

Introduction to Stochastic Processes

Assignment 2
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Exercise

We are given the following two stochastic matrices

$$P_1 = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{2}{3} & \frac{1}{3} & 0 \\ 0 & 0 & \frac{2}{3} & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}, \quad P_2 = \begin{pmatrix} \frac{1}{5} & \frac{3}{5} & \frac{1}{5} & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{2}{3} & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & \frac{3}{4} & \frac{1}{4} \end{pmatrix}$$

- (1) Find the recurrent states, and the irreducible classes of each matrix.
- (2) For P_1 , how does $N_4(4)$ distribute?
- (3) For P_2 , compute $f_{1 \rightarrow 4}$ and $f_{6 \rightarrow 5}$.

- (1) In P_1 the recurrent states are 1, 5, 6. 1 is obviously recurrent since if you get to it, you will always return immediately to it (it is absorbing). If you get to 5, you either go to 5 again or go to 6 and then 5, so you will always return to 5 eventually. And since $5 \rightarrow 6$, 6 is also recurrent. $2 \rightarrow 1$ but 2 is not reachable from 1, and so 2 is transient. And 5 is reachable from 3 but 3 is not reachable from 5, so 3 is transient and since $4 \rightarrow 3$, 4 is also transient. The irreducible classes are $\{1\}$ and $\{5, 6\}$ as they are obviously both closed and connected.

In P_2 the recurrent states are 2, 5, 6. 2 is recurrent since it is absorbing, and 5 and 6 are for the same reason as before. Since 2 is reachable from 4 but not vice versa, 4 is transient. And since $1 \rightarrow 3 \rightarrow 4$, 1 and 3 are also transient. The irreducible classes are $\{2\}$ and $\{5, 6\}$.

- (2) To compute this we must compute f_4 , which is the probability of returning to 4. The probability of not returning to 4 is if we go $4 \rightarrow 3 \rightarrow 5$ which has a probability of $\mathbb{P}(X_1 = 3 \mid X_0 = 4) \cdot \mathbb{P}(X_2 = 5 \mid X_1 = 3) = \frac{2}{3} \cdot \frac{1}{3} = \frac{2}{9}$. Thus $f_4 = \frac{7}{9}$ and so

$$N_4(4) \sim \text{Geo}(1 - f_4) - 1 = \text{Geo}\left(\frac{2}{9}\right) - 1$$

- (3) To go from 1 to 4, we can visit 1 an arbitrary amount of times and then go to 3 and then 4, so the probability is

$$\begin{aligned} \sum_{n=0}^{\infty} \mathbb{P}(X_{n+2} = 4, X_{n+1} = 3, X_n = \dots = X_1 = 1 \mid X_0 = 1) &= P(3 \rightarrow 4) \cdot P(1 \rightarrow 3) \cdot \sum_{n=0}^{\infty} P(1 \rightarrow 1)^n \\ &= \frac{1}{10} \cdot \sum_{n=0}^{\infty} \frac{1}{5} = \frac{1}{10} \cdot \frac{5}{4} = \frac{1}{8} \end{aligned}$$

Thus $f_{1 \rightarrow 4} = \frac{1}{8}$. And $f_{6 \rightarrow 5} = 1$ since 6 and 5 are recurrent and we showed in lecture that this means $f_{6 \rightarrow 5} = 1$.

Exercise

Yaron collects pokemon. He wants to collect all N pokemon in a series. In every package there is a pokemon which is uniformly distributed and independent of other packages. Let us denote by X_n the number of different pokemon Yaron has after purchasing n packages.

- (1) Show that X_n is a Markov chain over the state space $\{0, 1, \dots, N\}$ and compute the probability transitions.
- (2) Assuming $X_0 = 0$, let $R_k = T_k - T_{k-1}$. Find the distribution of R_k and explain why R_k are all independent.

- (3) Find a formula for the expected time until Yaron will collect all N distinct pokemon and show that it is approximately $N \log N$.

- (1) If $X_{n-1} = a_{n-1}, \dots, X_0 = a_0$ then the probability that $X_n = a_{n-1}$ is $\frac{a_{n-1}}{N}$ since this is the probability of choosing one of the pokemon already collected. And the probability that $X_n = a_{n-1} + 1$ is $\frac{N - a_{n-1}}{N}$, as this is the probability of choosing a new pokemon. This is independent of X_0, \dots, X_{n-2} and so X_n is indeed a Markov chain. And the transition probabilities are

$$\mathbb{P}(X_n = i \mid X_{n-1} = j) = \begin{cases} \frac{j}{N} & i = j \\ \frac{N-j}{N} & i = j + 1 \\ 0 & \text{else} \end{cases}$$

- (2) R_k is the number of packages opened between finding the $k-1$ th and k th distinct pokemon. Let us denote $T_{k-1} = t$ then $X_t = k-1$ and so

$$\begin{aligned} \mathbb{P}(R_k = a) &= \mathbb{P}(T_k = a + t) = \mathbb{P}(X_{a+t} = k, X_{a+t-1} = k-1, \dots, X_t = k-1) \\ &= \mathbb{P}(X_{a+t} = k \mid X_{a+t-1} = k-1) \cdot \mathbb{P}(X_{a+t-1} = k-1 \mid X_{a+t-2} = k-1) \cdots \mathbb{P}(X_{t+1} = k-1 \mid X_t = k-1) \end{aligned}$$

By the previous subquestion this is equal to

$$\frac{N - k + 1}{N} \cdot \left(\frac{k-1}{N} \right)^{a-1}$$

and thus $R_k \sim \text{Geo}\left(1 - \frac{k-1}{N}\right)$. R_k are all independent since the time it takes to find the k th pokemon after finding the $k-1$ th has no bearing on how long it takes to find the j th after finding the $j-1$ th.

