S&DS 351: Stochastic Processes - Homework 3

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Problem 1

(10 points) Is it possible for a transient state to be periodic? If so, construct an example of such a Markov chain; otherwise, give a mathematical proof why not.

Note: I (fortunately) solved this after proving problem 3, so for a more thorough proof on how this example is transient, please see Problem 3.

Yes, it is possible for a transient state to be periodic. Consider a 1-dimensional asymmetric random walk on \mathbb{Z} :

$$X_n = X_{n-1} + Z_n$$
, where $\mathbb{P}(Z_n = +1) = p$ and $\mathbb{P}(Z_n = -1) = 1 - p$,

for some $p \in (0,1)$ with $p \neq \frac{1}{2}$. Starting at state 0, state 0 is transient (see Problem 3).

Define the period as $d_i = \gcd\{n: P^n(i,i) > 0\}$, where P is the transition matrix.

In the random walk, the walk must trivially take as many +1 steps as -1 steps to reach the initial state. Thus one can only return to state x starting from x in an even number of steps. Note that this holds for all integers. Hence for each integer x,

$$(P^n)(x,x) > 0 \implies n \text{ is even.}$$

 $(P^n)(x,x) = 0 \implies n \text{ is odd.}$

Therefore, the greatest common divisor of all such n is 2, and every state $x \in \mathbb{Z}$ has period 2.

Problem 2

Let X_0, X_1, \ldots be a Markov chain with transition matrix P. Let $k \geq 1$ be an integer.

- 1. (5 points) Prove that $Y_n = X_{kn}$ is also a Markov chain. Find its transition matrix.
- 2. (10 points) Suppose that the original chain $\{X_n\}$ is irreducible. Is $\{Y_n\}$ irreducible? If so, prove it; if not, provide a counterexample.
- 3. (10 points) Suppose that the original chain $\{X_n\}$ is aperiodic. Is $\{Y_n\}$ aperiodic? If so, prove it; if not, provide a counterexample.
- 4. (10 points) Suppose that the original chain $\{X_n\}$ is transient. Is $\{Y_n\}$ transient? If so, prove it; if not, provide a counterexample.
- 5. (15 points) Suppose that the original chain $\{X_n\}$ is recurrent. Is $\{Y_n\}$ recurrent? If so, prove it; if not, provide a counterexample.

6. (5 points) Suppose that the original chain X_n is irreducible and that it has period d. What is the period of each state i in the new Markov chain Y_n for k = d?

Problem 3

(Asymmetric random walk, 15 points) Consider the asymmetric random walk on \mathbb{Z} , that is, $X_n = X_{n-1} + Z_n$, where Z_1, Z_2, \ldots are iid and $\mathbb{P}(Z_n = +1) = p$ and $\mathbb{P}(Z_n = -1) = 1 - p$, with $p \in [0, 1]$ and $p \neq \frac{1}{2}$. Show that the state 0 is a transient state.

In Lecture 7 we saw/will see that when $p = \frac{1}{2}$ this is not true anymore and the state 0 is recurrent. Can you explain intuitively why this is the case?

Hint: You may want to use Stirling's formula that $\lim_{n\to\infty} \frac{n!}{(n/e)^n\sqrt{2\pi n}} = 1$.

Starting from $X_0 = 0$, the random walk is at state 0 again at t = n only when it has taken an equal number of +1 steps as -1 steps. As such, n must be even.

Suppose n = 2k, and k is the number of Z_i that are +1,

$$\mathbb{P}(X_{2k} = 0 \mid X_0 = 0) = \binom{2k}{k} p^k (1-p)^k$$

Note that $\mathbb{P}(X_n = 0 \mid X_0 = 0) = 0$ if n is odd

Hence the series of return probabilities at 0 is

$$\sum_{n=0}^{\infty} \mathbb{P}(X_n = 0 \mid X_0 = 0) = 1 + \sum_{k=1}^{\infty} {2k \choose k} p^k (1-p)^k,$$

accounting for the initial state of 0. Using Stirling's approximation,

$$n! \sim \left(\frac{n}{e}\right)^n \sqrt{2\pi n}$$
 as $n \to \infty$,

applying to this case,

$$\binom{2k}{k} = \frac{(2k)!}{k! \, k!} \approx \frac{\sqrt{4\pi k} \left(\frac{2k}{e}\right)^{2k}}{2\pi k \left(\frac{k}{e}\right)^k \left(\frac{k}{e}\right)^k} = \frac{4^k}{\sqrt{\pi k}}$$

