggle information about X_1 into Y, so that $X_n + Y_n$ depends on X_1 (contradicting the Markov property) but X_n and Y_n are both Markov.

Idea: Let X be a random walk on the integers, so $X_1=\pm 1$ with some probabilities. Define Y_n as a symmetric 'boosted' random walk:

$$Y_n = \begin{cases} Y_{n-1} + 1 + 3X_1 & \text{with probability } 1/2 \\ Y_{n-1} - 1 + 3X_1 & \text{with probability } 1/2 \end{cases}$$

Then $X_n + Y_n$ depends on X_1 , not just X_{n-1}, Y_{n-1}

(i) Let the probability of the flea being on the original vertex after n hops be p_n . Then

$$p_n = \begin{cases} 0 & \text{if flea on first vertex after } n-1 \text{ hops} \\ \frac{1}{2} & \text{otherwise} \end{cases}$$

Hence we set up the recurrence, with $p_0 = 1$

$$p_n = 0 \cdot p_{n-1} + \frac{1}{2}(1 - p_{n-1})$$
$$= \frac{1}{2}(1 - p_{n-1})$$

Solving this difference equation gives general solution $p_n = \frac{1}{3} + \frac{2}{3} \left(-\frac{1}{2}\right)^n$.

(ii) Let p_n be defined as before. This can be though of as a Markov chain on the state space of triangle vertices $\{1, 2, 3\}$, with transition matrix

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

Given $p_{1,1}(0)=1$ (the flea starts on vertex 1) we want to find $p_{1,1}(n)$ The eigenvalues κ of P are the roots of the equation $\det(P-\kappa I)$, with solutions

$$\kappa_1 = 1, \quad \kappa_2, \kappa_3 = -\frac{1}{2} \pm \frac{i}{2\sqrt{3}}$$

Therefore

$$P = U^{-1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \kappa_2 & 0 \\ 0 & 0 & \kappa_3 \end{pmatrix} U$$

for some invertible matrix U. It follows that

$$P^{n} = U^{-1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \kappa_{2}^{n} & 0 \\ 0 & 0 & \kappa_{3}^{n} \end{pmatrix} U$$

Now note that $\kappa_2, \kappa_3 = \frac{1}{\sqrt{3}} e^{\pm i\pi/6}$, so

$$p_{1,1}(n) = A + Be^{i\pi n/6} + Ce^{-i\pi n/6}$$
$$= A + B' \left(3^{-1/2}\right)^n \cos(\pi n/6) + C' \left(3^{-1/2}\right)^n \sin(\pi n/6)$$

Using the b.c.s,

$$p_{1,1}(0) = 1, p_{1,1}(1) = 0, p_{1,1}(2) = 4/9$$
 and hence

$$p_{1,1}(n) = \frac{1}{3} + \frac{2}{3} \left(3^{-1/2}\right)^n \cos(\pi n/6)$$

Let X_n be the score of the die on the n^{th} roll; this is a Markov Chain. Define Y_n as

$$Y_n = \begin{cases} 1 & \text{if } X_n \neq 6 \\ 2 & \text{if } X_n = 6 \end{cases}$$

This has transition matrix

$$P = \begin{pmatrix} \frac{4}{5} & \frac{1}{5} \\ 1 & 0 \end{pmatrix}$$

We wish to find $P(Y_n = 1 \mid Y_1 = 1)$. Diagonalising P, we have

$$P^{n} = U^{-1} \begin{pmatrix} 1 & 0 \\ 0 & (-1/5)^{n} \end{pmatrix} U$$

for some invertible matrix U, hence

$$p_{2,2}(n) = A + B\left(-\frac{1}{5}\right)^n$$

for probability of 6 on the $n^{\rm th}$ throw. Given $p_{2,2}(0)=1$ and $p_{2,2}(1)=0$, this yields

