

## MA41031 Stochastic Processes in Finance

Wed 10-11, Thurs 9-10, Fri 11-12, 12-1

CT-1 10/9 Fri ; CT-2 8/10 Fri ; CT-3 18/11 Thurs  
Books 1) Probability Models SM Ross

2) Sto. Process SM Ross

3) Prob., Stat. and Queuing Theory  
by Arnold O. Allen

4) Intro. to Stochastic Modelling,  
by Pinsky, Karlin

5) Stochastic Calculus in Finance I, II by Shreve

### Review: Introduction to probability

#### Stochastic Process (S.P.)

Defn A S.P. is a family of r.v.'s  $\{X(t), t \in T\}$ , defined  
on a given probability space, indexed by the  
parameter  $t$ , where  $t \in T$   $\rightarrow$  parameter/time class

$\rightarrow$  The values assumed by r.v.  $X(t)$  are called states,  
and set of all possible values form the state  
space  $(S)$  of the process.  
—X—

1) discrete state, discrete parameter S.P.

2) " " , continuous " S.P.

3) " " , continuous " S.P.

3) Cont. " , " " or

4) " , discrete " S.P.  
— x —

Example: Consider a queueing system with jobs arriving at random point in time, queuing for service and departing from the system after service completion.

a)  $X(t)$  # of jobs in the system at time  $t$

$$S = \{0, 1, 2, \dots\} ; T = \{t \mid 0 \leq t < \infty\}$$

$[X(t)]$  discrete state, continuous parameter S.P.

b)  $W_k$  time that the  $k^{\text{th}}$  customer has to wait in the system before receiving service.

$$S = \{x \mid 0 \leq x < \infty\} , T = \{1, 2, 3, \dots\}$$

$[W_k]$  continuous state, discrete parameter S.P.

c)  $Y(t)$  cumulative service requirement (experience) of all jobs in the system at time  $t$ .

$$S = [0, \infty) , T = [0, \infty)$$

$[Y(t)]$  cont. state, cont. parameter S.P.

d)  $N_k$  # of jobs in the system at the time of the departure of the  $k^{\text{th}}$  customer (after service completion)

$$S = \{0, 1, 2, \dots\}, T = \{1, 2, 3, \dots\}$$

(Nk) discrete state, discrete parameter/time S.P.

—X—

Discrete Time Markov Chain (DTMC):

discrete state, discrete time/parameter S.P.

$$\{X_n, n=0, 1, 2, \dots\}$$

$$\text{state} \rightarrow S = \{x_0, x_1, x_2, \dots\}$$

$$\text{space} = \{0, 1, 2, \dots\}$$

$X_n = i \Leftrightarrow$  process is in state  $i$  at time/step  $n$ .  
 $\{i, j, i_{n-1}, \dots, i_0\} \in S$

$\{X_n\}$  is DTMC iff

$$P(X_{n+1}=j \mid X_n=i, X_{n-1}=i_{n-1}, \dots, X_0=i_0)$$

$$= P(X_{n+1}=j \mid X_n=i)$$

$$= p_{ij}^{(n)} \rightarrow \text{transition probability}$$

$$= p_{ij}^{(1)} \rightarrow \text{stationary transition probability (Homogeneous M.C.)}$$

$$= p_{ij} \rightarrow \begin{aligned} &P(X_1=j \mid X_0=i) \\ &= P(X_{n+1}=j \mid X_n=i) \end{aligned}$$

$$p_{ij} = P(\underset{\text{final state}}{X_{n+1}=j} \mid \underset{\text{initial state}}{X_n=i}) \quad i, j \in S$$

$$0 \leq p_{ij} \leq 1, \forall i, \forall j$$

For fixed  $i$ ;

$$\sum_{j=0}^{\infty} p_{ij} = 1$$

0    1    2    —

$$i \begin{pmatrix} 0 & p_{00} & p_{01} & p_{02} & \dots \\ 1 & p_{10} & p_{11} & p_{12} & \dots \\ 2 & p_{20} & p_{21} & p_{22} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix} = P = P^{(1)} \checkmark$$

Transition  
Probability  
Matrix  
(TPM or tpm)

Example Consider a game of ladder climbing. There are 5 levels in the game, level 1 is lowest (bottom) and level 5 is the highest (top). A player starts at the bottom. Each time, a fair coin is tossed. If it turns up heads, the player moves up one rung. If tails, the player moves down to the very bottom. Once at the top level, the player moves to the very bottom if tails turns up and stays at the top if head turns up.

Let  $X_n$  be the level of the game in the  $n^{\text{th}}$  step/transition.

$$X_n \in \{1, 2, 3, 4, 5\} = S$$

State space  $\mathcal{Q}$   $\mathbb{P}$

$$P(X_{n+1} = j \mid X_n = i, X_{n-1} = i_{n-1}, \dots, X_0 = i_0)$$

$$= P(X_{n+1} = j \mid X_n = i) = b_{ij}$$

5  $\mathbb{P}$   
4  $\mathbb{P}$   
3  $\mathbb{P}$

$(X_n)$  DTMC  
 $i, j \in S$   
 $X_{n+1} = j$

$Tpm$

$((p_{ij})) = P =$

$X_n = i$

$p = q = \frac{1}{2}$   
 fair coin

$$\begin{matrix}
 & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\
 \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} q & p & 0 & 0 & 0 \\ q & 0 & p & 0 & 0 \\ q & 0 & 0 & p & 0 \\ q & 0 & 0 & 0 & p \\ q & 0 & 0 & 0 & p \end{bmatrix}
 \end{matrix}$$

$$= \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & 0 & 0 & \frac{1}{2} & 0 \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} \end{bmatrix}$$

Example(2) Let  $\{X_n\}_{n=0,1,2,\dots}$  be a sequence of i.i.d. discrete r.v. with  $P(X_1 = j) = \left(\frac{1}{2}\right)^{j+1} \forall j = 0, 1, 2, \dots$ .  
 $p_j \leq \frac{1}{2}$

Determine whether each of the following chain is Markovian or not. If so find its corresponding state space  $(S)$  and  $tpm$

- (i)  $\{S_n\}_{n=0,1,2,\dots}$  where  $S_n = \sum_{i=1}^n X_i$
- (ii)  $\{m_n\}_{n=0,1,2,\dots}$

... where  $1/n = \max \{X_1, X_2, \dots, X_n\}$

Sol (i)  $S_n = \sum_{i=1}^n X_i$

$\{S_n\}$  is MC with state space  $S = \{0, 1, 2, \dots\}$

$$p_{ij} = P(S_{n+1} = j | S_n = i), \quad \forall i, j \in S$$

$$P = (p_{ij}) = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & \dots \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ \vdots \\ i \end{matrix} & \begin{bmatrix} p_0 & p_1 & p_2 & p_3 & \dots \\ 0 & p_0 & p_1 & p_2 & \dots \\ 0 & 0 & p_0 & p_1 & \dots \\ 0 & 0 & 0 & p_0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \end{matrix}$$

$$= \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{8} & \frac{1}{16} & \dots \\ 0 & \frac{1}{2} & \frac{1}{4} & \frac{1}{8} & \dots \\ 0 & 0 & \frac{1}{2} & \frac{1}{4} & \dots \\ 0 & 0 & 0 & \frac{1}{2} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

Example (3) (Transformation of a process into M.C.)

Suppose that whether or not it rains today depends on previous weather conditions through the last two days.

Suppose that if it has rained for the past two days, then it will rain tomorrow with prob (wk) 0.7; if

if it has rained today but not yesterday, then it will rain tomorrow w.p 0.5; if it has rained yesterday but not today, then it will rain tomorrow w.p 0.4; if it has not rained in the past two days, then it will rain tomorrow w.p 0.2.

$X_n$  weather condition of that day  
 $(X_n)$  not a m.c

$X_n$ : state at any time is determined by the weather conditions during both that day and the previous day.

$(X_n)$  m.c.

