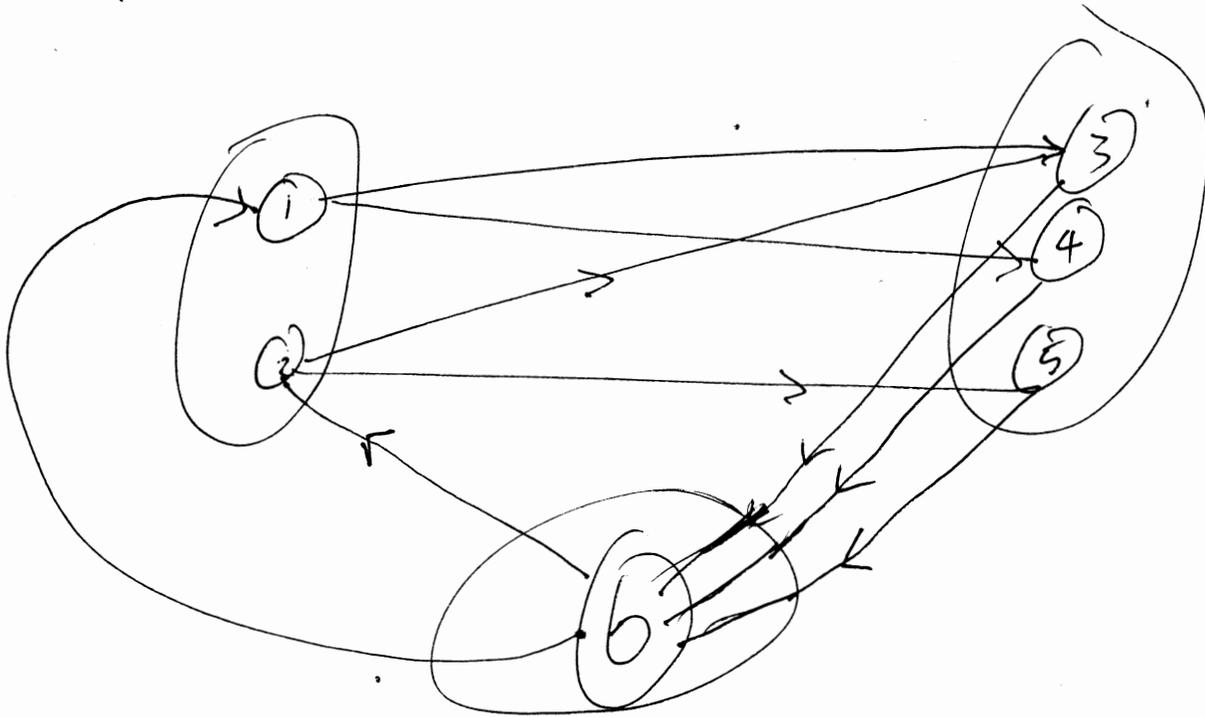


# Periodicity

- A recurrent class is periodic if its states can be grouped in  $d > 1$  disjoint subsets  $S_1, \dots, S_d$ .  
s.th. all transitions from one subset leads to the next subset

EX:



3 subsets  
(1, 2)  
(3, 4, 5)  
(6)

Mathematically:

$$\text{If } i \in S_k \text{ and } p_{ij} > 0 \Rightarrow \begin{cases} j \in S_{k+1} & k=1, \dots, d-1 \\ j \in S_1 & k=d \end{cases}$$

Aperiodic recurrent class : not periodic.

Intuitively: in a periodic recurrent class, move through the sequence of subsets in order; after  $d$  steps end up in the same subset

### Steady State Behavior

- what happens to  $r_{ij}(n)$  as  $n \rightarrow \infty$
- Does it converge to something?
- If so, is convergence depend on initial condition?

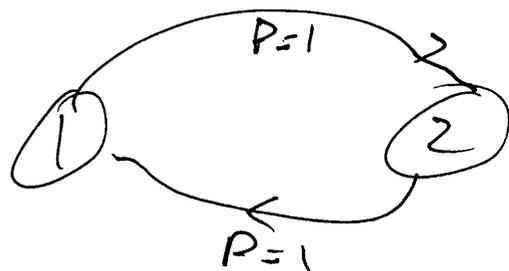
- observation 1 :  $\lim_{n \rightarrow \infty} r_{ij}(n)$  depends on initial

state if there are 2 or more recurrent state

why? depends on whether  $i$  and  $j$  are in the same class!

