

## UNIT - IV

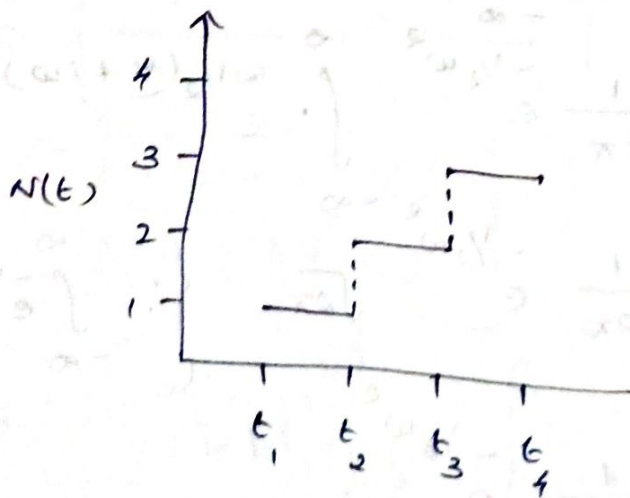
### POISSON PROCESS :

$\gamma_t$  is an example for a s.p with discrete state space and continuous parametric space.

Example :

- (i) NO. of incoming telephone calls at a telephone booth on  $[0, t]$  interval
- (ii) Arrival of customer for a service at a counter on  $[0, t]$  interval
- (iii) Occurance of an accident at a certain place on  $[0, t]$  interval.

Let  $N(t)$  denotes the total no. of occurrence of the events in an interval of duration 't'.  $\gamma_t$  may be given in the following graph



If the r.v  $N(t)$  assumes the value 'i', then the prob. of the event  $P[N(t) = i]$  is denoted by  $P_i(t)$ , where

$P_i(t)$  is a fn. of 't' such that  
 $\sum_{i=0}^{\infty} P_i(t) = 1$ , then  $P_i(t)$  is a prob. dist.  
 dist. fn. of  $N(t)$  for every value of 't'.

### Definition:

If  $\{N(t), t \geq 0\}$  is assumed as a counting process with  $N(0) = 0$ ,  $N(t)$  is said to be a Poisson process. Eg: any 2

### Postulates (or) Assumptions:

With respect to the Poisson process the following assumptions are made

(i) The increments are all independent

(ii) for the time points  $t_1, t_2, \dots, t_n$   $\exists$   
 $t_1 \leq t_2 \leq \dots \leq t_n$

$x(t_1), x(t_2) - x(t_1), x(t_3) - x(t_2), \dots$  are independent.

(ii) Transition prob. satisfy the stationary prob.

(ii) the dist. of  $x(t) - x(s)$  for  $s < t$  depends only on the time interval but not on  $s$  the time point where we started.

(iii) Prob. that atleast one event happen in the small interval of time  $h, h > 0$  is given by

$$P_1(h) = \lambda h + o(h)$$

where  $\frac{o(h)}{h} \rightarrow 0$  as  $h \rightarrow 0$

Prob. that the more than one event happens in the small interval is  $o(h), h > 0$ .



## Derivation of Poisson process:

$$\text{Let } P_m(t) = P[X(t) = m]$$

which gives the prob. that 'm' events happen in the interval  $[0, t]$

By postulates,

$$P_1(h) = \lambda h + o(h)$$

(i) i.e. prob. that the happening of at least one event in the interval of time 'h'

(ii) prob. that no events happened in the interval 'h' is

$$P_0(h) = 1 - P_1(h)$$

(iii) prob. that 2 (or) more events happened in the interval of time 'h' is

$$\sum_{m=2}^{\infty} P_m(h) = 0.$$

Now consider  $P_m(t) = 0$  for  $m = 0, 1, \dots$

for  $h > 0$ , the small interval

$$P_0(t+h) = P_0(t) \cdot P_0(h); \quad h > 0$$

$$= P_0(t) [1 - P_1(h)]$$

$$= P_0(t) [1 - \lambda h - o(h)]$$

$$= P_0(t) - \lambda h P_0(t) - o(h) \cdot P_0(t)$$

$$P_0(t+h) - P_0(t) = -\lambda h P_0(t) - o(h) P_0(t)$$

Dividing by 'h' and  $h \rightarrow 0$

$$\frac{P_0(t+h) - P_0(t)}{h} = \frac{-\lambda P_0(t)}{h} - \frac{o(h)P_0(t)}{h}$$

$$P_0'(t) = -\lambda P_0(t) \text{ as } h \rightarrow 0$$

which is a differential eqn. whose soln. is given by

$$P_0(t) = C \cdot e^{-\lambda t}$$

where  $C$  is to be determined from initial conditions.