Therefore,

$$\binom{2k}{k} p^k (1-p)^k \; \approx \; \frac{4^k}{\sqrt{\pi k}} \left[p(1-p) \right]^k \; = \; \frac{\left[4 \, p(1-p) \right]^k}{\sqrt{\pi k}}.$$

If $p \neq \frac{1}{2}$, then 4p(1-p) < 1 (If f(x) = x(1-x), then f'(x) = -x + 1 - x = -2x + 1. Solving for the max when $f'(x) = 0, x = \frac{1}{2}$).

Note, that as $k \to \infty$, $\left[4 \, p (1-p)\right]^k$ decays exponentially. Therefore,

$$\binom{2k}{k} p^k (1-p)^k = O([4p(1-p)]^k)$$
 and $\sum_{k=1}^{\infty} \binom{2k}{k} p^k (1-p)^k < \infty$.

Thus,

$$\sum_{n=0}^{\infty} \mathbb{P}(X_n = 0 \mid X_0 = 0) = 1 + \sum_{k=1}^{\infty} {2k \choose k} p^k (1-p)^k < \infty.$$

which defines a transient state.

However, when $p = \frac{1}{2}$,

$${2k \choose k} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^k \approx \frac{\left[4 \cdot 0.5(1 - 0.5)\right]^k}{\sqrt{\pi k}} = \frac{1}{\sqrt{\pi k}},$$

so

$$\sum_{k=1}^{\infty} \frac{1}{\sqrt{k}} = \infty.$$

which defines a recurrent state when $p = \frac{1}{2}$.

Exercise 1.8

Consider a Markov chain on the integers with

$$P(i, i + 1) = 0.4$$
 and $P(i, i - 1) = 0.6$ for $i > 0$,
$$P(i, i + 1) = 0.6$$
 and $P(i, i - 1) = 0.4$ for $i < 0$,
$$P(0, 1) = P(0, -1) = \frac{1}{2}.$$

This is a chain with infinitely many states, but it has a sort of probabilistic "restoring force" that always pushes back toward 0. Find the stationary distribution.

From the "resisting force" there is a single stationary distribution, expected to be symmetric around 0, with a geometric decay away from 0. Step 2. Stationarity equations and ratio method. Denote the stationary distribution by $\{\pi_i\}_{i\in\mathbb{Z}}$, satisfying

$$\sum_{j \in \mathbb{Z}} \pi_j P(j, i) = \pi_i \quad \text{for all } i \in \mathbb{Z},$$

and $\sum_{i\in\mathbb{Z}} \pi_i = 1$. Because this is a (two-sided) birth-death type chain, one may use the standard balance equations:

$$\pi_i P(i, i+1) = \pi_{i+1} P(i+1, i).$$

Concretely, for $i \geq 1$:

$$\pi_i \times 0.4 = \pi_{i+1} \times 0.6 \implies \frac{\pi_{i+1}}{\pi_i} = \frac{0.4}{0.6} = \frac{2}{3}.$$

For $i \leq -1$:

$$\pi_i \times 0.6 = \pi_{i+1} \times 0.4 \implies \frac{\pi_{i+1}}{\pi_i} = \frac{0.6}{0.4} = \frac{3}{2}.$$

We also need to handle the special transitions at i = 0. The balance equation between 0 and 1 yields:

$$\pi_0 \times 0.5 = \pi_1 \times 0.6 \implies \frac{\pi_1}{\pi_0} = \frac{0.5}{0.6} = \frac{5}{6}.$$

And for i = -1:

$$\pi_{-1} \times 0.6 = \pi_0 \times 0.5 \implies \frac{\pi_{-1}}{\pi_0} = \frac{0.5}{0.6} = \frac{5}{6},$$

