

CS2 107 12-4-25

- Final Exam: Mon. 12/8 12:00-2:00 PM
- Lab 6: Th 12/4 10:00 PM
- SETS: Sun 12/7 11:59 PM

Defn

A state $i \in S$ is called Recurrent if for every j accessible from i , we also have i accessible from j .

Equivalently

$$j \in A(i) \text{ iff } i \in A(j)$$

i.e.

$$(i) \rightarrow \dots (j) \text{ iff } (j) \rightarrow \dots (i)$$

Thus if we depart a recurrent state, there is a Positive Probability that we return, and so after a long enough time, we will return.

Hence a recurrent state, if it is visited at all, will be visited infinitely often.

Defn

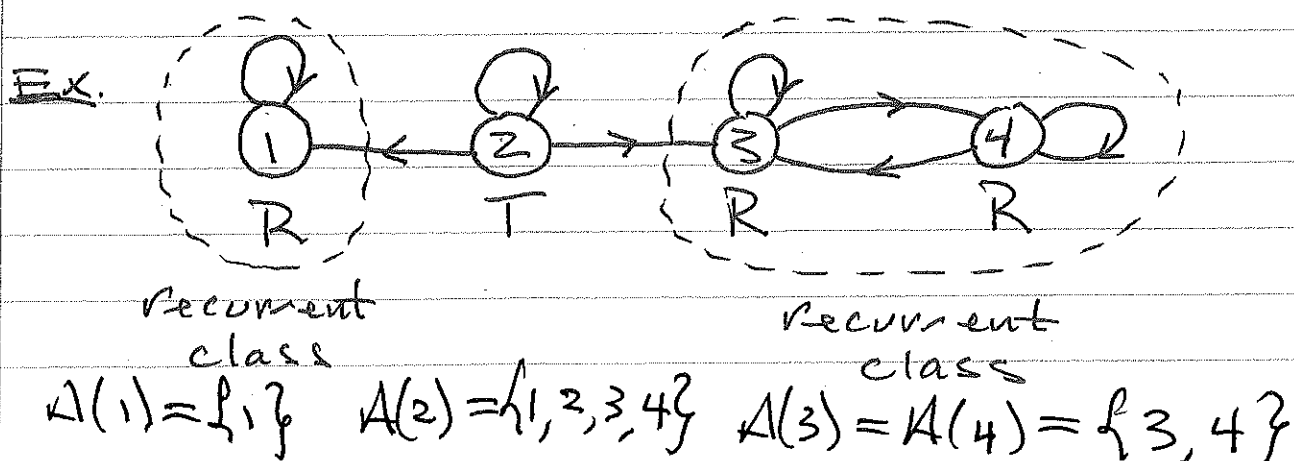
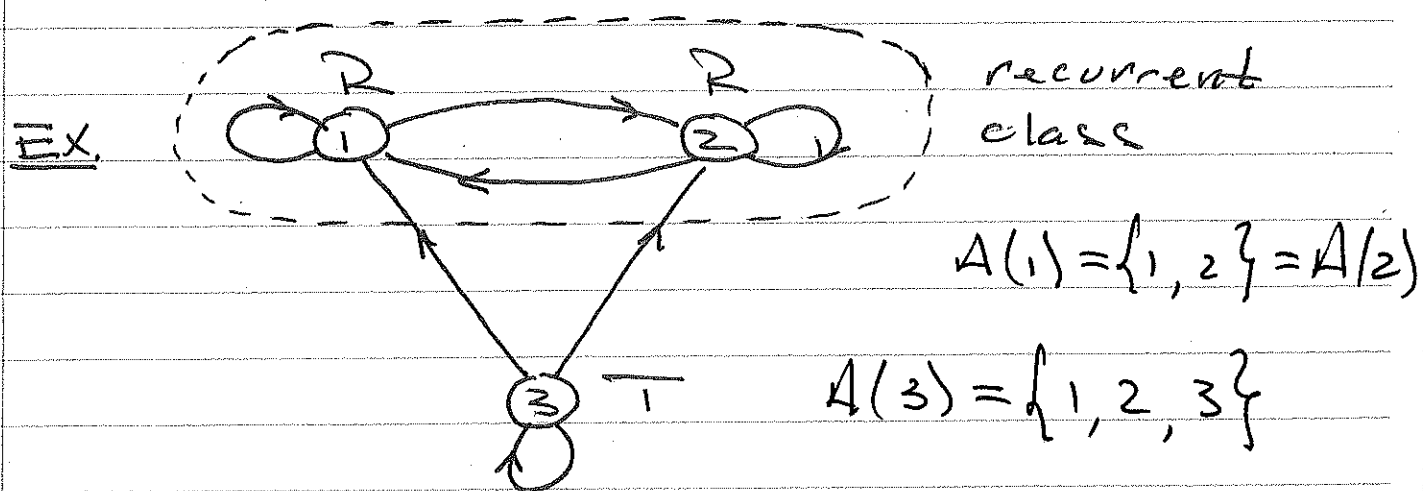
A state i is called transient iff it is not recurrent. Thus i is transient iff there exists a state j with