Put  $t=0$ ,

$$P_0(0) = C \cdot e^{-\lambda(0)}$$

$$P_0(0) = C$$

$$\boxed{1 = C} \quad \left\{ \because P_0(0) = 1 \right\}$$

$$\therefore P_0(t) = e^{-\lambda t} \quad \text{--- (1)}$$

Now consider,

$$P_m(t+h) = P_0(h)P_m(t) + P_1(h)P_{m-1}(t) + P_2(h)P_{m-2}(t) + \dots$$

$$= P_0(h)P_m(t) + P_1(h)P_{m-1}(t) + \sum_{i=2}^{\infty} P_i(h)P_{m-i}(t)$$

$$= [1 - \lambda h - o(h)]P_m(t) + [\lambda h + o(h)]P_{m-1}(t) + o(h)$$

$$P_m(t+h) - P_m(t) = -\lambda h P_m(t) - o(h)P_m(t) + \lambda h P_{m-1}(t) + o(h)P_{m-1}(t)$$

$$P_m(t+h) - P_m(t) = -\lambda h P_m(t) - o(h)P_m(t) + \lambda h P_{m-1}(t) + o(h)P_{m-1}(t)$$

Dividing by  $h$  &  $h \rightarrow 0$



$$\frac{P_m(t+h) - P_m(t)}{h} = \frac{-\lambda h P_m(t)}{h} + \frac{\lambda h P_{m-1}(t)}{h}$$

$$P_m'(t) = -\lambda [P_m(t) - P_{m-1}(t)] \quad \text{--- (2)}$$

which is a differential eqn. This eqn. can be solved either by Induction method (or) by Prob. Generating method.

(i) Induction method:

$$\text{Let, } Q_m(t) = P_m(t) \cdot e^{\lambda t} \rightarrow (1)$$

$$P_m(t) = Q_m(t) \cdot e^{-\lambda t} \rightarrow (2)$$

From (1),

$$t=0$$

$$Q_m(0) = P_m(0) \cdot e^{\lambda(0)} = 0 \rightarrow (3)$$

$$m=0 \Rightarrow Q_0(0) = P_0(0) \cdot e^{\lambda(0)} = 1 \rightarrow (4)$$

Diff equ. (2) w.r. to 't'.

$$\frac{dP_m(t)}{dt} = P_m'(t) = Q_m'(t) \cdot e^{-\lambda t} - \lambda Q_m(t) \cdot e^{-\lambda t} \rightarrow (5)$$

But

$$P_m'(t) = -\lambda [P_m(t) - P_{m-1}(t)] \\ = -\lambda [Q_m(t) e^{-\lambda t} - Q_{m-1}(t) e^{-\lambda t}] \rightarrow (6)$$

From (5) and (6),

$$Q_m'(t) \cdot e^{-\lambda t} - \lambda Q_m(t) \cdot e^{-\lambda t} = -\lambda Q_m(t) \cdot e^{-\lambda t} + \lambda Q_{m-1}(t) \cdot e^{-\lambda t}$$

$$Q_m'(t) \cdot e^{-\lambda t} = \lambda Q_{m-1}(t) \cdot e^{-\lambda t} \rightarrow (7)$$

Put  $m = 1$  in eqn (7),

$$Q_1'(t) = \lambda Q_0(t) = \lambda t^0 = \lambda \quad (\because Q_0(t) = 1)$$

Integrating 't' both sides, in eqn (7).

$$Q_1(t) = \lambda t + C$$

To determine 'c', consider  $Q_1(0) = 0 + C$

$$\text{Put } t=0 \quad 0 = 0 + C \Rightarrow C=0$$

$$\therefore Q_1(t) = \lambda t$$

$$P_1(t) e^{\lambda t} = \lambda t \quad \text{from (1)}$$

$$P_1(t) = e^{-\lambda t} (\lambda t)$$

When  $m = 2$ , in eqn (7)

$$Q_2'(t) = \lambda Q_1(t) = \lambda \cdot \lambda t$$

Integrating on both sides,

$$Q_2(t) = \lambda^2 \frac{t^2}{2} + C$$

$C = 0$  as in previous case,

$$\Rightarrow Q_2(t) = \frac{(\lambda t)^2}{2}$$

$$P_2(t) e^{\lambda t} = \frac{(\lambda t)^2}{2} \quad \text{from eqn (1)}$$

$$P_2(t) = \frac{e^{-\lambda t} (\lambda t)^2}{2!}$$

$$P_m(t) = \frac{e^{-\lambda t} (\lambda t)^m}{m!}$$

which is poisson process.