State $X_n$	Rained yesterday	Rained today
0	✓	✓
1	X	✓
2	✓	X
3	X	X

$$S = \{0, 1, 2, 3\} \equiv \{VV, XV, VX, XX\}$$

tpm

$$P = \begin{matrix} & \begin{matrix} \text{yesterday} & \text{today} & \text{today} & \text{tomorrow} & X_{n+1}=j \end{matrix} \\ \begin{matrix} X_n=i \\ \downarrow \\ \text{yesterday} & \text{today} & \text{today} & \text{tomorrow} \end{matrix} & \begin{matrix} V & V & X & X \\ \downarrow & \downarrow & \downarrow & \downarrow \end{matrix} & \begin{matrix} V & V & X & X \\ \downarrow & \downarrow & \downarrow & \downarrow \end{matrix} & \begin{matrix} V & V & X & X \\ \downarrow & \downarrow & \downarrow & \downarrow \end{matrix} & \begin{matrix} V & V & X & X \\ \downarrow & \downarrow & \downarrow & \downarrow \end{matrix} \end{matrix} \begin{matrix} 0.7 & 0 & 0.3 & 0 \\ 0.5 & 0 & 0.5 & 0 \\ 0 & 0.4 & 0 & 0.6 \\ 0 & 0.2 & 0 & 0.8 \end{matrix} \end{matrix}$$

$$p_{ij} = P(X_{n+1}=j | X_n=i), \forall i, j \in S$$

n-Step transition probability:  $\{X_n\}$  DTMC ,  $S = \{0, 1, 2, \dots\}$   
 State space

$$p_{ij}^{(n)} = P(X_{m+n}=j | X_m=i)$$

$$= P(X_n=j | X_0=i) \quad , i, j \in S$$

$$0 \leq p_{ij}^{(n)} \leq 1$$

For fixed  $i$

$$\sum_j p_{ij}^{(n)} = 1$$

n-step tpm

$$P^{(n)} = (p_{ij}^{(n)}) = \begin{pmatrix} p_{00}^{(n)} & p_{01}^{(n)} & \dots \\ p_{10}^{(n)} & p_{11}^{(n)} & \dots \\ \dots & \dots & \dots \\ \dots & \dots & \dots \end{pmatrix}$$

Chapman kolomogorov equation:  $\{X_n\}$  M.C

$$p_{ij}^{(m+n)} = \sum_k p_{ik}^{(m)} p_{kj}^{(n)} = \sum_k p_{ik}^{(n)} p_{kj}^{(m)}$$

Let  $(i, j)^{th}$  element of  $P^{(m+n)}$  is

$$p_{ij}^{(m+n)} = P(X_{m+n}=j | X_0=i)$$

$$= \sum_k P(X_{m+n}=j, X_n=k | X_0=i)$$

$$= \sum_k P(X_{m+n}=j | X_n=k, X_0=i)$$

$$\cdot P(X_n=k | X_0=i)$$

[using thm of total prob]

$$\left\{ \begin{aligned} P(A|B|C) &= \frac{P(ABC)}{P(B|C)} \times \frac{P(B|C)}{P(C)} \\ &= P(A|BC) \cdot P(B|C) \end{aligned} \right.$$



$$= \sum_k P(X_{m+n}=j | X_n=k) \cdot P(X_n=k | X_0=i)$$

$\because (X_n) \text{ is m.c.}$

$$= \sum_k p_{kj}^{(m)} p_{ik}^{(n)}$$

$$= \sum_k p_{ik}^{(n)} p_{kj}^{(m)}$$

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$$p_{ij}^{(m+n)} = \sum_k p_{ik}^{(n)} p_{kj}^{(m)}$$


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$p^{(n)}$ 
 $p^{(m)}$

in

$$i \begin{pmatrix} \begin{matrix} \text{---} & \text{---} & \text{---} & \text{---} \\ \text{---} & \text{---} & \text{---} & \text{---} \\ p_{i0}^{(n)} & p_{ij}^{(n)} & \text{---} & \text{---} \\ \text{---} & \text{---} & \text{---} & \text{---} \end{matrix} \end{pmatrix} \begin{pmatrix} \begin{matrix} \text{---} & \text{---} & \text{---} & \text{---} \\ \text{---} & \text{---} & p_{0j}^{(m)} & \text{---} \\ \text{---} & \text{---} & p_{ij}^{(m)} & \text{---} \\ \text{---} & \text{---} & \vdots & \text{---} \end{matrix} \end{pmatrix}$$

$$p^{(m+n)} = p^{(n)} p^{(m)}$$

We know  $p^{(1)} = P$

$$m=1, n=1 \quad p^{(2)} = p^{(1)} p^{(1)} = P \cdot P = P^2$$

⋮

$$p^{(n)} = P^n$$


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$$P^n = p^{(n)} = (p_{ij}^{(n)})$$


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$(X_n)$  DTMC

$S = \{0, 1, 2, \dots\}$

$P$

$\uparrow$   $p_{ij}$

$X_0$

$$P(X_0=i) = p_i^{(0)}$$

$i \in S$

$$p^{(0)} = (p_0^{(0)}, p_1^{(0)}, p_2^{(0)})$$

$\rightarrow$  initial state ...

$\dots, r_1, r_2, \dots$  same prob. dist

$$X_n \quad P(X_n=i) = p_i^{(n)} \quad i \in S$$

$$\underline{\underline{pmf}} \quad (p_0^{(n)}, p_1^{(n)}, p_2^{(n)}, \dots) = \tilde{p}^{(n)}$$

$$\checkmark \quad \tilde{p}^{(n)} = \tilde{p}^{(0)} P^n$$

$$\checkmark \quad \boxed{\tilde{p}^{(n)} = \tilde{p}^{(n-1)} P}$$

$$(p_0^{(n)}, p_1^{(n)}, \dots, p_i^{(n)}, \dots) = (p_0^{(n-1)}, p_1^{(n-1)}, \dots) \begin{pmatrix} p_{00} & p_{01} & \dots & p_{0i} \\ p_{10} & p_{11} & \dots & p_{1i} \\ \vdots & \vdots & \ddots & \vdots \end{pmatrix}$$

$$\boxed{p_i^{(n)} = \sum_k p_k^{(n-1)} p_{ki}}$$

$$\text{Let } p_i^{(n)} = P(X_n=i)$$

$$= \sum_k P(X_n=i, X_{n-1}=k)$$

| using the total prob

$$= \sum_k P(X_n=i | X_{n-1}=k) P(X_{n-1}=k)$$

$$= \sum_k p_{ki}^{(1)} p_k^{(n-1)}$$

$$= \sum_k p_{ki} p_k^{(n-1)}$$

Example: Suppose that the chance of rain tomorrow

depends on previous weather conditions only through whether or not it is raining today and not on past weather conditions. Suppose also that if it rains today, then it will rain tomorrow with prob.  $\alpha$ ; and if it does not rain today, then it will rain tomorrow with prob.  $\beta$ .

Let  $X_n$ : weather condition on  $n^{\text{th}}$  day

$$S = \{0, 1\}$$

0 rain, 1 not raining

$$P = \begin{matrix} & \begin{matrix} X_{n+1}=j \\ 0 & 1 \end{matrix} \\ \begin{matrix} X_n=i \\ 0 & 1 \end{matrix} & \begin{pmatrix} \alpha & 1-\alpha \\ \beta & 1-\beta \end{pmatrix} \end{matrix}$$

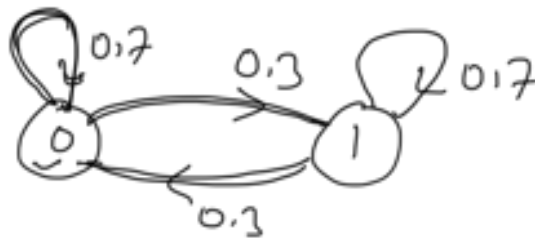
$$\alpha = 0.7, \beta = 0.3$$

$$P = \begin{pmatrix} 0.7 & 0.3 \\ 0.3 & 0.7 \end{pmatrix}$$

$$p_{00}^{(2)} = P(X_2=0 | X_0=0)$$

$$\begin{aligned} p^{(2)} = P^2 &= P \cdot P = \begin{pmatrix} 0.7 & 0.3 \\ 0.3 & 0.7 \end{pmatrix} \begin{pmatrix} 0.7 & 0.3 \\ 0.3 & 0.7 \end{pmatrix} \\ &= \begin{pmatrix} 0.58 & 0.42 \\ 0.42 & 0.58 \end{pmatrix} = \begin{pmatrix} p_{00}^{(2)} & p_{01}^{(2)} \\ p_{10}^{(2)} & p_{11}^{(2)} \end{pmatrix} \\ p_{00}^{(2)} &= 0.58 \end{aligned}$$

$$p_{10}^{(2)} = \sum_{k=0}^1 p_{1k}^{(1)} p_{k0}^{(1)} = p_{10} p_{00} + p_{11} p_{10} = 0.42$$



$$p_{00}^{(2)} = 0.7 \times 0.7 + 0.3 \times 0.3 = 0.58$$

Example:  $(X_n)$  DTMC  $S = \{1, 2, 3\}$  tpm

$$t_{pm} \quad P = \begin{pmatrix} 1 & 2 & 3 \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{pmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & \underline{0.2} \\ \underline{0.3} & \underline{0.4} & \underline{0.3} \end{pmatrix} \end{pmatrix}$$

$$\underline{X_0} \quad p_{mf} \quad (0.7, 0.2, 0.1) = (p_1^{(0)}, p_2^{(0)}, p_3^{(0)}) = p^{(0)}$$