$$p_{1,1}(n) = -\frac{1}{4} + \frac{5}{4} \left(-\frac{1}{5}\right)^n$$

Then $p_{2,1}(n) = 1 - p_{2,2}(n) = 1/4 - (5/4)(5^{-1})^n$

Next part: Then all other scores have probability 1/4, and transition matrix becomes

$$P = \begin{pmatrix} 0 & 0 & 1/4 & 1/4 & 1/4 & 1/4 \\ 1/4 & 0 & 0 & 1/4 & 1/4 & 1/4 \\ 1/4 & 1/4 & 0 & 0 & 1/4 & 1/4 \\ 1/4 & 1/4 & 1/4 & 0 & 0 & 1/4 \\ 1/4 & 1/4 & 1/4 & 1/4 & 0 & 0 \\ 0 & 1/4 & 1/4 & 1/4 & 1/4 & 0 \end{pmatrix}$$

I can't spot a way to simplify this. Want $\{1, 2, 3, 4\}, \{5\}, \{6\}$ as states but not sure what the first row of the transition matrix is.

(a) For p=1/16, the characteristic equation $\det(P-\kappa\iota)=0$ gives roots $\kappa_1=1,\ \kappa_2=-1/4,\ \kappa_3=-1/12.$ Thus

$$P^{n} = U^{-1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & -1/4 & 0 \\ 0 & 0 & -1/12 \end{pmatrix} U$$

for some invertible matrix U, and so

$$p_{1,1}(n) = A + B\left(-\frac{1}{4}\right)^n + C\left(-\frac{1}{12}\right)^n$$

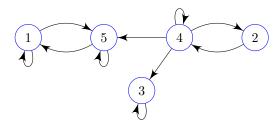
Using $p_{1,1}(0) = 1$, $p_{1,1}(1) = 0$, $p_{1,1}(2) = 0$ This gives A = 1/65, B = -2/5 and C = 18/13

(b) For p=1/6, the characteristic equation $\det(P-\kappa\iota)=0$ gives roots $\kappa_1=1$, $\kappa_2,\kappa_3=-1/6\pm i/6=\frac{1}{\sqrt{18}}e^{\pm i\pi/4}$, so similarly,

$$p_{1,1}(n) = A + \left(18^{-1/2}\right)^n B\cos(\pi n/4) + \left(18^{-1/2}\right)^n C\sin(\pi n/4)$$

yielding A = 1/13, B = 12/13, C = -18/13

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.

Suppose S contains no closed communicating classes.

Pick some $i_0 \in S$. If i is a singleton communicating class, then $\{i_0\}$ is closed. So $\exists i_1 \in S$ st. $i_0 \to i_1$. Similarly, $\exists i_2 \in S$ st. $i_1 \to i_2$. $\{i_2 \neq i_0, \text{ or } \{i_0, i_1\} \text{ is a closed communicating class}\}$. Continuing in this fashion, we eventually hit the 'last' element in the state space i_n , as it is finite. Then $S = \{i_0, \dots, i_n\}$ itself is a closed communicating class.

For a transition matrix with no communicating classes a simple example is:

$$\begin{pmatrix} 0 & 1 & 0 & 0 & \cdots \\ 0 & 0 & 1 & 0 & \cdots \\ 0 & 0 & 0 & 1 & \ddots \\ \vdots & \vdots & & \ddots \end{pmatrix}$$

ie. $i_0 \rightarrow i_1 \rightarrow i_2 \rightarrow \cdots$

If $X_0=2$, following the strategy we see the state space produced is $\{0,2,4,6,8,10\}$. The transition matrix P is given by

$$P = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 1/2 & 0 & 1/2 & 0 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 1/2 & 0 \\ 0 & 1/2 & 0 & 0 & 0 & 1/2 \\ 0 & 0 & 0 & 1/2 & 0 & 1/2 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

With $\{0\}$ and $\{10\}$ being absorbing states. Want to show $\mathbb{P}(X_n=10|X_0=2)=1/5$

Not sure about this bit.