

# 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  $\mathbb{Z}$ :

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

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

(a) (5 points) Prove that  $Y_n = X_{kn}$  is also a Markov chain. Find its transition matrix.

Since  $Y_n = X_{kn}$ , the conditional probability for  $Y_{n+1}$  can be defined as

$$\mathbb{P}(X_{k(n+1)} = y_{n+1} \mid X_{kn} = y_n, X_{k(n-1)} = y_{n-1}, \dots, X_0 = y_0)$$

Because  $\{X_n\}$  is a Markov chain, it satisfies the Markov property,

$$P(X_{m+1} = x_{m+1} \mid X_m = x_m, \dots, X_0 = x_0) = P(X_{m+1} = x_{m+1} \mid X_m = x_m).$$

Applying for the  $k$ -steps from time  $kn$  to time  $k(n+1)$ ,

$$\mathbb{P}(X_{k(n+1)} = y_{n+1} \mid X_{kn} = y_n, X_{k(n-1)} = y_{n-1}, \dots, X_0 = y_0) = \mathbb{P}(X_{k(n+1)} = y_{n+1} \mid X_{kn} = y_n)$$

Therefore,

$$\mathbb{P}(Y_{n+1} = y_{n+1} \mid Y_n = y_n, Y_{n-1} = y_{n-1}, \dots, Y_0 = y_0) = \mathbb{P}(Y_{n+1} = y_{n+1} \mid Y_n = y_n)$$

Thus,  $\{Y_n\}$  satisfies the Markov property and is a Markov chain.

Solving for the transition matrix, note that for any state  $i, j \in S$ ,

$$\mathbb{P}(Y_{n+1} = j \mid Y_n = i) = \mathbb{P}(X_{k(n+1)} = j \mid X_{kn} = i)$$

In the Markov chain  $\{X_n\}$ ,

$$\mathbb{P}(X_{k(n+1)} = j \mid X_{kn} = i) = (P^k)_{ij}$$

Thus, the one-step transition probability for  $\{Y_n\}$  is

$$\mathbb{P}(Y_{n+1} = j \mid Y_n = i) = (P^k)_{ij}.$$

Therefore, the transition matrix for  $\{Y_n\}$  is  $P^k$ .

- (b) (10 points) Suppose that the original chain  $\{X_n\}$  is irreducible. Is  $\{Y_n\}$  irreducible? If so, prove it; if not, provide a counterexample.

Consider a Markov chain, with state space  $S = \{X_0, X_1\}$ . Define its transition matrix as

$$P_X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Since  $X_0$  and  $X_1$  communicate, this chain is trivially irreducible.

Suppose  $k = 2$ . The transition matrix of  $Y_m = X_{2m}$  is

$$P_Y = P_X^2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$X_0$  and  $X_1$  do not communicate in the new chain, as  $X_0$  and  $X_1$  are closed subsets on  $S$ . Therefore,  $Y_n$  is not irreducible.

Note that this can be generalized to any Markov chain  $X_0, X_1, \dots, X_k$ , where  $\mathbb{P}_X(i, j) = 1$  for  $j = (i + 1) \bmod k$  and  $\mathbb{P}_X(i, j) = 0$  for all other  $i, j$ . This chain is trivially irreducible, and yields a transition matrix  $P^k$  which is a  $k \times k$  identity matrix. As such, the chain  $Y_n$  is not irreducible, as the states  $X_0, X_1, \dots, X_k$  are closed subsets on  $S$ .

- (c) (10 points) Suppose that the original chain  $\{X_n\}$  is aperiodic. Is  $\{Y_n\}$  aperiodic? If so, prove it; if not, provide a counterexample.

- A Markov chain with transition matrix  $P$  is called *aperiodic* if every state  $i$  has period 1. Concretely, the period of a state  $i$  is defined as

$$d(i) = \gcd\{n \geq 1 : (P^n)_{ii} > 0\}.$$

If this gcd equals 1 for every  $i$ , then the chain is aperiodic.

- In the chain  $\{Y_n\}$  defined by  $Y_n := X_{kn}$ , its transition probabilities arise from the  $k$ -step transition probabilities of the original chain. In particular,

$$\mathbb{P}(Y_{m+1} = j \mid Y_m = i) = \mathbb{P}(X_{k(m+1)} = j \mid X_{km} = i) = (P^k)_{ij}.$$

Hence the transition matrix of  $\{Y_n\}$  is exactly  $P^k$ , and the  $m$ -step transition probabilities in  $\{Y_n\}$  are given by  $(P^k)^m = P^{km}$ .

- Therefore, the period of state  $i$  as a state of the chain  $\{Y_n\}$  is

$$d_Y(i) = \gcd\{m \geq 1 : (P^{km})_{ii} > 0\}.$$

We must show that  $d_Y(i) = 1$  whenever  $d(i) = 1$  in the original chain.