$P(X_0=2)$   $P(X_0=i) = p_i^{(0)}$

$$a) \quad P(X_1=3, X_2=3, X_0=2, X_3=2)$$

$$= P(X_3=2, X_2=3, X_1=3, X_0=2)$$

$$\quad \quad \quad | \quad P(ABCD) = \underline{P(A|BCD)} \underline{P(B|CD)} \underline{P(C|D)} P(D)$$

$$= P(X_3=2 | X_2=3, X_1=3, X_0=2) \cdot P(X_2=3 | X_1=3, X_0=2) \\ \cdot P(X_1=3 | X_0=2) \cdot P(X_0=2)$$

$$= P(\underline{X_3=2} | \underline{X_2=3}) \cdot P(\underline{X_2=3} | \underline{X_1=3}) \cdot P(\underline{X_1=3} | \underline{X_0=2}) \\ \cdot P(\underline{X_0=2})$$

$\because (X_n) \text{ MC}$

$$= p_{32} \, p_{33} \, p_{23} \, P(X_0=2)$$

$$= 0.4 \times 0.3 \times 0.2 \times 0.2 = 0.0048$$

$$(b) \quad P(X_2=3, X_1=3 | X_0=2)$$

$$= P(X_2=3 | X_1=3, X_0=2) \cdot P(X_1=3 | X_0=2)$$

$$\left| \begin{aligned} P(A|C) &= \frac{P(A \cap C)}{P(C)} \times \frac{P(C)}{P(C)} \\ &= P(A \cap C) \cdot P(C) \end{aligned} \right.$$

$$= P(X_2=3 | X_1=3) \cdot P(X_1=3 | X_0=2)$$

$$= p_{33}^{(1)} p_{23}^{(1)} = 0.3 \times 0.2 = 0.06$$

$$(c) \quad P(X_3=2, X_0=2, X_1=3)$$

$$= P(X_3=2, X_1=3, X_0=2)$$

$$= P(\underline{X_3=2} | \underline{X_1=3}, \underline{X_0=2}) \cdot P(X_1=3 | X_0=2) \cdot P(X_0=2)$$

$$= P(X_3=2 | X_1=3) \cdot P(X_1=3 | X_0=2) \cdot P(X_0=2)$$

$$= \underbrace{p_{32}^{(2)}}_{\text{where}} \underbrace{p_{23}^{(1)}}_{\text{where}} P(X_0=2) = 0.35 \times 0.2 \times 0.2 = 0.0048,$$

where

$$\begin{aligned} p_{32}^{(2)} &= \sum_{k \in \{1,2,3\}} p_{3k}^{(1)} p_{k2}^{(1)} = p_{31} p_{12} + p_{32} p_{22} + p_{33} p_{32} \\ &= 0.3 \times 0.5 + 0.4 \times 0.2 + 0.3 \times 0.4 \\ &= 0.35 \end{aligned}$$

$$(d) \quad p_3^{(2)} = P(X_2=3)$$

$$X_2 \text{ pmf } \underline{p}^{(2)} = \left( \underline{p_1^{(2)}}, \underline{p_2^{(2)}}, \underline{p_3^{(2)}} \right)$$

$$X_0 \text{ pmf } \underline{p}^{(0)} = (0.7, 0.2, 0.1)$$

$$X_1 \quad \underline{p}^{(1)} = \underline{p}^{(0)} P = (0.7, 0.2, 0.1) \begin{pmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{pmatrix} \\ = (0.22, 0.43, 0.35)$$

$$X_2 \quad \underline{p}^{(2)} = \underline{p}^{(1)} P = (0.22, 0.43, 0.35) \begin{pmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{pmatrix} \\ = (\underline{0.385}, \underline{0.336}, \underline{0.279})$$

$$P(X_2=3) = 0.279 = p_3^{(2)}$$

—X—  
Classification of States:

$(X_n)$  MC  $S = \{0, 1, 2, \dots\}$ , tpm  $P = ((p_{ij}))$   
state  $i, j \in S$

Def  $i \rightarrow j$ , state  $j$  is accessible from state  $i$  if  $p_{ij}^{(n)} > 0$  for some  $n$ .

Def  $i \leftrightarrow j$ , state  $i$  and  $j$  communicate with each other  
if  $i \rightarrow j$  and  $j \rightarrow i$

Remark  $i \leftrightarrow j, j \leftrightarrow k \Rightarrow i \leftrightarrow k$   
Sel  $\exists n, m$   
 $\underline{p_{ij}^{(n)} > 0, p_{jk}^{(m)} > 0} \quad p_{ik}^{(n+m)} = \sum_l p_{il}^{(n)} p_{lk}^{(m)} \geq p_{ij}^{(n)} p_{jk}^{(m)}$

> 0

Def M.C is irreducible/connected if every state communicate with every other state otherwise M.C. reducible.

Def period of state  $i : d(i)$

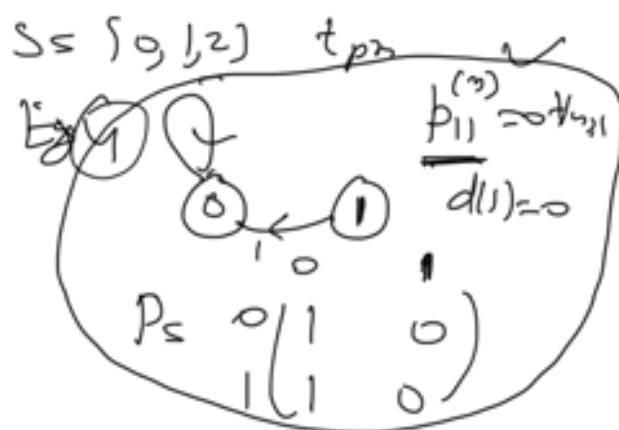
$$d(i) = \gcd \{ n^+ = 1, 2, \dots \mid n \text{ st. } p_{ii}^{(n)} > 0 \}$$

( If  $p_{ii}^{(n)} = 0 \forall n \geq 1$ , define  $d(i) = \infty$  )

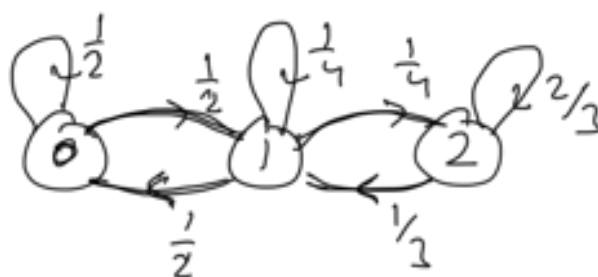
If  $d(i) = 1 \quad i \rightarrow \text{aperiodic}$

Example: (1)  $\{X_n\}$  DTMC

$$P = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ 1 & \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ 2 & 0 & \frac{1}{3} & \frac{2}{3} \end{pmatrix}$$



$$\begin{aligned} f_0 &= 1 \\ f_0^{(1)} &= \\ f_0^{(2)} &= \\ f_0^{(3)} &= \\ &\vdots \end{aligned}$$



0, 1, 2 recurrent

$0 \leftrightarrow 1 \leftrightarrow 2$   
 Class  $\{0, 1, 2\}$

M.C. irreducible/connected

$$d(0) = \gcd \{ 1, 2, 3, \dots \} = 1 = d(1) = d(2)$$

(2)

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$S = \{0, 1\}$

$p^{(1)}$

1

$$f_0^{(1)} = 0$$

$$f_0^{(2)} = 1$$

$$f_0^{(3)} = 0$$

$$f_0^{(4)} = 0$$

Class  $\{0, 1\}$

$$p_{00}^{(n)} > 0$$

MC irreducible



0, recurrent

$$d(0) = \gcd\{2, 4, 6, \dots\} = 2 \leq d(1)$$

Fig (3)



$P =$

Reducible MC

$$\begin{array}{c|cccc} & -2 & -1 & 0 & 1 & 2 \\ \hline -2 & 0 & 1 & 0 & 0 & 0 \\ -1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 \\ 2 & 0 & 0 & 0 & 0 & 0 \end{array}$$

$$p_{00}^{(n)} = 0, \forall n \geq 1 \quad d(0) = 0$$

$$d(i) = 0$$

$X_n$  DTMC,  $P = (p_{ij})$ ,  $S$  —  $p_{ij} = P(X_{n+1} = j | X_n = i)$   
For state  $i$

$$f_{ii}^{(n)} \text{ or } f_i^{(n)} = P(X_n = i, X_k \neq i, k=1, 2, \dots, n-1 | X_0 = i)$$