$\Rightarrow$  Better focus only on single recurrent class.

Observation 2 Consider



Never Converges!

$$r_{ij}(n) = \begin{cases} 1 & \text{if } n \text{ is even} \\ 0 & \text{if } n \text{ is odd} \end{cases}$$

Why? It is periodic recurrent class

Can show that for P. r. C.  $r_{ij}(n)$  oscillates.

upshot: Assume A single recurrent class that is not periodic

① Then  $\lim_{n \rightarrow \infty} r_{ij}(n) = \pi_j \quad \forall i$  i.e. independent of initial state  $i$ .

$\pi_j$  is called steady state probability for  $j$ .

②  $\pi_j$  are the unique solution to the systems of Equations

~~balance~~  
balance equations  $\rightarrow$   $\pi_j = \sum_{k=1}^m \pi_k P_{kj} \quad j=1, \dots, m$

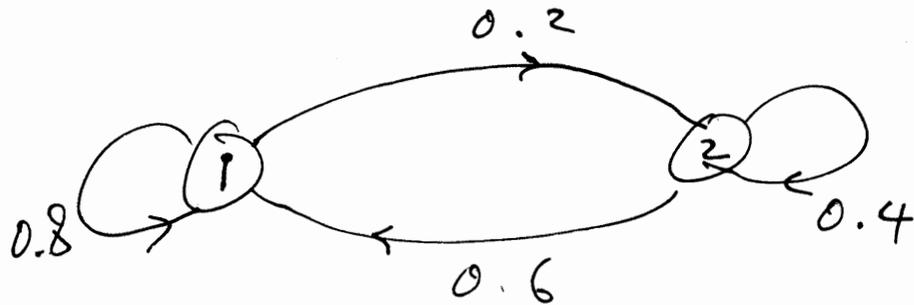
$$1 = \sum_{k=1}^m \pi_k$$

$\rightarrow$  can use these to solve for  $\pi_j$  from  $P_{ij}$ .

③  $\pi_j = 0$  if  $j$  transient  
 $\pi_j > 0$  if  $j$  recurrent.

Let  $n \rightarrow \infty$  in  $\rightarrow$   $r_{ij}(n) = \sum_{k=1}^m r_{ik}(n-1) P_{kj} \Rightarrow \pi_j = \sum_{k=1}^m \pi_k P_{kj}$

EX Student up to date / Behind.



$$P_{11} = 0.8$$

$$P_{12} = 0.2$$

$$P_{21} = 0.6$$

$$P_{22} = 0.4$$

Balance Equations

Linearly dependent.

$$\left\{ \begin{array}{l} \pi_1 = \pi_1 P_{11} + \pi_2 P_{21} \\ \pi_2 = \pi_1 P_{12} + \pi_2 P_{22} \end{array} \right.$$

$$\pi_1 + \pi_2 = 1$$

Typically one of the  $m$  Balance Equs can be ~~written~~ derived

$\Rightarrow$  2 Linearly indep. Equ, 2 unknown  $\Rightarrow$

$$\pi_1 = 0.75 \quad \pi_2 = 0.25$$

---

EX Prof commutes from office to home & back

- has 2 umbrellas.

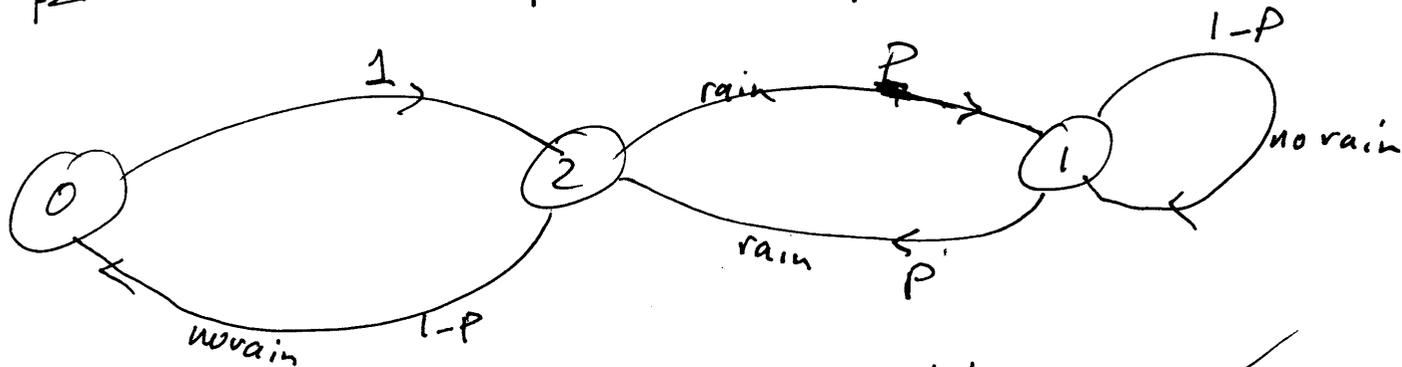
- If it rains and has an umbrella in her location, takes it.