Hence,

$$\pi_1 = \frac{5}{6} \pi_0, \quad \pi_{-1} = \frac{5}{6} \pi_0.$$

This shows the symmetry $\pi_1 = \pi_{-1}$ indeed. (a) For i > 0:

$$\frac{\pi_{i+1}}{\pi_i} = \frac{2}{3} \implies \pi_i = \left(\frac{2}{3}\right)^{i-1} \pi_1 \quad \text{for } i \ge 1.$$

But $\pi_1 = \frac{5}{6} \pi_0$, so

$$\pi_i \ = \ \left(\frac{2}{3}\right)^{i-1} \cdot \frac{5}{6} \, \pi_0 \quad \text{for } i \ge 1.$$

(b) For i < 0:

$$\frac{\pi_{i+1}}{\pi_i} = \frac{3}{2} \quad \Longrightarrow \quad \pi_i = \frac{2}{3} \,\pi_{i+1} \quad \text{for } i \le -2,$$

stepping upward until i=-1, for which we already have $\pi_{-1}=\frac{5}{6}\,\pi_0$. Iterating gives

$$\pi_{-2} = \frac{2}{3}\pi_{-1} = \frac{2}{3} \cdot \frac{5}{6}\pi_0 = \frac{5}{9}\pi_0,$$

$$2 \qquad 2 \qquad 5 \qquad 10$$

$$\pi_{-3} = \frac{2}{3}\pi_{-2} = \frac{2}{3} \cdot \frac{5}{9}\pi_0 = \frac{10}{27}\pi_0,$$

and so on. In fact, a direct pattern emerges, and for i < 0,

$$\pi_i = \frac{5}{6} \left(\frac{2}{3}\right)^{|i|-1} \pi_0.$$

Step 3. A unified formula. To summarize, set |0| - 1 = -1 in the exponents interpreted carefully, or write piecewise. A concise way is:

$$\pi_0 = \pi_0, \quad \pi_i = \frac{5}{6} \left(\frac{2}{3}\right)^{|i|-1} \pi_0 \quad \text{for } i \neq 0.$$

We still must determine the constant π_0 by requiring

$$\sum_{i=-\infty}^{\infty} \pi_i = 1.$$

Hence

$$\pi_0 + \sum_{i \neq 0} \frac{5}{6} \left(\frac{2}{3}\right)^{|i|-1} \pi_0 = 1.$$

Factor out π_0 :

$$\pi_0 \left[1 + \frac{5}{6} \sum_{i \neq 0} \left(\frac{2}{3} \right)^{|i|-1} \right] = 1.$$

Next we split the sum at i > 0 and i < 0, noticing the symmetry:

$$\sum_{i \neq 0} \left(\frac{2}{3}\right)^{|i|-1} = 2\sum_{i=1}^{\infty} \left(\frac{2}{3}\right)^{j-1} = 2\sum_{k=0}^{\infty} \left(\frac{2}{3}\right)^{k} = 2 \cdot \frac{1}{1 - \frac{2}{3}} = 2 \cdot 3 = 6.$$

Thus

$$\frac{5}{6} \cdot 6 = 5,$$

and so

$$\pi_0 \left[1+5 \right] = 6 \pi_0 = 1 \implies \pi_0 = \frac{1}{6}$$

Therefore,

$$\pi_0 = \frac{1}{6}, \quad \pi_i \ = \ \frac{5}{6} \left(\frac{2}{3}\right)^{|i|-1} \frac{1}{6} \ = \ \frac{5}{36} \left(\frac{2}{3}\right)^{|i|-1}, \quad i \neq 0.$$

Rewriting compactly,

$$\pi_i = \begin{cases} \frac{1}{6}, & i = 0, \\ \frac{5}{36} \left(\frac{2}{3}\right)^{|i|-1}, & i \neq 0. \end{cases}$$

Exercise 1.10

On page 13 we argued that a limiting distribution must be stationary. This argument was clear in the case of a finite state space. For you fans of mathematical analysis, what happens in the case of a countably infinite state space? Can you still make the limiting argument work?