Prob of first visit to state  $i$  in  $n$  steps

/transitions starting from state  $i$

or recurrence time prob

$$f_{ii}^{(0)} \text{ or } f_i^{(0)} = 1$$

$$f_{ii} \text{ or } f_i = f_i^{(1)} + f_i^{(2)} + f_i^{(3)} + \dots$$

Prob of ever visiting state  $i$



$f_i = 1 \Leftrightarrow i$  recurrent state  $\Leftrightarrow$  return to state  $i$  is certain

$f_i < 1 \Leftrightarrow i$  transient state  $\Leftrightarrow$  return to state  $i$  is uncertain

$$\text{Let } I_n = \begin{cases} 1 & \text{if } X_n = i \\ 0 & \text{if } X_n \neq i \end{cases}$$

$\sum_{n=1}^{\infty} I_n \rightarrow \# \text{ of time period, the process is in state } i$

$$E\left(\sum_{n=1}^{\infty} I_n \mid X_0 = i\right) = \sum_{n=1}^{\infty} E(I_n \mid X_0 = i)$$

$$= \sum_{n=1}^{\infty} (1 \cdot P(X_n = i \mid X_0 = i) + 0 \cdot P(X_n \neq i \mid X_0 = i))$$
$$= \sum_{n=1}^{\infty} p_{ii}^{(n)}$$

$$i \text{ recurrent} \Leftrightarrow f_i = 1 \Leftrightarrow \sum_{n=1}^{\infty} p_{ii}^{(n)} = \infty \quad \checkmark$$

$$i \text{ transient} \Leftrightarrow f_i < 1 \Leftrightarrow \sum_{n=1}^{\infty} p_{ii}^{(n)} < \infty$$

—x—

Def  $i$  recurrent

$$m_i = \sum_{n=1}^{\infty} n f_i^{(n)} \quad \text{mean recurrence time / expected time the process returns to state } i$$

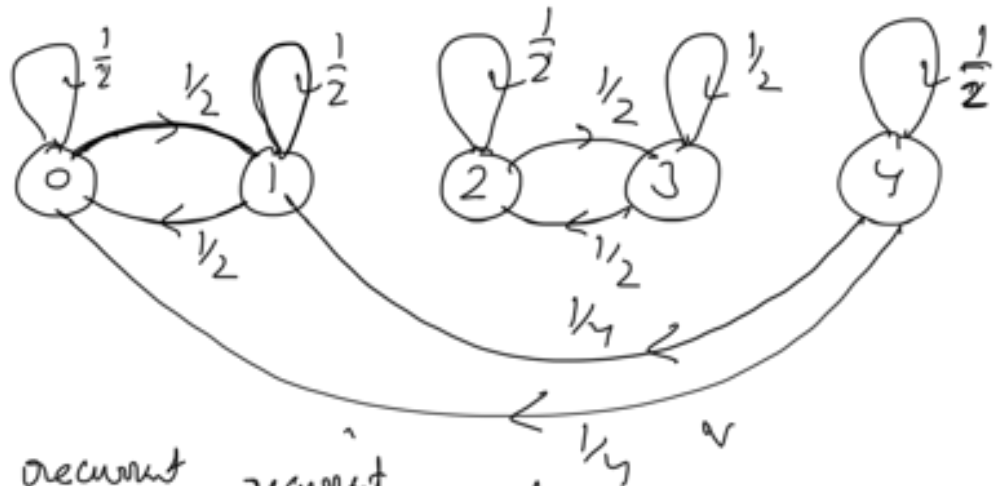
$$m_i = \infty, \quad i \text{ null recurrent}$$

$$\underline{m_i < \infty}, \quad i \text{ non-null recurrent / positive recurrent.}$$

-x-

Example: M.C. having  $S = \{0, 1, 2, 3, 4\}$  and tpm

	0	1	2	3	4
0	$\frac{1}{2}$	$\frac{1}{2}$	0	0	0
1	$\frac{1}{2}$	$\frac{1}{2}$	0	0	0
2	0	0	$\frac{1}{2}$	$\frac{1}{2}$	0
3	0	0	$\frac{1}{2}$	$\frac{1}{2}$	0
4	$\frac{1}{4}$	$\frac{1}{4}$	0	0	$\frac{1}{2}$



Classes  $C_1 = \{0, 1\}$  (recurrent),  $C_2 = \{2, 3\}$  (transient),  $\{4\} = C_3$  (recurrent)

$d(0) = 1 = d(1) = d(2) = d(3) = d(4)$  aperiodic

$$f_0 = f_0^{(1)} + f_0^{(2)} + f_0^{(3)} + \dots$$

$$= \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots = \frac{1}{2} (1 + \frac{1}{2} + \frac{1}{4} + \dots)$$

$$= \frac{1}{2} \times \frac{1}{1 - \frac{1}{2}} = 1.$$

0  $\rightarrow$  recurrent state.

P1  $i \leftrightarrow j$  ,  $i$  recurrent  $\Rightarrow$   $j$  recurrent

Given  $\text{Sel } \left[ \begin{array}{l} \exists n, m \\ p_{ij}^{(n)} > 0, p_{ji}^{(m)} > 0, \sum_{v=1}^{\infty} p_{ii}^{(v)} = \infty \end{array} \right.$

$$p_{jj}^{(m+n+v)} \geq p_{ji}^{(m)} p_{ii}^{(n)} p_{ij}^{(v)} \quad \text{using } C_k = 1$$

$$\sum_v p_{jj}^{(m+n+v)} \geq p_{ji}^{(m)} p_{ij}^{(n)} \left( \sum_v p_{ii}^{(v)} \right) = \infty$$

$$\Rightarrow \sum_n p_{jj}^{(n)} = \infty$$

$j$  recurrent state.

P2  $i \leftrightarrow j$  ,  $i$  transient  $\Rightarrow$   $j$  transient.

$\text{Sel}$  On contrary suppose  $j$  recurrent, since  $i \leftrightarrow j$   
 $\Rightarrow i$  recurrent (using P1)  $\nRightarrow$  a contradiction

P3 In a finite state M.C. all states can not be transient.

P4 In a finite <sup>state</sup>, irreducible M.C., all states are recurrent.

P5 In a irreducible M.C., all states are recurrent or transient.

$\rightarrow$   $x \leftrightarrow y \Rightarrow d(x) = d(y)$

$\text{Sel}$

$$d(x) = \inf \{ n \geq 1 : p_{xx}^{(n)} > 0 \}$$

$$x \leftrightarrow y \\ \exists m, n \quad p_{x,y}^{(m)} > 0, \quad p_{y,x}^{(n)} > 0 \\ p_{xx}^{(s)} > 0$$

$$p_{yy}^{(n+m)} \geq p_{yx}^{(n)} p_{xy}^{(m)} > 0$$

$$p_{yy}^{(n+s+m)} \geq p_{yx}^{(n)} p_{xx}^{(s)} p_{xy}^{(m)} > 0$$

$d(y)$  divides both  $n+m$  and  $n+s+m$

$\Rightarrow d(y)$  divides every  $s$  with  $p_{xx}^{(s)} > 0$

$\Rightarrow d(y)$  divides gcd of such  $s$

$\Rightarrow d(y)$  divides  $d(x)$

Repeat by changing the role of  $x$  and  $y$

$d(x)$  divides  $d(y)$

$$d(x) = d(y)$$

Example

	M.C.	$S = \{1, 3, 3, 4\}$				tpm
	1	2	3	4		
1	0	0	1	0	}	
2	0	0	0	1		
3	0	1	0	0		
4	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{2}$		



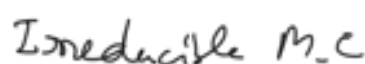
(2)



All states are recurrent. (using pg)

$$S = \{ \dots, -3, -1, 0, 1, 2, \dots \}$$

$$0 < p < 1$$



$$\sum_i p_{ii}^{(n)} = \sum_m a_m$$

$$p_{ii}^{(n)} = P(X_n = i | X_0 = i)$$

$$= \begin{cases} \binom{2m}{m} p^m q^m, & n=2m \\ 0, & n=2m+1 \end{cases}, m=1, 2, 3, \dots$$

$$= \begin{cases} a_m, & n=2m \\ 0, & n=2m+1 \end{cases}$$

1922-23

$$\frac{a_{m+1}}{a_m} = \frac{\binom{2m+2}{m+1} p^{m+1} q^{m+1}}{\binom{2m}{m} p^m q^m}$$

$$= \frac{(2m+2)(2m+1)}{(m+1)(m+1)} p q$$

$$\lim_{m \rightarrow \infty} \frac{a_{m+1}}{a_m} = 4 p q$$

$$= \begin{cases} 1 & p = \frac{1}{2} \\ < 1 & p \neq \frac{1}{2} \end{cases} \quad p \in (0,1)$$