- If ~~rains~~ not raining doesn't take an umbrella. (5)

What is the steady state prob. that she gets wet during a commute?

State  $i$  :  $i$  umbrellas available in her location.  
 $i = 0, 1, 2.$

Draw ~~the~~ Transition prob. graph.



- Single recurrent class, ~~not~~ aperiodic  $\Rightarrow$  ✓

Balance Equs:

$$\pi_0 = (1-p)\pi_2$$

$$\pi_1 = (1-p)\pi_1 + p\pi_2$$

$$\pi_2 = \pi_0 + p\pi_1$$

$$\pi_0 + \pi_1 + \pi_2 = 1$$

$$\Rightarrow \pi_0 = \frac{1-p}{3-p}$$

$$\pi_1 = \frac{1}{3-p}$$

$$\pi_2 = \frac{1}{3-p}$$

$\pi_0 =$  steady state prob for being without umbrella

$$\text{Prob getting wet} = \pi_0 \cdot p(\text{rain}) = \pi_0 p$$

---

### Long Term Freq. Interpretation:

- Think of probability as relative frequency in  $\infty$  # of trials
- EX machine at the end of each day either OK or broken
  - Each repair cost \$2.
  - Expected cost of repair per day??

---

Approach ① pick a random day - find expected repair cost for that day.  
"  $\pi_{\text{Broken}}$  = steady state prob of broken

Approach ② 
$$\frac{\text{Total expected repair cost in } n \text{ days}}{n} \quad n \rightarrow \infty$$

Should get same answer either way.

Let  $v_{ij}(n) =$  expected # of visits to  $j$  within  $n$  steps, starting from  $i$

Assume M.C. is single class aperiodic.

Then  $\Rightarrow \pi_j = \lim_{n \rightarrow \infty} \frac{v_{ij}(n)}{n}$

---

$\pi_j =$  fraction of time in state  $j$

$P_{jk} =$  prob of going from  $j$  to  $k$ .

$\Rightarrow \pi_j P_{jk} =$  fraction of transitions from  $j$  to  $k$ .

---

Let  $q_{jk}(n) =$  expected # of transitions from  $j$  to  $k$  over  $n$  steps

Then  $\lim_{n \rightarrow \infty} \frac{q_{jk}(n)}{n} = \pi_j P_{jk}$



Can show

$$\pi_i = \pi_0 \frac{b_0 b_1 \dots b_{i-1}}{d_1 d_2 \dots d_i} \quad i=1, \dots, m$$

$$\sum_i \pi_i = 1$$

→ Compute all  $\pi_i$ .

EX

- person walks along a straight line.

- step to right prob  $b$

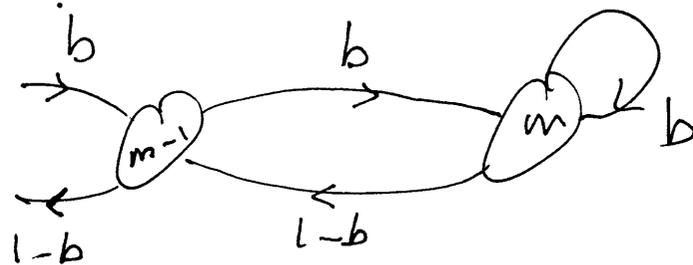
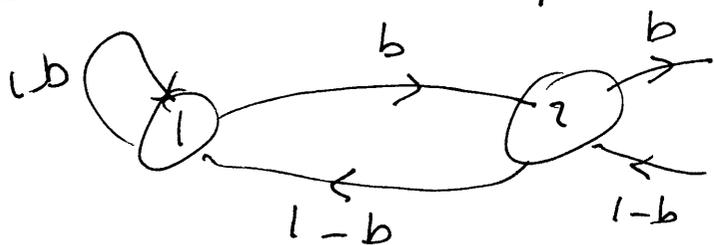
- " " left prob  $1-b$

