

Probabilities with mc:  $X(t) \ a_0, a_1, a_2, \dots, a_n$

$$(I) P(X_0 = a_0, X_1 = a_1, X_2 = a_2, \dots, X_n = a_n) \\ = P(X_n = a_n | X_{n-1} = a_{n-1}) \cdot P(X_{n-1} = a_{n-1} | X_{n-2} = a_{n-2}) \\ \dots \dots P(X_1 = a_1 | X_0 = a_0) P(X_0 = a_0)$$

$$(II) P(X_n = a_j) \equiv \begin{matrix} \downarrow & \downarrow \\ \text{Step} & \text{State} \end{matrix} j^{\text{th}} \text{ state in } \begin{matrix} \downarrow \\ \text{Initial Probability Vector} \end{matrix} V_0 P^n$$

$$\text{Given } V_0 \Rightarrow V_1 = V_0 P, V_2 = V_1 P, \dots$$

$$\dots V_{n+1} = V_n \cdot P \\ \equiv V_{n+1} = [ \dots \dots \dots \underbrace{\quad}_{j^n} \dots \dots ]_{n+1}$$

### Classification of States:

The state of a M.C can be classified as recurrent or transient based on whether they are visited again on that state.

1) Recurrent: State  $a_i$  of a MC is said to be recurrent if starting from  $a_i$ , the return to state  $a_i$  is certain.

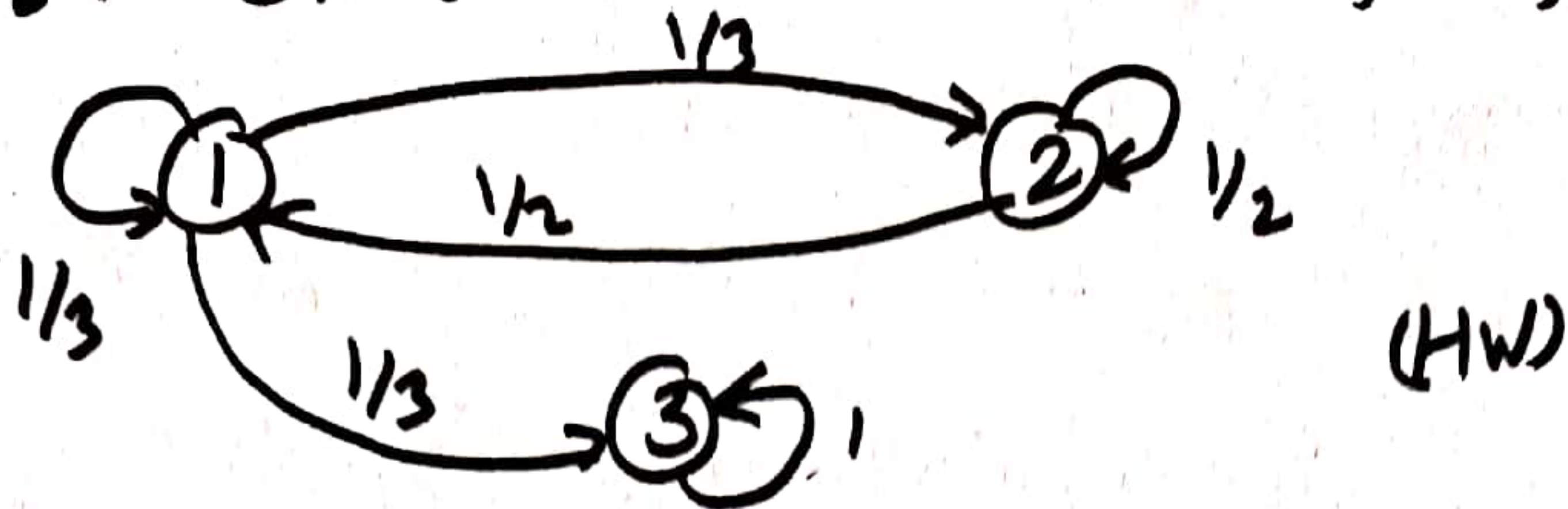
2) Transient: State  $a_i$  of a MC is said to be transient if starting from  $a_i$ , the return to state  $a_i$  is uncertain.

