

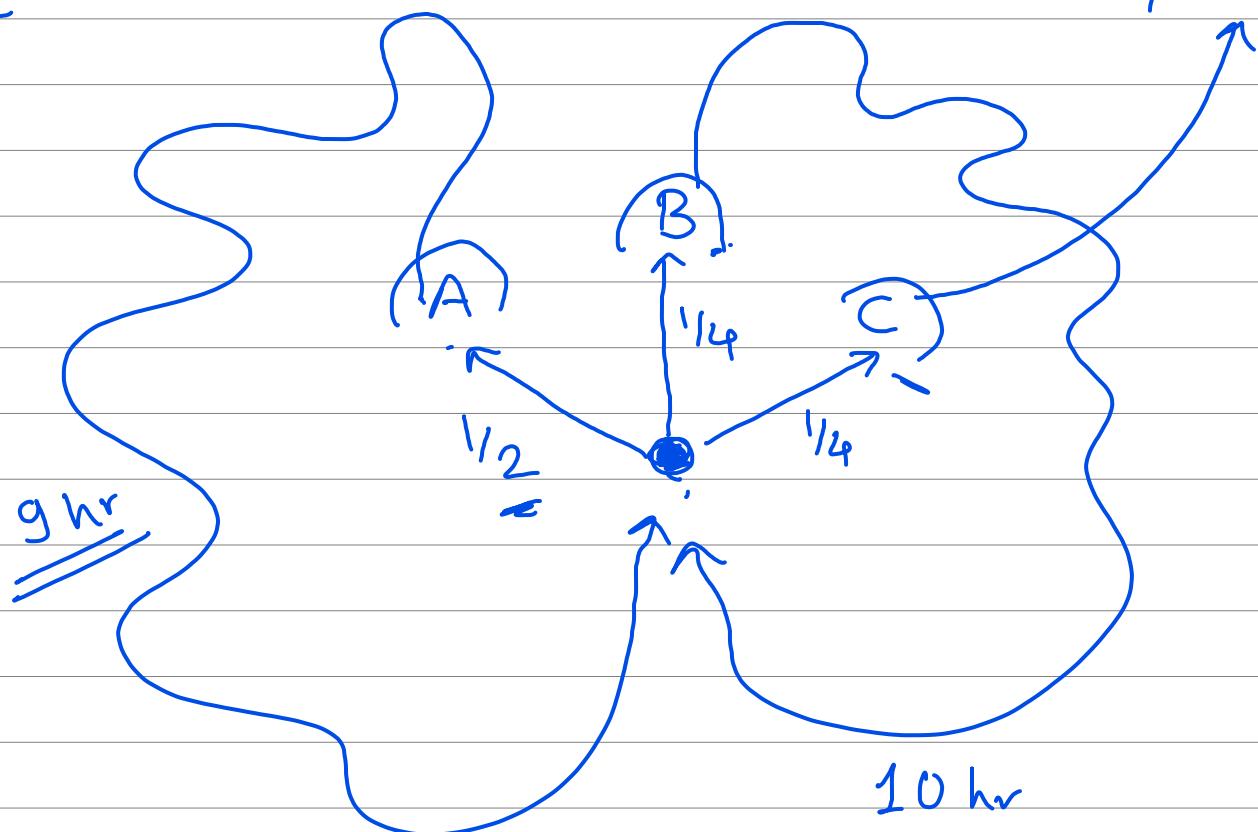
# ① Distributions Revision

Discrete

Rat

Rat Cave

Continuous  
outside  
7 hr



A  $\frac{1}{2}$   $\rightarrow$  9 hr return

B  $\frac{1}{4}$   $\rightarrow$  10 hr return

C  $\frac{1}{4}$   $\rightarrow$  7 hr leave

$E(X) = ?$

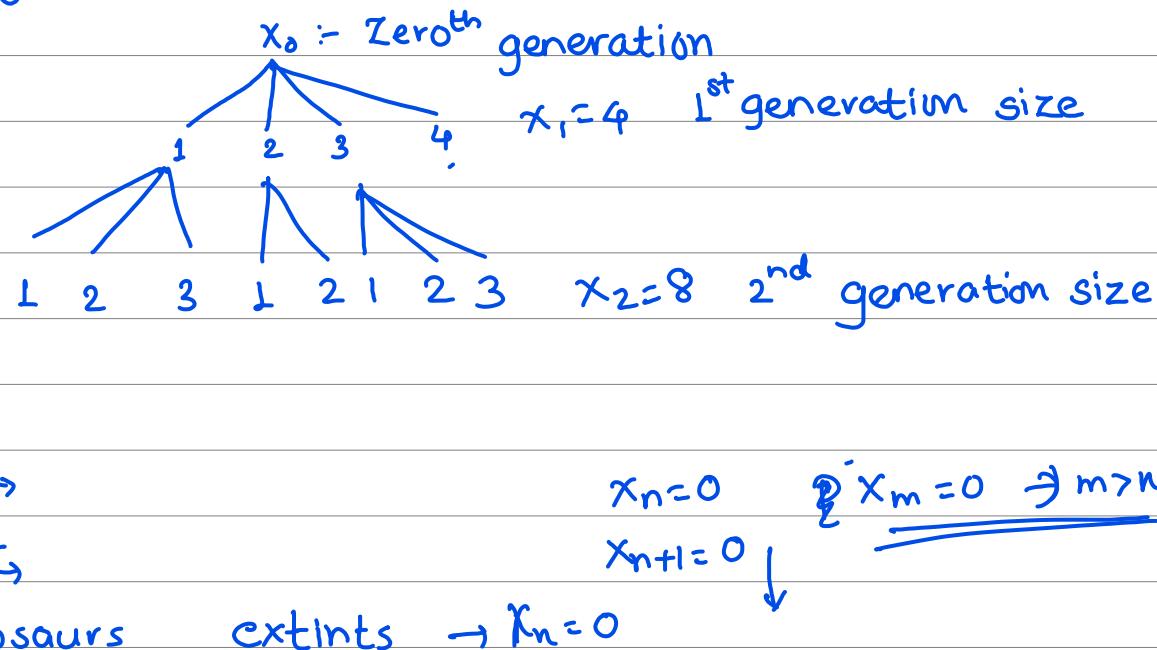
Assume

$$\begin{array}{l} A \frac{1}{2} \rightarrow \frac{g+\mu}{2} \\ B \frac{1}{4} \rightarrow \frac{10+\mu}{4} \\ C \frac{1}{4} \rightarrow 7 \end{array}$$

$$\mu = \frac{1}{2} \underline{\underline{(g+\mu)}} + \frac{1}{4} \underline{\underline{(10+\mu)}} + \underline{\underline{7}}$$

=

## Offspring Distribution:-



$$\underline{x_0 = 0} \quad x_1 = 3, \quad x_2 = 5$$

$$\lim_{n \rightarrow \infty} x_n = 0$$

Sum of expectations

$$y = \sum_{i=1}^n x_i \leftarrow \begin{array}{l} \text{fixed} \\ \text{Random} \end{array} \Rightarrow E(y) = n \cdot E(x) = \sum_{i=1}^n E(x_i)$$

$$\text{Random sum of Random number } v(y) = \sum_{i=1}^n v(x_i) + \sum_{i,j} \sum_{i \neq j} \text{cov}(x_i, x_j)$$

↳ compound distributions

$$Y \quad \textcircled{1} \quad N \sim \text{Bino} \rightarrow y \quad \text{Comp. Bino.-distribution} \quad \begin{matrix} (k, p) \\ E(y) = k \cdot p \cdot E(x) \end{matrix}$$

$$Y \quad \textcircled{2} \quad \sim \text{Pois} \sim \text{Comp. Pois. distribution} \quad E(y) = \lambda \cdot E(x)$$

likewise

$$\underline{E(Y)} = E\left(\sum_{i=1}^n x_i\right) = E_N\left(E_{Y|N}\left(\frac{\sum_{i=1}^n x_i}{N=n}\right)\right) \quad \begin{matrix} \text{Comp. Geometric, Negative Binomial} \\ E(x_i) = \mu \\ v(x_i) = v(x) \end{matrix}$$

$$v(x_i) = v(x) \quad E(x_i) = \mu \quad E(x_i) = E(x) = \mu \quad E(x_i) = E(x) = \mu$$

$$v(y) = v\left(\sum_{i=1}^n x_i\right) = E_N\left(\sum_{i=1}^n E(x_i)\right) = E_N(n \cdot E(x)) = \underline{E(x)} \cdot \underline{E(N)} \quad \checkmark$$

$$v(y) = E\left(v\left(\sum_{i=1}^n x_i\right)\right) + v\left(E\left(\sum_{i=1}^n x_i\right)\right)$$

$$\checkmark = E_N\left(N v(x)\right) + v\left(N \cdot E(x)\right) \Rightarrow \underline{E(N)} \underline{v(x)} + \underline{[E(x)]^2} \underline{v(N)}$$

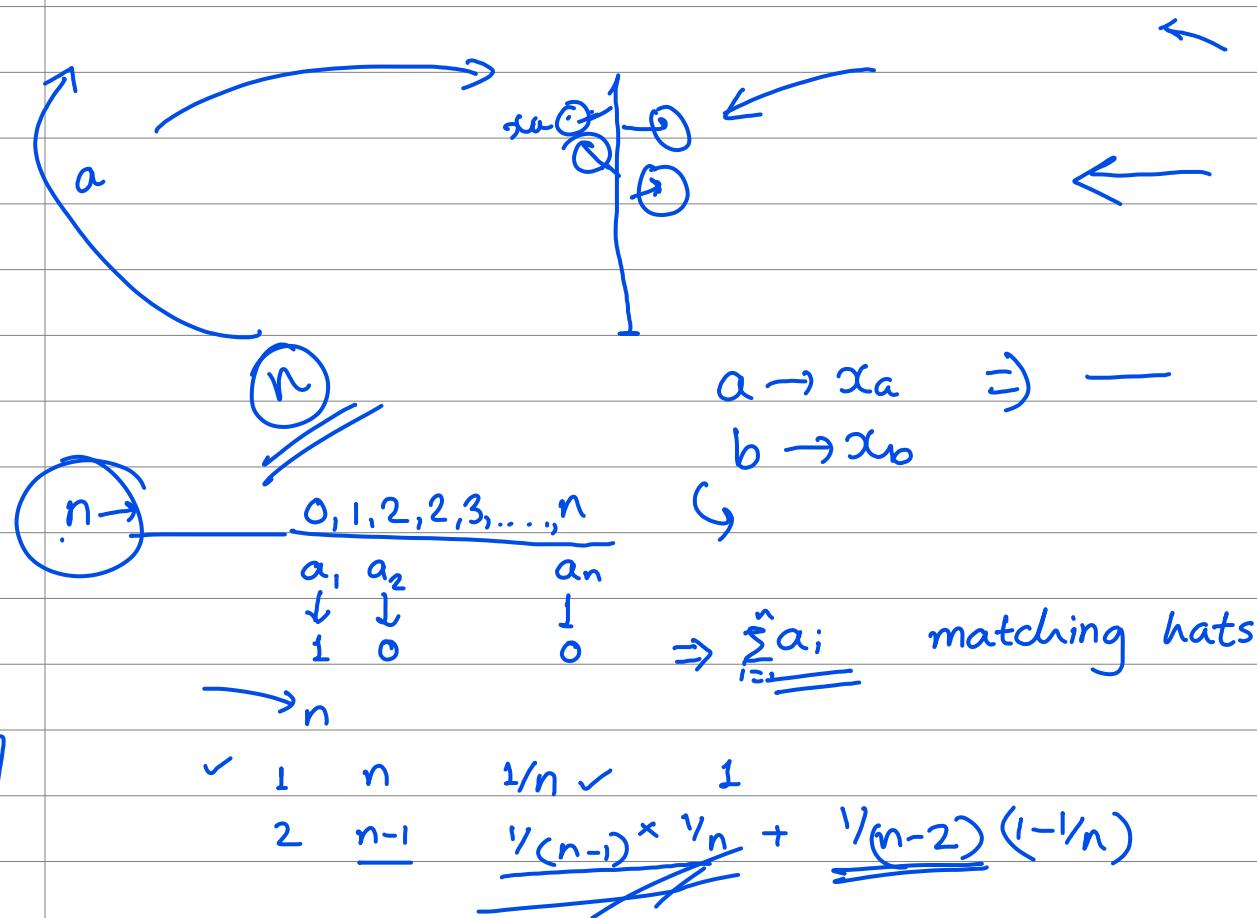
$$\text{Comp. Bino.} \Rightarrow V(Y) = kp \cdot V(X) + [E(X)]^2 (kpq) \quad \begin{cases} \Rightarrow \text{Comp. Poisson} \Rightarrow V(Y) = \lambda [V(X) + E(X)^2] \\ \Rightarrow \lambda \cdot E(X^2) \end{cases}$$