Ratio test

$$\lim_{m \rightarrow \infty} \frac{a_{m+1}}{a_m} \begin{cases} < 1, \sum a_m \text{ converges} \\ > 1, \sum a_m \text{ diverges} \end{cases}$$

$$\text{If } p \neq \frac{1}{2}, \sum_m a_m = \sum_n p_{ii}^{(n)} < \infty \Rightarrow i \text{ transient state}$$

$$\text{Stirling approx } n! \sim n^{n+\frac{1}{2}} e^{-n} \sqrt{2\pi}$$

$$\text{where } a_n \sim b_n \text{ when } \lim_{n \rightarrow \infty} \frac{a_n}{b_n} = 1$$

$$p_{ii}^{(2n)} \sim \frac{(4p(1-p))^n}{\sqrt{\pi n}} \sim \frac{1}{\sqrt{\pi n}} \text{ if } p = \frac{1}{2}$$

$$\sum p_{ii}^{(2n)} = \sum_{n=1}^{\infty} \frac{1}{\sqrt{\pi n}^{1/2}} = \infty \quad i \text{ recurrent if } p = \frac{1}{2}$$

$$\sum_{n=1}^{\infty} \frac{1}{n^p} \left\{ \begin{array}{l} \text{converges if } p > 1 \\ \text{diverges if } p \leq 1 \end{array} \right.$$

All states are recurrent if  $p = \frac{1}{2}$

" " " transient if  $p \neq \frac{1}{2}$ .  
—X—

Gambler's Ruin Problem:  $i = 0, 1, \dots, N$

initial capital  $R_0 i$

aim  $R_N N$

$Z_i$   $i^{\text{th}}$  bet / step / transition / time

$$P(Z_i = 1) = p, P(Z_i = -1) = q = 1 - p$$

$Z_1, Z_2, \dots$  are independent

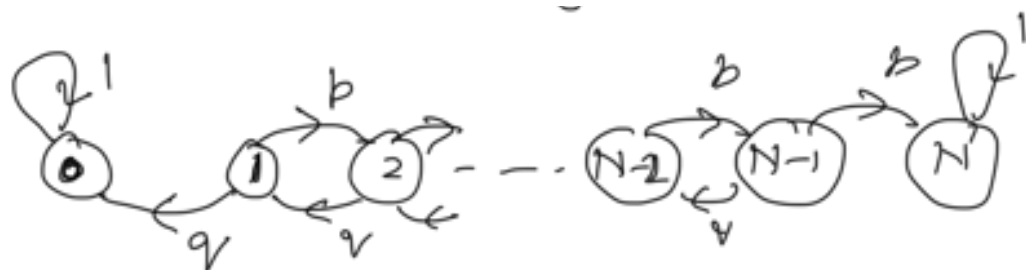
$X_n$ : Fortune of the gambler after  $n$  steps

$$X_n = i + Z_1 + Z_2 + \dots + Z_n \quad X_{n+1}$$

$$X_n \in \{0, 1, \dots, N\} = S \quad |X_n| \text{ is a M.C.}$$

$$p_{ij} = P(X_{n+1} = j | X_n = i) \quad i, j \in S$$

$$\begin{aligned} & \left[ \begin{array}{l} p_{00} = 1 = p_{NN} \\ p_{i,i+1} = p, p_{i,i-1} = q, i=1, \dots, N-1 \end{array} \right] \\ & \text{tpm} \quad P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & \dots & N \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ \vdots \\ N \end{matrix} & \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ q & 0 & p & 0 & \dots & 0 \\ 0 & q & 0 & p & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix} \end{matrix} \end{aligned}$$



Class  $\{0\} \quad \{1, 2, \dots, N-1\} \quad \{N\}$   
 recurrent or absorbing      transient      recurrent or absorbing

$T_0 =$  time he broke  $= \inf \{n : X_n = 0\}$

$T_N =$  time he has  $\$N = \inf \{n : X_n = N\}$

$P_i = P(T_N < T_0)$  Prob. starting with  $i$  units, the gambler's fortune will reach  $N$  before reaching 0.

$$P_i = P(T_N < T_0 | Z_1 = 1) \cdot P(Z_1 = 1) + P(T_N < T_0 | Z_1 = -1) \cdot P(Z_1 = -1)$$

$$P_i = P_{i+1} p + P_{i-1} q$$

$$\Rightarrow \underbrace{p P_{i+1} + q P_{i-1}}_{P_i} = P_{i+1} p + P_{i-1} q$$

$$\Rightarrow (P_{i+1} - P_i) p = q (P_i - P_{i-1})$$

$$\Rightarrow P_{i+1} - P_i = \frac{q}{p} (P_i - P_{i-1})$$

$$i=1 \quad P_2 - P_1 = \frac{q}{p} (P_1 - P_0) = \frac{q}{p} P_1$$

$$i=2 \quad P_3 - P_2 = \frac{q}{p} (P_2 - P_1) = \left(\frac{q}{p}\right)^2 P_1$$

$$\left. \begin{array}{l} P_0 = 0 \\ P_N = 1 \end{array} \right\}$$



$$P_i - P_{i-1} = \left(\frac{q}{p}\right)^{i-1} P_1$$

$$P_i - P_1 = \left[ \frac{q}{p} + \left(\frac{q}{p}\right)^2 + \dots + \left(\frac{q}{p}\right)^{i-1} \right] P_1$$

$$P_i = \left[ 1 + \frac{q}{p} + \left(\frac{q}{p}\right)^2 + \dots + \left(\frac{q}{p}\right)^{i-1} \right] P_1$$

$$P_i = \begin{cases} \frac{1 - \left(\frac{q}{p}\right)^i}{1 - \frac{q}{p}} P_1, & \frac{q}{p} \neq 1 \\ i P_1, & \frac{q}{p} = 1 \end{cases} \quad (1)$$

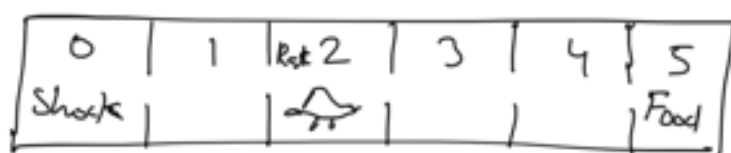
$$\underline{i=N} \text{ in } (1) \quad P_N = 1 \quad P_1 = \begin{cases} \frac{1 - \frac{q}{p}}{1 - \left(\frac{q}{p}\right)^N}, & \frac{q}{p} \neq 1 \\ \frac{1}{N}, & \frac{q}{p} = 1 \end{cases}$$

$$\Rightarrow P_i = \begin{cases} \frac{1 - \left(\frac{q}{p}\right)^i}{1 - \left(\frac{q}{p}\right)^N} & \text{if } \frac{q}{p} \neq 1 \Leftrightarrow p \neq \frac{1}{2} \\ \frac{i}{N} & \text{if } \frac{q}{p} = 1 \Leftrightarrow p = \frac{1}{2} \end{cases}$$

$N \rightarrow \infty$

$$P_i = \begin{cases} 1 - \left(\frac{q}{p}\right)^i & \text{if } \frac{q}{p} < 1 \Leftrightarrow p > \frac{1}{2} \\ i & \text{if } \frac{q}{p} \geq 1 \Leftrightarrow p \leq \frac{1}{2} \end{cases}$$

Example A rat is put into the linear maze as shown below



At each step the rat moves to the right with prob  $3/4$  and to the left with prob.  $1/4$ .

What is the prob. that the rat finds the food before getting shocked?

Sol. Gambler's ruin problem

$$i = 2, N = 5, p = \frac{3}{4}, q = \frac{1}{4}$$

$$\frac{q}{p} = \frac{1}{4} \times \frac{4}{3} = \frac{1}{3} \neq 1$$

$$P_2 = \frac{1 - \left(\frac{q}{p}\right)^2}{1 - \left(\frac{q}{p}\right)^5} = \frac{1 - \left(\frac{1}{3}\right)^2}{1 - \left(\frac{1}{3}\right)^5} = 0.892$$

$$\text{Prob. that rat will reach 0 before 5} = 1 - P_2 \\ = 1 - 0.892$$

2. The prob. of the thrower winning in the dice game called "Craps" is  $p = 0.49$ . Suppose Player A is the thrower and begins the game with \$5, and Player B, his opponent, begins with \$10. What is the probability that player A goes bankrupt

before Player B? Assume that the bet is \$1 per round.