$$\rightarrow (1) \text{ Find } F_{ii} \equiv \text{Probability of returning to state } a_i \\ = f_{ii}^{(1)} + f_{ii}^{(2)} + \dots = \sum_{n=1}^{\infty} f_{ii}^{(n)}$$

(2) if  $F_{ii} = 1$ , then recurrent / if  $F_{ii} < 1$ , then transient



Q1 Categorized state 1 and 3 in the following MC



State 1 (1)

No of Step	Path	$f_{ii}^{(n)}$
1	1 → 1 (loop)	$\frac{1}{3} = f_{11}^{(1)}$
2	1 → 2 → 1	$\frac{1}{3} \cdot \frac{1}{2} = f_{11}^{(2)}$
3	1 → 2 → 2 → 1	$\frac{1}{3} \cdot (\frac{1}{2})^2 = f_{11}^{(3)}$
4	1 → 2 → 2 → 2 → 1	$\frac{1}{3} (\frac{1}{2})^3 = f_{11}^{(4)}$
⋮	⋮	⋮

$$\begin{aligned}
 F_{11} &= \sum_{n=1}^{\infty} f_{11}^{(n)} = \frac{1}{3} + \frac{1}{3} \left( \frac{1}{2} \right) + \frac{1}{3} \left( \frac{1}{2} \right)^2 + \frac{1}{3} \left( \frac{1}{2} \right)^3 + \dots \\
 &= \frac{1}{3} \left( 1 + \frac{1}{2} + \left( \frac{1}{2} \right)^2 + \left( \frac{1}{2} \right)^3 + \dots \right) \\
 &= \frac{1}{3} \cdot \frac{1}{1 - \frac{1}{2}} = \frac{2}{3} < 1 \quad \left[ \because \underbrace{1 + x + x^2 + \dots}_{\frac{1}{1-x}} = \frac{1}{1-x} \right]
 \end{aligned}$$

State 1 is transient.

State 3 (3)

No. of Step	Path	$f_{ii}^{(n)}$
1	3 → 3 (loop)	1 = $f_{33}^{(1)}$
2	No path exist	0 = $f_{33}^{(2)}$
⋮	⋮	⋮

