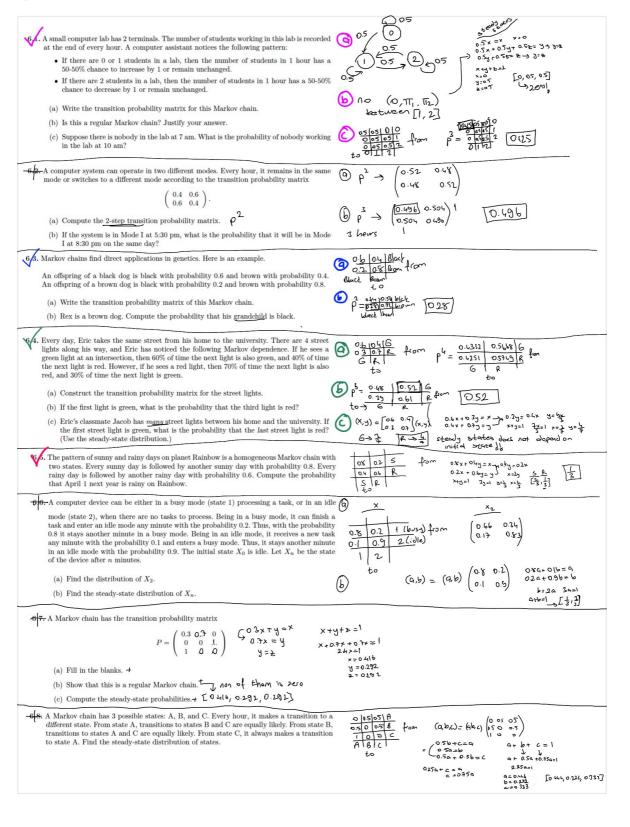
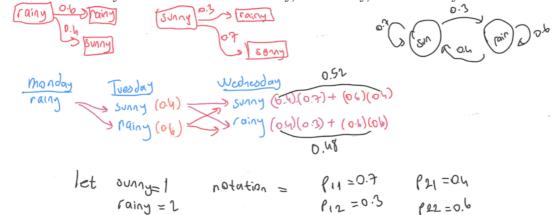
5 - markov chain



Example 6.7 (Weather forecasts). In some town, each day is either sunny or rainy. A sunny day is followed by another sunny day with probability 0.7, whereas a rainy day is followed by a sunny day with probability 0.4.

It rains on Monday. Make forecasts for Tuesday, Wednesday, and Thursday.



Example 6.8 (Weather, continued). Suppose now that it does not rain yet, but meteorologists predict an 80% chance of rain on Monday. How does this affect our forecasts?

matrix approach
$$\rho = 060.4 R \text{ from } \rho^{2} = 0.480.52 R$$
0.39 061 S
$$R S$$
1.4 monday is $R \rightarrow toology$, wednesday $R \rightarrow 0.48$

$$R \rightarrow 0.48$$

$$R \rightarrow 0.52$$

$$R \rightarrow 0.52$$
Moday $R \rightarrow 0.8$

$$S \rightarrow 0.2$$

$$Luesday = \left[0.48 + 0.06\right] \cdot 0.92 + 0.14$$

$$0.54 \cdot 0.94$$

$$0.54 \cdot 0.94$$

$$P = \begin{pmatrix} p_{11} + p_{12} + \cdots + p_{1n} & 1 \\ p_{21} + p_{22} + \cdots + p_{2n} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{pmatrix} \begin{pmatrix} p_{11} & p_{12} + \cdots + p_{1n} & 1 \\ p_{21} & p_{22} + \cdots + p_{2n} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{pmatrix}$$
To state: 1 2 \cdots n

Example 6.9 (Shared device). A computer is shared by 2 users who send tasks to a computer remotely and work independently. At any minute, any connected user may disconnect with probability 0.5, and any disconnected user may connect with a new task with probability 0.2. Let X(t) be the number of concurrent users at time t (minutes). This is a Markov chain with 3 states: 0, 1, and 2.