(ii) Generating function method:

Consider,

$$P_m'(t) = -\lambda [P_m(t) - P_{m-1}(t)]$$

$$\frac{\partial P_m(t)}{\partial t} = -\lambda [P_m(t) - P_{m-1}(t)]$$

Multiplying on both sides by  $\Delta^m$  and taking summation over  $m$ ,

$$\frac{\partial \sum_{m=1}^{\infty} P_m(t) \Delta^m}{\partial t} = -\lambda \left[ \sum_{m=1}^{\infty} P_m(t) \Delta^m - \sum_{m=1}^{\infty} P_{m-1}(t) \Delta^m \right] \quad (1)$$

Consider the probability generating function,

$$\begin{aligned} P(\Delta, t) &= \sum_{m=0}^{\infty} P_m(t) \Delta^m \\ &= P_0(t) \Delta^0 + \sum_{m=1}^{\infty} P_m(t) \Delta^m \end{aligned}$$

$$\Rightarrow \sum_{m=1}^{\infty} P_m(t) \Delta^m = P(\Delta, t) - e^{-\lambda t} \rightarrow (2)$$

Consider,

$$\sum_{m=1}^{\infty} P_{m-1}(t) \Delta^m = \Delta \cdot \sum_{m=1}^{\infty} P_{m-1}(t) \Delta^{m-1}$$

$$= \Delta \cdot P(\Delta, t) \rightarrow (3)$$

Diff. w.r. to  $t$ ,

From (2)

$$\frac{\partial \sum_{m=1}^{\infty} P_m(t) \Delta^m}{\partial t} = P'(\Delta, t) + \lambda e^{-\lambda t} \rightarrow (4)$$

using (1) & (4)

$$P'(s, t) + \lambda e^{-\lambda t} = -\lambda \left[ \sum_m P(t) s^m - \sum_{m-1} P(t) s^m \right]$$

$$= -\lambda \left[ P(s, t) - e^{-\lambda t} - s P(s, t) \right]$$

{ using (2) & (3) }

$$P'(s, t) + \lambda e^{-\lambda t} = -\lambda P(s, t) + \lambda e^{-\lambda t} + \lambda s P(s, t)$$

$$P'(s, t) = \lambda P(s, t) [s - 1] \quad \text{--- (5)}$$

which is a differential eqn. whose soln is

$$P(s, t) = C \cdot e^{\lambda(s-1)t}$$

to find 'c'

$$\text{consider, } P(s, t) = \sum_{m=0}^m P(t) s^m$$

$$= P_0(t) s^0 + \sum_{m=1}^m P(t) s^m$$

$$P(s, 0) = P_0(0) s^0 + 1$$

$$= 1 \quad \text{--- (6)}$$

now consider,

$$P(s, t) = C e^{\lambda(s-1)t}$$

$$= C \cdot 1 \quad \text{--- (7)}$$

From (6) & (7),  $C = 1$ .

$$\text{Thus } P(s, t) = e^{\lambda(s-1)t}$$

$$= e^{\lambda s t} e^{-\lambda t}$$



which can be written as

$$\sum_{m=0}^{\infty} P_m(t) \lambda^m = e^{-\lambda t} \cdot e^{\lambda t}$$

$$= e^{-\lambda t} \sum_{m=0}^{\infty} \frac{(\lambda t)^m \lambda^m}{m!}$$

$$\left\{ \begin{aligned} \therefore e^x &= 1 + \frac{x}{1!} + \frac{x^2}{2!} + \dots \\ &= \sum_{r=0}^{\infty} \frac{x^r}{r!} \end{aligned} \right.$$

consider the coeff. of  $\lambda^m$

$$P_m(t) = \frac{e^{-\lambda t} (\lambda t)^m}{m!}$$

which is the required Poisson process

Properties of Poisson process:  
[Application (or) Features]

(i) The sum of two Poisson processes is also a Poisson process.