$$F_{33} = \sum_{n=1}^{\infty} F_{33}^{(n)}$$

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

$$= 1$$

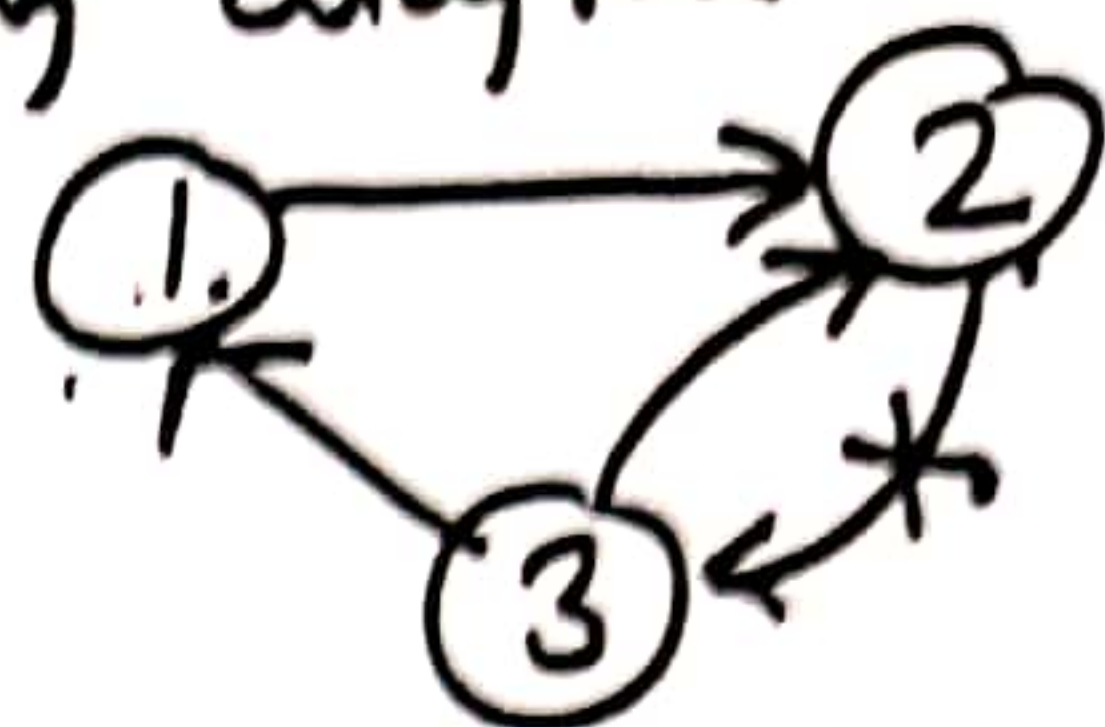
⇒ 3 is recurrent



\* Irreducible MC: If all states are reachable from all other states, then MC is called irreducible.

eg Consider the MC given in following diagram

Prove that it is irreducible.



(1) 2:  $2 \rightarrow 3 \rightarrow 1$   
3:  $3 \rightarrow 1$

(2) 1:  $2 \rightarrow 1$   
3:  ~~$3 \rightarrow 2$~~   $3 \rightarrow 2$

(3) 1:  $1 \rightarrow 2 \rightarrow 3$   
2:  $2 \rightarrow 3$

$\Rightarrow$  All states are reachable from all other states of MC  
 $\Rightarrow$  Given MC is irreducible.

Not loop

$3 \rightarrow 3$

(HW)

HW  
 Clarify the State



## Poisson Process:

If  $X(t)$  be the no of occurrence of a <sup>D.R.P</sup> certain event in  $(0, t)$ , then the underlying process is called Poisson Process if it satisfy the following postulates:

(i)  $P(1 \text{ occurrence in } (t, t+\Delta t)) = \lambda \Delta t + o(\Delta t)$   
parameter

(ii)  $P(2 \text{ or more occurrence in } (t, t+\Delta t)) = o(\Delta t) \rightarrow 0$

(iii)  $X(t)$  is independent of occurrences of the event in any interval prior or after  $(0, t)$ .

(iv) Probability that the event occurs a specified no of time in  $(t_0, t_0+t)$  depends only on  $t$  but not on  $t_0$ .

## Probability Law for PP:

1st order PMF:  $P_n(t) = P(X(t) = n) = \frac{e^{-\lambda t} (\lambda t)^n}{n!} = \phi(\lambda t)$   
 $n = 0, 1, 2, \dots$

PP  $\rightarrow \frac{\lambda t}{1}$

2nd order PMF  $P_{m,n}(t) = P[X(t_1) = m, X(t_2) = n]$   
 $= \begin{cases} \frac{e^{-\lambda t_1} (\lambda t_1)^m (t_2 - t_1)^{n-m}}{m! (n-m)!} & ; n \geq m \\ 0 & \text{o.w} \end{cases}$

Statistics of PP: (1)  $E(X(t)) = \lambda t$  (2)  $\text{Var}(X(t)) = \lambda t$