- starts in one of  $1, 2, \dots, m$

- if in state  $1$  ~~stays~~ stays in  $1$  with prob  $1-b$

- " " " "  $m$  " " " "  $m$  " " " "  $b$

~~Transition~~ Transition prob. graph:



Find steady state prob  $\pi_i$

Apply (\*)  $\pi_i b = \pi_{i+1} (1-b) \Rightarrow \pi_{i+1} = \rho \pi_i$   
 $\rho = \frac{b}{1-b}$

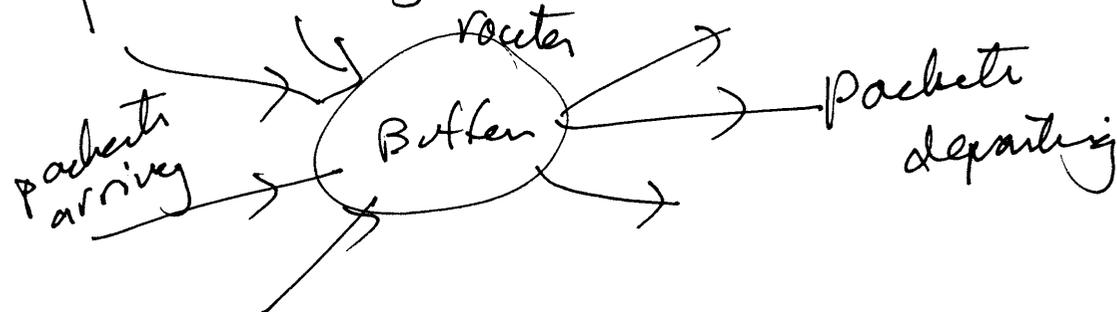
$\Rightarrow \pi_i = \rho^{i-1} \pi_1 \quad i=1, \dots, m$

Apply normalization  $1 = \pi_1 + \dots + \pi_m \Rightarrow$   
 $1 = \pi_1 (1 + \rho + \rho^2 + \dots + \rho^{m-1})$

$\Rightarrow \pi_1 = \frac{1}{1 + \rho + \rho^2 + \dots + \rho^{m-1}}$

$\Rightarrow \pi_i = \frac{\rho^{i-1}}{\sum_{i=0}^{m-1} \rho^i}$

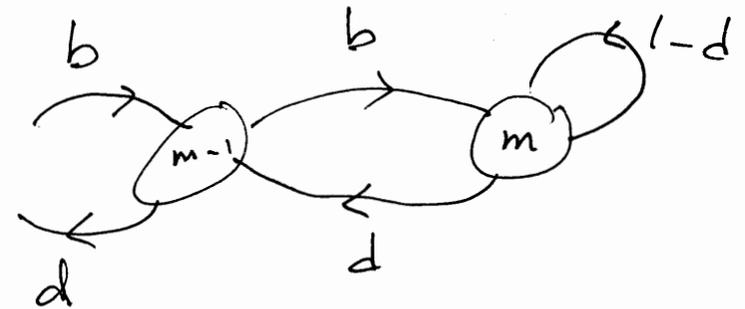
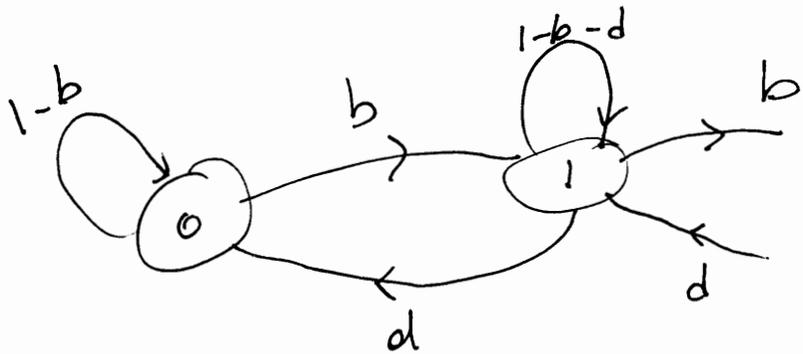
EX Queueing Theory:



- Buffer capacity  $m$
- If buffer full (has  $m$  packets), new packets dropped
- Discretize time into small periods.
- Each period, one of 3 things can happen.