Us if 2 useres are connected at $\rho_0 \rightarrow (0,0,1)$ Us probability of 1 connected user at $t=2 \rightarrow \rho_0=(\frac{1}{3},\frac{1}{3},\frac{1}{3}). \ \rho \rightarrow (\frac{1}{3},\frac{1}{5},\frac{1}{5})$ steady state

computed
$$\pi P = \pi$$

$$\rho = \begin{bmatrix} 0.7 & 0.3 \\ 0.h & 0.6 \end{bmatrix}$$

$$(\pi_1, \pi_2) = (\pi_1, \pi_1) \begin{pmatrix} 0.7 & 0.5 \\ 0.4 & 0.6 \end{pmatrix}$$

$$\rho = \begin{cases} 0.7 & 0.3 \\ 0.4 & 0.6 \end{cases}$$

$$(\pi_1, \pi_2) = (\pi_1, \pi_2) \begin{pmatrix} 0.7 & 0.5 \\ 0.4 & 0.6 \end{pmatrix}$$

$$(\pi_1, \pi_2) = (\pi_1, \pi_2) \begin{pmatrix} 0.7 & 0.5 \\ 0.4 & 0.6 \end{pmatrix}$$

$$0.3 \pi_1 + 0.6 \pi_2 = \pi_2$$

$$0.3 \pi_1 = 0.4 \pi_1$$

$$\pi_1 + \frac{1}{4} \widehat{\Pi}_1 = \frac{1}{4} \approx 0.57 \ \text{m} \ (2) \ \Pi_1 + \Pi_2 = 1$$
 $\Pi_2 = \frac{1}{4} \approx 0.43$

Given the transition probability matrix P for a three-state Markov chain, in which the set of states is {1, 2, 3} and 0.0x+0.4y+ 0.8=x

$$P = \begin{bmatrix} 0.15 & 0.60 & 0.25 \\ 0.40 & 0.40 & 0.20 \\ 0.80 & 0.10 & 0.10 \end{bmatrix}$$

compute the steady-state probabilities.

$$\pi_1 = \boxed{0.334}$$

$$\pi_1 = 0.334$$
 \mathbf{x}
 $\pi_2 = 0.334$ \mathbf{x}

 $\pi_3 = 0.334$

Write your answers by providing up to 4 digits after the decimal point.

2=0.199261

DEFINITION 6.9 -

A Markov chain is regular if

for some h and all i, j. That is, for some h, matrix $P^{(h)}$ has only non-zero entries, and h-step transitions from any state to any state are possible.

Any regular Markov chain has a steady-state distribution.

Example 6.15. A Markov chain with transition probability matrix

$$P = \left(\begin{array}{cccc} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0.9 & 0 & 0 & 0.1 \end{array}\right).$$

is also regular. Matrix P contains zeros, and so do P^2 , P^3 , P^4 , and P^5 . The 6-step transition probability matrix

$$P^{(6)} = \begin{pmatrix} .009 & .090 & .900 & .001 \\ .001 & .009 & .090 & .900 \\ .810 & .001 & .009 & .180 \\ .162 & .810 & .001 & .027 \end{pmatrix}$$

contains no zeros and proves regularity of this Markov chain.

also be seen from the transition diagram in Figure 6.4. Based on this diagram, any state j can be reached in 6 steps from any state i. Indeed, moving counterclockwise through this

Example 6.16 (IRREGULAR MARKOV CHAIN, ABSORBING STATES). When there is a state i with $p_{ii} = 1$, a Markov chain cannot be regular. There is no exit from state i; therefore,

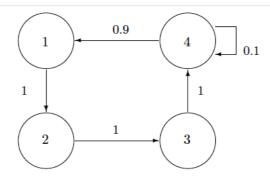


FIGURE 6.4: Transition diagram for a regular Markov chain in Example 6.15.

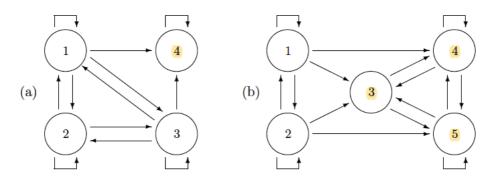


FIGURE 6.5: Absorbing states and absorbing zones (Example 6.16).

Notice that both Markov chains do have steady-state distributions. The first process will eventually reach state 4 and will stay there for good. Therefore, the limiting distribution of X(h) is $\pi = \lim P_h = (0,0,0,1)$. The second Markov chain will eventually leave states 1 and 2 for good, thus its limiting (steady-state) distribution has the form $\pi = (0,0,\pi_3,\pi_4,\pi_5)$. \Diamond