$$j \in A(i) \text{ and } i \notin A(j)$$

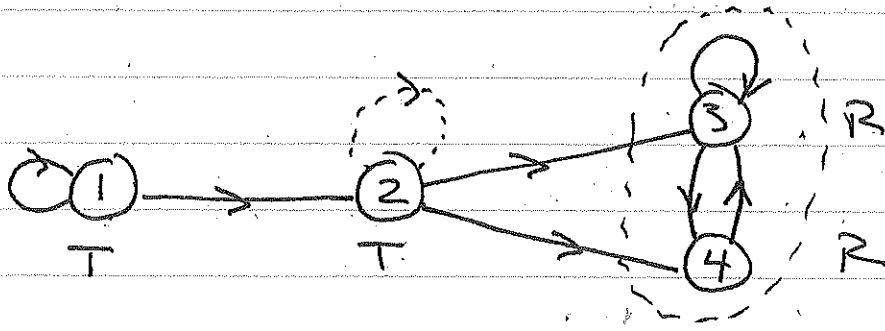
Thus after a visit to a transient state i , there is a positive probability of entering such a state i . Given enough time, this is certain to happen.

Hence a transient state will only be visited a finite number of times.

Note: Transience and recurrence are determined by the arcs in a state transition diagram, not by the actual probabilities P_{ij} on each arc.



note: is i always accessible from i ,
i.e. is $i \in A(i)$? not necessarily



2 is not accessible from 2 unless the self loop $\curvearrowright$ exists, i.e. has Positive Probability. observe 4 is accessible from 4 with or without a self loop.