Manoj C Patil

$$V(X) = E(X^2) - [E(X)]^2$$

Lecture:

## Matching hat problem

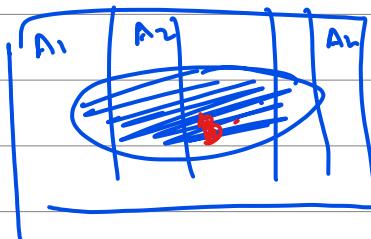


Bayes' Rule

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A_i|B) = \frac{P(B|A_i)}{\sum_{i=1}^n P(B|A_i)}$$

$$\sum A_i$$



$$\cup A_i = \Omega$$

e.g. Family 2 children

$$X_i = 0, \text{ if } i \text{ is female} \quad \frac{1}{2} \\ = 1, \text{ if } i \text{ is male} \quad - \frac{1}{2}$$

$$P\left[\sum_{i=1}^2 X_i = 2 / \sum_{i=1}^2 X_i \geq 1\right] = ?$$

$$E \quad \Omega = \{BB, BG, GB, GG\}$$

$$BA = \{ BB, BG, GB \} \checkmark$$

$$B = \{BB\}^*$$

$$P(B/A) = \frac{P(B \cap A)}{P(A)} = \frac{\frac{1}{4}}{\frac{3}{4}} = \frac{1}{3}$$

e.g. Maths → A →  $\frac{1}{2}$   
Stats → A →  $\frac{1}{3}$

$$P[A/\text{stats}] = 1/3$$

$$P[\text{Stats}] = \frac{1}{2}$$

$$P[\text{Stats}/A] = ?$$

$$\cdot P[A \text{ Maths}] = 1/2$$

$$P[\text{Maths}] = 1/2$$

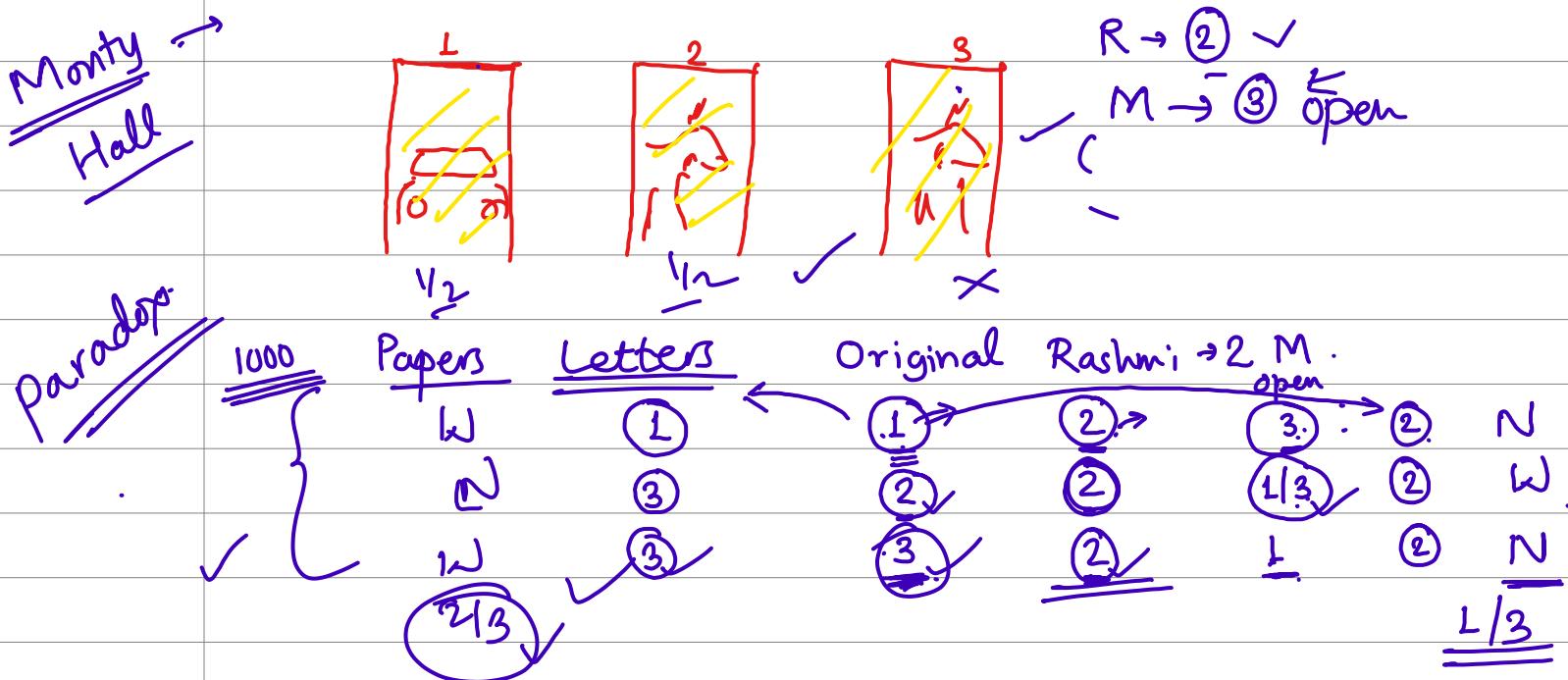
$$P[\text{Stats} | A] = \frac{P[\text{Stats}, A]}{P[\text{Stats}, A] + P[\text{Maths}, A]}$$

$$P[\text{Stats/A}] = \frac{\frac{1}{3}}{\frac{1}{3} + \frac{1}{2}} = \frac{2}{2+3} = \frac{2}{5}$$

# Monty Hall Problem ? Paradox ?

## Show

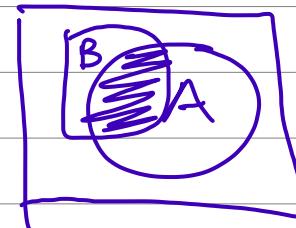
80



$$E(X|Y)$$

$$= V(X|Y) \cdot E(X) + E(X|Y) \cdot V(X)$$

$$P(B/A) = \frac{P(B \cap A)}{P(A)}$$



$X, Y$  ind //

$x/y$	1	2	3
prob	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

$x$	1	2	3
1			
2			
3			

$$E(X/S)$$

$$\underline{S=3} \rightarrow (x=1) \frac{1}{4}, (x=2) \frac{1}{2}, (x=3) \frac{1}{4}$$

$x$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$
$y$	1	2	3
1	<u><u><math>\frac{1}{4}</math></u></u>	<u><u><math>\frac{1}{2}</math></u></u>	<u><u><math>\frac{1}{4}</math></u></u>
2	<u><u><math>\frac{1}{2}</math></u></u>	<u><u><math>\frac{1}{4}</math></u></u>	<u><u><math>\frac{1}{4}</math></u></u>
3	<u><u><math>\frac{1}{4}</math></u></u>	<u><u><math>\frac{1}{4}</math></u></u>	<u><u><math>\frac{1}{4}</math></u></u>

$$P(S) = \frac{2}{16} = \frac{3}{16} - \frac{\frac{1}{4} \times \frac{1}{4} + \frac{1}{2} \times \frac{1}{2} + \frac{1}{4} \times \frac{1}{4}}{\frac{1}{16} + \frac{1}{4} + \frac{1}{16}}$$

$$= \frac{1}{16} + \frac{1}{4} + \frac{1}{16} = \frac{2}{16} + \frac{1}{4} = \frac{3}{8}$$

$$S = \frac{1}{16}, \frac{1}{4}, \frac{3}{8}, \frac{1}{8}, \frac{1}{16}$$

$$x = 1, 2, 3, 4, 5$$

$$x=1 \quad 1 \quad \frac{1}{2} \quad \parallel \quad \frac{1}{6}$$

$$x=2 \quad \quad \quad \frac{1}{2} \quad \frac{2}{3}$$

$$\frac{P(x=1, S=3)}{P(S=3)} = \frac{P(x=1, y=2)}{P(x=1) P(y=2)} = \frac{1/4}{1/2} = 1/2$$

$$= \frac{1/4 \times 1/2}{3/8} \approx$$

$$\cancel{\frac{P(x=2, S=3)}{P(S=3)}} = \cancel{\frac{1/4 \times 1/2}{3/8}}$$

$$\frac{P(x=1, S=4)}{P(S=4)} = \frac{1/4 \times 1/4}{3/8} = \frac{3/8}{3/8} = \frac{1/16}{3/8}$$

$$\frac{P(X=1, S=4)}{P(S=4)} = \frac{P(X=1), P(Y=3)}{3/8} = \frac{\frac{1}{4} \times \frac{1}{4} - 2}{3/8} = \frac{\frac{1}{3} \times \frac{1}{2}}{3/8} = \frac{1}{6}$$

$$\frac{P(X=2, S=4)}{P(S=4)} = \frac{P(X=2) \cdot P(Y=2)}{P(S=4)} = \frac{\frac{1}{2} \times \frac{1}{2}}{3/8} = \frac{\frac{1}{4} \times 1}{3/8} = \frac{2}{3}$$

$$\frac{1}{6} + \frac{2}{3} = \frac{1+4}{6} = \frac{5}{6} \rightarrow P(X=3|S=4)^{1/6}$$

$$E(X|S=2) = 1$$

$$E(X|S=3) = 1 \times \frac{1}{2} + 2 \times \frac{1}{2} = 1.5$$

$$E( \quad | S=4) = \underline{\quad}$$

↗ ↗  $E(X) = E[\cancel{E[X|Y]}] = E_Y [E_{X|Y}(X|Y)]$

MGF of compound distributions  $Y = \sum_{i=1}^N X_i$

$$M_Y(t) = E(e^{tY}) = E_N(E_{Y|N}(e^{tY}|N=n))$$

$$= E_N(E(e^{t \sum_{i=1}^N X_i} | N=n))$$

$X_i$  = i.i.d.