(2)  $R(t_1, t_2) = \lambda^2 t_1 t_2 + \lambda \min(t_1, t_2) \mid R \begin{matrix} t_1 \leq t_2 \\ \lambda^2 t_1 t_2 + \lambda t_1 \end{matrix}$



$$(4) C(t_1, t_2) = \lambda \min(t_1, t_2) = \begin{cases} \lambda t_1 & t_1 \leq t_2 \\ \lambda t_2 & t_2 \leq t_1 \end{cases}$$

$$(5) \gamma(t_1, t_2) = \frac{C(t_1, t_2)}{\sqrt{\text{Var}(X(t_1))} \sqrt{\text{Var}(X(t_2))}}$$

$$= \frac{\lambda \min(t_1, t_2)}{\sqrt{\lambda t_1} \sqrt{\lambda t_2}} = \begin{cases} \sqrt{t_1/t_2} & t_1 \leq t_2 \\ \sqrt{t_2/t_1} & t_2 \leq t_1 \end{cases}$$

### Properties:

- (1) PP is a MP.  $\Rightarrow$  Memoryless
- (2) Sum of two independent PP is a PP.
- (3) Diff " " " " is not a PP.
- (4) The interarrival time of a PP (interval between two successive occurrence) with parameter  $\lambda$  has an exponential distribution.
- \* The two consecutive occurrence of events be  $E_i$  and  $E_{i+1}$  with  $T$  be the interval b/w occurrence of  $E_i$  and  $E_{i+1}$ . Then,  $T$  be a RV and follow exponential distribution.

PDF of  $T$ :  $f(t) = \lambda e^{-\lambda t} \quad (t \geq 0)$

- (5) If the # of occurrence of an event  $E$  is an interval of length of  $t$  is a PP with  $\lambda$  and if each occurrence of  $E$  has a constant probability  $p$  of being recorded. If every recording is independent, then number  $N(t)$  of recorded occurrence in  $t$  is also a PP with  $(\lambda p)$ .



$$P(N(t)=n) = \frac{e^{-\lambda t} (\lambda t)^n}{n!} \quad \lambda t = \lambda t$$

$\downarrow$   
 $\lambda t \equiv (\lambda t)$

Q) For a PP with arrival rate 30 car/hour. Find the following:

- (1) Expected No of arrival in first 10 min of an hour.
- (2) Prob of exactly 4 arrival in first 10 min of an hour.
- (3) Prob of 35 or more arrival in an hour given 8 arrival in first 10 min.

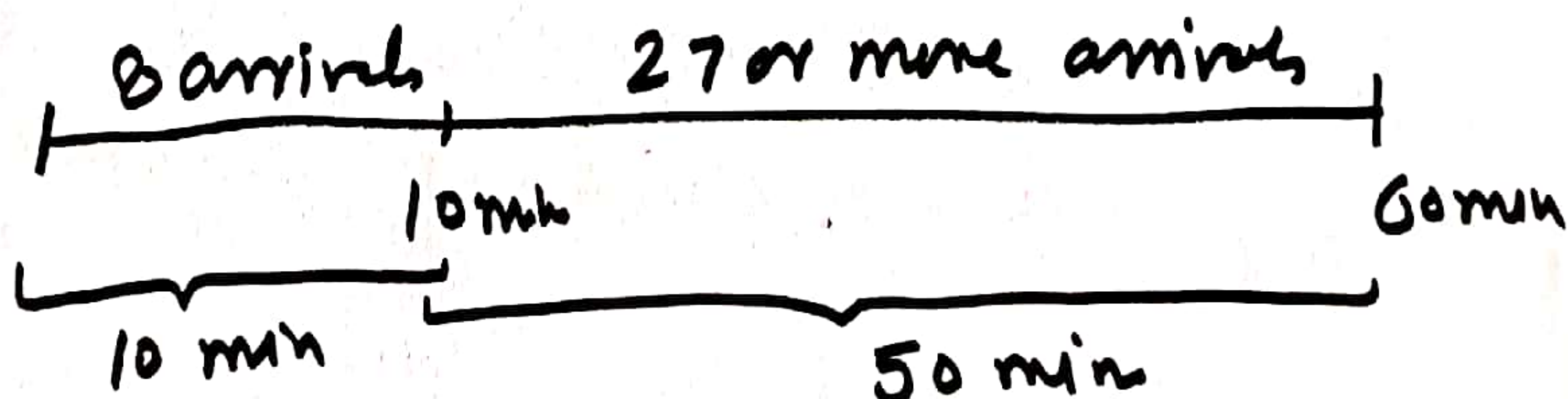
Sol<sup>n</sup> (i)  $\lambda = 30$   $t = 10 \text{ min in an hour} = \frac{1}{6} \text{ hour}$

