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This problem asks us to find the **mean return time** to a specific state (State 1) in a given Markov chain. This concept is a part of the study of **recurrence** in Markov chains. The mean return time to a state i, denoted as r_i , is the expected number of steps required to return to state i, given that the process started in state i.

Problem Description:

A Markov chain has states $S=\{0,1,2\}$ with the following transition matrix P:

$$P = egin{pmatrix} 0.8 & 0.1 & 0.1 \ 0.3 & 0.5 & 0.2 \ 0.2 & 0.4 & 0.4 \end{pmatrix}$$

We need to find the mean return time to state 1, r_1 , and round the answer to the nearest integer.

Key Concept: Mean Return Time and Stationary Distribution

For a recurrent state i in an irreducible Markov chain, the mean return time r_i is inversely related to its steady-state probability π_i :

$$r_i = rac{1}{\pi_i}$$

where π_i is the i-th component of the steady-state probability vector $\pi=(\pi_0,\pi_1,\pi_2)$

Steps to Solve:

1. Find the Steady-State Probability Vector $\pi=(\pi_0,\pi_1,\pi_2)$:

The steady-state probability vector satisfies two conditions:

•
$$\pi P=\pi$$
 (or $\pi(I-P)=0$)

•
$$\sum_i \pi_i = 1$$

Let's write out the equations from $\pi P=\pi$:

$$(\pi_0,\pi_1,\pi_2) egin{pmatrix} 0.8 & 0.1 & 0.1 \ 0.3 & 0.5 & 0.2 \ 0.2 & 0.4 & 0.4 \end{pmatrix} = (\pi_0,\pi_1,\pi_2)$$

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This gives us the following system of linear equations:

•
$$\pi_0(0.8) + \pi_1(0.3) + \pi_2(0.2) = \pi_0$$

•
$$\pi_0(0.1) + \pi_1(0.5) + \pi_2(0.4) = \pi_1$$

•
$$\pi_0(0.1) + \pi_1(0.2) + \pi_2(0.4) = \pi_2$$

And the sum constraint:

•
$$\pi_0 + \pi_1 + \pi_2 = 1$$

Let's simplify the first two equations:

$$1.0.8\pi_0 + 0.3\pi_1 + 0.2\pi_2 = \pi_0 \implies -0.2\pi_0 + 0.3\pi_1 + 0.2\pi_2 = 0$$

2.
$$0.1\pi_0 + 0.5\pi_1 + 0.4\pi_2 = \pi_1 \implies 0.1\pi_0 - 0.5\pi_1 + 0.4\pi_2 = 0$$

From equation (1), multiply by 10: $-2\pi_0+3\pi_1+2\pi_2=0$ (Eq. A) From equation (2), multiply by 10: $1\pi_0-5\pi_1+4\pi_2=0$ (Eq. B)

From (Eq. B), we can express π_0 in terms of π_1 and π_2 :

$$\pi_0 = 5\pi_1 - 4\pi_2$$

Substitute this into (Eq. A):

$$-2(5\pi_1-4\pi_2)+3\pi_1+2\pi_2=0$$

$$-10\pi_1 + 8\pi_2 + 3\pi_1 + 2\pi_2 = 0$$

$$-7\pi_1 + 10\pi_2 = 0$$

So,
$$10\pi_2=7\pi_1 \implies \pi_2=rac{7}{10}\pi_1$$

Now, substitute π_2 back into the expression for π_0 :

$$\begin{array}{l} \pi_0 = 5\pi_1 - 4\left(\frac{7}{10}\pi_1\right) = 5\pi_1 - \frac{28}{10}\pi_1 = 5\pi_1 - \frac{14}{5}\pi_1 \\ \pi_0 = \frac{25}{5}\pi_1 - \frac{14}{5}\pi_1 = \frac{11}{5}\pi_1 \end{array}$$

Finally, use the normalization condition $\pi_0+\pi_1+\pi_2=1$:

$$\frac{11}{5}\pi_1 + \pi_1 + \frac{7}{10}\pi_1 = 1$$

To sum these fractions, use a common denominator of 10:

$$egin{array}{l} rac{22}{10}\pi_1 + rac{10}{10}\pi_1 + rac{7}{10}\pi_1 = 1 \ rac{(22+10+7)}{10}\pi_1 = 1 \ rac{39}{10}\pi_1 = 1 \ \pi_1 = rac{10}{20} \end{array}$$

Now find π_0 and π_2 :

$$\pi_0 = \frac{11}{5}\pi_1 = \frac{11}{5} \cdot \frac{10}{39} = \frac{11 \cdot 2}{39} = \frac{22}{39}$$
$$\pi_2 = \frac{7}{10}\pi_1 = \frac{7}{10} \cdot \frac{10}{39} = \frac{7}{39}$$

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So, the steady-state probabilities are:

$$\pi_0=rac{22}{39}, \pi_1=rac{10}{39}, \pi_2=rac{7}{39}.$$
 (Check: $22/39+10/39+7/39=39/39=1.$ This is correct.)

2. Calculate the Mean Return Time to State 1 (r_1):

Using the formula
$$r_i=rac{1}{\pi_i}$$
: $r_1=rac{1}{\pi_1}=rac{1}{10/39}=rac{39}{10}=3.9$

3. Round to the nearest integer:

Rounding 3.9 to the nearest integer gives **4**.

Topics Covered:

- Markov Chains: A mathematical model describing a sequence of possible events
 where the probability of each event depends only on the state attained in the
 previous event.
- **Transition Matrix:** A matrix whose elements represent the probabilities of transitioning from one state to another.
- Steady-State Probabilities (Stationary Distribution): The long-run probabilities of being in each state, which are independent of the initial state for ergodic chains.

 This is a crucial concept for understanding long-term behavior.
- System of Linear Equations: The method used to solve for the unknown steadystate probabilities.
- **Mean Return Time:** The expected number of steps to return to a given state. It's an important measure of how often a system visits a particular state in the long run.

WolframAlpha/Computational Check:

You can use WolframAlpha to verify the steady-state probabilities and the mean return time.

1. Find the steady-state probabilities of the given matrix:

Input: steady state probabilities of
$$\{\{0.8, 0.1, 0.1\}, \{0.3, 0.5, 0.2\}, \{0.2, 0.4, 0.4\}\}$$

WolframAlpha will output:

$$\left\{\frac{22}{39}, \frac{10}{39}, \frac{7}{39}\right\}$$

This confirms our calculated π_0, π_1, π_2 .

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2. Calculate the inverse of π_1 :

Input: 1 / (10/39)

WolframAlpha will give 39/10 or 3.9.

3. Round to the nearest integer:

round(3.9) will give 4.

These checks confirm the accuracy of our calculations.

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