$$= E_N(\prod_{i=1}^N \underline{E(e^{tX_i})})$$

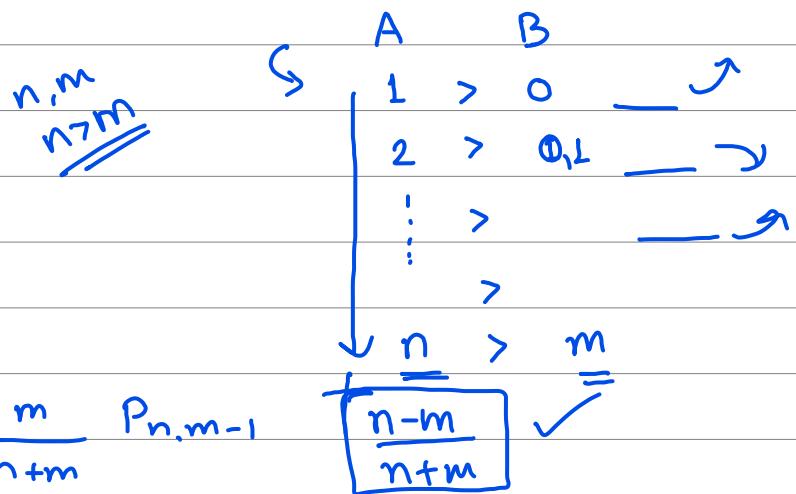
$$= E_N(\underline{[E(e^{tX})]^N})$$

$E(e^{tx})$   
 $E(x)$   
 $e^{\log t}$

$$M_Y(t) = E_N([M_X(t)]^N) = P_N(M_X(t))$$

$$= E_N(e^{N \cdot \log M_X(t)}) = M_N(\log M_X(t))$$

## Ballot Problem



$$P_{n,m} = \frac{n}{n+m} P_{n-1,m} + \frac{m}{n+m} P_{n,m-1}$$

$$= \frac{n}{n+m} \cdot \frac{(n-1)-m}{(n-1)+m} + \frac{m}{n+m} \cdot \frac{(n)-(m-1)}{n+m-1}$$

$$= \frac{n^2 - n - mn + mn - m^2 + m}{(n+m)(m+n-1)} = \frac{n^2 - n - m^2 + m}{(n+m)(m+n-1)}$$

$$= \frac{(n^2 - m^2) - (n-m)}{(n+m)(n+m-1)} = \frac{(n-m)(n+m) - (n-m)}{(n+m)(n+m-1)}$$

$$= \frac{(n-m)}{(n+m)} \frac{(n+m)}{(n+m)}$$

**Exercise 1.8**  $X_1 \sim \text{Poi}(\lambda_1)$   $X_2 \sim \text{Poi}(\lambda_2)$   $X_1 + X_2 \sim ?$

$$Z = X_1 + X_2$$

$$M_Z(t) = M_{X_1+X_2}(t) = M_{X_1}(t) \cdot M_{X_2}(t) = e^{\lambda_1(e^t-1)} \cdot e^{\lambda_2(e^t-1)}$$

$$= \frac{(e^{\lambda_1+e^{\lambda_2}})(e^t-1)}{e^t}$$

$$Z \sim \text{Poi}(\lambda_1 + \lambda_2)$$

Conditional distribution of  $X_1 / X_1 + X_2$

$$P(X_1=x / X_1 + X_2) = \frac{P(X_1=x, X_1+X_2=z)}{P(X_1+X_2=z)} = \frac{P(X_1=x, X_2=z-x)}{P(z)}$$

$$= \frac{P(X_1=x) \cdot P(X_2=z-x)}{P(z)}$$

$$= \frac{e^{-\lambda_1} \frac{\lambda_1^x}{x!}}{x!} \cdot \frac{e^{-\lambda_2} \frac{\lambda_2^{z-x}}{(z-x)!}}{(z-x)!} \Bigg/ \frac{e^{-(\lambda_1+\lambda_2)} (\lambda_1+\lambda_2)^z}{z!}$$

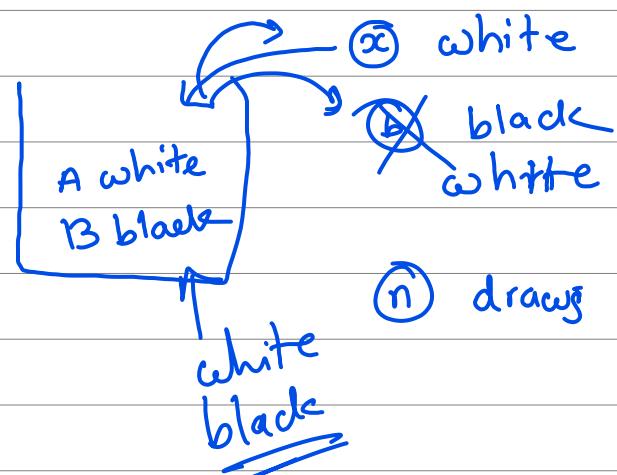
$$= \frac{z!}{x!(z-x)!} \left( \frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^x \left( \frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{z-x}$$

$$= Z_{Cx}$$

$$X_1 / X_1 + X_2 \sim \text{Bino} \left( z, \frac{\lambda_1}{\lambda_1 + \lambda_2} \right)$$

1.19 exercise

Urn

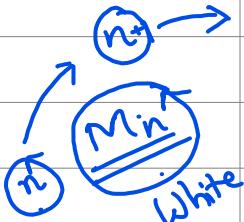


$$a \quad b \quad M_0 = a \quad \rightarrow$$

$$\begin{aligned} 1. \quad & \cancel{WB} \quad \cancel{W} \uparrow \quad \cancel{B} \uparrow \\ & M_1 = \underline{\underline{a}} \times \underline{\underline{\frac{a}{a+b}}} + (a+1) \underline{\underline{\frac{b}{a+b}}} = \underline{\underline{\frac{a(a+b)}{(a+b)}}} + \underline{\underline{\frac{b}{(a+b)}}} \\ & M \end{aligned}$$

 $M_0 = a$  white ball

$$M_1 = \underline{\underline{a}} \left( \frac{a}{a+b} \right) + (a+1) \left( \frac{b}{a+b} \right) \quad \rightarrow \underline{\underline{\bar{z} + \gamma}}$$



$$\begin{aligned} M_{n+1} &= \underline{\underline{M_n}} \left( \frac{?}{a+b} \right) + (M_n + 1) \left( \frac{(a+b) - ?}{a+b} \right) \\ &= M_n \cdot \left( \frac{M_n}{a+b} \right) + (M_n + 1) \left( \frac{(a+b) - M_n}{a+b} \right) \quad \leftarrow 1 - \frac{M_n}{a+b} \\ &= \cancel{\frac{M_n^2}{(a+b)}} + (M_n + 1) - \cancel{\frac{M_n^2}{(a+b)}} - \frac{M_n}{(a+b)} \end{aligned}$$

$$\textcircled{1} \quad \underline{M_{n+1}} = 1 + M_n \left( 1 - \frac{1}{a+b} \right) \quad \textcircled{2} \quad M_0 = a$$

$$M_n = a + b - b \left( 1 - \frac{1}{a+b} \right)^n$$

$$M_1 = 1 + a \left( 1 - \frac{1}{a+b} \right) = 1 + ax$$

$$M_2 = 1 + (1+ax) \cdot x = 1 + x + ax^2$$

$$\begin{aligned} M_3 &= 1 + (1+x+ax^2)x = 1 + x + x^2 + ax^3 \\ &= \underline{1+x+x^2+x^3} + (a-1)x^3 \end{aligned}$$

$$\begin{aligned} \sum_{n=0}^{\infty} x^n &= \frac{1}{1-x} \\ \sum_{n=0}^j x^n &= \frac{1-x^{j+1}}{1-x} \\ M_n &= \frac{(1-x^{n+1})}{1-x} + (a-1)x^n \\ x &= 1 - \frac{1}{a+b} \Rightarrow \frac{1}{1-x} = (a+b) \end{aligned}$$

$$\begin{aligned} M_n &= (a+b) - \frac{x^{n+1}}{1-x} + (a-1)x^n \\ &= (a+b) - (a+b)x^{n+1} + (a-1)x^n \\ &= (a+b) - x^n [(a+b) \cdot x - a+1] \\ &= (a+b) - \left( 1 - \frac{1}{a+b} \right)^n \left[ (a+b) \left( 1 - \frac{1}{a+b} \right) - a+1 \right] \end{aligned}$$