**Key idea: Aperiodicity implies eventually-positive returns in all large times.** A standard characterization of aperiodicity is that for each state  $i$ , there exists some integer  $N$  (possibly depending on  $i$ ) such that

$$(P^n)_{ii} > 0 \quad \text{for all } n \geq N.$$

(Equivalently, the set  $\{n : (P^n)_{ii} > 0\}$  is *co-finite*, i.e. it contains all sufficiently large  $n$ .) We use this fact to see what happens when we look only at  $k$ -step transitions, i.e. the set

$$\{m : (P^{km})_{ii} > 0\}.$$

If  $(P^n)_{ii} > 0$  for all  $n \geq N$ , then in particular  $(P^{km})_{ii} > 0$  whenever  $km \geq N$ . Hence for all integers

$$m \geq \lceil N/k \rceil,$$

we have  $(P^{km})_{ii} > 0$ . Thus

$$\{m : (P^{km})_{ii} > 0\}$$

contains *all* sufficiently large integers  $m$ . **Period calculation for the chain  $\{Y_n\}$ .** For the chain

$\{Y_n\}$ , the period of  $i$  is

$$d_Y(i) = \gcd\{m \geq 1 : (P^{km})_{ii} > 0\}.$$

Since this set includes *all* sufficiently large integers  $m$ , its greatest common divisor must be 1. Indeed, any infinite set of consecutive integers has gcd equal to 1. Formally:

- Because  $\{X_n\}$  is aperiodic at  $i$ , there is an  $N$  such that  $(P^n)_{ii} > 0$  for every  $n \geq N$ .
- Consequently, for every integer  $m \geq \lceil N/k \rceil$ , we have

$$(P^{km})_{ii} > 0.$$

- Hence  $\{m : (P^{km})_{ii} > 0\}$  contains the tail set  $\{\lceil N/k \rceil, \lceil N/k \rceil + 1, \lceil N/k \rceil + 2, \dots\}$  of *all* sufficiently large integers.
- The greatest common divisor of any infinite set of consecutive integers is 1.
- Therefore  $d_Y(i) = 1$ .

Since  $d_Y(i) = 1$  for every state  $i$ , the subsampled chain  $\{Y_n\}$  is indeed aperiodic.

- (d) (10 points) Suppose that the original chain  $\{X_n\}$  is transient. Is  $\{Y_n\}$  transient? If so, prove it; if not, provide a counterexample.
- (e) (15 points) Suppose that the original chain  $\{X_n\}$  is recurrent. Is  $\{Y_n\}$  recurrent? If so, prove it; if not, provide a counterexample.
- (f) (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  $k = d$ ?

Since the original chain is irreducible with period  $d$ , for each state  $i$ ,

$$(P_X^d)_{ii} > 0 \quad \text{and} \quad (P_X^{d-b})_{ii} = 0, \quad \forall b = 1, 2, \dots, d-1.$$

Therefore, for any multiple of  $d$ ,  $a \in \mathbb{Z}$  such that  $a \geq 1$ ,

$$(P_X^d)_{ii}^a > 0$$

As such, all returns to  $i$  can occur only at multiples of  $d$  steps.

Define the transition matrix for  $Y_n$  where  $k = d$  as

$$\mathbb{P}(Y_{m+1} = j \mid Y_m = i) = (P_X^k)_{ij} = (P_X^d)_{ij}$$

When  $j = i$ ,

$$(P_Y^1)_{ii} = (P_X^d)_{ii} > 0$$

As such, state  $i$  can reach itself in a single step in chain  $Y$ , forming a self-loop in  $Y$ .

Defining the period of  $i$  in  $Y$ ,

$$d_Y(i) = \gcd \{ m \geq 1 : (P^{dm})_{ii} > 0 \}.$$

Note that  $(P^d)_{ii} > 0$ , so at  $m = 1$ ,  $(P^{d \cdot 1})_{ii} = (P^d)_{ii} > 0$ . Since the chain is irreducible, raising  $P^d$  to a higher power trivially cannot change its positivity:

$$(P^{d^2})_{ii} = (P^d)_{ii}^2 > 0, \quad (P^{d^3})_{ii} > 0, \quad \dots$$

$(P^{dn})_{ii} > 0$  holds for all  $n \geq 1$ , so the set of possible  $m$  is  $\{1, 2, 3, \dots\}$ . Thus,

$$\{ m : (P^{dm})_{ii} > 0 \} = \{1, 2, 3, \dots\},$$

with a trivial gcd of 1. Note that since the  $d$  period was a state property in the original chain,  $\{X_n\}$ , it holds for all states in the old chain, and thus all states in the new chain.

As such, the period of each state in the new chain  $\{Y_n\}$  is 1. In other words, the chain  $\{Y_n\}$  is aperiodic.