$$(i) E(X(t)) = \lambda t = 30 \times \frac{1}{6} = 5$$

(ii)  $\lambda = 30$   $t = 10 \text{ min in an hour} = \frac{1}{6} \text{ hour}$

$$P(X(t)=4) = \frac{e^{-\lambda t} (\lambda t)^4}{4!} = \frac{e^{-5} (5)^4}{4!} = 0.1755$$

(iii)



$$\lambda t = 30 \cdot \frac{1}{6} = 5$$

$X_1(t)$

$$\lambda t = 30 \cdot \frac{5}{6} = 25$$

$$X_2(t) = P(X_2(t) \geq 27)$$

$$P(X_1(t)=8) = \frac{e^{-5} (5)^8}{8!}$$

(HW)



$$P(X_2(t) \geq 27) = P(X_2(t) = 27) + P(X_2(t) = 28) + \dots \checkmark$$

$$\Downarrow$$

$$1 - P(X_2(t) \leq 26) = \sum_{i=27}^{\infty} P(X_2(t) = i)$$

$$= 1 - \sum_{i=0}^{26} P(X_2(t) = i)$$

$$= 0.3706$$

### Birth and Death Process:

$X(t) \rightarrow$  population at time  $t$

$P_n(t) \rightarrow$  Prob.  $n$  ~~individuals~~ birth/death at time  $t$

Birth  $\rightarrow$  increases the population

Death  $\rightarrow$  decreases the population

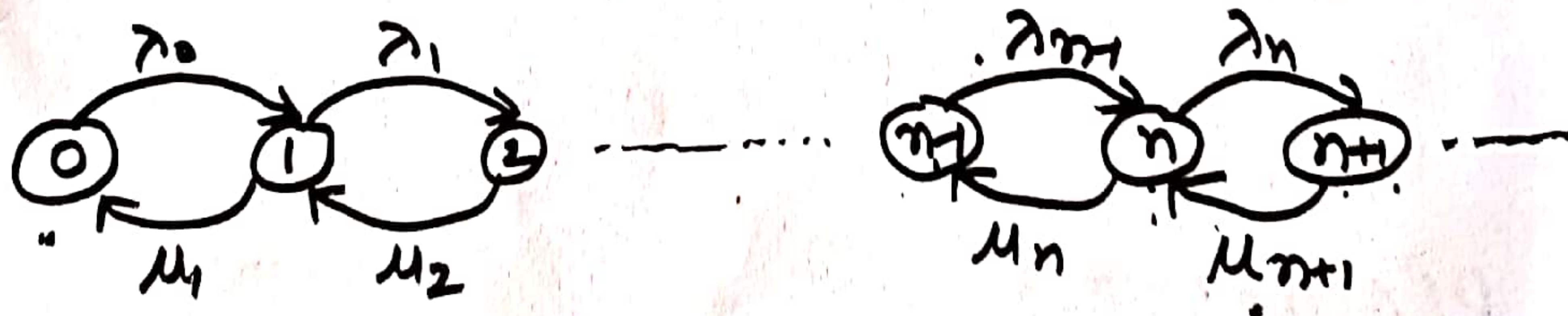
$T_B$ : time until next birth  $\equiv$  Birth process follows the PP