$\{X(t)\}$  is a Poisson process

$$\begin{aligned} \phi_{X(t)}(s) &= E[e^{isX(t)}] \\ &= e^{\lambda t [e^{is} - 1]} \end{aligned}$$

$\{X_1(t)\}$  and  $\{X_2(t)\}$  are two Poisson processes with parameter  $\lambda_1, \lambda_2$  respectively.

Consider,

$$\begin{aligned}
 Q_{x_1(t) + x_2(t)} &= E \left[ e^{is [x_1(t) + x_2(t)]} \right] \\
 &= E \left[ e^{is x_1(t)} \right] \cdot E \left[ e^{is x_2(t)} \right] \\
 &= e^{\lambda_1 t [e^{is} - 1]} \cdot e^{\lambda_2 t [e^{is} - 1]} \\
 &= e^{(\lambda_1 + \lambda_2)t [e^{is} - 1]} \\
 &= e^{t (e^{is} - 1) \lambda} \quad \text{where } \lambda = \lambda_1 + \lambda_2 \text{ which is}
 \end{aligned}$$

similar to the c.f. of poisson process.

(ii) Binomial distn /- in relation with poisson process:

$\{x_1(t)\}$  is a poisson process with parameter  $\lambda_1$ .  
 $\{x_2(t)\}$  is a poisson process with parameter  $\lambda_2$ .

For an  $m < n$ , consider,

$$P[x_1 = m / x_1 + x_2 = n] = \frac{P[x_1 = m \cap x_1 + x_2 = n]}{P[x_1 + x_2 = n]}$$

$$= \frac{P[x_1 = m \cap x_2 = n - m]}{P[x_1 + x_2 = n]}$$

$$= \frac{P[x_1 = m] \cdot P[x_2 = n - m]}{P[x_1 + x_2 = n]}$$



$$= \frac{e^{-\lambda_1 t} (\lambda_1 t)^m}{m!} \cdot \frac{e^{-\lambda_2 t} (\lambda_2 t)^{n-m}}{(n-m)!}$$

$$\frac{e^{-(\lambda_1 + \lambda_2)t} [(\lambda_1 + \lambda_2)t]^n}{n!}$$

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

Let  $\frac{\lambda_1}{\lambda_1 + \lambda_2} = p$ ,  $\frac{\lambda_2}{\lambda_1 + \lambda_2} = q$

$$= \binom{n}{m} p^m q^{n-m}$$

which is the p.d.f. of binomial distn. /-

(iii)  $x(t)$  is a poisson r.v., then mean of

$$x(t) = E[x(t)]$$

$$E[x(t)] = \sum_{n=0}^{\infty} x(t) \cdot P_n(t)$$

$$= \sum_{n=0}^{\infty} x(t) \frac{e^{-\lambda t} (\lambda t)^n}{n!}$$

$$= \sum_{n=0}^{\infty} n \cdot \frac{e^{-\lambda t} (\lambda t)^n}{n!}$$

$$= \sum_{n=0}^{\infty} n \cdot \frac{e^{-\lambda t} (\lambda t)^n}{n(n-1)!}$$

$$= (\lambda t) e^{-\lambda t} \sum_{n=0}^{\infty} \frac{(\lambda t)^{n-1}}{(n-1)!}$$

$$= \lambda t e^{-\lambda t} \cdot e^{\lambda t}$$

$$E[x(t)] = \lambda t$$

$$\therefore \text{Var}[x(t)] = E[x(t)^2] - [E(x(t))]^2$$

$$E[x(t)^2] = \sum x(t)^2 P_n(t)$$

$$= \sum_{n=0}^{\infty} x(t)^2 \frac{e^{-\lambda t} (\lambda t)^n}{n!}$$

$$= \sum_{n=0}^{\infty} n^2 \frac{e^{-\lambda t} \lambda t^n}{n!}$$

$$= \sum_{n=0}^{\infty} \left[ \frac{n}{n!} + \frac{n(n-1)}{n!} \right] e^{-\lambda t} (\lambda t)^n$$

$$= \sum_{n=0}^{\infty} \frac{n}{n(n-1)!} + \frac{n(n-1)}{n(n-1)(n-2)!} e^{-\lambda t} (\lambda t)^n$$

$$= \sum_{n=0}^{\infty} \left[ \frac{1}{(n-1)!} + \frac{1}{(n-2)!} \right] e^{-\lambda t} (\lambda t)^n$$

