

# ECMM450 Stochastic Processes

## Coursework Assignment 2

March 2023

1. Consider customers arriving independently at a bar that has two bartenders serving independently. Customers are assigned to an available bartender on a first come, first served basis but walk away if there are more than  $K > 1$  customers waiting or being served at the bar. You may assume that arrivals are a Poisson process with rate  $\lambda$  and each of the bartenders serves customers according to a Poisson process with rate  $\mu$ .

(a) The Kendall notation for the above Markov chain is M/M/2

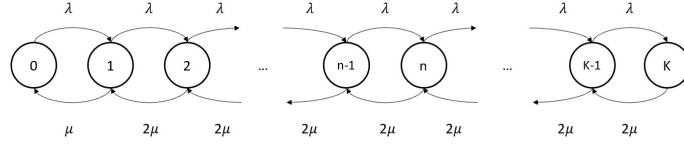


Figure 1: Diagram for Markov Chain. Numbers represent size of the system, i.e. how many people there are in the cafe

(b) Let  $\rho = \frac{\lambda}{2\mu}$ . The steady state probability equations are as follows:

$$\begin{aligned}
 \lambda P_0 &= \mu P_1 \implies P_1 = \frac{\lambda}{\mu} P_0 = 2\rho P_0 \\
 \lambda P_1 &= 2\mu P_2 \implies P_2 = \frac{\lambda}{2\mu} P_1 = 2\rho^2 P_0 \\
 &\dots \\
 \lambda P_{n-1} &= 2\mu P_n \implies P_n = \frac{\lambda}{2\mu} P_{n-1} = 2\rho^n P_0 \\
 &\dots \\
 \lambda P_{K-1} &= 2\mu P_K \implies P_K = \frac{\lambda}{2\mu} P_{K-1} = 2\rho^K P_0
 \end{aligned}$$

We know that the sum of probabilities must add up to 1.

$$\begin{aligned}
 P_0 + P_1 + \dots + P_K &= 1 \\
 \implies P_0(1 + 2\rho + 2\rho^2 + \dots + 2\rho^K) &= 1 \\
 \implies P_0(2 + 2\rho + 2\rho^2 + \dots + 2\rho^K - 1) &= 1 \\
 \implies P_0(2(1 + \rho + \rho^2 + \dots + \rho^K) - 1) &= 1 \\
 \implies P_0(2 \cdot \frac{1 - \rho^{K+1}}{1 - \rho} - 1) &= 1, \text{ using given hint} \\
 \implies P_0(2 - 2\rho^{K+1} - 1 + \rho) &= 1 - \rho \\
 \implies P_0(1 + \rho - 2\rho^{K+1}) &= 1 - \rho \\
 \implies P_0 &= \frac{1 - \rho}{1 + \rho - 2\rho^{K+1}} \\
 \implies P_n &= 2\rho^n \frac{1 - \rho}{1 + \rho - 2\rho^{K+1}}, \text{ for } n \geq 1
 \end{aligned}$$

(c) At steady state, expected number of customers at the bar is given by  $E[N]$ .

$$\begin{aligned}
E[N] &= \sum_{n=0}^K n P_n \\
E[N] &= \sum_{n=1}^K n \cdot 2\rho^n \frac{1-\rho}{1+\rho-2\rho^{K+1}} \\
E[N] &= \frac{2(1-\rho)}{1+\rho-2\rho^{K+1}} \sum_{n=0}^K n \rho^n \\
E[N] &= \frac{2(1-\rho)}{1+\rho-2\rho^{K+1}} [\rho + 2\rho^2 + 3\rho^3 + \dots + K\rho^K] \\
E[N] &= \frac{2(1-\rho)}{1+\rho-2\rho^{K+1}} [\rho \cdot \frac{d}{d\rho} (1 + \rho + \rho^2 + \rho^3 + \dots + \rho^K)] \\
E[N] &= \frac{2\rho(1-\rho)}{1+\rho-2\rho^{K+1}} \frac{d}{d\rho} \left( \frac{1-\rho^{K+1}}{1-\rho} \right), \text{ using given hint as } K \text{ is a positive integer} \\
E[N] &= \frac{2\rho(1-\rho)}{1+\rho-2\rho^{K+1}} \frac{(1-\rho)(-(K+1)\rho^K) - (1-\rho^{K+1})(-1)}{(1-\rho)^2} \\
E[N] &= \frac{2\rho(1-\rho)}{1+\rho-2\rho^{K+1}} \frac{-(K+1)\rho^K + (K+1)\rho^{K+1} + 1 - \rho^{K+1}}{(1-\rho)^2} \\
E[N] &= \frac{2\rho}{1+\rho-2\rho^{K+1}} \frac{1 - (K+1)\rho^K + K\rho^{K+1}}{(1-\rho)}
\end{aligned}$$