$T_B \equiv$  Exponential distributed RV  
with parameter  $(\lambda_n)$

$T_D$ : time until next death / busy time

$\equiv$  Exponential distributed RV with parameter  $(\mu_n)$

### Population Transition Diagram:





Rate of change in the population: ( $n$  state)

$$P_n'(t) = \underbrace{P_{n-1}(t) \lambda_{n-1} + P_{n+1}(t) \mu_{n+1}}_{\text{inflow at } n} - \underbrace{P_n(t) (\lambda_n + \mu_n)}_{\text{outflow}} \quad (1)$$

$$P_0'(t) = P_1(t) \mu_1 - P_0(t) \lambda_0 \quad (2)$$

eq<sup>n</sup>(1) & eq<sup>n</sup>(2) together is called Birth and Death process model.

Steady State of BDP: No rate of change.

$$P_n'(t) = 0 \quad \forall n$$

$$\Rightarrow P_0'(t) = 0$$

from eq<sup>n</sup> (1) & (2)

$$P_{n-1}(t) \lambda_{n-1} + P_{n+1}(t) \mu_{n+1} - P_n(t) (\lambda_n + \mu_n) = 0 \quad (3)$$

$$P_1(t) \mu_1 - P_0(t) \lambda_0 = 0 \quad (4)$$

ES in the BDP.

Eq<sup>n</sup> (3) & (4)  $AX = 0$

\* Birth  $\rightarrow$  Arrival of a customer

Death  $\rightarrow$  Departure of a served customer.



## Queueing theory:

QT tries to provide a balance b/w cost associated and the waiting time of customers.

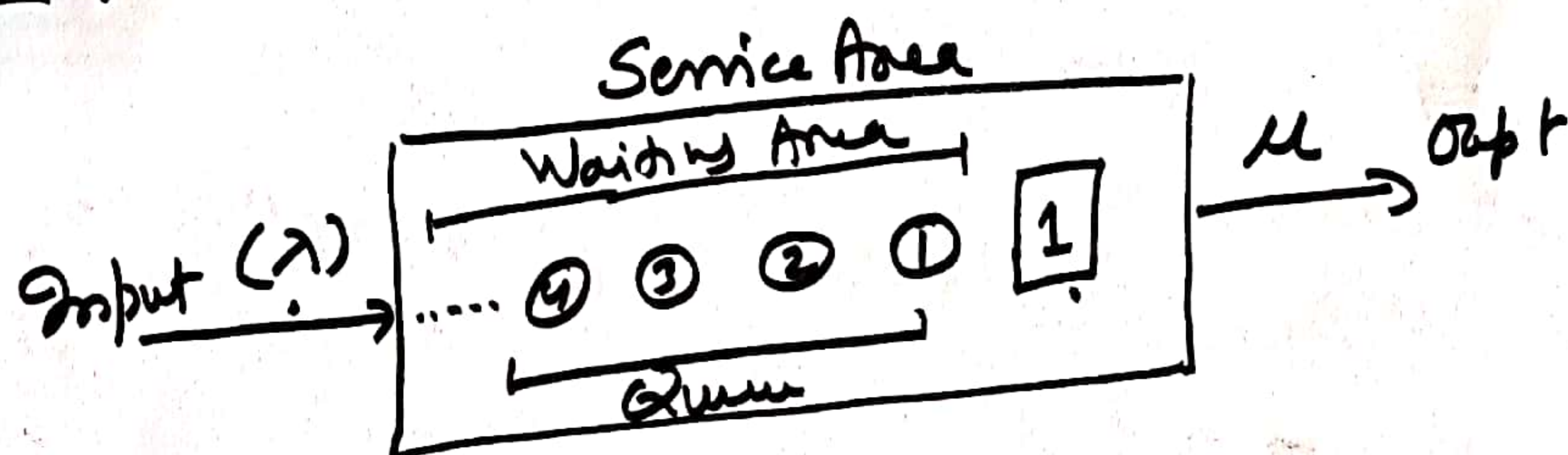
OR

balance b/w current demand of a particular service and the capacity to provide the service.