- ① new packet arrives w/ prob  $b > 0$
- ② an existing packet leaves w/ prob  $d > 0$   
(if there is ~~a packet~~ at least one packet)
- ③ ~~neither~~ <sup>neither</sup> 1 nor 2.  
prob  $1-b-d$  if at least one packet  
prob  $1-b$  otherwise

Draw Transition prob graph. States  $0, 1, \dots, m = \# \text{ of packets in the buffer}$



Compute steady state probabilities

$$(*) \Rightarrow \pi_i b = \pi_{i+1} d \quad i = 0, 1, \dots, m-1$$

Define  $\rho = \frac{b}{d} \Rightarrow \pi_{i+1} = \rho \pi_i$

$$\pi_i = \rho^i \pi_0 \quad i = 0, 1, \dots, m \quad \left. \vphantom{\pi_i} \right\} \Rightarrow$$

Normalization

$$1 = \pi_0 + \pi_1 + \dots + \pi_m$$

$$1 = \pi_0 (1 + \rho + \dots + \rho^m)$$

$$\Rightarrow \pi_0 = \begin{cases} \frac{1 - \rho}{1 - \rho^{m+1}} & \rho \neq 1 \\ \frac{1}{m+1} & \rho = 1 \end{cases}$$

$$\pi_i = \begin{cases} \frac{1 - \rho}{1 - \rho^{m+1}} \rho^i & \rho \neq 1 \\ \frac{1}{m+1} & \rho = 1 \end{cases}$$

what happens as buffer size  $m \rightarrow \infty$ ?

Two cases.

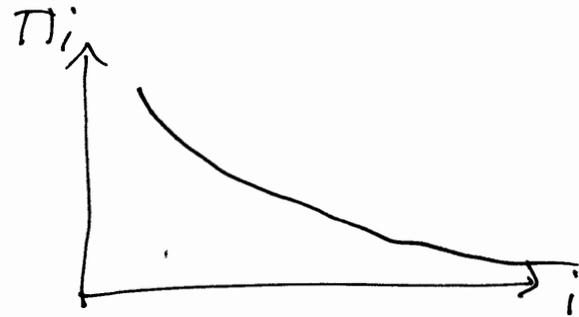
①  $b < d$  i.e.  $\rho < 1$

$$1 - \rho^{m+1} \rightarrow 1$$

i.e. more departures than arrivals

$$\pi_i \rightarrow \rho^i (1 - \rho) \quad \forall i$$

$\Rightarrow$  # of packets in the buffer does not grow because  $b < d$



②  $b > d$  i.e.  $\rho > 1$

$\Rightarrow$  more arrivals than departure

as  $m \rightarrow \infty$

$$\pi_i \rightarrow 0$$

$\forall i \Rightarrow$  all states transient  $\Rightarrow$  no steady state distribution.

$\pi_i$  increases with  $i$

$\Rightarrow$  # of packets in the buffer increases

# Absorption properties

Def. Recurrent state  $\hat{k}$  is absorbing if

$$P_{kk} = 1 \quad P_{kj} = 0 \quad \forall j \neq k$$

→ Suppose M.C. starts at a transient state.

- What is the first recurrent state it enters?

- How long does it take for this to happen?

- Observe: Answer is same regardless of whether we use first recurrent or first absorbing.

Why? Don't care what happens after recurrent state is reached

⇒ Focus on the case where every recurrent state is absorbing.