- (3) This is asking for the expected value of T_N . By definition $T_N = R_N + \dots + R_1$ and so by linearity of the expected value, and the expected value of a geometric distribution we get

$$\mathbb{E}[T_N] = \sum_{k=1}^N \mathbb{E}[R_N] = \sum_{k=1}^N \frac{N}{N - k + 1} = N \cdot \sum_{k=1}^N \frac{1}{k}$$

Now, it is known that $\sum_{k=1}^N \frac{1}{k} \sim \log N$ and thus $\mathbb{E}[T_N] \sim N \log N$ as required.

Exercise

Let X_n be a Markov chain over the state space \mathbb{Z} where

$$\mathbb{P}(X_{n+1} = j \mid X_n = i) = \frac{1}{2} \delta(|i - j| = 1)$$

($\delta\varphi = 1$ if and only if φ is true/is satisfied in the structure.) This describes a random walk on \mathbb{Z} where at each step the subject walks either left or right by one unit.

- (1) Show that there exists a constant $c > 0$ such that $\mathbb{P}_0(X_{2n} = 0) \geq \frac{c}{\sqrt{n}}$.
- (2) Show that $\mathbb{E}[N_0(0)] = \sum_{n=1}^{\infty} \mathbb{P}_0(X_{2n} = 0) = \infty$.
- (3) Show that in this chain, all states are recurrent.

- (1) Notice that at every step, X_n changes by ± 1 . So let us define the indicator

$$I_n = X_n - X_{n-1} = \begin{cases} 1 & X_n = X_{n-1} + 1 \\ -1 & X_n = X_{n-1} - 1 \end{cases}$$

These indicators are all independent of one another as the movement of each step are all independent. By definition, since $X_0 = 0$, $X_n = \sum_{k=1}^n I_k$. Now, notice that $\frac{I_n+1}{2} \sim \text{Ber}\left(\frac{1}{2}\right)$ and thus $\frac{X_n+n}{2} = \sum_{k=1}^n \frac{I_k+1}{2} \sim \text{Bin}\left(n, \frac{1}{2}\right)$.

$$\mathbb{P}_0(X_{2n} = 0) = \mathbb{P}_0\left(\frac{X_{2n} + 2n}{2} = n\right) = \binom{2n}{n} \cdot \frac{1}{2^{2n}}$$

Now, using Stirling's approximation, recall that $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$ ($f(n) \sim g(n)$ means the limit of their quotient is 1). This asymptotically,

$$\mathbb{P}_0(X_{2n} = 0) \sim \frac{\sqrt{2\pi n} \left(\frac{2n}{e}\right)^{2n}}{2\pi n \left(\frac{n}{e}\right)^{2n}} \cdot \frac{1}{2^{2n}} = \frac{1}{\sqrt{\pi}} \cdot \frac{1}{\sqrt{n}}$$

And so this means that $\mathbb{P}_0(X_{2n} = 0) \in \Omega\left(\frac{1}{\sqrt{n}}\right)$. Thus eventually $\mathbb{P}_0(X_{2n} = 0)$ will be greater than $\frac{c}{\sqrt{n}}$ for some c . Then by looking at the range of ns (which is finite) for which $\mathbb{P}_0(X_{2n} = 0) < \frac{c}{\sqrt{n}}$, we can reduce the c until it holds for all n (since the probability is never zero, we can indeed reduce c to such a point).

- (2) Let us define new indicators

$$I_n = \begin{cases} 1 & X_{2n} = 0 \\ 0 & \text{else} \end{cases}$$

Notice that for odd n , $I_n = 0$ (there are a few ways to show this, $\mathbb{P}_0(X_n = 0) = \mathbb{P}_0\left(\frac{X_n+n}{2} = \frac{n}{2}\right) = 0$ since $\frac{X_n+n}{2}$ is binomial which takes on only integers. Alternatively if r is the number of steps $+1$ and ℓ is the number of steps -1 , then we must have $r + \ell = n$ and $r - \ell = 0$ so $r + \ell \equiv_2 n \equiv_2 1$ and $r - \ell \equiv_2 0$ but $r - \ell \equiv_2 r + \ell$ in contradiction.) By definition since I_n denote a return to 0,

$$N_0(0) = \sum_{n=1}^{\infty} I_{2n}$$

and so by the σ -linearity of the expected value:

$$\mathbb{E}[N_0(0)] = \sum_{n=1}^{\infty} \mathbb{E}[I_{2n}] = \sum_{n=1}^{\infty} \mathbb{P}_0(X_{2n} = 0)$$

as $\mathbb{E}_0[I_{2n}] = \mathbb{P}_0(I_{2n} = 1) = \mathbb{P}_0(X_{2n} = 0)$.

- (3) Since every state is reachable from every other state, if one state is recurrent they all are. This is also obvious due to symmetry. Now since $N_0(0) \sim \text{Geo}(1 - f_0) - 1$, if f_0 were less than one it would have a finite expected value (as a finite geometric distribution), contradicting the previous subquestion. And so $f_0 = 1$ and thus 0 is recurrent. And as said previously, since every n is reachable from 0, every n must be recurrent as required.