

# 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  $\mathbb{Z}$ :

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

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. 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.

Recall that 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  $Y_n = X_{kn}$ , the transition probabilities are given by

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

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

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

A standard characterization of aperiodicity is that for each state  $i$ , there exists some integer  $N$  st

$$(P^n)_{ii} > 0 \quad \forall n \geq N$$

In other words,  $\{n : (P^n)_{ii} > 0\}$  is co-finite, or contains all sufficiently large  $n$ .

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  $m \geq \lceil N/k \rceil$ ,  $(P^{km})_{ii} > 0$ .

Thus,  $\{m : (P^{km})_{ii} > 0\}$  contains the tail set  $\{\lceil N/k \rceil, \lceil N/k \rceil + 1, \lceil N/k \rceil + 2, \dots\}$  of all sufficiently large integers  $m$ .

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

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

As this set includes all sufficiently large integers  $m$ , the gcd of any infinite set must be 1.

Since  $d_Y(i) = 1$  for every state  $i$ ,  $\{Y_n\}$  is 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.

Recall that a Markov chain  $\{X_n\}$  on  $S$  is transient if there exists at least one state  $i \in S$  st

$$\mathbb{P}_i(\text{the chain ever returns to } i) < 1$$

Equivalently,

$$\mathbb{P}_i(\exists n \geq 1 : X_n = i) < 1$$

Consider state  $i \in S$  and suppose we start from  $X_0 = i$ .

By transience of  $\{X_n\}$ ,

$$\mathbb{P}_i(\exists m \geq 1 : X_m = i) < 1$$

Define the event

$$A = \{\exists m \geq 1 : X_m = i\}$$

and the event

$$B = \{\exists n \geq 1 : Y_n = i\} = \{\exists n \geq 1 : X_{kn} = i\}$$

Note that if  $X_{kn} = i$  for some  $n$ ,  $X_m = i$  for  $m = kn$ , i.e.  $B \subseteq A$ . Hence,

$$\mathbb{P}_i(B) \leq \mathbb{P}_i(A)$$

Since  $A$  has probability strictly less than 1,

$$\mathbb{P}_i(\exists n \geq 1 : Y_n = i) = \mathbb{P}_i(B) < 1$$

Therefore,  $i$  is transient for the chain  $\{Y_n\}$  as well. Since this applies to every state  $i \in S$  that is transient in  $\{X_n\}$ , it shows that no state  $i$  can become recurrent under the  $k$ -step sampling. The probability that  $\{Y_n\}$  returns to  $i$  is bounded above by the probability that  $\{X_n\}$  returns to  $i$ , and the latter is less than 1 for transient states.

- (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.

Let  $\{X_n\}_{n \geq 0}$  be a Markov chain on a state space  $\mathcal{S}$  with transition matrix  $P$ , and assume that  $\{X_n\}$  is recurrent. Fix state  $i \in \mathcal{S}$  that is recurrent, st

$$\mathbb{P}_i(\{n \geq 1 : X_n = i\} = \infty) = 1$$

For  $k \geq 1$ , define the chain  $\{Y_n\}_{n \geq 0}$  by

$$Y_n = X_{kn}, \quad n = 0, 1, 2, \dots$$

Since  $i$  is recurrent for  $\{X_n\}$ , define the set  $A$  as

$$A = \{n \geq 1 : X_n = i\} = \infty$$

Now, note that every positive integer belongs to one of the  $k$  residue classes modulo  $k$ , or

$$\mathbb{N} = \bigcup_{r=0}^{k-1} \{n \in \mathbb{N} : n \equiv r \pmod{k}\}$$

Thus, the return times can be partitioned as

$$A = \bigcup_{r=0}^{k-1} A_r, \quad \text{where } A_r = \{n \in A : n \equiv r \pmod{k}\}$$

Because  $A$  is countably infinite, by the pigeonhole principle at least one of the sets  $A_r$  must be countably infinite. As such, there exists an  $r_0 \in \{0, 1, \dots, k-1\}$  st

$$\mathbb{P}_i(A_{r_0} = \infty) = 1$$

Consider the two cases,

Case 1: If  $r_0 = 0$ , then infinitely many returns to  $i$  occur at times that are multiples of  $k$ . In other words,

$$A_0 = \{n \geq 1 : X_n = i \text{ and } n \equiv 0 \pmod{k}\} = \infty$$

Hence,  $i$  is visited infinitely often by the chain  $\{Y_n\}$  and is therefore recurrent.

Case 2: If  $r_0 \neq 0$ , define a time-shifted chain

$$\tilde{Y}_n = X_{kn+r_0}$$

Since  $\tilde{Y}_n$  is merely a shifted version of the  $k$ -chain, it is also a Markov chain, with the same state space and transition matrix  $P^k$ , and has the same recurrence properties, as recurrence is invariant under a finite time shift. Since  $A_{r_0}$  is infinite,

$$\{n \geq 0 : \tilde{Y}_n = i\} = \{n \geq 0 : X_{kn+r_0} = i\} = \infty$$

Therefore,  $i$  is recurrent for the chain  $\{\tilde{Y}_n\}$ . Since  $\tilde{Y}_n$  and  $Y_n$  differ only by a fixed time shift, it follows that recurrence holds for both chains.

Hence,  $i$  is recurrent for  $\{Y_n\}$  as well. In either case, if  $i$  is recurrent for  $\{X_n\}$ , then it is also recurrent for  $\{Y_n\}$ .

Since recurrence is a class property, the entire chain  $\{Y_n\}$  must be recurrent.

- (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:

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

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

So 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  $\{\pi_i\}_{i \in \mathbb{Z}}$

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

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.