When  $K \rightarrow \infty$  and  $\rho < 1$ ,

$$\begin{aligned}
\lim_{K \rightarrow \infty} E[N] &= \lim_{K \rightarrow \infty} \frac{2\rho}{1+\rho-2\rho^{K+1}} \frac{1 - (K+1)\rho^K + K\rho^{K+1}}{(1-\rho)} \\
\lim_{K \rightarrow \infty} E[N] &= \frac{2\rho}{1+\rho-0} \frac{1-0+0}{(1-\rho)}
\end{aligned}$$

Note: We see terms with  $\rho^K$  or  $\rho^{K+1}$  becoming 0 as such terms drop to 0 exponentially fast thus offsetting their coefficients, which tend to infinity. Thus, we see that for  $K \rightarrow \infty$  and  $\rho < 1$ ,  $E[N] \rightarrow \frac{2\rho}{1-\rho^2}$

(d) Using Little's Law, we find the average waiting time as  $W_s = \frac{E[N]}{\lambda_{eff}}$ . To derive an expression for  $\lambda_{eff}$ , we have

$$\begin{aligned}
\lambda_{eff} &= \sum_{n=0}^K \lambda_n P_n = \sum_{n=0}^{K-1} \lambda_n P_n = \lambda(1 - P_K), \text{ (as } \lambda_K = 0) \\
\lambda_{eff} &= \lambda \left[ 1 - \frac{2\rho^K(1-\rho)}{1+\rho-2\rho^{K+1}} \right] \\
\lambda_{eff} &= \lambda \left[ \frac{1+\rho-2\rho^{K+1}-2\rho^K+2\rho^{K+1}}{1+\rho-2\rho^{K+1}} \right] \\
\lambda_{eff} &= \lambda \left[ \frac{1+\rho-2\rho^K}{1+\rho-2\rho^{K+1}} \right]
\end{aligned}$$

(e) Probability at steady state of having 2 or more customers at the bar is found as

$$\begin{aligned}
P(N \geq 2) &= P(N = 2) + P(N = 3) + \dots + P(N = k) \\
&= P_2 + P_3 + \dots + P_K \\
&= P_0(2\rho^2 + 2\rho^3 + \dots + 2\rho^K) \\
&= \frac{(1 - \rho)}{1 + \rho - 2\rho^{K+1}}(2\rho^2 + 2\rho^3 + \dots + 2\rho^K) \\
&= \frac{(1 - \rho)}{1 + \rho - 2\rho^{K+1}} \cdot 2\rho^2(1 + \rho + \dots + \rho^{K-2}) \\
&= \frac{2(1 - \rho)\rho^2}{1 + \rho - 2\rho^{K+1}} \cdot \left[ \frac{1 - \rho^{K-1}}{1 - \rho} \right] \\
&= \frac{2\rho^2}{1 + \rho - 2\rho^{K+1}} \cdot [1 - \rho^{K-1}]
\end{aligned}$$

When  $K \rightarrow \infty$ ,

$$\begin{aligned}
\lim_{K \rightarrow \infty} P(N \geq 2) &= \frac{2\rho^2}{1 + \rho} < \frac{1}{3} \\
&\implies 6\rho^2 < 1 + \rho \\
&\implies 6\rho^2 - \rho - 1 < 0
\end{aligned}$$

Solving the above, we get  $\frac{-1}{3} < \rho < \frac{1}{2}$ . However, the first part of this is non-sensical as  $\rho$  cannot be less than 0 (both  $\lambda$  and  $\mu$  are positive). Taking that into account, the possible range of values for  $\rho$  is given as  $0 < \rho < \frac{1}{2}$ .

