

Sheet(6)

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Problem: 1

a) Following scenario,

Given the matrix P :

$$P = \begin{pmatrix} \frac{2}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{4}{5} \end{pmatrix}$$

Since we assume today is sunny (day 0), the initial vector is:

$$V_0 = [1 \ 0]$$

We can get Day 1 by multiplying V_0 by P ,

$$V_1 = V_0 \cdot P = [1, 0] \cdot \begin{pmatrix} \frac{2}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{4}{5} \end{pmatrix} = \begin{bmatrix} \frac{2}{3} & \frac{1}{5} \end{bmatrix}$$

Likewise, for Day 2,

$$V_2 = V_1 \cdot P = \begin{bmatrix} \frac{2}{3} & \frac{1}{5} \end{bmatrix} \cdot \begin{pmatrix} \frac{2}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{4}{5} \end{pmatrix}$$

Entries of V_2 ,

$$\begin{aligned} V_2 &= \begin{bmatrix} \frac{2}{3} \cdot \frac{2}{3} + \frac{1}{5} \cdot \frac{1}{3} & \frac{2}{3} \cdot \frac{1}{5} + \frac{1}{5} \cdot \frac{4}{5} \end{bmatrix} \\ &= \begin{bmatrix} \frac{4}{9} + \frac{1}{15} & \frac{2}{15} + \frac{4}{25} \end{bmatrix} \end{aligned}$$

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By converting a common ^{Deno} denominator,

$$V_2 = \left[\frac{20}{45} + \frac{3}{45} \quad \frac{10}{75} + \frac{12}{75} \right] \quad \text{(Denominator 45, 75)}$$

$$= \left[\frac{23}{45} \quad \frac{22}{75} \right]$$

So, here, the probability that it will be raining the day after tomorrow is the second element of V_2 .

Hence, the probability of raining day on the day after tomorrow is $\frac{22}{75}$.

b) To find the relative frequency, we need to find stationary distribution of Markov Chain.

Given that,

$$P = \begin{pmatrix} \frac{2}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{4}{5} \end{pmatrix}$$

we know,

$$\pi = [\pi_S \quad \pi_R]$$

and it satisfies,

$$\pi P = \pi, \quad \text{Normalized condition } [\pi_S + \pi_R = 1]$$

So, From $\pi P = \pi$,

$$[\pi_S \quad \pi_R] \begin{pmatrix} \frac{2}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{4}{5} \end{pmatrix} = [\pi_S \quad \pi_R]$$

From this we found following system of linear equation,

$$\pi_S \cdot \frac{2}{3} + \pi_R \cdot \frac{1}{3} = \pi_S \quad \text{--- (i)}$$

$$\pi_S \cdot \frac{1}{5} + \pi_R \cdot \frac{4}{5} = \pi_R$$

$$\pi_S + \pi_R = 1$$

Now, from (i),

$$\pi_S \cdot \frac{2}{3} + \pi_R \cdot \frac{1}{3} = \pi_S$$

$$\Rightarrow \pi_R \cdot \frac{1}{3} = \pi_S \left(1 - \frac{2}{3}\right) \quad \left[\text{rearrange to isolate } \pi_R \right]$$

$$\pi_R \cdot \frac{1}{3} = \pi_S \cdot \frac{1}{3}$$

$$\pi_R = \pi_S$$

Normalize condition is

$$\pi_S + \pi_R = 1$$

$$\pi_S + \pi_S = 1 \quad \left[\pi_R = \pi_S \right]$$

$$2\pi_S = 1$$

$$\pi_S = \frac{1}{2}$$

Since, $\pi_R = \pi_S$

so, the stationary distribution $\pi = \left[\frac{1}{2} \quad \frac{1}{2} \right]$

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This means that over an infinitely long period, the relative frequency of sunny day is $\pi_s = 1/2$

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c) To find the invariant measure for the given Markov chain, we need to solve,

$$\pi = (x, y), \quad \pi P = \pi$$

Where P is,

$$P = \begin{pmatrix} 2/3 & 1/5 \\ 1/3 & 4/5 \end{pmatrix}$$

This gives the following equation,

$$2/3 x + 1/3 y = x \quad \text{--- (I)}$$

$$1/5 x + 4/5 y = y \quad \text{--- (II)}$$

From (I) equation,

$$2/3 x + 1/3 y = x$$

By simplify the it,

$$\Rightarrow 1/3 x = 1/3 y$$

$$\Rightarrow x = y$$

Now, Substituting $(x=y)$ into equation (II),

$$1/5 x + 4/5 y = x$$

$$\Rightarrow 5/5 x = x$$

$$\Rightarrow \cancel{x} x = 1$$

So, the invariant measure for the given Markov chain is $\pi = (1, 1)$

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Since, $\pi = (1,1)$, now, to determine if the chain converges to its invariant measure, we need to verify the conditions of irreducibility, aperiodicity and finiteness.