Sol

$$i=5, N=15, p \leq 0.49, q=0.51, \frac{q}{p} \neq 1$$

$$1 - P_5 = 1 - \frac{1 - \left(\frac{q}{p}\right)^5}{1 - \left(\frac{q}{p}\right)^{15}} \quad i=5, N=15$$

$$=$$

$\begin{matrix} A \\ \$5 \end{matrix}$

$\begin{matrix} B \\ \$10 \end{matrix}$

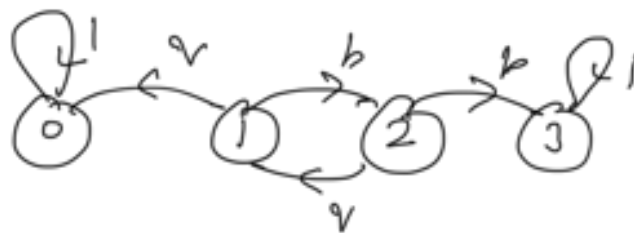
$$p^* = q^* = 0.51$$

$$p^* = q^* = 0.49$$

$$P_{10}^* = 1 - \frac{\left(\frac{q^*}{p^*}\right)^{10}}{1 - \left(\frac{q^*}{p^*}\right)^{15}}$$

$$\frac{q^*}{p^*} \neq 1$$

Example (1)



0, 3 → absorbing

1, 2 → transient

$$f_{1,0} = f_{2,0} = \sum_{n=0}^{\infty} f_{1,0}^{(n)}$$

$$f_{0,1} = 0 \leftarrow$$

$$f_{0,0} = 1$$

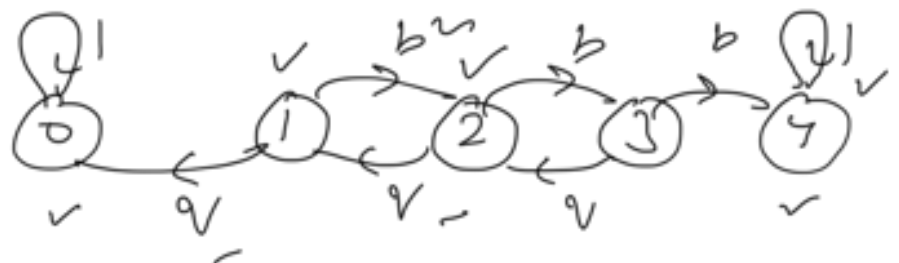
recurrent

$$f_{1,0} = f_{1,0}^{(1)} + f_{1,0}^{(2)} + f_{1,0}^{(3)} + \dots$$

transient recurrent

$$= q + 0 + p q^2 + 0 + (p q)^2 q + \dots$$

(2)



$$f_{2,1} = q + pv^2 + p^2v^3 + p^3v^4 + \dots$$

$$= q + pv(q + pv^2 + \dots)$$

$$= q + pv f_{2,1}$$

$$\Rightarrow f_{2,1} = \frac{q}{1-pv}$$

$$\text{If } p=0.4$$

$$f_{2,1} = \frac{0.6}{1-0.6 \times 0.4} = \frac{0.6}{0.76} = 0.78$$

using Gambler's ruin problem

$f_{2,1}^x \rightarrow$  start in 2, visit 1 before 4

$\rightarrow$  Start in 1 goes broke before reaching 3 \*

$$= 1 - \frac{1 - \left(\frac{0.6}{0.4}\right)^1}{1 - \left(\frac{0.6}{0.4}\right)^3} = 0.78$$

Mean time spent in transient states?

Finite state (S) M.C.  $\{X_n\}$

$T = \{1, 2, \dots, t\}$  set of transient states

$$P_T = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1t} \\ \vdots & \vdots & \ddots & \vdots \\ p_{t1} & p_{t2} & \dots & p_{tt} \end{bmatrix} \quad \text{sum of rows} < 1$$

$$i, j \in T$$

$\delta_{ij}$ : expected # of time periods that the M.C. is in state  $j$ , given that it starts in state  $i$ .

$$\text{Let } \delta_{ij} = \begin{cases} 1 & \text{when } i=j \\ 0 & \text{o.w.} \end{cases} ; I_{n,j} = \begin{cases} 1 & \text{if } X_n=j \\ 0 & \text{o.w.} \end{cases}$$

$$\delta_{ij} = \delta_{ij} + E\left(\sum_{n=1}^{\infty} I_{n,j} \mid X_0=i\right)$$

$$\begin{aligned} &= \delta_{ij} + \sum_{n=1}^{\infty} E(I_{n,j} \mid X_0=i) \\ &= \delta_{ij} + \sum_{n=1}^{\infty} P(X_n=j \mid X_0=i) \\ &= \delta_{ij} + \sum_{n=1}^{\infty} p_{ij}^{(n)} \rightarrow \textcircled{X} \end{aligned}$$

$$= \delta_{ij} + \sum_{n=1}^{\infty} \sum_k p_{ik} p_{kj}^{(n-1)}$$

$$= \delta_{ij} + \sum_k p_{ik} \sum_{n=1}^{\infty} p_{kj}^{(n-1)}$$

$$= \delta_{ij} + \sum_k p_{ik} \left[ \delta_{kj} + \sum_{n=2}^{\infty} p_{kj}^{(n-1)} \right]$$

$$= \delta_{ij} + \sum_k p_{ik} \left( \delta_{kj} + \sum_{n=1}^{\infty} p_{kj}^{(n)} \right)$$

$$= \delta_{ij} + \sum_k p_{ik} \underline{\delta_{kj}} \quad (\text{using } \textcircled{X})$$

$$= \delta_{ij} + \sum_{k=1}^t p_{ik} \delta_{kj} ,$$

since it is impossible to go from a recurrent to

transient state  $\Rightarrow \delta_{kj} = 0$  when  $k$  is a recurrent state

$$\boxed{\delta_{ij} = \delta_{ij} + \sum_{k=1}^t p_{ik} \delta_{kj}} \quad , i, j \in T$$

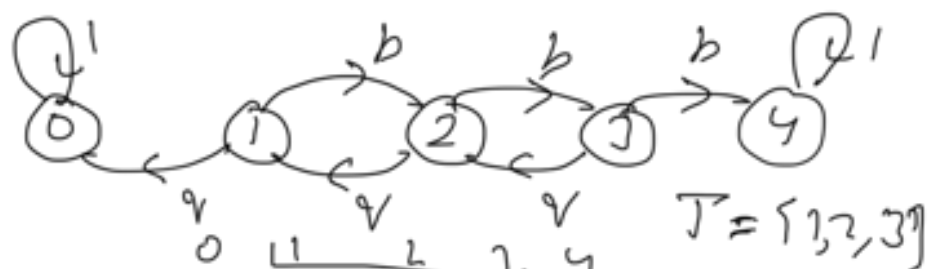
$$S = \begin{bmatrix} \delta_{11} & \delta_{12} & \dots & \delta_{1t} \\ - & - & - & - \\ \delta_{t1} & \delta_{t2} & \dots & \delta_{tt} \end{bmatrix} \quad , P_T$$

$$\Rightarrow S = I + P_T S$$

$$\Rightarrow (I - P_T) S = I$$

$$\Rightarrow \boxed{S = (I - P_T)^{-1}}$$

Example! (A) Gambler's ruin problem  $p=0.4, N=4$



$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0.6 & 0 & 0.4 & 0 & 0 \\ 0 & 0.6 & 0 & 0.4 & 0 \\ 0 & 0 & 0.6 & 0 & 0.4 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \end{matrix} \rightarrow P_T$$

$$I - P_T = \begin{pmatrix} 1 & -0.4 & 0 \\ -0.6 & 1 & -0.4 \\ 0 & -0.6 & 1 \end{pmatrix}$$

$$S = (I - P_T)^{-1} = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 1.46 & 0.76 & 0.31 \\ 1.15 & 1.92 & 1.71 \end{bmatrix} \end{matrix}$$

$$3 \left( \begin{array}{ccc} 0.69 & 1.15 & 1.46 \end{array} \right)$$

→ Starting with 2 units, determine the expected amt of time the gambler has 3 units

$$s_{23} = 0.76$$

→ " " 2 " , " " " " " " " " " " " " 1 unit

$$s_{21} = \underline{1.15}$$

$$i, j \in T$$

$f_{ij}$  : prob. that the M.C. ever makes a transition into state  $j$  given that it starts in state  $i$ .