Sheldon Ross

$$= (a+b) - b \left( 1 - \frac{1}{a+b} \right)^n$$

(a+b) - x - a + x

J. Medhi ↴

## Stochastic Process

Seq<sup>n</sup> of Random Variable

✓  $X_n$  = fortune of gambler after  $n^{\text{th}}$  game

$$X_n = X_{n-1} + Z_n$$

$$Z_n = \begin{cases} +1 & \text{w.p. } p \\ -1 & \text{w.p. } 1-p \end{cases}$$

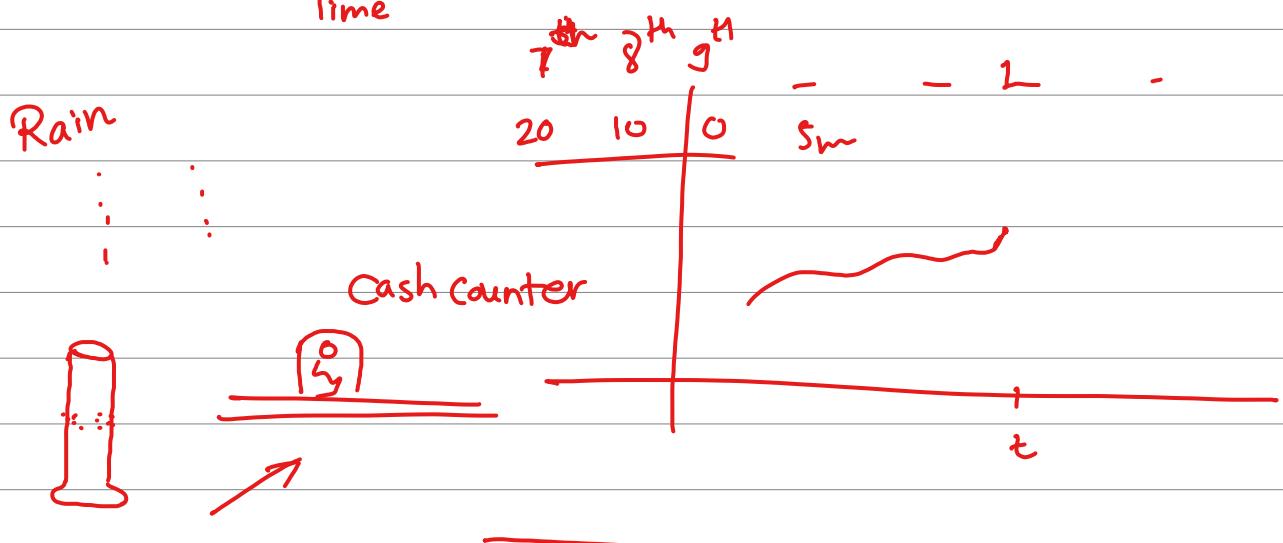
100  
99  
101

$x_0 \ x_1 \ x_2 \dots$

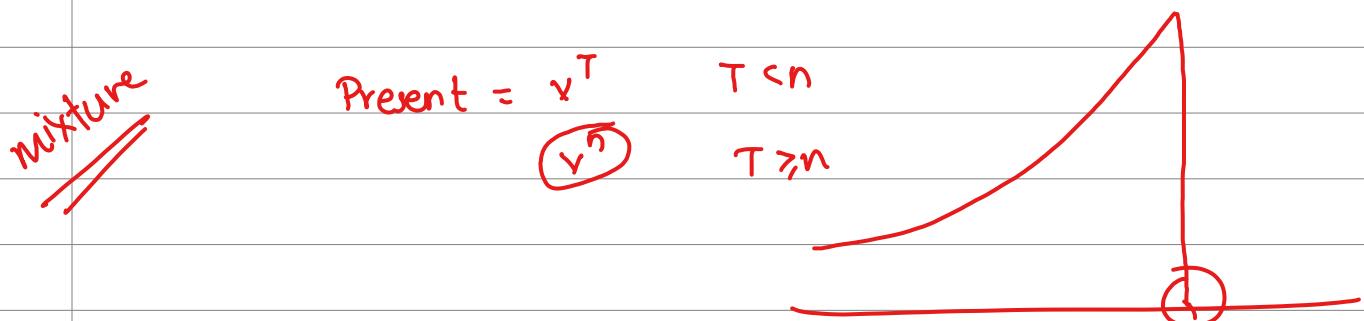
$x_n \dots$

Discrete Seq<sup>n</sup> of Random Variables  
 $\{x_i\}_{i=1}^{\infty}$

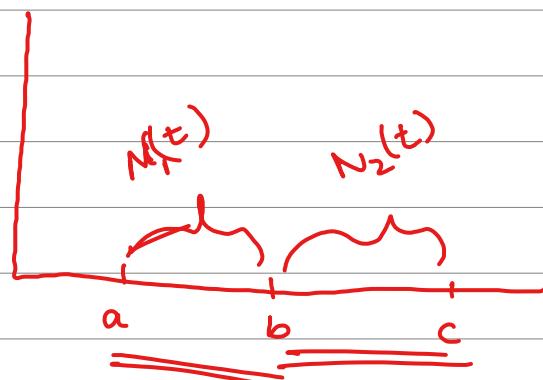
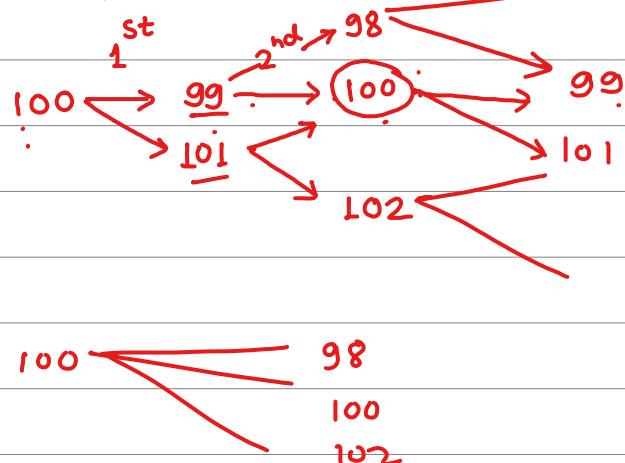
9.00 → 9.12, 9.12 → 9.15      g-30      3:30



	Statespace	Time domain	Example
1	Discrete	Discrete	* No. of covid patients on $n^{\text{th}}$ day * No. of customers on a shop - ..
2	Discrete	Conti	* Pop <sup>n</sup> size $\rightarrow$ continuous time discrete pop * No. of accidents up to time $t$
3	Continuous	Discrete	* Milk by cow on $n^{\text{th}}$ day * Min <sup>m</sup> /Max <sup>m</sup> temp on $n^{\text{th}}$ day
4	Continuous	Contin.	* Fever - body temp on $t$ time * Blood conc up to time $t$
5	Mixture	Discrete Cont	* Speed of internet * Rainfall up to time $t$



Independent Increments



disjoint time interval  
 $N_1(t)$  &  $N_2(t)$   
dist<sup>n</sup> indep

### independent increments

$$\begin{matrix} t_0 < t_1 < t_2 < \dots < t_{n-1} < t_n \\ \rightarrow \underline{x_{t_0}}, \underline{x_{t_1}}, \underline{x_{t_2}}, \dots, \underline{x_{t_{n-1}}}, \underline{x_{t_n}} \end{matrix}$$

$x_{t_1} - x_{t_0}, x_{t_2} - x_{t_1}, \dots, x_{t_n} - x_{t_{n-1}}$  indep. inc.

12-2

$$\frac{\underline{12-1}}{12.30 - 1.30}$$



Joint same  
Joint

$x_{t_0+u}$	$x_{t_1+u}$	$x_{t_2+u}$
$t_0$	$t_1$	$t_2$
$(x_{t_0}, x_{t_1}, x_{t_2},$		
$x_{t_1}-x_{t_0}$	$x_{t_2}-x_{t_1}$	

$x_{t_{n-1}}, x_{t_n})$

$x_{t_n}-x_{t_{n-1}}$

independent      Stationarity      Strictly

increment  $\underline{(s,t)}$        $\underline{x(t) - x(s)}$        $\underline{u > 0}$

✓ interval length  $(t-s)$        $\frac{s < t}{\downarrow}$        $\frac{12-2}{\downarrow}$  . No. of students

$(s+u, t+u]$        $\underline{x(t+u) - x(s+u)}$        $\underline{7-9}$ .

✓ interval length  $(t-s)$

$x_{t_0}$

weak stationary  $\Rightarrow \left\{ \begin{array}{l} E(x(t)) = \underline{ll} \text{ constant } \forall t \in I \\ \text{Cov}(x(t), x(s)) = \text{fun}(t-s) \end{array} \right.$



Find the p.g.f. of the sum  $S_n = X_1 + \dots + X_n$  of  $n$  independent and identical zero-truncated Poisson variates. Find  $E(S_n)$  and  $\Pr\{S_n = m\}$ ,  $m = n, n+1, n+2\dots$

$X_i \sim \text{Zero-truncated poisson}$

$$P[X=x] = \frac{e^{-\lambda} \lambda^x}{x!} / (1-e^{-\lambda}) \rightarrow E(X)$$

$$S_n = \sum_{i=1}^n X_i \geq E(S_n) = n \cdot E(X) = n \cdot \frac{\lambda}{1-e^{-\lambda}}$$

$$P_s(t) = E(t^S) = E(t^{\sum_{i=1}^n X_i}) = [P_X(t)]^n$$

$$\begin{aligned} P_X(t) &= \frac{1}{1-e^{-\lambda}} e^{-\lambda} \sum_{x=1}^{\infty} \frac{(\lambda t)^x}{x!} \\ &= \frac{e^{-\lambda}}{1-e^{-\lambda}} \left[ e^{\lambda t} \left( \sum_{x=0}^{\infty} \frac{e^{-\lambda t} (\lambda t)^x}{x!} \right) - e^{-\lambda t} \right] \end{aligned}$$

$$= \frac{e^{-\lambda}}{1-e^{-\lambda}} [e^{\lambda t} - e^{-\lambda t}]$$

## Markov Chains:-

discrete statespace & discrete time domain stochastic process following Markov property.

Markov Property :-

Future depends on present not on past

Notation

$$P_{ij}^{(n)} = P[X_n=j / X_0=i]$$



Starting from state  $i$   
process reaches to  
state  $j$  in  $n$ -steps

Stationary  $\rightarrow$  Time-homogeneous process

Non-stationary

$P_{ij}^{(m,n)}$  = Starting from  $i$  process reaches to state  $j$   
from  $m$  to  $n$  steps (No. of steps -  $n-m$ )

$\checkmark P_{ij}^{(m,n+m)} = P[X_{n+m}=j / X_m=i]$

statespace =  $S = \{1, 0\} \Rightarrow 1$  if it rains today 0 o.w.

0 0  
1 2 3

1 1  
8 9

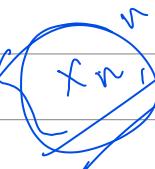
1  
15

0  
30

$\nearrow P_{10}^{(15)}$   
 $i j$

$\nearrow P_{10}^{(15,30)}$

$\cancel{\text{if } P_{ij}}$   $P[X_{n+1}=j / X_n=i] = P_{ij}^+ = P_{ij}$



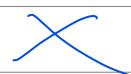
$P[X_{n+1}=j / X_n=i, X_{n-1}=i_{n-1}, \dots, X_1=i_1, X_0=i_0]$

$P[X_{30}=1 / \underline{X_{29}=1}, \underline{X_{28}=1}, \dots, \underline{X_0=0}] = P[X_{30}=1 / \underline{X_{29}=1}]$

Future

present

past



$P_{ij} = P[X_{n+1}=j | X_n=i, X_{n-1}=i_{n-1}, \dots, X_0=i_0] = P[X_{n+1}=j | X_n=i]$

↳  $\{X_n, n \geq 0\}$  as Markov chain.

one-step transition prob. matrix

0	1	← future state
Current state	0	$\begin{bmatrix} P_{00} & P_{01} \\ P_{10} & P_{11} \end{bmatrix}$
1		

$P_{00} \Rightarrow P[X_{n+1}=0 | X_n=0]$

$P_{10} \Rightarrow P[X_{n+1}=0 | X_n=1]$

starting ending

$P^{(2)}$  = Two step transition prob. matrix

↳

n-step

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

$P^{(2)}$

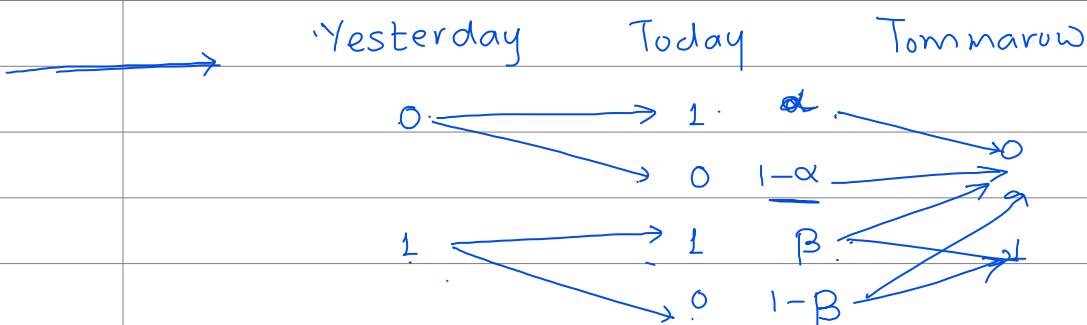
current state

future state

0	$P_{00}^{(2)}$	$P_{01}^{(2)}$
1	$P_{10}^{(2)}$	$P_{11}^{(2)}$

$P^{(n)}$

$P_{00}^{(n)}$	$P_{01}^{(n)}$
$P_{10}^{(n)}$	$P_{11}^{(n)}$

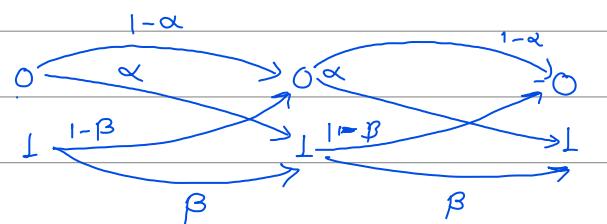


one-step

$$P = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1-\alpha & \alpha \\ 1-\beta & \beta \end{bmatrix}$$

Yesterday Today Tomorrow

Two-step

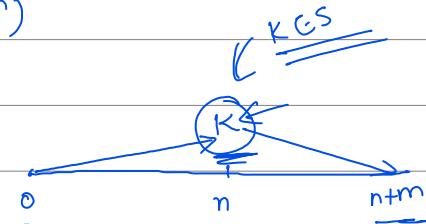


$$\begin{array}{c}
 \text{Today} \\
 \begin{matrix} 0 & 1 \\ 0 & P_{00}^{(2)} & P_{01}^{(2)} \\ 1 & P_{10}^{(2)} & P_{11}^{(2)} \end{matrix}
 \end{array}
 = P_{00} \cdot P_{00} + P_{01} P_{10} \\
 = (1-\alpha)^2 + \alpha(1-\beta)$$

n-step transition probabilities

Chapman-Kolmogorov's - (CK-equation)

$$P_{ij}^{(n+m)} = \sum_{k \in S} P_{ik}^{(n)} P_{kj}^{(m)}$$



$\{X_n, n \geq 0\}$  is Markov chain?