### 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  $Z_1, Z_2, \dots$  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  $\lim_{n \rightarrow \infty} \frac{n!}{(n/e)^n \sqrt{2\pi n}} = 1$ .

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

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

accounting for the initial state of 0.

Using Stirling's approximation,

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

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,

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

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  $k \rightarrow \infty$ ,  $[4p(1-p)]^k$  decays exponentially. Therefore,

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

Thus,

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

which defines a transient state.

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

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

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

$$P(i, i+1) = 0.4 \text{ and } P(i, i-1) = 0.6 \text{ for } i > 0,$$

$$P(i, i+1) = 0.6 \text{ and } P(i, i-1) = 0.4 \text{ for } i < 0,$$

$$P(0, 1) = P(0, -1) = \frac{1}{2}.$$

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.

Denote the stationary distribution by  $\{\pi_i\}_{i \in \mathbb{Z}}$

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

and  $\sum_{i \in \mathbb{Z}} \pi_i = 1$ . As this is a two-sided birth-death chain,

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

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  $i \leq -1$ :

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

When  $i = 0$ ,

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

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

Hence,

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

As the probabilities of a jump remain the same, generalizing for all  $i \geq 1$ ,

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

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

Generalizing for all  $i \leq -1$ ,

$$\pi_{i-1} = \frac{2}{3}\pi_i \implies \pi_i = \frac{2^{-i-1}}{3} \pi_{-1} \quad \forall i \leq -1,$$

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

Combining the two cases,

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

Solving for  $\pi_0$ , first recall that,

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

Hence,

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

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

Exploiting the symmetry of the chain,

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

Thus

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

Therefore,

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

### Exercise 1.16

Show that if an irreducible Markov chain has a state  $i$  such that  $P(i, i) > 0$ , then the chain is aperiodic. Also show by example that this sufficient condition is not necessary.

Let  $\{X_n\}$  be an irreducible Markov chain on a countable state space  $S$ . Suppose there is a state  $i \in S$  such that

$$P(i, i) > 0.$$

The period of state  $i$  is defined as

$$d(i) = \gcd\{n \geq 1 : P^n(i, i) > 0\},$$

Since  $P(i, i) = P^1(i, i) > 0$ , there is a positive probability of returning to  $i$  in exactly 1 step. As such,  $1 \in \{n : P^n(i, i) > 0\}$ , so any common divisor of all  $n$  must trivially divide 1.

Therefore,

$$d(i) = \gcd\{n \geq 1 : P^n(i, i) > 0\} = 1.$$

Thus,  $i$  is an aperiodic state.

By irreducibility, for any other state  $j \in S$ ,  $\exists m, k \in \mathbb{Z}$  such that  $m, k \geq 1$ ,  $P^m(j, i) > 0$ , and  $P^k(i, j) > 0$ . Therefore as proven in lecture, for any  $n \geq 1$ ,

$$P^{m+n+k}(j, j) \geq P^m(j, i) P^n(i, i) P^k(i, j).$$

Since  $P^n(i, i) > 0$  for all  $n \geq 1$ , for arbitrarily many  $n$ , the probability  $P^{m+n+k}(j, j)$  is strictly positive. As such,

$$d(j) = \gcd\{n \geq 1 : P^n(j, j) > 0\} = 1.$$

Thus, every state in an irreducible chain with a self-loop, i.e.  $P(i, i) > 0$ , is aperiodic.

To show that this sufficient condition is not necessary, consider a chain on three states  $\{1, 2, 3\}$  with transition matrix:

$$P(1, 1) = 0, \quad P(1, 2) = 1, \quad P(1, 3) = 0,$$

$$P(2, 1) = \frac{1}{2}, \quad P(2, 2) = 0, \quad P(2, 3) = \frac{1}{2},$$

$$P(3, 1) = \frac{1}{2}, \quad P(3, 2) = \frac{1}{2}, \quad P(3, 3) = 0.$$

Note that the chain is irreducible, but does not contain any self-loops, i.e.  $P(i, i) = 0$  for  $i = 1, 2, 3$ .

$$P^2(1, 1) = P(1, 2)P(2, 1) = (1) \left(\frac{1}{2}\right) = \frac{1}{2} > 0,$$

$$P^3(1, 1) = P(1, 2)P(2, 3)P(3, 1) = 1 \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} > 0.$$

Therefore, state 1 can return to itself in both 2 steps and 3 steps, which comprises a subset of the set of all possible return times back to state 1. The gcd of 2 and 3 is 1, or a period of 1. Note that  $P^2(2, 2) > 0$ ,  $P^2(3, 3) > 0$ , and  $P^3(2, 2) > 0$ ,  $P^3(3, 3) > 0$ , so every state also has period 1, making the entire chain aperiodic. Thus,  $P(i, i) > 0$  for some  $i$  is not a necessary condition for a chain to be aperiodic.