2. Consider the following Markov chain with transition matrix:

$$\begin{pmatrix}
0 & 1 & 0 & 0 & 0 \\
1 - \alpha & 0 & \alpha & 0 & 0 \\
0 & 1 & 0 & 0 & 0 \\
\beta & 0 & 0 & 0 & 1 - \beta \\
0 & 0 & 0 & \frac{1}{2} & \frac{1}{2}
\end{pmatrix}$$

Here  $T_{ij} = P(X_n = j | X_{n-1} = i)$  for all  $i, j \in \{1, 2, 3, 4, 5\}$ , and  $\alpha$  and  $\beta$  do not depend on  $n$ .

(a) The state transition diagram is given below

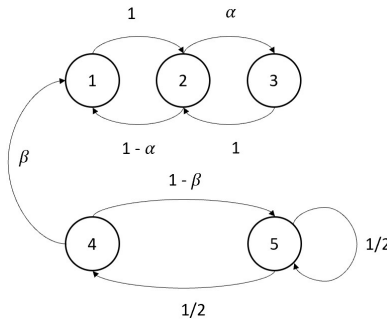


Figure 2: State transition diagram

**State 2** To prove that state 2 is positively recurrent, we need to prove that the mean recurrence

time  $\mu = \sum_{n=1}^{\infty} n f_i^{(n)} < \infty$ . First, we check the times of first return.

$$\begin{aligned} f_2^{(1)} &= 0 \\ f_2^{(2)} &= 1 - \alpha + \alpha = 1 \\ f_2^{(3)} &= 0 \\ f_2^{(4)} &= 0 \\ f_2^{(5)} &= 0 \\ &\vdots \end{aligned}$$

Note that the process cannot return to state 2 for the *first* time after more than 2 steps - by the 3rd or 4th time (and indeed any step more than that), it will already be *revisiting* state 2. Calculate the mean recurrence time  $\mu = 1 \cdot f_2^{(1)} + 2 \cdot f_2^{(2)} + 3 \cdot f_2^{(3)} + 4 \cdot f_2^{(4)} + \dots$

$$\implies \mu = 1 \cdot 0 + 2 \cdot 1 + 3 \cdot 0 + 4 \cdot 0 + \dots$$

$$\implies \mu = 2 < \infty, \text{ Therefore, state 2 is positively recurrent.}$$

**State 5** To see if state 5 is transient, we check if sum of times of first return is  $< 1$ .

$$\begin{aligned} f_5^{(1)} &= \frac{1}{2} \\ f_5^{(2)} &= \frac{1}{2}(1 - \beta) \\ f_5^{(3)} &= 0 \\ f_5^{(4)} &= 0 \\ f_5^{(5)} &= 0 \\ &\vdots \\ \sum_{n=1}^{\infty} f_5^{(n)} &= \frac{1}{2}(1 - \beta) \end{aligned}$$

It is not dependent on  $\alpha$ , but for  $0 < \beta < 1$ , the sum is lesser than 1, which means that state 5 is transient. We notice that similar to state 2, the process cannot return to state 5 for the *first* time after more than 2 steps - it will have already visited it before (on the first or second step).

(b) For fixed  $\alpha, \beta > 0$ , it can be seen that there are two sub-chains: 1-2-3 and 4-5.

Following are the classifications of the states:

**State 1:** It is periodic as for  $k = 2$ ,  $(T^n)_{11} > 0$  for all  $n = k, 2k, 3k, \dots$  and  $(T^n)_{11} = 0$  for all  $n \neq k, 2k, 3k, \dots$ . To elaborate, starting at state 1, the system can return to state 1 after 2, 4, 6, ... steps, but cannot return to it after 1, 3, 5, ... steps.