$$\begin{aligned}
 P_{ij}^{(n+m)} &= P[X_{n+m}=j / X_0=i] \\
 &= \sum_{k \in S} \underbrace{P[X_{n+m}=j / X_m=k]}_{\text{future}} \cdot \underbrace{P[X_m=k / X_0=i]}_{\text{present past}} \\
 &= \sum_{k \in S} \underbrace{P[X_{n+m}=j / X_m=k]}_{\downarrow} \cdot \underbrace{P[X_m=k / X_0=i]}_{\text{stationary processes}}
 \end{aligned}$$

$$= \sum_{k \in S} \underbrace{P[X_m=j / X_0=k]}_{\text{past}} \cdot \underbrace{P[X_m=k / X_0=i]}_{\text{present}}$$

$$= \sum_{k \in S} P_{ik}^{(n)} \cdot P_{kj}^{(m)}$$

if we have used ~~it~~ Matrix form

$$P^{(n+m)} = P^{(n)} \cdot P^{(m)}$$

if  $n=m=1$ ,

$$P^{(1+1)} = P^{(2)} = P^{(1)} \cdot P^{(1)}$$

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

$$= \begin{bmatrix} (1-\alpha)^2 + \alpha(1-\beta) & \alpha(1-\alpha) + \beta\alpha \\ (1-\alpha)(1-\beta) + \beta(1-\beta) & (1-\alpha)\beta + \beta^2 \end{bmatrix}$$

$\{x_n, n \geq 0\}$  denotes whether it rains on  $n^{\text{th}}$  day or not  
 $x_n = \begin{cases} 0 & \text{No rain} \\ 1 & \text{Rain} \end{cases}$

onestep transition  
 $x_{n+1} = \begin{cases} 0 & \text{Monday} \\ 1 & \text{Tuesday} \\ 2 & \text{Wednesday} \\ 3 & \text{Thursday} \\ 4 & \text{Friday} \\ 5 & \text{Saturday} \end{cases}$

Monday  $\Rightarrow 0$   
 $\begin{array}{c|cc} & \text{Monday} & \dots \\ \hline & 1 & \dots \\ \text{Tuesday} & \vdots & \ddots \\ & 5 & \end{array}$

fivestep  $P^{(5)}$

$$P[x_5 = 1 | x_0 = 1]$$

$$\begin{aligned} &= P^{(5)} = P^{(4)} \cdot P \\ &= P^{(3)} \cdot P^{(2)} \\ &= P^{(3)} \cdot \left[ \begin{array}{cc} 2/3 & 1/3 \\ 1/2 & 1/2 \end{array} \right] \left[ \begin{array}{cc} 2/3 & 1/3 \\ 1/2 & 1/2 \end{array} \right] \end{aligned}$$

$$\begin{aligned} &= P^{(3)} \cdot \frac{1}{6} \left[ \begin{array}{cc} 4 & 2 \\ 3 & 3 \end{array} \right] \cdot \frac{1}{6} \left[ \begin{array}{cc} 4 & 2 \\ 3 & 3 \end{array} \right] \\ &= \frac{1}{6^2} P^{(3)} \left[ \begin{array}{cc} 22 & 14 \\ 21 & 15 \end{array} \right] \end{aligned}$$

$$= \frac{1}{6^2} P \cdot P^{(2)} \cdot \left[ \begin{array}{cc} 22 & 14 \\ 21 & 15 \end{array} \right]$$

$$\begin{aligned} &= \frac{1}{6^4} P \cdot \left[ \begin{array}{cc} 22 & 14 \\ 21 & 15 \end{array} \right] \left[ \begin{array}{cc} 22 & 14 \\ 21 & 15 \end{array} \right] \\ &= \frac{1}{6^4} P \left[ \begin{array}{cc} 778 & 518 \\ 777 & 519 \end{array} \right] \xrightarrow{\text{Row sum}} \begin{array}{l} = 1 \\ \hline 1296 \end{array} \end{aligned}$$

$$= \frac{1}{6^4} \left[ \begin{array}{cc} 778 & 518 \\ 777 & 519 \end{array} \right] \times \frac{1}{6} \left[ \begin{array}{cc} 4 & 2 \\ 3 & 3 \end{array} \right]$$

$$= \frac{1}{6^5} \left[ \begin{array}{cc} 4666 & 3110 \\ 4665 & 3111 \end{array} \right]$$

5-step  
transition  
prob-matrix

$$P[x_5 = 1 | x_0 = 1] = \frac{3111}{6^5} = \underline{\underline{0.4}}$$

MC  $\{X_n, n \geq 0\}$  Markov chain with statespace  $S = \{0, 1, 2\}$

One-step TPM

$$P = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 1 & 0 & \frac{1}{3} & \frac{2}{3} \\ 2 & \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$$

initial prob:

$$\textcircled{1} \quad P[X_0=0] = P[X_0=1] = \frac{1}{2} \quad ?$$

$$\textcircled{2} \quad X_1=?$$

$$P[X_3=0]$$

$$P[X_3=1]$$

$$P[X_3=2]$$

$$E[X_3] = ?$$

$$P[X_1=1] = ? = P[X_1=1/X_0=0] \cdot P[X_0=0] + \left( \frac{1}{3} \times \frac{1}{2} + \right)$$

$$= \sum_k P[X_1=1/X_0=k] \cdot \underline{P[X_0=k]} = P[X_1=1/X_0=1] \cdot P[X_0=1] + \left( \frac{1}{3} \times \frac{1}{2} + \right)$$

$$P[X_1=1/X_0=2] \cdot P[X_0=2] \quad \frac{1}{2} \times 0$$

$$= \frac{1}{3}$$

$$P[X_1=?] = \underline{\alpha \cdot P} \quad \alpha = [\alpha_0 \ \alpha_1 \ \alpha_2]$$

$$= \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \frac{1}{6} \begin{bmatrix} 3 & 2 & 1 \\ 0 & 2 & 4 \\ 3 & 0 & 3 \end{bmatrix}$$

$$P[X_2=?] = \frac{1}{12} [3 \ 4 \ 5] \underbrace{P[X_1=0]}_{P[X_1=1]} \underbrace{P[X_1=2]}_{4/12} \underbrace{P[X_1=2]}_{5/12}$$

$$P[X_3=?] \Rightarrow [P(X_3=0) \ P(X_3=1) \ P(X_3=2)] = \alpha \cdot P^3$$

$$= \frac{1}{2} \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \frac{1}{6} \begin{bmatrix} 3 & 2 & 1 \\ 0 & 2 & 4 \\ 3 & 0 & 3 \end{bmatrix} \cdot P^2 = \frac{1}{12} [3 \ 4 \ 5] P^2$$

$$= \frac{1}{12} [3 \ 4 \ 5] \frac{1}{6} \begin{bmatrix} 3 & 2 & 1 \\ 0 & 2 & 4 \\ 3 & 0 & 3 \end{bmatrix} P$$

$$\frac{9}{15} \quad \frac{6}{8} \quad \cancel{3+16+15}$$

$$= \frac{1}{72} [24 \ 14 \ 34] \frac{1}{6} \begin{bmatrix} 3 & 2 & 1 \\ 0 & 2 & 4 \\ 3 & 0 & 3 \end{bmatrix}$$

$$+ \frac{24}{56}$$

$X_3$

$$= \frac{1}{432} [174 \ 76 \ 182] \checkmark$$

$$E(X_3) = \frac{1}{432} [0 \times 174 + 1 \times 76 + 2 \times 182]$$

$$= \frac{1}{432} [76 + 364] = \frac{1}{432} [440] = 1.0885$$

Let the TPM of two state markov chain is  $P = \begin{bmatrix} P & 1-P \\ 1-P & P \end{bmatrix}$

Show that, by mathematical induction

$$P^{(n)} = \begin{bmatrix} \frac{1}{2} + \frac{1}{2}(2P-1)^n & \frac{1}{2} - \frac{1}{2}(2P-1)^n \\ \frac{1}{2} - \frac{1}{2}(2P-1)^n & \frac{1}{2} + \frac{1}{2}(2P-1)^n \end{bmatrix}$$

Put  $n=1$ .

$$P^{(1)} = P = \begin{bmatrix} \frac{1}{2} + \frac{1}{2}(2P-1)^1 & \frac{1}{2} - \frac{1}{2}(2P-1)^1 \\ \frac{1}{2} - \frac{1}{2}(2P-1)^1 & \frac{1}{2} + \frac{1}{2}(2P-1)^1 \end{bmatrix} = \begin{bmatrix} P & 1-P \\ 1-P & P \end{bmatrix}$$

Assume that it is true for  $n=k$

$$P^{(k)} = \begin{bmatrix} \frac{1}{2} + \frac{1}{2}(2P-1)^k & \frac{1}{2} - \frac{1}{2}(2P-1)^k \\ \frac{1}{2} - \frac{1}{2}(2P-1)^k & \frac{1}{2} + \frac{1}{2}(2P-1)^k \end{bmatrix}$$

$$\begin{aligned} P^{(k+1)} &= P^{(k)} \cdot P = \left[ \begin{array}{cc} & \begin{bmatrix} P & 1-P \\ 1-P & P \end{bmatrix} \\ \begin{bmatrix} P(\frac{1}{2} + \frac{1}{2}(2P-1)^k) + (1-P)(\frac{1}{2} - \frac{1}{2}(2P-1)^k) & (1-P)(\frac{1}{2} + \frac{1}{2}(2P-1)^k) + P(\frac{1}{2} - \frac{1}{2}(2P-1)^k) \\ P(\frac{1}{2} - \frac{1}{2}(2P-1)^k) + (1-P)(\frac{1}{2} + \frac{1}{2}(2P-1)^k) & (1-P)(\frac{1}{2} - \frac{1}{2}(2P-1)^k) + P(\frac{1}{2} + \frac{1}{2}(2P-1)^k) \end{array} \end{bmatrix} \end{aligned}$$

$$= \begin{bmatrix} \frac{1}{2} + \frac{1}{2}(2P-1)^{k+1} & \frac{1}{2} - \frac{1}{2}(2P-1)^{k+1} \\ \frac{1}{2} - \frac{1}{2}(2P-1)^{k+1} & \frac{1}{2} + \frac{1}{2}(2P-1)^{k+1} \end{bmatrix}$$

$$S = \{1, 2\} \quad P = \frac{1}{2} \begin{bmatrix} 2/3 & 1/3 \\ 1/3 & 2/3 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \quad P_{12}^{(3)-} = ?$$

$$\begin{aligned} P^{(3)} &= P^{(2)} \cdot P = \frac{1}{3^2} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} P \\ &= \frac{1}{3^2} \begin{bmatrix} 5 & 4 \\ 4 & 5 \end{bmatrix} \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \\ &= \frac{1}{3^3} \begin{bmatrix} 14 & 13 \\ 13 & 14 \end{bmatrix} \end{aligned}$$

$$(P^{(3)})_{12} = \frac{13}{27} = 0.48$$

$$P = \frac{1}{2} \begin{bmatrix} 0 & 1 \\ 0.7 & 0.3 \\ 0.5 & 0.5 \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix}, P_{11}^{(4)} = ?$$

$$P^{(2)} = \frac{1}{10^2} \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix} \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix} = \frac{1}{10^2} \begin{bmatrix} 64 & 36 \\ 60 & 40 \end{bmatrix}$$

$$\begin{aligned} P^{(4)} &= P^{(2)} \cdot P^{(2)} = \frac{1}{10^4} \begin{bmatrix} 64 & 36 \\ 60 & 40 \end{bmatrix} \begin{bmatrix} 64 & 36 \\ 60 & 40 \end{bmatrix} \\ &= \frac{1}{10^4} \begin{bmatrix} 6256 & 3744 \\ 6240 & \boxed{3760} \end{bmatrix}. \end{aligned}$$

$$P_{11}^{(4)} = \frac{\boxed{3760}}{10^4}$$

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

$$\underline{\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}} \stackrel{P^4}{=} \underline{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_1$$

$$\frac{1}{10^2} \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix} P^2 \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}_1$$

$$\Rightarrow \frac{1}{10^2} \begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix} P^2 \begin{bmatrix} 3 \\ 5 \end{bmatrix} = \frac{5}{10^4} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 7 & 3 \\ 7 & 3 \\ 5 & 5 \end{bmatrix} \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

$$= \frac{5}{10^4} [12 \ 8] \begin{bmatrix} 36 \\ 40 \end{bmatrix}$$

$$= \frac{5}{10^4} [752] = \frac{3760}{10^4}$$

$$P = \frac{1}{6} \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 1 \\ 1 & 2 & 3 \\ 3 & 2 & 1 \end{bmatrix} \times \frac{1}{6}$$