$$A(2) = \{3, 4\}$$

$$A(1) = \{1, 2, 3, 4\}$$

Note that for any states i, j we have

$$j \in A(i) \Rightarrow A(j) \subseteq A(i)$$

and

$$i \in A(j) \Rightarrow A(i) \subseteq A(j)$$

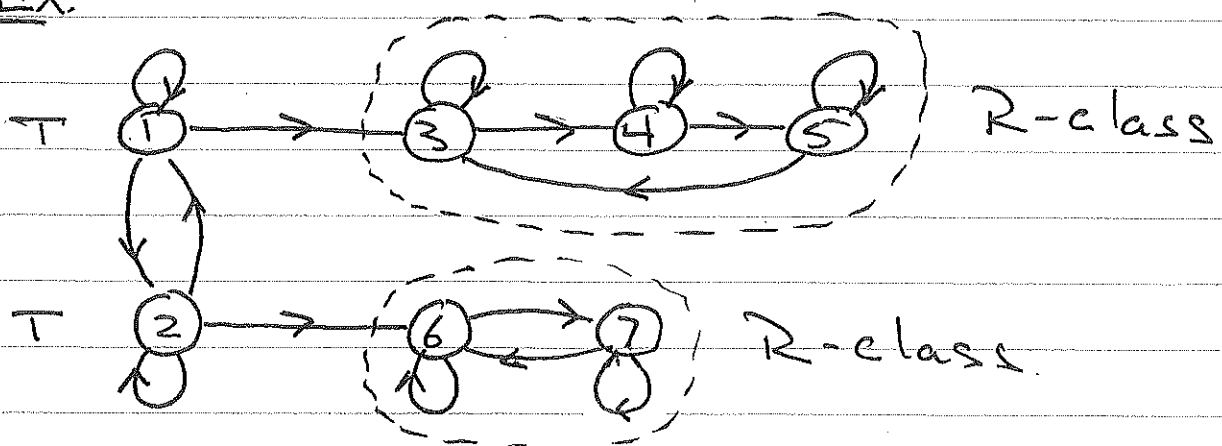
Therefore if i is recurrent, then

$$\left. \begin{array}{l} i \in A(i) \Rightarrow A(i) \subseteq A(i) \\ \updownarrow \\ i \in A(i) \Rightarrow A(i) \subseteq A(i) \end{array} \right\} \Rightarrow A(i) = A(i)$$

Defn

If i is a recurrent state, the set $A(i) \subseteq S$ is called a recurrent class.

EX.



Problem

Exercise (see #6 on p. 381)

Let i be a transient state. Prove there exists a recurrent state j such that

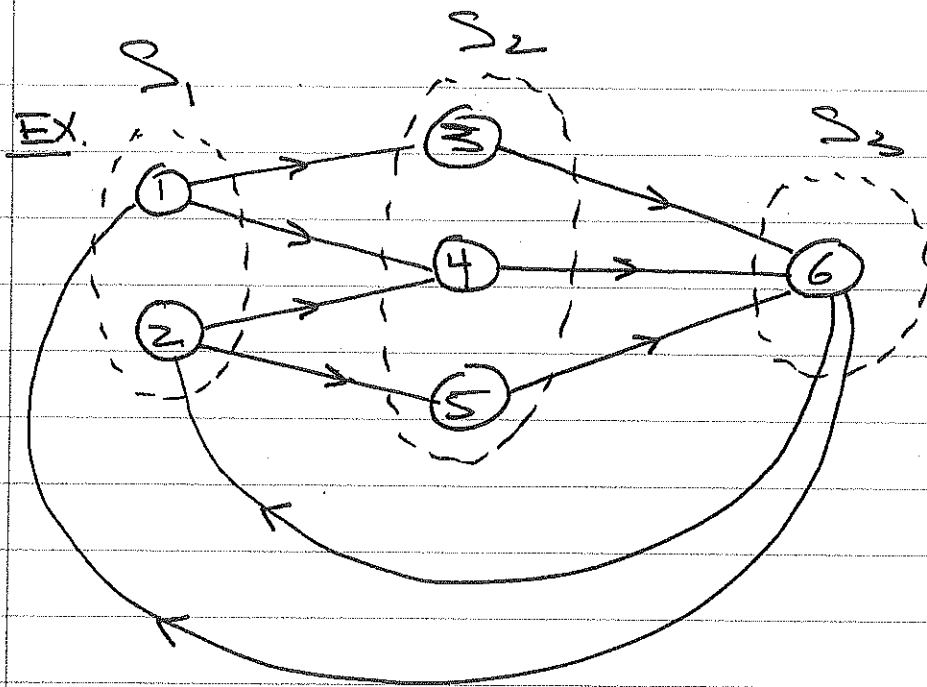
$$T \text{ (i) } \dashrightarrow \text{ (j) } R$$

i.e.

$$j \in A(i)$$

Theorem: Markov Chain Decomposition

- A Markov chain can be decomposed into one or more recurrent classes, and possibly some transient states.
- A recurrent state is accessible from all states in its class, but not from recurrent states in other classes.
- A transient state is not accessible from any recurrent state.
- At least one recurrent state is accessible from a given transient state.