Assume: Birth  $\equiv$  Arrival of a customer

Death  $\equiv$  Departure of a served customer.

## Single Service Desk Case:



## M/M/1 Queueing Model

$\rightarrow$  M stand for 'Markov' indicating the no of arrival at it and completed service i'n it follow the PP/MP/MC.

M/M/1 model  $\begin{cases} \infty / \text{FIFO} \\ k / \text{FIFO} \end{cases}$



Rate of change in the population: ( $n$  State)

$$P_n'(t) = \underbrace{P_{n-1}(t) \lambda_{n-1} + P_{n+1}(t) \mu_{n+1}}_{\text{inflow at } n} - \underbrace{P_n(t) (\lambda_n + \mu_n)}_{\text{outflow}} \quad (1)$$

$$P_0'(t) = P_1(t) \mu_1 - P_0(t) \lambda_0 \quad (2)$$

eq<sup>n</sup> (1) & eq<sup>n</sup> (2) together is called Birth and Death process model.

Steady state of BDP: No rate of change.

$$P_n'(t) = 0 \quad \forall n$$

$$\Rightarrow P_0'(t) = 0$$

from eq<sup>n</sup> (1) & (2)

$$P_{n-1}(t) \lambda_{n-1} + P_{n+1}(t) \mu_{n+1} - P_n(t) (\lambda_n + \mu_n) = 0 \quad (3)$$

$$P_1(t) \mu_1 - P_0(t) \lambda_0 = 0 \quad (4)$$

ES in the BDP.

Eq<sup>n</sup> (3) & (4)

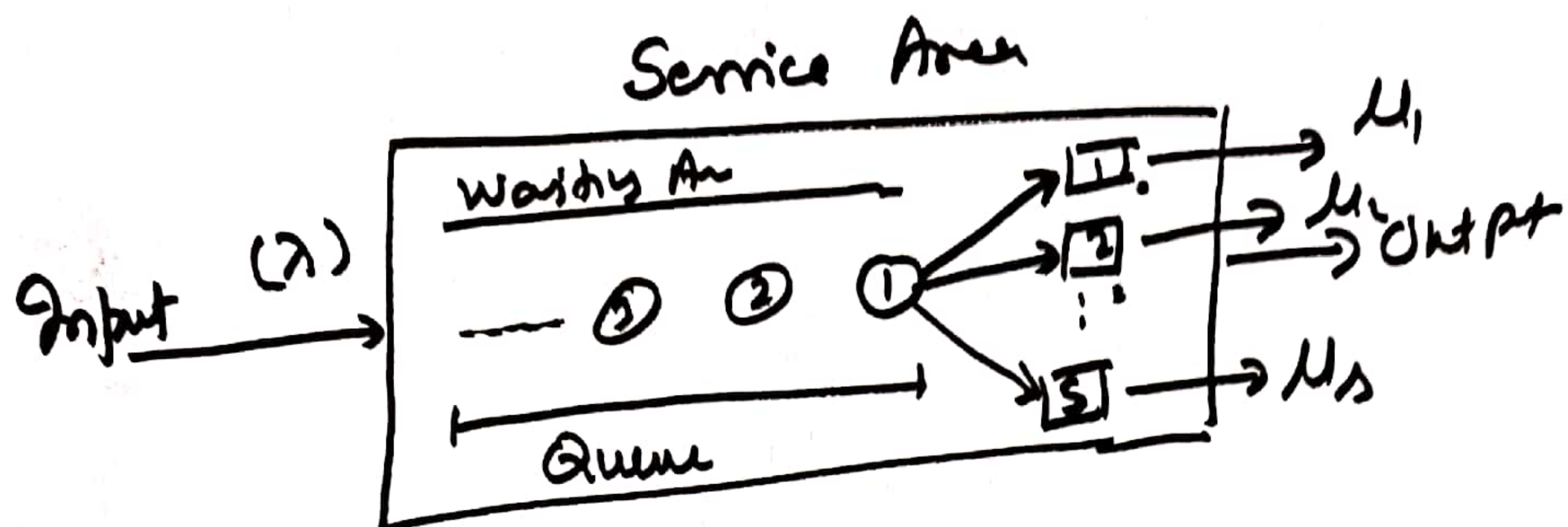
$$\boxed{AX = 0}$$

\* Birth  $\rightarrow$  Arrival of a customer

Death  $\rightarrow$  Departure of a served customer.