**State 2:** Following the above logic, it can be seen that state 2 is recurrent. The difference from state 1 is that, for state 1, the system can return to it after any even step size (2, 4, 6, 8, ...), whereas, for state 2, the system must return to it every 2 steps. Mean recurrence time for state 2 is therefore  $\mu = 2$ , as seen earlier in part (a).

**State 3:** Following exactly the same logic as state 1, it can be seen that state 3 is also periodic with the system returning to it after any even-numbered time step(s).

**States 4 & 5:** There is a probability of  $\beta$  of the system leaving state 4 (and thus the subchain) and never being able to return to it. So states 4 and 5 are transient.

The criteria for a subchain to be *ergodic* is for it to be both aperiodic and positively recurrent. Subchain 1-2-3 violates the condition for aperiodicity. Subchain 4-5 violates the condition of positively recurrent. Thus, neither subchain is ergodic.

(c) To find the steady state probability vector in terms of  $\alpha$  and  $\beta$ , we can solve the eigenvector equation  $\tilde{P} = \tilde{P}T$ .

$$\Rightarrow [P_1, P_2, P_3, P_4, P_5] = [P_1, P_2, P_3, P_4, P_5] \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 1-\alpha & 0 & \alpha & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ \beta & 0 & 0 & 0 & 1-\beta \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

From the above equation, the probabilities can be arranged as

$$P_1 = P_2(1 - \alpha) + P_4\beta \quad (1)$$

$$P_2 = P_1 + P_3 \quad (2)$$

$$P_3 = P_2\alpha \quad (3)$$

$$P_4 = \frac{P_5}{2} \quad (4)$$

$$P_5 = P_4(1 - \beta) + \frac{P_5}{2} \quad (5)$$

Equation 5 can be rewritten as  $P_5 = \frac{P_5}{2}(1 - \beta) + \frac{P_5}{2}$ , solving which we get  $P_5 = 0$  for  $\beta > 0$ . It follows from 4 that  $P_4 = \frac{P_5}{2} = 0$ . This also reduces 1 to  $P_1 = P_2(1 - \alpha)$ .

We can now say that  $P_1 + P_2 + P_3 = 1$  as the sum of the steady state probabilities must be 1. Solving,

$$\begin{aligned} P_1 + P_2 + P_3 &= 1 \\ P_2(1 - \alpha) + P_2 + P_2\alpha &= 1 \\ P_2(1 - \alpha + 1 + \alpha) &= 1 \\ P_2 &= \frac{1}{2} \\ \Rightarrow P_1 &= \frac{1 - \alpha}{2} \text{ \& } P_3 = \frac{\alpha}{2} \end{aligned}$$

(d) Suppose now  $\alpha := \alpha_n = \frac{n}{n+1}$ , so that now  $T \equiv T(n)$  is a time dependent transition matrix. This amends the state transition diagram to

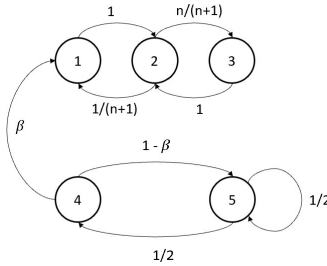


Figure 3: State transition diagram for time-dependent Markov chain

To find the classification of state 1, we need to compute  $\sum_{n=1}^{\infty} f_1^{(n)}$  and  $\mu = \sum_{n=1}^{\infty} n f_1^{(n)} < \infty$ .