Period = 3

observe that in this example, if $X_0 \in S_1$, then

$$X_n \in S_1 \text{ iff } n \equiv 0 \pmod{3}$$

$$X_n \in S_2 \text{ iff } n \equiv 1 \pmod{3}$$

$$X_n \in S_3 \text{ iff } n \equiv 2 \pmod{3}$$

Defn

A recurrent class is called aperiodic iff it is not periodic. Equivalently, there exists a time $n \geq 0$ such that

$$r_{ij}(n) > 0 \text{ for all } i, j \in R$$

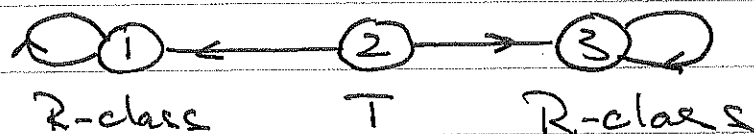
Ex. add a self loop to any state in the previous example.

7.3 steady-state behavior

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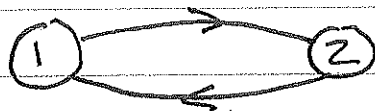
we do not expect the n -step transition probabilities to have limits (not depending on initial state) when there are either

- multiple recurrent classes



or

- periodic recurrent classes



we will write

$$\lim_{n \rightarrow \infty} P_{ij}^{(n)} = \pi_j \quad (\text{for all } i)$$

when the limit exists, and we call π_j the steady-state probability of j .

In the case where these limits exist (independent of i) we interpret

$$\pi_j \approx P(X_n = j) \quad (\text{for } n \text{ large})$$

when do the π_j exist?

Theorem (steady-state convergence)

Consider a Markov chain with a single recurrent class, which is aperiodic.

Then for each $j \in S$:

$$\lim_{n \rightarrow \infty} P_{ij}(n) = \pi_j \quad (\text{for all } i)$$

exists. Furthermore, π_j are solutions to the linear system

$$(1) \quad \sum_{k=1}^m \pi_k P_{kj} = \pi_j \quad (j=1, 2, \dots, m)$$

$$(2) \quad \sum_{k=1}^m \pi_k = 1$$

Also $\pi_j = 0$ if j is transient, and $\pi_j > 0$ if j is recurrent.

Equations (1) are called the balance equations, and follow directly from the Chapman-Kolmogorov equations

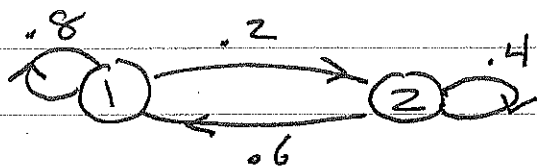
$$r_{ij}(n) = \sum_{k=1}^m r_{ik}(n-1) p_{kj}$$

By taking limits $r_{ij}(n) \rightarrow \pi_j$ and $r_{ik}(n-1) \rightarrow \pi_k$ as $n \rightarrow \infty$.

Equation (2) is called the normalization equation, and follows from the fact that

$$\sum_{j=1}^m P(X_n = j) = 1 \quad (\text{any } n \geq 0)$$

Ex



$$P_{11} = .8, \quad P_{12} = .2$$

$$P_{21} = .6, \quad P_{22} = .4$$

$$(1) \quad \begin{cases} \pi_1 P_{11} + \pi_2 P_{21} = \pi_1 \\ \pi_1 P_{12} + \pi_2 P_{22} = \pi_2 \end{cases}$$

$$(2) \quad \pi_1 + \pi_2 = 1$$

$$\begin{cases} (.8)\pi_1 + (.6)\pi_2 = \pi_1 \Leftrightarrow (.6)\pi_2 = (.2)\pi_1 \\ (.2)\pi_1 + (.4)\pi_2 = \pi_2 \Leftrightarrow (.2)\pi_1 = (.6)\pi_2 \\ \pi_1 + \pi_2 = 1 \end{cases}$$