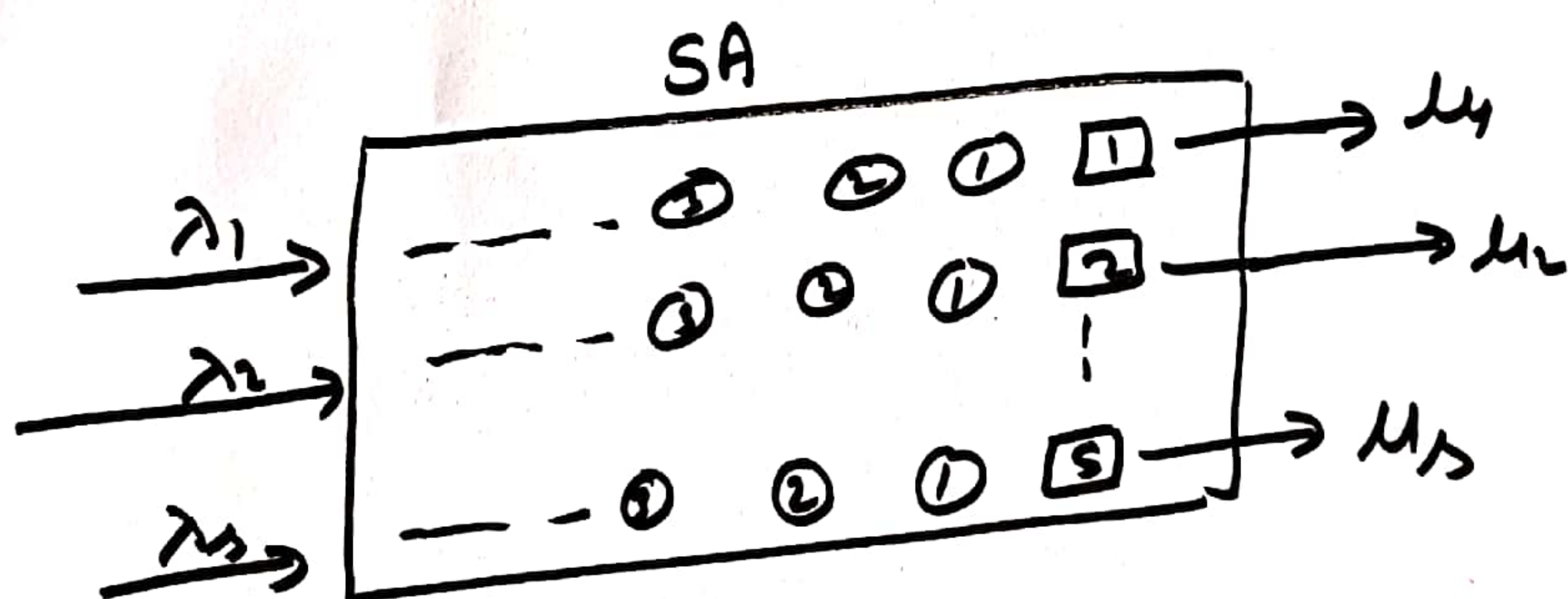


## Multiple Service Desk (in Parallel)



M/M/1s Model  $\swarrow$   $\infty$ /FIFO  
 $\searrow$   $K$ /FIFO

## Multiple Service Desk (in Series)



M/M/1s Model (with Series)

$\swarrow$   
 $\infty$ /FIFO  $\searrow$   $K$ /FIFO