$P_{2,3}^{(3)}$

$$= \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} P^3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$= [3 \ 2 \ 2] P \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$$

$$= \frac{1}{6^3} [3 \ 2 \ 2] \begin{bmatrix} 2 & 3 & 1 \\ 1 & 2 & 3 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$$

$$= \frac{1}{6^3} [14 \ 17 \ 11] \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$$

$$= \frac{1}{6^3} [76] = \frac{76}{216}$$

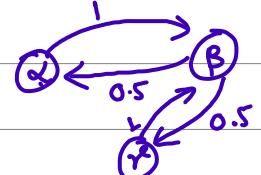
$C(\omega) = \{\alpha, \beta, \gamma\}$  Irreducible

$\alpha \neq \beta \neq \gamma$

$$\textcircled{a} \quad [0.25 \ 0.25 \ 0.5]$$

$$\xrightarrow{x_0} = \frac{1}{4} [1 \ 1 \ 2]$$

$$\alpha \begin{bmatrix} \alpha & \beta & \gamma \\ 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \end{bmatrix}$$



$$\textcircled{b} \quad P[X_4 = \gamma] = ? \Rightarrow \frac{1}{4} [1 \ 1 \ 2] P^4 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

✓

P. 14

$$P[(\textcircled{X}_1 = 1) / (\textcircled{X}_2 = 2)] \Rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & \checkmark & \\ \end{bmatrix} P^2 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad \checkmark$$

Classification

✓ Periodicity is class property ie.  $i \leftrightarrow j \Rightarrow d(i) = d(j)$

$\rightarrow i \rightarrow j \quad p_{ij}^n > 0 \dots \text{for some } n$

$j \rightarrow i \quad p_{ji}^m > 0 \quad \text{for some } m$

$$d(i) = \underline{\gcd} \{ t / p_{ii}^t > 0 \} \quad d(j) = \underline{\gcd} \{ t / p_{jj}^t > 0 \}$$

$$\begin{aligned} p_{ii}^{(n+m)} &= \sum_{k \in S} p_{ik}^n p_{ki}^m \\ &\geq p_{ij}^n p_{ji}^m \\ &> 0 \end{aligned}$$

$$\begin{aligned} p_{jj}^{(n+m)} &= \sum_{k \in S} p_{jk}^m p_{kj}^n \\ &\geq p_{ji}^m p_{ij}^n \\ &> 0 \end{aligned}$$

$\Rightarrow \underline{n+m}$  is multiple of  $d(i) \Rightarrow n+m = \underline{d(i)} \times \text{something}$

Similarly,  $p_{jj}^{(n+m)} > 0 \Rightarrow n+m$  is multiple of  $d(j) \Rightarrow n+m = \underline{d(j)} \times \text{something}$

✓ for some s,  $p_{ii}^{(s)} > 0 \Rightarrow s = \underline{d(i)} \times \text{something}$  —①

$$\begin{aligned} p_{jj}^{(n+m+s)} &\geq p_{ji}^{(m)} p_{ii}^{(s)} p_{ij}^{(n)} \\ &> 0 \quad > 0 \quad > 0 \\ &> 0 \end{aligned}$$

$\Rightarrow \underline{n+m+s} \neq \underline{d(j)} \times \text{something}$

$\Rightarrow s = \underline{d(j)} \times \text{something}$  —② (as  $d(j)$  divides  $n+m$ )

From ① & ②  
 $\Rightarrow d(i) = d(j)$

Transitivity :-  $i \leftrightarrow j, j \leftrightarrow k \Rightarrow i \leftrightarrow k$

$i \leftrightarrow j \quad P_{ij}^n > 0, P_{ji}^m > 0 \quad \text{for some } n, m$

$j \leftrightarrow k \quad P_{jk}^s > 0, P_{kj}^t > 0 \quad \text{for some } s, t$

To show  $i \leftrightarrow k \Rightarrow$

$$P_{ik}^{n+s} \geq P_{ij}^n P_{jk}^s \\ > 0 \quad > 0$$

$$P_{ki}^{t+m} \geq P_{kj}^t P_{ji}^m \\ > 0 \quad > 0$$

$$\begin{matrix} i \rightarrow k \\ \vdots \\ \rightarrow \end{matrix} \quad i \leftrightarrow k$$

$$k \rightarrow i$$

Recurrent

Starting from state  $i$ ,

Process reenters in the same state with probability 1

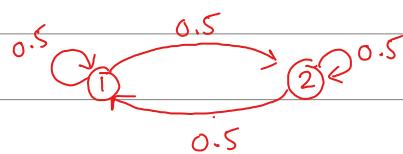
$$\begin{array}{l} \text{① } F_i = 1 \\ \text{② } \sum P_{ii} < 0 \end{array}$$

For any state  $i$ ,  $F_i$  denote the probability that the process, starting from state  $i$ , will ever reaches state  $i$ .

State  $i$  is said to recurrent if  $F_i = 1$ , & transient if  $F_i < 1$

$f_i^{(n)}$   $\Rightarrow$  Prob. that starting from state  $i$  process reaches to state  $i$  in  $n$  steps first time

$$F_i = \sum_n f_i^n$$



$$f_1^1 = 0.5$$

$$f_1^2 = 0.5^2$$

$$f_1^3 = 0.5^3$$

$$\Rightarrow f_1^n = 0.5^n$$

~~1-1-1~~

~~1-2-1~~

~~1-2-2-1~~

$$F_i = \sum_n f_i^n \Rightarrow = 0.5 + 0.5^2 + \dots$$

$$= 0.5 (1 + 0.5 + 0.5^2 + \dots)$$

$$= 0.5 \left( \frac{1}{1-0.5} \right) = 0.5 \cdot \frac{1}{0.5} = 1$$

1 & 2 recurrent



$$f_1^{(1)} = 0$$

$$f_1^{(2)} = 1$$

$$f_1^{(3)} = 0$$

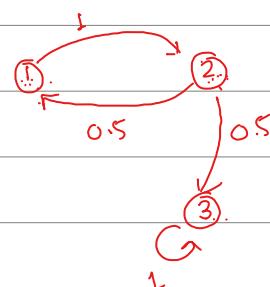
$$F_1 = f_1^{(1)} + f_1^{(2)} + 0 \dots + 0$$

$$= 0 + 1 + 0$$

$$= 1$$

$\Rightarrow 1 \& 2$  recurrent

Reducible



$$c(1) = \{1, 2\} = c(2)$$

$$c(3) = \{3\}$$

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

$$d(3) = 1$$

$$= 2$$

$$= d(2)$$

(Periodicity is class property)

$$\begin{aligned} f_2^1 &= 0, & f_2^2 &= 0.5, & f_2^3 &= 0, & f_2^4 &= 0 \dots \\ f_1^1 &= 0 & f_1^2 &= 0.5 & f_1^3 &= 0 & f_1^4 &= 0 \dots \end{aligned}$$

$F_2 = 0.5 < 1 \Rightarrow 1 \& 2$  are transient states

$$F_3 = f_3^{(1)} + f_3^{(2)} + f_3^{(3)} + \dots = 1 + 0 + 0 + \dots = 1 \Rightarrow 3 \text{ is recurrent.}$$

persistent

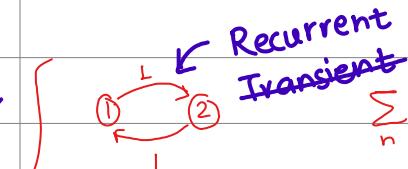
$P_{ii}^{(n)}$   $\rightarrow$  Starting from i, process reaches to state i, in n steps

To prove

i recurrent iff  $\sum_n P_{ii}^{(n)} = \infty$

Transient iff  $\sum_n P_{ii}^{(n)} < \infty$

$c(1) = \{1, 2\}$   
 $c(2) = \{1, 2\}$   
Irreducible MC

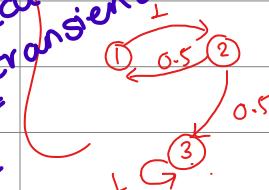


$$\begin{aligned} \sum_n P_{11}^{(n)} &= P_{11}^1 + P_{11}^2 + P_{11}^3 + \dots \\ &= 0 + 1 + 0 + 1 + \dots \\ &= \infty \end{aligned}$$

$$= \sum_n P_{11}^{2n}$$

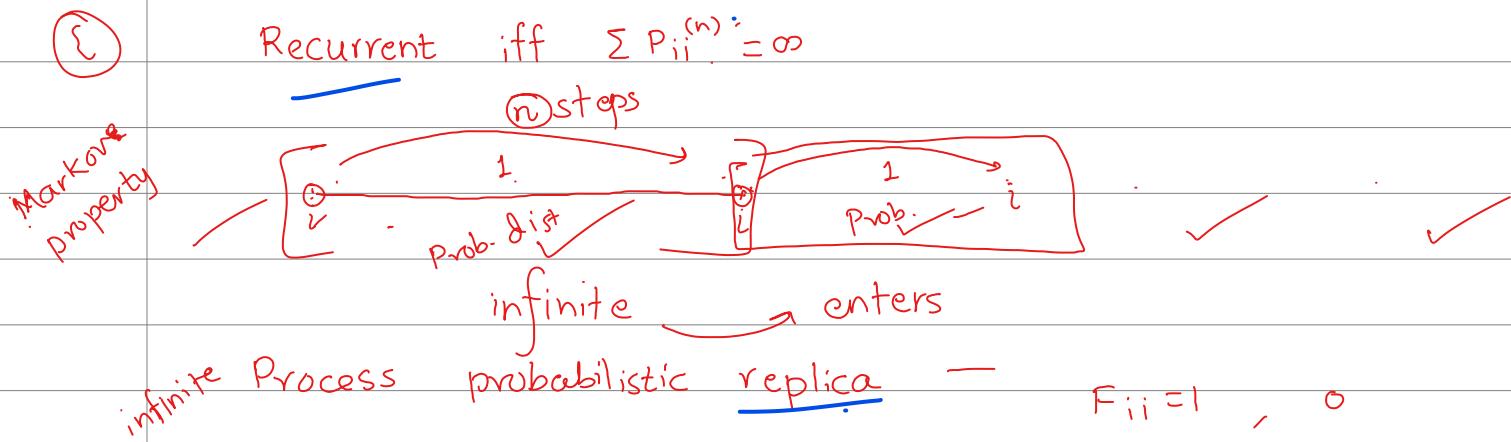
infinite - Recurrent

$c(1) = \{1, 2\}$   
 $c(2) = \{1, 2\}$   
 $c(3) = \{3\}$   
Transient  
Reducible MC



$$\begin{aligned} \sum_n P_{11}^{(n)} &= P_{11}^1 + P_{11}^2 + P_{11}^3 + P_{11}^4 + \dots \\ &= 0 + 0.5 + 0 + 0.5^2 + \dots \\ &= \frac{1}{1-0.5} = \frac{1}{0.5} = 2 < \infty \end{aligned}$$