The times of first return are

$$\begin{aligned}
f_1^{(1)} &= 0 \\
f_1^{(2)} &= \frac{1}{n_2 + 1} = \frac{1}{3} \\
f_1^{(3)} &= 0 \\
f_1^{(4)} &= \frac{n_2}{n_2 + 1} \frac{1}{n_4 + 1} = \frac{2}{3} \cdot \frac{1}{5} \\
f_1^{(5)} &= 0 \\
f_1^{(6)} &= \frac{n_2}{n_2 + 1} \frac{n_4}{n_4 + 1} \frac{n_6}{n_6 + 1} = \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{1}{7} \\
f_1^{(7)} &= 0 \\
&\vdots
\end{aligned}$$

Unsurprisingly, the probability of every odd-step return is 0.

$$\begin{aligned}
\sum_{n=1}^{\infty} f_5^{(n)} &= \frac{1}{3} + \left(\frac{2}{3} \cdot \frac{1}{5}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{1}{7}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \frac{1}{9}\right) + \dots \\
\sum_{n=1}^{\infty} f_5^{(n)} &= \frac{1}{3} + \left(\frac{2}{3} \cdot \left(1 - \frac{4}{5}\right)\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \left(1 - \frac{6}{7}\right)\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \left(1 - \frac{8}{9}\right)\right) + \dots \\
\sum_{n=1}^{\infty} f_5^{(n)} &= \frac{1}{3} + \left(\frac{2}{3} - \frac{2}{3} \cdot \frac{4}{5}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} - \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} - \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \frac{8}{9}\right) + \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \frac{8}{9} + \dots
\end{aligned}$$

After the second term, every subsequent pair of terms cancels each other out, leaving behind  $\frac{1}{3} + \frac{2}{3} = 1$ . Therefore, state 1 is recurrent. To further investigate if it is positively recurrent or null recurrent, we compute mean recurrence time  $\mu$ . Recall that every odd time-step probability of first return is 0, so we will only consider the even terms.

$$\begin{aligned}
\mu &= \sum_{n=1}^{\infty} n f_1^{(n)} = 2 \cdot \frac{1}{3} + 4 \cdot \left(\frac{2}{3} \cdot \frac{1}{5}\right) + 6 \cdot \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{1}{7}\right) + 8 \cdot \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \frac{1}{9}\right) + \dots \\
\mu &= \sum_{n=1}^{\infty} n f_1^{(n)} = \frac{2}{3} + \left(\frac{4}{3} \cdot \frac{1}{5}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7}\right) + \left(\frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdot \frac{8}{9}\right) + \dots
\end{aligned}$$

The  $n$ th term can be represented as  $t_n = \prod_{i=1}^n (1 - a_i)$ , where  $a_n = (2n + 1)^{-1}$

As a series, we can rewrite  $t_n$

$$\begin{aligned}
t_n &= \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdots (1 - a_n) = \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdots \left(1 - \frac{1}{2n + 1}\right) = \frac{2}{3} \cdot \frac{4}{5} \cdot \frac{6}{7} \cdots \frac{2n}{2n + 1} \\
\Rightarrow t_n &= 2^n \frac{1}{3} \cdot \frac{2}{5} \cdot \frac{3}{7} \cdots \frac{n}{2n + 1} = \frac{2^n n!}{3 \cdot 5 \cdot 7 \cdots (2n + 1)} = \frac{2^n n! \cdot 2 \cdot 4 \cdot 6 \cdots 2n}{1 \cdot 2 \cdot 3 \cdot 4 \cdot 5 \cdot 6 \cdot 7 \cdots 2n \cdot (2n + 1)} \\
\Rightarrow t_n &= \frac{2^n n! \cdot 2^n (1 \cdot 2 \cdot 3 \cdot 4 \cdots n)}{1 \cdot 2 \cdot 3 \cdot 4 \cdot 5 \cdot 6 \cdot 7 \cdots 2n \cdot (2n + 1)} = \frac{2^n n! \cdot 2^n n!}{(2n + 1)!} = \frac{(2^n n!)^2}{(2n + 1)!}
\end{aligned}$$