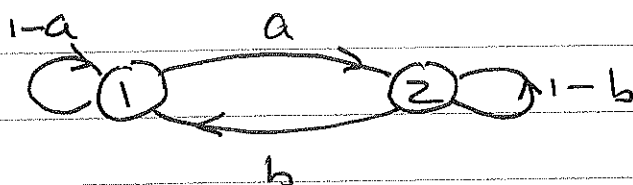
The 1st two give $\pi_1 = 3\pi_2$. The 3rd yields

$$3\pi_2 + \pi_2 = 1 \rightarrow 4\pi_2 = 1 \rightarrow \boxed{\pi_2 = \frac{1}{4}}$$

hence

$$\boxed{\pi_1 = \frac{3}{4}}$$

Ex.



$$\begin{cases} \pi_1(1-a) + \pi_2 b = \pi_1 \\ \pi_1 a + \pi_2(1-b) = \pi_2 \end{cases}$$

$$\rightarrow \pi_2 b = \pi_1 a$$

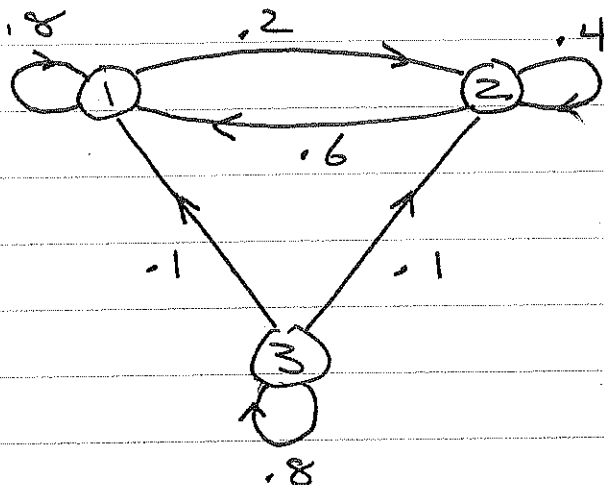
$$\therefore \pi_1 = \frac{b}{a} \pi_2$$

$$\therefore \frac{b}{a} \pi_2 + \pi_2 = 1$$

$$\therefore \pi_2 = \frac{1}{\frac{b}{a} + 1} \rightarrow \boxed{\pi_2 = \frac{a}{a+b}}$$

$$\therefore \boxed{\pi_1 = \frac{b}{a+b}}$$

Ex.



$$\begin{cases} \pi_1 p_{11} + \pi_2 p_{21} + \pi_3 p_{31} = \pi_1 \\ \pi_1 p_{12} + \pi_2 p_{22} + \pi_3 p_{32} = \pi_2 \\ \pi_1 p_{13} + \pi_2 p_{23} + \pi_3 p_{33} = \pi_3 \end{cases}$$

$$\begin{cases} (.8)\pi_1 + (.6)\pi_2 + (.1)\pi_3 = \pi_1 \\ (.2)\pi_1 + (.4)\pi_2 + (.1)\pi_3 = \pi_2 \end{cases}$$

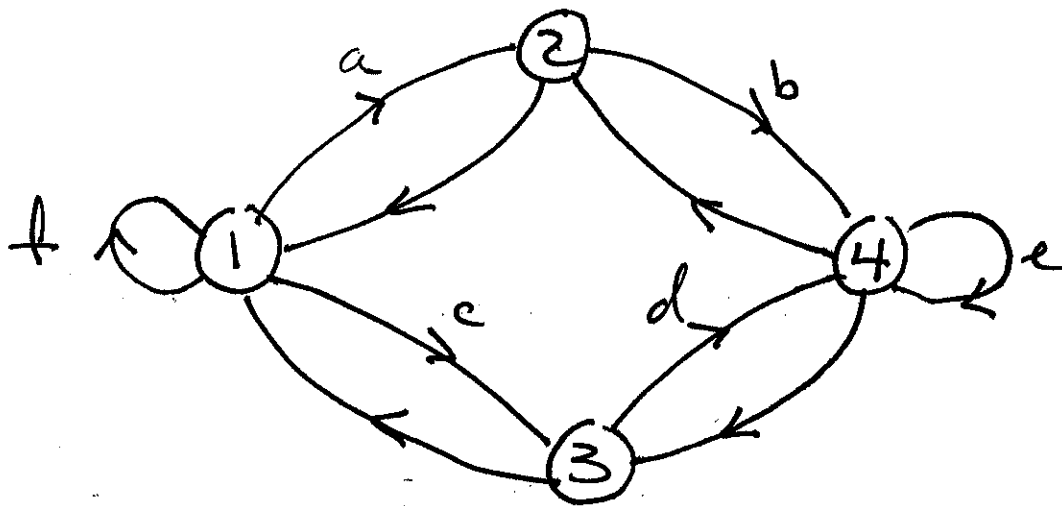
$$\begin{cases} -2\pi_1 + 6\pi_2 + \pi_3 = 0 \\ 2\pi_1 - 6\pi_2 + \pi_3 = 0 \end{cases} \rightarrow \begin{aligned} &2\pi_3 = 0 \\ &\rightarrow \boxed{\pi_3 = 0} \end{aligned}$$