$$s_{ij} = E(\text{time in } j \mid \text{start in } i)$$

$$= \underbrace{E(\text{time in } j \mid \text{start in } i, \text{ ever transit to } j)}_{s_{ij}} \cdot f_{ij} + E(\text{time in } j \mid \text{start in } i, \text{ never transit to } j) \cdot (1 - f_{ij})$$

$$= (\underbrace{s_{ij}}_{s_{ij}} + s_{jj}) \cdot f_{ij} + s_{ij} \cdot (1 - f_{ij})$$

$$= s_{ij} + f_{ij} s_{jj}$$

$$\Rightarrow f_{ij} = \frac{s_{ij} - s_{ij}}{s_{jj}} \quad \checkmark$$

Example A (cont) Starting with 2 units, what is the prob. that the gambler ever has a fortune of 1?

$$f \leftarrow s_{21} - s_{21} \quad 1.15 \quad \sim$$

$$2,1) = \frac{1}{s_{11}} = \frac{1.13 - 0}{1.46} = 0.78$$

$$|X_n| \text{ M.C. } S = \{0, 1, 2, \dots\}$$

Def<sup>n</sup> M.C. is said to have stationary prob. dist<sup>n</sup>

$$\underline{\pi} = (\pi_0, \pi_1, \dots) \text{ s.t.}$$

$$\underline{\pi} = \underline{\pi} P$$

$$\sum_{i=0}^{\infty} \pi_i = 1$$

$$P \rightarrow \text{TPM}$$

Def<sup>n</sup> M.C. have limiting probability,

$$\lim_{n \rightarrow \infty} p_j^{(n)} = \lim_{n \rightarrow \infty} P(X_n = j) = \pi_j \quad \forall j$$

Example

$$S = \{0, 1\}$$

$$P = \begin{pmatrix} \alpha & 1-\alpha \\ \beta & 1-\beta \end{pmatrix}$$

Stationary prob. dist<sup>n</sup>?

$$\underline{\pi} = (\pi_0, \pi_1)$$

$$\underline{\pi} = \underline{\pi} \hat{P} \Rightarrow \begin{aligned} \pi_0 &= \alpha \pi_0 + \beta \pi_1 \\ \pi_0 + \pi_1 &= 1 \end{aligned}$$

$$\pi_0 = \frac{\beta}{1+\beta-\alpha}, \quad \pi_1 = 1 - \pi_0 = \frac{1-\alpha}{1+\beta-\alpha}$$

Example:

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad S = \{1, 2\}$$



(i) Find stationary prob. dist

$$\underline{\pi} = (\pi_1, \pi_2)$$

$$\underline{\pi} = \underline{\pi} P \Rightarrow \begin{matrix} \pi_1 = \pi_2 \\ \pi_1 + \pi_2 = 1 \end{matrix} \Rightarrow \underline{\pi} = \left(\frac{1}{2}, \frac{1}{2}\right)$$

(ii)  $\underline{p}^{(0)} = (\alpha, 1-\alpha)$

$$\underline{p}^{(1)} = \underline{p}^{(0)} P = (\alpha, 1-\alpha) \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = (1-\alpha, \alpha)$$

$$\underline{p}^{(2)} = \underline{p}^{(1)} P = (\alpha, 1-\alpha)$$

Here limiting prob. dist. does not exist.

—X—

Ergodic M.C. M.C. irreducible, aperiodic (period 1)

and all states are positive recurrent.

→ Finite state M.C.  $(X_n)$  that is irreducible and aperiodic is ergodic

→ For ergodic M.C., the limiting prob. dist. and stationary prob. dist. are same and

$$\pi_j = \frac{1}{m_j} \quad \forall j, \quad \text{where } m_j = \sum_{n=1}^{\infty} n f_j^{(n)}$$

—X—

Example NCD (No Claim Discount) systems

NCD class	$E_0$	$E_1$	$E_2$
% discount	0	20	40
annual premium	100	80	60

Movement in the system is determined by the rule whereby one steps back one discount level (or stays in  $E_0$ ) with one claim in a year, and returns to a level of no discount if more than one claim is made. A claim-free year results in a step up to a higher discount level (or one remains in  $E_2$  if already there).

If we suppose that for anyone in this scheme the prob. of one claim in a year is 0.2 while the prob. of two or more claims is 0.1, Find

(i) tpm of system

(ii) In long run, what proportion of time is the process in each of the three discount classes

(iii) Find the av. annual premium paid.

$E_i \rightarrow i \text{ state}, i=0,1,2$

$$\begin{array}{c}
 \text{tpm} \\
 P = \begin{array}{c|ccc}
 & 0 & 1 & 2 \\
 0 & 0.3 & 0.7 & 0 \\
 1 & 0.2 & 0.7 & 0.1 \\
 2 & 0.1 & 0.1 & 0.8
 \end{array}
 \end{array}$$

$$2 \begin{pmatrix} 0.3 & 0 & 0.4 \\ 0.1 & 0.2 & 0.7 \end{pmatrix}$$

$S = \{0, 1, 2\}$ , irreducible,  
aperiodic M.C.

$\Rightarrow$  ergodic M.C.

$$\underline{\pi} = (\pi_0, \pi_1, \pi_2)$$

$$\left. \begin{array}{l} \underline{\pi} = \underline{\pi} P \\ \pi_0 + \pi_1 + \pi_2 = 1 \end{array} \right\} \Rightarrow \begin{array}{l} 0.3\pi_0 + 0.3\pi_1 + 0.1\pi_2 = \pi_0 \\ 0.7\pi_0 + 0.2\pi_2 = \pi_1 \\ \pi_0 + \pi_1 + \pi_2 = 1 \end{array}$$

$$\Rightarrow \underline{\pi} = (0.1860, 0.2442, 0.5698)$$

$$\text{su. annual premium} = 100 \times 0.1860 + 80 \times 0.2442 + 60 \times 0.5698 = 72.324$$

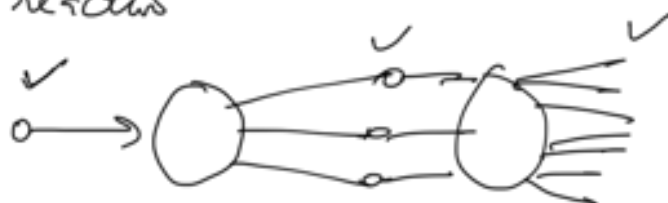
—X—

Branching Process:

electron multipliers



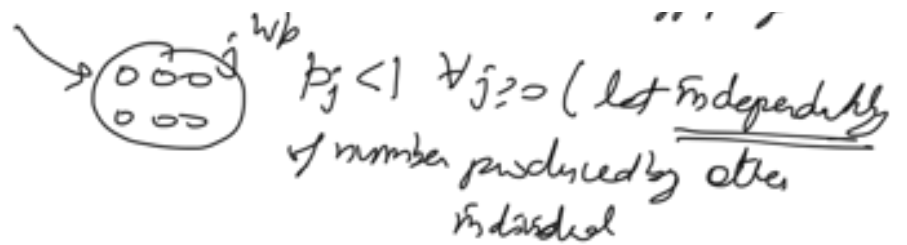
neutron chain reaction



Survival of family name

by the end of its lifetime

o each individual produce grow offspring



$X_0$  size of zeroth generation

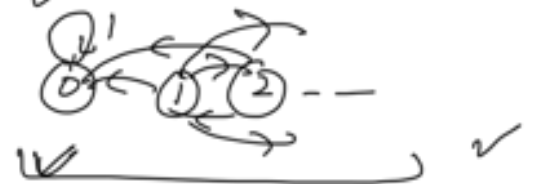
$X_1$  all offsprings of zeroth generation  
or first generation

$X_n$  size of  $n^{\text{th}}$  generation

$\{X_n : n \geq 0\}$  M.C.  $S = \{0, 1, 2, \dots\}$

$p_0 > 0$

State 0 recurrent  
all other states are transient



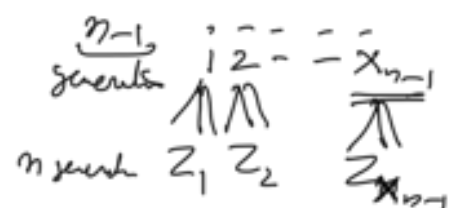
If  $p_0 > 0$  then the popn will either die out or its size will converge to  $\infty$ .