Expanding  $t_n$ , we can write

$$\begin{aligned}
t_n &= \frac{(2^n n!)^2}{(2n+1)!} = 2^{2n} \frac{(n!)^2}{(2n+1)!} = 4^n \frac{(n!)^2}{(2n+1)!} \\
\Rightarrow t_n &= 4^n \frac{n^2}{(2n+1)(2n)} \cdot \frac{(n-1)^2}{(2n-1)(2n-2)} \cdot \frac{(n-2)^2}{(2n-3)(2n-4)} \cdots \frac{2^2}{5 \cdot 4} \cdot \frac{1^2}{3 \cdot 2} \\
\Rightarrow t_n &= 4^n \frac{n^2}{(2n+1)(2n)} \cdot \frac{(n-1)^2}{(2n-1)(2n-1)} \cdot \frac{(n-2)^2}{(2n-3)(2n-2)} \cdots \frac{2^2}{5 \cdot 2 \cdot 2} \cdot \frac{1^2}{3 \cdot 2 \cdot 1} \\
\Rightarrow t_n &= 4^n \frac{n}{(2n+1)(2)} \cdot \frac{(n-1)}{(2n-1)(2)} \cdot \frac{(n-2)}{(2n-3)(2)} \cdots \frac{2}{5 \cdot 2} \cdot \frac{1}{3 \cdot 2} \\
\Rightarrow t_n &= 4^n \frac{n}{(2(n+\frac{1}{2}))(2)} \cdot \frac{(n-1)}{(2(n-\frac{1}{2}))(2)} \cdot \frac{(n-2)}{(2(n-\frac{3}{2}))(2)} \cdots \frac{2}{2 \cdot \frac{5}{2} \cdot 2} \cdot \frac{1}{2 \cdot \frac{3}{2} \cdot 2} \\
\Rightarrow t_n &= \frac{4^n}{4^n} \frac{n}{(n+\frac{1}{2})} \cdot \frac{(n-1)}{(n-\frac{1}{2})} \cdot \frac{(n-2)}{(n-\frac{3}{2})} \cdots \frac{2}{\frac{5}{2}} \cdot \frac{1}{\frac{3}{2}} \\
\Rightarrow t_n &= \frac{n}{(n+\frac{1}{2})} \cdot \frac{(n-1)}{(n-\frac{1}{2})} \cdot \frac{(n-2)}{(n-\frac{3}{2})} \cdots \frac{2}{\frac{5}{2}} \cdot \frac{1}{\frac{3}{2}}
\end{aligned}$$

Subtracting  $\frac{1}{2}$  from all the numerators inevitably leaves us with a term smaller than  $t_n$ . Therefore,

$$\begin{aligned}
t_n &\geq \frac{n-\frac{1}{2}}{(n+\frac{1}{2})} \cdot \frac{(n-\frac{3}{2})}{(n-\frac{1}{2})} \cdot \frac{(n-\frac{5}{2})}{(n-\frac{3}{2})} \cdots \frac{\frac{3}{2}}{\frac{5}{2}} \cdot \frac{\frac{1}{2}}{\frac{3}{2}} \\
\Rightarrow t_n &\geq \frac{\frac{1}{2}}{n+\frac{1}{2}}
\end{aligned}$$

The last line follows from alternate numerators and denominators cancelling out. We now have  $t_n \geq \frac{1}{2n+1}$ , which can be compared against the harmonic series to showcase divergence. By the limit comparison test, for two given series  $\sum a_n$  and  $\sum b_n$ , if  $\lim_{n \rightarrow \infty} \frac{a_n}{b_n} = c$  for  $0 < c < \infty$ , where  $a_n, b_n$  are respectively the  $n$ th terms of the series, then either both series converge or both series diverge.

We have  $a_n = \frac{1}{2n+1}$ ,  $b_n = \frac{1}{n}$ , the latter denoting the harmonic series.

$$\lim_{n \rightarrow \infty} \frac{a_n}{b_n} = \lim_{n \rightarrow \infty} \frac{\frac{1}{2n+1}}{\frac{1}{n}} = \lim_{n \rightarrow \infty} \frac{n}{2n+1} = \frac{1}{2}, \text{ L'Hospital rule}$$

As we know that the harmonic series diverges, the series with  $n$ th term  $a_n = \frac{1}{2n+1}$  also diverges. Therefore the series with  $n$ th term  $t_n$ , which is greater than  $a_n$ , must also diverge. Hence, the mean recurrence time  $\mu$  diverges. This proves that the state 1 is null recurrent.