# Absorption Probability Eqs

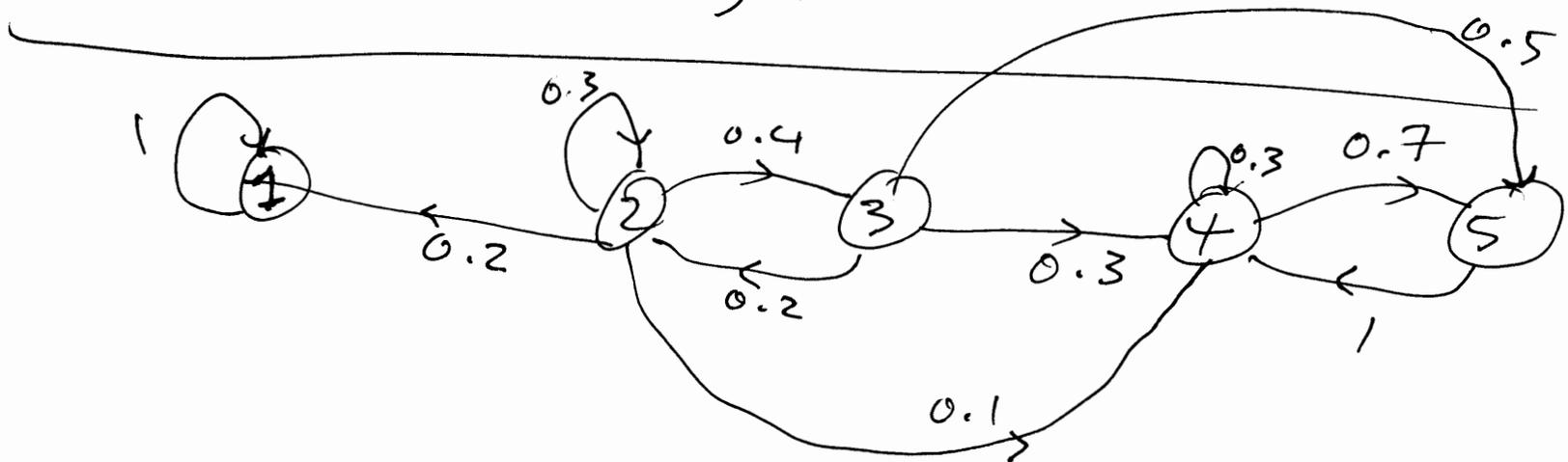
- Consider a M.C each state either Transient or absorbing
  - Fix an absorbing state  $s$
  - $a_i = \text{prob eventually reaching } s \text{ starting from } i$
- Then: can find  $a_i$  by solving ~~the equations~~

$$a_s = 1$$

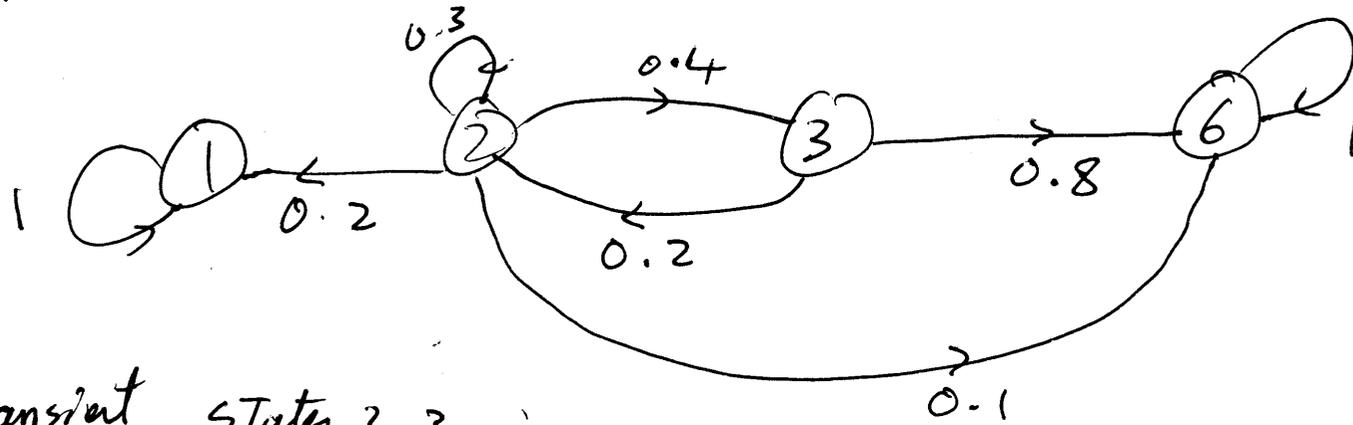
$$a_i = 0 \quad \forall \text{ absorbing } i \neq s$$

$$a_i = \sum_{j=1}^m p_{ij} a_j \quad \forall \text{ transient } i$$

Ex



- Find prob state reaches recurrent class  $\{4,5\}$  starting from Transient state
- Lump  $\{4,5\}$  into  $\{6\}$ .



For Transient States 2, 3 :

$$\begin{cases} a_2 = 0.2 a_1 + 0.3 a_2 + 0.4 a_3 + 0.1 a_6 \\ a_3 = 0.2 a_2 + 0.8 a_6 \end{cases} \rightarrow$$

Use  $a_1 = 0$        $a_6 = 1$

$$\begin{cases} a_2 = 0.3 a_2 + 0.4 a_3 + 0.1 \\ a_3 = 0.2 a_2 + 0.8 \end{cases} \rightarrow$$

$$a_2 = \frac{21}{31}$$

$$a_3 = \frac{29}{31}$$