mean # of offsprings of a single individual  $\mu = \sum_{j=0}^{\infty} j p_j$

var  $\sigma^2 = \sum_{j=0}^{\infty} (j - \mu)^2 p_j$

Let  $X_0 = 1$

$$X_n = \sum_{i=1}^{X_{n-1}} Z_i$$



$Z_i$  # of offsprings of  $i^{\text{th}}$  individual of the  $(n-1)^{\text{st}}$  generation

$$\text{... or } E(Z_i) = \mu, V(Z_i) = \sigma^2$$

$$\begin{aligned} E(X_n) &= E(E(X_n | X_{n-1})) \\ &= E\left(E\left(\sum_{i=1}^{X_{n-1}} Z_i \mid X_{n-1}\right)\right) \\ &\quad \left[ \begin{array}{l} \xrightarrow{X_{n-1}\mu} \\ E\left(\sum_{i=1}^{X_{n-1}} Z_i \mid X_{n-1} \neq 0\right) \\ = E\left(\sum_{i=1}^2 Z_i\right) = 2\mu \end{array} \right] \\ &= E(X_{n-1}\mu) = \mu E(X_{n-1}) \\ &= \mu^n E(X_0) = \mu^n \end{aligned}$$

$$\begin{aligned} \checkmark \quad V(X_n) &= E\left(\underbrace{V(X_n | X_{n-1})}_{\sigma^2 X_{n-1}}\right) + V\left(\underbrace{E(X_n | X_{n-1})}_{\mu X_{n-1}}\right) \\ &= \sigma^2 E(X_{n-1}) + \mu^2 V(X_{n-1}) \\ &= \sigma^2 \mu^{n-1} + \mu^2 V(X_{n-1}) \quad \leftarrow \checkmark \\ &= \sigma^2 \mu^{n-1} + \mu^2 [\sigma^2 \mu^{n-2} + \mu^2 V(X_{n-2})] \\ &= \sigma^2 (\mu^{n-1} + \mu^n) + \mu^4 \underbrace{V(X_{n-2})}_{\sigma^2 \mu^{n-3} + \mu^2 V(X_{n-3})} \\ &= \sigma^2 (\mu^{n-1} + \mu^n + \mu^{n+1}) + \mu^6 V(X_{n-3}) \\ &\quad \dots \\ &= \sigma^2 (\mu^{n-1} + \mu^n + \dots + \mu^{2n-2}) + \mu^{2n} V(X_0) \quad \rightarrow 0 \end{aligned}$$

$$= \sigma^2 \mu^{n-1} (1 + \mu + \dots + \mu^{n-1})$$

$$V(X_n) = \begin{cases} \sigma^2 \mu^{n-1} \left( \frac{1-\mu^n}{1-\mu} \right) & \text{if } \mu \neq 1 \\ n \sigma^2 & \text{if } \mu = 1 \end{cases}$$

$$\begin{aligned} \underline{u_{n+1}} &= P(X_{n+1}=0) = \sum_j P(X_{n+1}=0 | X_1=j) p_j \\ &= \sum_j (P(X_n=0))^j p_j \\ &= \sum_j u_n^j p_j \end{aligned}$$

Let  $\pi_0$  denote the prob. that popl<sup>n</sup> will eventually die out (under the assumption that  $X_0=1$ )  
i.e., prob. of ultimate extinction

$$\pi_0 = \lim_{n \rightarrow \infty} P(X_n=0 | X_0=1)$$

Note that

→  $\pi_0=1$  if  $\mu < 1$ , since

$$\begin{aligned} \mu^n = E(X_n) &= \sum_{j=0}^{\infty} j P(X_n=j) \geq \sum_{j=1}^{\infty} 1 \cdot P(X_n=j) \\ &= P(X_n \geq 1) \end{aligned}$$

Since  $\mu^n \rightarrow 0$  if  $\mu < 1, n \rightarrow \infty$

∴  $\frac{n \rightarrow \infty}{\therefore P(X_n \geq 1) < \mu \rightarrow 0 \text{ as } n \rightarrow \infty}$

$$\dots \rightarrow P(X_n \leq 0) = 0$$

$$\therefore P(X_n = 0) = 1$$

$$\pi_0 = 1$$

→ It can be shown that  $\pi_0 = 1$  even when  $\mu = 1$

→ When  $\mu > 1$ , it turns out  $\pi_0 < 1$

$$\pi_0 = P(\text{popl}^n \text{ dies out}) = \sum_{j=0}^{\infty} \underbrace{P(\text{popl}^n \text{ dies out} | X_0 = j)}_{\text{thn } j \text{ total prob}} p_j$$

$$\boxed{\pi_0 = \sum_{j=0}^{\infty} \pi_0^j p_j} \quad \text{--- (1)}$$

It can be shown that  $\pi_0$  is the smallest +ve number satisfying equation (1).

Example (1) If  $\underline{p_0 = \frac{1}{2}}$ ,  $\underline{p_1 = \frac{1}{4}}$ ,  $\underline{p_2 = \frac{1}{4}}$ ,  $\underline{X_0 = 1}$   
 $\pi_0$ ?

$$\mu = \frac{3}{4} \leq 1 \Rightarrow \underline{\underline{\pi_0 = 1}}$$

$$\underline{\underline{\pi_0^h = 1}} \quad \underline{X_0 = n}$$

(2) If  $\underline{p_0 = \frac{1}{4}}$ ,  $\underline{p_1 = \frac{1}{4}}$ ,  $\underline{p_2 = \frac{1}{2}}$ ,  $\underline{X_0 = 1}$

$$\mu = \frac{1}{4} + 1 = \frac{5}{4} > 1$$

$$\pi = \sum_{j=0}^{\infty} \pi^j p_j$$

$$\pi = \frac{1}{4} + \pi + \pi^2$$

$$\Rightarrow 2\pi^2 - 3\pi + 1 = 0$$

$$\Rightarrow \pi = \frac{1}{2}, 1$$

$$\boxed{\pi_0 = \frac{1}{2}}$$

—X—

Transformed M.C

Example A pensioner receives  $\text{£}2$  (0,000) at the beginning of each month. The amt of money he needs to spend during a month is independent of the amt he has and is equal to  $i$  with prob  $p_i$ ,  $i = 1, 2, 3, 4$ ,  $\sum_{i=1}^4 p_i = 1$ . If the pensioner has more than 3 at the end of a month, he gives the amt greater than 3 to his son. If, after receiving his payment at the beginning of a month, the pensioner has a capital of 5, what is the prob. that his capital is ever less than 1 at any time within four months?

Sol.

$X_n$  amt the pensioner has at the end of month

$$S = \{1, 2, 3\}$$

$\swarrow$   $\searrow$   $\rightarrow$   
 less than or equal to 2    greater than or equal to 3  
 10.  $\uparrow$  1



transformed  $\begin{matrix} \checkmark & 1^* & 2^* & 3^* \\ & 1^* & 2^* & 3^* \end{matrix}$   $|X_n|$   $\infty$   $M.C.$

$$Q = \begin{matrix} 1^* \\ \checkmark 2^* \\ 3^* \end{matrix} \begin{bmatrix} 1 & 0 & 0 \\ p_3+p_4 & p_2 & p_1 \\ p_4 & p_3 & p_1+p_2 \end{bmatrix}$$

make  $\frac{2^*}{2+2}$   $\frac{2+2}{2}$   $\begin{pmatrix} 4,3 \end{pmatrix}$

absorbing state  
(supremacy of  $Q_{nn}$ )  $\begin{matrix} 1^* \\ (0,1) \end{matrix}$

$\begin{matrix} 2^* & 2+2 \\ 2^* & (2) \end{matrix}$

$\begin{matrix} 3^* & 3+3 \\ 3^* & (4) \end{matrix}$

$3^* \xrightarrow{3+2=5} 3,2$   
 $\xrightarrow{3} 3,1$   
 $(3,3,1)$

$Q^4 = Q^{(4)}$   
 $\xrightarrow{\text{Ans}} Q_{3^*,1^*}^{(4)} = Q_{3^*,1^*}^4$

—X—

Example Suppose that a production process changes states in accordance with a M.C. having transition probabilities  $p_{ij}$  ;  $i, j = 1, \dots, n$ ; and suppose that certain of the states are considered acceptable and the remaining unacceptable ( $A^c$ )

The production process is said to be "up" when in an  $A$  and "down" when in  $A^c$ . Determine

1. the rate of breakdowns
2. the average number of breakdowns per unit time

When it goes down, and

] " " " " - - - - up  
- - goes up.

Sol  $\prod_k, k=1, \dots, \infty$  limiting prob.

$$\underline{i \in A}, \underline{j \in A^c}$$

$$\text{rate of breakdown} = \sum_{j \in A^c} \sum_{i \in A} \pi_i p_{ij}$$

$\overline{U}$  ,  $\overline{D} \rightarrow$  av. time the process remains down when it goes down  
av. time the process remains up when it goes up

Single breakdown every  $\overline{U} + \overline{D}$  time

$$\text{rate at which breakdown occur} = \frac{1}{\overline{U} + \overline{D}}$$

$$\therefore \frac{1}{\dots} = \sum \sum \pi_i b_{i,j} \dots$$