- 1) irreducibility: $\pi(1,1)$ is possible to transition between sunny and rainy, so, it meets the condition.
- 2) Aperiodicity: Along with, $\pi(1,1)$, it's possible to transition from sunny to rainy, it's also meets a vice versa transition. so its Aperiodicity
- 3) Finiteness: The chain is finite since there are only two possible states. (Sunny, Rainy)

~~Since~~, hence, it converge to its invariant measure
 $\pi = (1,1)$.

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Exercise 6

Lamia Islam

Problem 2

Let P_n be the probability distribution of X_n (in column vector). Then we have

$$P_{n+1} = P P_n$$

The Markov chain has a unique invariant measure

$$\left[\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4} \right]^T$$

① If $x_0 = 1$,

$$P_0 = [1, 0, 0, 0]^T$$

$$\therefore P_1 = P P_0$$

$$= \begin{pmatrix} 0 & 1/3 & 0 & 2/3 \\ 1/2 & 0 & 1/2 & 0 \\ 0 & 2/3 & 0 & 1/3 \\ 1/2 & 0 & 1/2 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1/2 \\ 0 \\ 1/2 \end{pmatrix}$$

$$\text{and } P_2 = P P_1 = \begin{pmatrix} 0 & 1/3 & 0 & 2/3 \\ 1/2 & 0 & 1/2 & 0 \\ 0 & 2/3 & 0 & 1/3 \\ 1/2 & 0 & 1/2 & 0 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 1/2 \\ 0 \\ 1/2 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 0 \\ 1/2 \\ 0 \end{pmatrix}$$

\therefore Probability Mass function of $X_2 = [1/2, 0, 1/2, 0]^T$

② It does not converge to any probability distribution.

It is straight forward that for all $n \geq 1$,

$$P_n = \begin{cases} [1/2, 0, 1/2, 0]^T & \text{if } n \text{ is even} \\ [0, 1/2, 0, 1/2]^T & \text{if } n \text{ is odd} \end{cases}$$

This shows an alternating pattern between the states, which does not settle down to a fixed distribution.

Since the probability distribution alternates between two vectors depending on whether n is even or odd, the Markov chain does not converge to the invariant measure.

The sum of probabilities for state 1 and 3 remains equal to the sum for states 2 and 4 at each step.

Therefore, since $P_0(1) + P_0(3) = P_0(2) + P_0(4)$, P_n will never converge.

Exercise 6

SANTAY RAJPUROHIT

Problem 3

a.) Given $X_0 = x_0$ we have,

$$X_1 = x_0(1 - \Delta t) + \sqrt{2\Delta t}N(0,1) \Rightarrow X_1 \sim N(x_0(1 - \Delta t), 2\Delta t)$$

Since a linear transformation of a normal r.v. is still a normal r.v.

$$b.) X_2 = X_1(1 - \Delta t) + \sqrt{2\Delta t}N(0,1) \Rightarrow X_2 \sim N(x_0(1 - \Delta t)^2, ((1 - \Delta t)^2 + 1)\Delta t)$$

Since the sum of two independent normal r.v. is still a normal r.v.

~~Problem~~

b.) Code added to pdf.

c.) We see that the distribution, except for the time $= 0$, where $X_0 = x_0$ a.s. at time $t \in [0.5, 1, 10]$, given $X_0 = x_0$ are all normal and they seem to converge to a standard normal distribution.

→ This makes sense since X_t has mean $x_0(1 - \Delta t)^t$ that if $t \rightarrow \infty$ goes to 0 and is normally distributed since the sum of independent normal r.v. is still a normal variable.

Therefore the distribution over time converge to a normal distribution with mean as 0 and variance 1, since most support is seen inside the values of -2.75 and 2.75.