Finite  $\Rightarrow$  Transient



i state transient,  $\left\{ \begin{array}{l} \text{Starting from } i \text{ stat process reaches to} \\ \text{state } i \text{ w.p. } < 1 \end{array} \right.$

$F_{ii} < 1$

$1 - F_{ii} > 0$  it will never return to state i

No. visits Finite no.  
~ Geo

$$E(\text{No. of visits}) \Rightarrow \frac{1}{1 - F_{ii}}$$

- State  $i$  is recurrent if, with probability 1, a process starting from state  $i$ , will eventually return.
- However, by Markovian Property, the process probabilistically restarts itself upon returning to state  $i$ . Hence with prob. 1 it will return to  $i$ .
- Repeating this argument, with probability 1, the no. of visits to state  $i$  will be infinite & will thus have infinite expectation.~

On the other hand, if state  $i$  is transient, there is positive prob.,  $1 - F_{ii} > 0$ , that it will never return again. Hence the no. of visits is Geometric with  $E(n) = \frac{1}{1 - F_{ii}}$

$\therefore$  State  $i$  is recurrent iff  $E(\text{no. of visits to } i / X_0 = i) = \infty$

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

$$\left| \begin{array}{l} E(I_n) = P(I_n = 1 / X_0 = i) \\ = P(X_n = i / X_0 = i) \\ = P_{ii}^n \end{array} \right.$$

$\therefore \sum_n I_n$  denotes no. of visits to  $i$ .  $\checkmark$

$$\begin{aligned} E\left(\sum_n I_n / X_0 = i\right) &= \sum_n E(I_n / X_0 = i) \\ &= \sum_n P_{ii}^n \end{aligned}$$

$\Rightarrow i$  is recurrent iff  $\sum_n P_{ii}^n = \infty \checkmark$

Theo

Recurrence is class property so does transience

Recurrence

$$\hookrightarrow F_{ii} = 1 \text{ or/and } \sum_n P_{ii}^n = \infty$$

Transience

$$F_{ii} < 1 \quad \sum_n P_{ii}^n < \infty$$

$i \leftrightarrow j$   $\Rightarrow$   $i$  is recurrent  $\Rightarrow$   $j$  is also recurrent

$\rightarrow$   $i \leftrightarrow j$ ,  $\exists i \rightarrow j$  for some  $s$ ,  $P_{ij}^s > 0$   
 $j \rightarrow i$  for some  $t$ ,  $P_{ji}^t > 0$

$i$  is recurrent  $\Rightarrow \sum_n P_{ii}^n = \infty$

$$P_{jj}^{s+t+n} = \sum_{k \in S} P_{jk}^t P_{kk}^n P_{kj}^s$$

$$< \sum_{n=0}^{\infty} P_{jj}^n = \infty$$

$$\geq P_{ji}^t P_{ii}^n P_{ij}^s$$

$$\sum_n P_{jj}^{s+t+n} \geq \sum_n P_{ji}^t P_{ii}^n P_{ij}^s$$

$$> P_{ji}^t \left( \sum_n P_{ii}^n \right) P_{ij}^s \\ > 0 \quad \infty \quad > 0$$

$$\Rightarrow \sum_n P_{jj}^n \geq \sum_n P_{jj}^{s+t+n} \geq \infty$$

~~$\leq \infty$~~

$$\Rightarrow \sum_n P_{jj}^n = \infty$$

Transience is class property.

$i \leftrightarrow j$ , if  $i$  is transient  $\Rightarrow j$  is also transient.

$$i \leftrightarrow j \cdot i \rightarrow j, \quad 1 \geq P_{ij}^s > 0 \\ j \rightarrow i, \quad 1 \geq P_{ji}^t > 0$$

$i$  is transient  $\Rightarrow \sum_n P_{ii}^n < \infty$

$$P_{jj}^{s+t+n} \geq P_{ji}^t P_{ii}^n P_{ij}^s$$

$$\sum_n P_{jj}^{s+t+n} \geq \boxed{P_{ji}^t \left( \sum_n P_{ii}^n \right) P_{ij}^s} \\ \leq 1 \quad < \infty \quad \leq 1$$

Recurrence  
 $i \leftrightarrow j$   $i$  transient  
 $\Rightarrow j$  transient

By method of contradiction  
 $i$  transient but  $j$  is recurrent

$$\sum_n P_{jj}^{s+t+n} \leq \infty$$

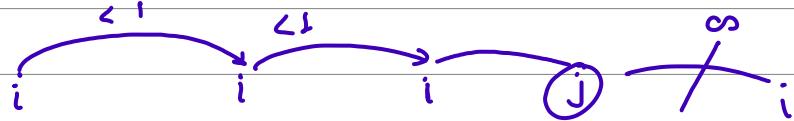
$j$  recurrent  $\Rightarrow i$  recurrent

which is contradiction

**Irreducible MC :-** Only one class , All states are communicating with each other .

Ergodic //  
Recurrent  
 $\circ f_i = 1$   
 $\circ \sum p_{ii} = 0$

let  $\mu_{jj}$  denotes the expected no. of transitions need to return to state  $j$



$$\mu_{jj} = \begin{cases} \infty & \text{if } j \text{ is transient.} \\ \sum n f_i^n & \text{if } j \text{ is recurrent.} \\ \infty & \text{null recurrent} \\ \mu_{jj} < \infty & \text{tve recurrent} \end{cases}$$

First time ↓

$$\begin{array}{c}
 \alpha \quad \beta \quad \gamma^* \\
 \alpha \left[ \begin{array}{ccc} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{array} \right] \\
 \beta \\
 \gamma^* 
 \end{array}$$
$$\begin{aligned}
 \mu_{\alpha\alpha} &= \sum_n n \cdot f_\alpha^n \\
 &= 1 \cdot 0 + 2 \cdot 0 + 3 \cdot 1 + 4 \cdot 0 + \dots \\
 &= 3
 \end{aligned}$$

$\mu_{\alpha\alpha} = 3 < \infty$

$c(\alpha) = \{\alpha, \beta, \gamma^*\} \Rightarrow$  Irreducible

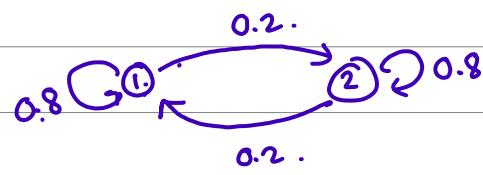
$$F_\alpha = f_\alpha^1 + f_\alpha^2 + f_\alpha^3 + f_\alpha^4$$

$$= 0 + 0 + 1 + 0 + \dots$$

= 1  $\Rightarrow \alpha, \beta, \gamma^*$  all are recurrent

$\mu_{\alpha\alpha} = 3 < \infty \Rightarrow \alpha$  is tve recurrent

$$P = \begin{bmatrix} 1 & 2 \\ 0.8 & 0.2 \\ 2 & 0.2 & 0.8 \end{bmatrix}$$



$C(L) = \{1, 2\} \Rightarrow$  Irreducible

$1 \rightarrow 2 \rightarrow 2 \rightarrow 1$

$$F_1 = f_1 + f_1^2 + f_1^3 + f_1^4 + \dots$$

$$= 0.8 + 0.2^2 + 0.2^2 \cdot 0.8^1 + 0.2^2 \cdot 0.8^2 + \dots + \underline{0.2^2 (0.8)^{n-2}} + \dots$$

$$= \underline{0.8} + 0.2^2 [1 + \underline{0.8 + 0.8^2 + \dots}] < \infty$$

$0.8 < 1$

$$\left\{ \begin{array}{l} = 0.8 + 0.2^2 \left[ \frac{1}{1-0.8} \right] \\ = 0.8 + 0.2 \cdot \\ = 1 \end{array} \right.$$

$\Rightarrow 1, 2$  are recurrent states

$$\mu_{11} = \sum_n n \cdot f_1^n = 1 \cdot 0.8 + 2 \cdot \underline{0.2^2} + 3 \cdot \underline{0.2^2 \cdot 0.8} + \dots + n \cdot 0.2^2 \cdot 0.8^{n-2} + \dots$$

$$= 0.8 + 0.2^2 [2 + 3 \cdot 0.8 + \dots + n \cdot 0.8^{n-2} + \dots]$$

$$\sum_{n=0}^{\infty} x^n = \frac{1}{1-x} \Rightarrow \underline{\frac{d}{dx} \sum_{n=0}^{\infty} x^n} = \frac{d}{dx} \cdot \frac{1}{1-x} = \frac{1}{(1-x)^2}$$

$$\sum_{n=1}^{\infty} n \cdot x^{n-1} = \frac{1}{(1-x)^2}$$

$$\begin{aligned} &= 0.8 + 0.2^2 \left[ \sum_n (n+1) x^{n-1} \right] = 0.8 + 0.2^2 \cdot \sum_{n=1}^{\infty} n \cdot x^{n-1} + 0.2^2 \sum_{n=0}^{\infty} x^n \\ &= 0.8 + (0.2^2) \cdot \frac{1}{(1-0.8)^2} + 0.2^2 \sum_{n=0}^{\infty} x^n = 0.8 + 1 + 0.2^2 \frac{n!}{(n-1)!} \end{aligned}$$

1, 2 +ve recurrent.

- \* Ergodic state: Aperiodic, positive recurrent  
in ergodic state  $\Rightarrow d(i) = 1, \bar{F}_i = 1$  or  $\sum P_{ii}^n = \infty$   
 $\mu_{ii} < \infty$

- \* Ergodic MC  $\Rightarrow$  All states are ergodic.

$$\begin{bmatrix} 0.8 & 0.2 \\ 0.2 & 0.8 \end{bmatrix} \quad C(1) = \{1, 2\} \quad 1, 2, \text{ positive recurrent}$$

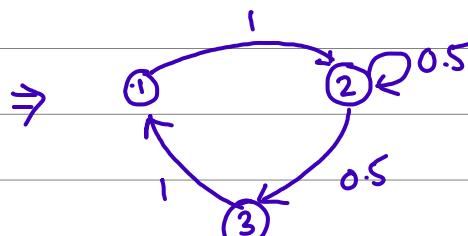
$$d(1) = \gcd(1, 2, \dots) = 1 = d(2)$$

1, 2 positive recurrent, aperiodic  $\Rightarrow$  ergodic

$$P = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0.5 & 0.5 \\ 1 & 0 & 0 \end{bmatrix}$$

States:  
 1: Aperiodic  
 2: Recurrent  
 3: Transient  
 tre  $\nwarrow$  null  
 Ergodic

$\Rightarrow$  MC Ergodic ?