$$\rightarrow \pi_1 = 3\pi_2$$

$$3\pi_2 + \pi_2 = 1 \rightarrow \boxed{\pi_2 = \frac{1}{4}} \therefore \boxed{\pi_1 = \frac{3}{4}}$$

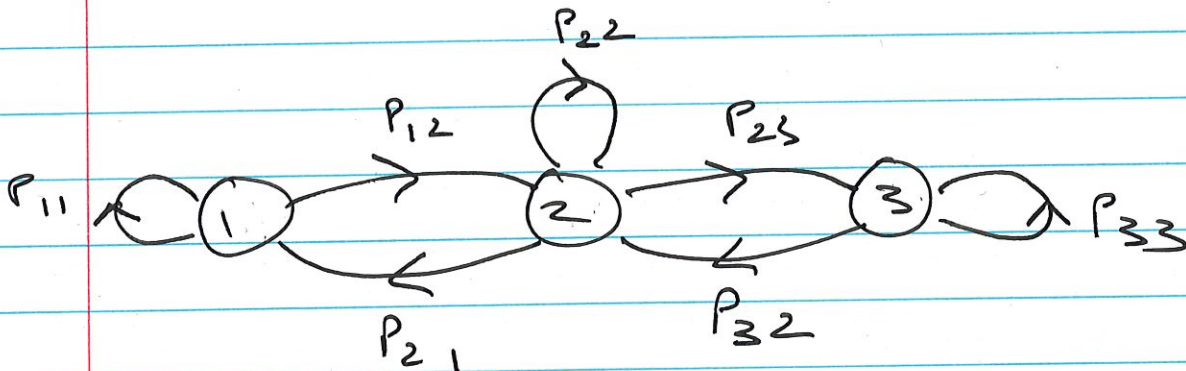
See examples 7.6 and 7.7 on
Pages 355 and 356, resp.

we should have
anticipated this since
we know state 3 is
transient

Ex.

$$r_{14}(3) = abe + cde \\ + fab + fcd$$

$\left\{ \begin{array}{l} 1, 2, 4, 4 \\ 1, 3, 4, 4 \\ 1, 1, 2, 4 \\ 1, 1, 3, 4 \end{array} \right.$
 all paths
 from 1 to 4
 of length 3

Ex.

$$P(X_{103} = 3 \mid X_{100} = 1)$$

$$= r_{13}(3)$$

$\left\{ \begin{array}{l} 1 \ 1 \ 2 \ 3 \\ 1 \ 2 \ 2 \ 3 \\ 1 \ 2 \ 3 \ 3 \end{array} \right.$
 i-3 Paths
 of len. 3

$$= P_{11} P_{12} P_{23} + P_{12} P_{22} P_{23} + P_{12} P_{23} P_{33}$$

$$= \underbrace{(P_{11} P_{12} + P_{12} P_{22})}_{r_{12}(2)} P_{23} + \underbrace{(P_{12} P_{23})}_{r_{13}(2)} P_{33}$$

$$= r_{12}(2) P_{23} + r_{13}(2) P_{33}$$

$$= r_{11}(2) \cdot \underbrace{P_{13}}_0 + r_{12}(2) P_{23} + r_{13}(2) P_{33}$$