Let $\{X_n\}$ be a Markov chain with countably infinite state space S, and let P be its one-step transition matrix. Suppose π_n is the distribution of X_n , i.e. for each $j \in S$,

$$\pi_n(j) = \mathbb{P}(X_n = j).$$

Assume that $\pi_{n+1} = \pi_n P$ for all n. If there is a distribution $\tilde{\pi}$ such that

$$\lim_{n \to \infty} \pi_n(i) = \tilde{\pi}(i) \quad \text{for all } i \in S,$$

then $\tilde{\pi}$ is a stationary distribution for P, i.e.

$$\tilde{\pi} = \tilde{\pi} P$$
.

By hypothesis, $\pi_{n+1} = \pi_n P$. In coordinates, that means

$$\pi_{n+1}(j) = \sum_{i \in S} \pi_n(i) P(i,j)$$
 for each $j \in S$.

Taking limits as $n \to \infty$, we want to show

$$\tilde{\pi}(j) = \sum_{i \in S} \tilde{\pi}(i) P(i, j)$$
 for all $j \in S$.

The delicate step is justifying that we can pass the limit inside the infinite sum. Step 1: Pointwise limits. By assumption, for each $i \in S$,

$$\lim_{n \to \infty} \pi_n(i) = \tilde{\pi}(i).$$

Thus, $\{\pi_n(i)\}$ is a convergent sequence of nonnegative numbers (since all π_n are probability distributions). Moreover, because $\sum_{i \in S} \pi_n(i) = 1$ for each n, we have $\{\pi_n(i)\}$ bounded above by 1. **Step 2: Passing the limit through the sum.** Consider the sum

$$\sum_{i \in S} \pi_n(i) P(i,j).$$

Since $P(i,j) \ge 0$ (transition probabilities are nonnegative) and $\sum_{i \in S} \pi_n(i) = 1$, we have

$$0 \le \pi_n(i) P(i,j) \le \pi_n(i).$$

Hence,

$$0 \le \sum_{i \in S} \pi_n(i) P(i,j) \le \sum_{i \in S} \pi_n(i) = 1.$$

We want to apply a standard result (often called the Dominated Convergence Theorem for series or the Monotone Convergence Theorem in the setting of nonnegative series) to interchange the limit and the infinite sum. Specifically, each term $\pi_n(i) P(i,j)$ is dominated by $\pi_n(i)$, and

$$\sum_{i \in S} \pi_n(i) = 1$$

is a convergent sum (it does not depend on i). Therefore,

$$\lim_{n \to \infty} \sum_{i \in S} \pi_n(i) \, P(i,j) \; = \; \sum_{i \in S} \lim_{n \to \infty} \left[\pi_n(i) \, P(i,j) \right] \; = \; \sum_{i \in S} \tilde{\pi}(i) \, P(i,j).$$

(The second equality uses $\lim_{n\to\infty} \pi_n(i) = \tilde{\pi}(i)$.) Step 3: Concluding $\tilde{\pi}$ is stationary. Since $\pi_{n+1}(j) = \sum_{i\in S} \pi_n(i) P(i,j)$, we also have

$$\lim_{n \to \infty} \pi_{n+1}(j) = \lim_{n \to \infty} \sum_{i \in S} \pi_n(i) P(i, j).$$

But $\lim_{n\to\infty} \pi_{n+1}(j) = \tilde{\pi}(j)$. Hence,

$$\tilde{\pi}(j) = \sum_{i \in S} \tilde{\pi}(i) P(i, j)$$
 for all $j \in S$,

which is precisely the condition

$$\tilde{\pi} = \tilde{\pi}P.$$

Thus, $\tilde{\pi}$ is a stationary (or invariant) distribution for the transition matrix P.

The crucial point in the above proof is the passage of the limit inside the (infinite) sum. We use the fact that $\sum_{i \in S} \pi_n(i) P(i,j)$ is a nonnegative series and is dominated by the convergent series $\sum_{i \in S} \pi_n(i) = 1$. Thus, standard convergence theorems guarantee that

$$\lim_{n \to \infty} \sum_{i \in S} \pi_n(i) \, P(i,j) \; = \; \sum_{i \in S} \lim_{n \to \infty} \left[\pi_n(i) \, P(i,j) \right].$$

This justifies exchanging "lim" and " \sum " and completes the proof that a limiting distribution must be stationary even in the countably infinite case.