$C(1) = \{1, 2, 3\} \Rightarrow$  Irreducible

$d(2) = \gcd\{1, 2, \dots\} = 1 \Rightarrow$  Aperiodic

Periodicity is class property  $\Rightarrow$

All states are aperiodic.

$$\begin{aligned}
 F_2 &= f_2^1 + f_2^2 + \dots = 0.5 + 0 + 0.5 \cdot 1 \cdot 1 + 0 + \dots \\
 &= 0.5 + 0.5 \\
 &= 1 \quad \Rightarrow \text{Recurrent}
 \end{aligned}$$

$\therefore$  Recurrence is class property.  $\Rightarrow$  All states are recurrent.

$$\mu_2 = 1 \cdot f_2^1 + 2 \cdot f_2^2 + 3 \cdot f_2^3 + 4 \cdot f_2^4 + \dots$$

$$\begin{aligned}
 &= 1 \cdot 0.5 + 0 + 3 \cdot 0.5 + 0 + \dots \\
 &= 0.5 + 1.5 \\
 &= 2 < \infty
 \end{aligned}$$

$\Rightarrow$  Positive Recurrent

All states are positive recurrent & aperiodic, i.e. all states are ergodic  $\therefore$  MC is ergodic MC.

Limit Theo. :-

If  $j$  is transient, then  $\sum_n P_{jj}^n < \infty$

$$P_{jj}^n \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

We have only finite no. of transitions returning to  $j$ .

$$\underline{\pi_j} = \lim_{n \rightarrow \infty} P_{jj}^n$$

$$\underline{\pi_{ij}} = \lim_{n \rightarrow \infty} P_{ij}^n$$

✓ tve recurrent state  $\pi_j > 0$

✗ null recurrent  $\pi_j = 0 ?$  —

Stationary

CK eq<sup>n</sup>

$$P^{n+1} = P^n \cdot P$$

$$\lim_{n \rightarrow \infty} P_{ij}^{(n+1)} = \lim_{n \rightarrow \infty} \sum_{k \in S} P_{ik}^{(n)} \cdot P_{kj}^{(1)}$$

$$\textcircled{e} \quad \underline{\underline{\pi_{ij}}} = \sum_{k \in S} \underline{\pi_{ik}} P_{kj}$$

$$P^{n+1} = P^n \cdot P$$

$$\left( \lim_{n \rightarrow \infty} P^{n+1} \right) = \left( \lim_{n \rightarrow \infty} P^n \right) \cdot P$$

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

In Matrix form

Recurrent limiting

$$\underline{\underline{\pi}} = \text{Stationary dist}^*$$

$$P = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 2 & 0 & 0.5 & 0.5 \\ 3 & 1 & 0 & 0 \end{bmatrix} \quad \underline{\pi} = [\pi_1, \pi_2, \pi_3]'$$

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

$$[\pi_1, \pi_2, \pi_3] = [\pi_1, \pi_2, \pi_3] \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0.5 & 0.5 \\ 1 & 0 & 0 \end{bmatrix}$$

$$\pi_1 = \pi_3$$

$$\pi_2 = \pi_1 + \pi_2/2 \Rightarrow \pi_1 = \pi_2/2$$

$$\pi_3 = \pi_2/2$$

$$\underline{\pi} = \underline{\left[ \frac{\pi_2}{2}, \pi_2, \frac{\pi_2}{2} \right]} \Rightarrow \sum_{i=1}^3 \pi_i = 1 \Rightarrow$$

$$\Rightarrow \frac{\pi_2}{2} + \pi_2 + \frac{\pi_2}{2} = 2\pi_2 = 1$$

$$\pi_2 = 1/2$$

$$\checkmark \quad \underline{\pi}_r = [1/4, 1/2, 1/4]$$

$$P = \frac{1}{10} \begin{bmatrix} 3 & 4 & 3 \\ 4 & 3 & 3 \\ 3 & 3 & 4 \end{bmatrix}$$

$$\pi = ?$$

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

$$[\pi_1, \pi_2, \pi_3] = \left[ \begin{bmatrix} 3 & 4 & 3 \\ 4 & 3 & 3 \\ 3 & 3 & 4 \end{bmatrix} \frac{1}{10} \right]$$

$$10\pi_1 = 3\pi_1 + 4\pi_2 + 3\pi_3 \Rightarrow 7\pi_1 = 4\pi_2 + 3\pi_3 \Rightarrow 7\pi_1 = 4\pi_2 + \frac{3}{2}(\pi_1 + \pi_2)$$

$$10\pi_2 = 4\pi_1 + 3\pi_2 + 3\pi_3 \Rightarrow 7\pi_2 = 4\pi_1 + 3\pi_3$$

$$10\pi_3 = 3\pi_1 + 3\pi_2 + 4\pi_3 \Rightarrow 6\pi_3 = 3\pi_1 + 3\pi_2 \Rightarrow 2\pi_3 = \pi_1 + \pi_2$$

$$11\pi_2 + \pi_1 = 8\pi_2 + 3\pi_1 + 3\pi_2 \Rightarrow \pi_1 = \pi_2$$

$$\underline{\pi} = [\pi_1, \pi_2, \pi_3] = [\pi_1, \pi_1, \pi_1]$$

$$\sum \pi_i = 1, \quad \pi_1 + \pi_1 + \pi_1 = 1 \Rightarrow \pi_1 = \frac{1}{3}$$

$$\pi = [\frac{1}{3} \quad \frac{1}{3} \quad \frac{1}{3}]$$

①  $L \begin{bmatrix} 0.5 & 0.5 & 0 & 0 \\ 0.5 & 0.5 & 0 & 0 \\ 6 & 0 & 0.5 & 0.5 \\ 0 & 0 & 0.5 & 0.5 \end{bmatrix} \Rightarrow ? \quad [x, x, y, y] \quad \begin{array}{l} x, y \in (0, 1) \\ x+y = 1 \end{array}$

Stationary distribution may or may not  
be unique.

②  $\frac{1}{6} \begin{bmatrix} 3 & 3 & 0 & 0 \\ 6 & 0 & 0 & 0 \\ 2 & 0 & 4 & 0 \\ 0 & 0 & 4 & 2 \end{bmatrix}$

Transition Graph

$C(1) = \{1, 2\}, \quad C(3) = \{3\}, \quad C(4) = \{4\} \Rightarrow$  Reducible MC

$d(1) = 1 \quad d(3) = 1 \quad d(4) = 1 \Rightarrow$  aperiodic

$$F_1 = f_1^1 + f_1^2 + f_1^n + \dots$$

$$= \frac{1}{2} + \frac{1}{2} \cdot 1 + 0 + \dots$$

$\Rightarrow F_1 = 1 \Rightarrow 1, 2$  are recurrent

$$F_3 = f_3^1 + f_3^2 + f_3^3 + \dots$$

$$= \frac{2}{3} + 0 + 0 + \dots$$

$= \frac{2}{3} < 1 \Rightarrow$  State 3 is transient

$$F_4 = f_4^1 + f_4^2 + f_4^3 + \dots$$

$$= \frac{1}{3} + 0 + \dots$$

$= \frac{1}{3} < 1 \Rightarrow$  State 4 is transient

$$\mu_{11} = \sum_n n f_1^n = 1 \cdot f_1^1 + 2 \cdot f_1^2 + 0 = 1 \times \frac{1}{2} + 2 \times \frac{1}{2} = 1.5 < \infty$$



Recurrence  
is class  
property

$$\begin{aligned}
 F_2 &= f_2^1 + f_2^2 + f_2^3 + f_2^4 + \dots + f_2^n + \dots \\
 &= 0 + \frac{1}{2} + \frac{1 \cdot 1 \cdot \frac{1}{2}}{\underline{\underline{\frac{1}{2}}}} + \frac{1}{2} \cdot 1 \cdot \left(\frac{1}{2}\right)^{n-2} = \frac{1}{2} \left(1 + \frac{1}{2} + \frac{1}{2^2} + \dots\right) \\
 &= \frac{1}{2} \cdot \left(\frac{1}{1 - \frac{1}{2}}\right) = 1 \Rightarrow 2 \text{ is recurrent.}
 \end{aligned}$$

$$\begin{aligned}
 \mu_2 &= \sum n f_2^n \Rightarrow 2 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2^2} + \dots + n \cdot \frac{1}{2^{n-1}} + (n+1) \frac{1}{2^n} \\
 &\Rightarrow \sum (n+1) \frac{1}{2^n} = \sum n \cdot \frac{1}{2^n} + \underline{\underline{\frac{1}{2^n}}} < \infty \\
 &\Rightarrow 2 \text{ is positive recurrent}
 \end{aligned}$$

1,2, aperiodic, positive recurrent  $\Rightarrow$  Ergodic States

Lets find out limiting distribution / Stationary

$$[\pi_1, \pi_2, \pi_3, \pi_4] = [\pi_1, \pi_2, \pi_3, \pi_4] \underset{-G}{\perp} \begin{bmatrix} 3 & 3 & 0 & 0 \\ 6 & 0 & 0 & 0 \\ 2 & 0 & 4 & 0 \\ 0 & 0 & 4 & 2 \end{bmatrix}$$

$$6\pi_1 = 3\pi_1 + \underline{\underline{6\pi_2}} + 2\pi_3 \rightarrow 6\pi_1 = 3\pi_1 + 3\pi_2 \Rightarrow \pi_1 = 2\pi_2 -$$

$$6\pi_2 = 3\pi_1 \Rightarrow \pi_2 = 0$$

$$6\pi_3 = 4\pi_3 + 4\pi_4 \Rightarrow \pi_3 = 0$$

$$\underline{6\pi_4} = 2\pi_4 \Rightarrow \pi_4 = 0 \Rightarrow 4\pi_4 = 0 \Rightarrow \pi_4 = 0$$

$$\pi_1 + \pi_2 + \pi_3 + \pi_4 = 1 \Rightarrow 2\pi_2 + \pi_2 + 0 + 0 = 1 \Rightarrow \pi_2 = 1/3$$

$$\Rightarrow \pi_1 = 2/3$$

Stationary  
Distribution

$$\begin{bmatrix} 2/3 & 1/3 & 0 & 0 \end{bmatrix}$$

$$\frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \underline{\underline{\pi}} = [\pi_1, \pi_1, \pi_3, \pi_3]$$

$$\pi_1 + \pi_1 + \pi_3 + \pi_3 = 1$$

$$\Rightarrow \underline{\pi_1 + \pi_3 = 1}$$

$\pi_1 = x$

$[x \ x \ 1-x \ 1-x]$

$$\lim_{n \rightarrow \infty} p_{ij}^n \quad i \rightarrow j \text{ in } n \text{ steps}$$

$\pi_j$  ✓       $i \rightarrow \text{doesn't matter}$

$$\pi' = \pi P \quad \text{eq}^n \rightarrow \checkmark$$

$$\pi' = \pi' P$$

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

$$\Rightarrow (P' - I) \underline{\pi} = 0$$

$$\pi_1 + \pi_2 + \pi_3 = 1$$

$$[1 \ 1 \ 1] \pi_1 = 1$$

$$\pi_3$$

$$A = [P' - I] \quad \text{comes}$$

$$b = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

No limiting distribution exists

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

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

$$P^3 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = P$$

$$\text{G} \quad \underline{\pi' = \overline{\pi' P}} \\ \underline{\text{initial} = \overline{\pi}}$$

$$\underline{P' = \overline{\pi P P^2}}$$

$$P(X_0=x) = \underline{\alpha} \quad \leftarrow \lim_{n \rightarrow \infty} P(X_n=x) \rightarrow \underline{\pi}$$

$$\underline{\alpha = \pi} \quad \begin{matrix} x_1 \\ \underline{x} \\ x_2 \end{matrix} \quad \frac{\alpha P}{\pi P} = P \quad \frac{\alpha \cdot P^2}{\pi \cdot P^2} \\ \frac{\pi \cdot P^2}{\pi \cdot P \cdot P^2} \quad \underline{\underline{\frac{\pi \cdot P \cdot P^2}{\pi \cdot P^2}}}$$







































































































































































































































































































