$$= \sum_{n=0}^{\infty} \frac{e^{-\lambda t} (\lambda t)^n}{(n-1)!} + \sum_{n=0}^{\infty} \frac{e^{-\lambda t} (\lambda t)^n}{(n-2)!}$$

$$= \lambda t + (\lambda t)^2$$

$$\text{Var}[x(t)] = \lambda t + (\lambda t)^2 - (\lambda t)^2$$

$$\text{Var}[x(t)] = \lambda t$$

(iv) C.F. of the poisson process:

$$\phi_{x(t)}(\lambda) = E[e^{i s x(t)}]$$

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



## Birth and Death process:

It is one of the important classes of Markov process. Its state space is countable such that  $(0, 1, 2, \dots)$   $P_n(t)$ . Let  $x(t)$  denote the pop/- size at time 't'. It can be increased by birth, decrease by death. Multiple birth (or) multiple death are ruled out. They are assumed as zero. The birth and death rates depends on 't'. But certainly depends pop size. If they do not depend on 't', we get a time homogenous transition probability.

$$P_{ij}(t) = P\{x(t) = j / x(0) = i\}$$
$$= P\{x(t+u) = j / x(u) = i\} \forall u.$$

## Assumptions

- (i)  $P\{x(t+h) = k-1 / x(t) = k\} = \mu_k h + o(h)$  for  $k=1, 2, \dots$
- (ii)  $P\{x(t+h) = k+1 / x(t) = k\} = \lambda_k h + o(h)$
- (iii)  $P\{x(t+h) = k / x(t) = k\} = 1 - \lambda_k h - \mu_k h + o(h)$
- (iv)  $P\{|x(t+h) - x(t)| > 1 / x(t) = k\} = o(h)$ .

The stochastic process  $x(t)$  is called Birth and death process if it satisfies the assumption given in (1).

Thus  $x(t)$  is a Markov process with state space  $\{0, 1, 2, \dots\}$  and intensity rates are,

$$q_{ijk} = \begin{cases} \mu_j & ; \text{ for } k = j-1 \\ \lambda_j & ; \text{ for } k = j+1 \\ -(\lambda_j + \mu_j) & ; \text{ for } k = j \end{cases} \quad (2)$$

for  $j = 1, 2$  and  $\mu_0 = 0$

for  $j = 0 ; k = 1$   
 $q_{01} = \lambda_0$

for  $j = 0, k = 0$

$$q_{00} = -\lambda_0 \quad [ \because \mu_0 = 0 ]$$

From eq. (2) the forward Kolmogorov differential eq. of B and D. process are obtained as follows.

$$P_j'(t) = [ -(\lambda_j + \mu_j) P_j(t) + \lambda_{j-1} P_{j-1}(t) + \mu_{j+1} P_{j+1}(t) ; \text{ for } j \geq 1 ] \rightarrow (3)$$

$$P_0'(t) = [ -\lambda_0 P_0(t) + \mu_1 P_1(t) ]$$



### Derivation of (3):

Let us consider three time points  $0, t, t+h$  such that  $0 < t < t+h$ .

$$(0, t+h) = (0, t) \cup (t, t+h)$$

The no. of popl. at time  $t+h$ ,

$$P_{ij}(t+h) = P_{ij}(t) \cdot P\{X(t+h) = j / X(t) = j\} + P_{i, j-1}(t)$$

$$\cdot \{P\{X(t+h) = j / X(t) = j-1\} + P_{i, j+1}(t)$$

$$P\{X(t+h) = j / X(t) = j+1\}$$

$$P_{ij}(t+h) = \sum_{r=j, j-1, j+1} P_{ir}(t) P_{rj}(h)$$

$$= P_{ij}(t) P_{jj}(h) + P_{i, j-1}(t) P_{j-1, j}(h)$$

$$+ P_{i, j+1}(t) P_{j+1, j}(h)$$

$$= P_{ij}(t) [1 - (\lambda_j + \mu_j)h + P_{i, j-1}(t) \lambda_{j-1}(h)$$

$$+ P_{i, j+1}(t) \mu_{j+1}(h) + o(h)]$$

$$= P_{ij}(t) - P_{ij}(t) (\lambda_j + \mu_j)h + \lambda_{j-1}(h) \cdot P_{i, j-1}(t)$$

$$+ \mu_{j+1}(h) \cdot P_{i, j+1}(t) + o(h)$$

$$P_{ij}(t+h) - P_{ij}(t) = -P_{ij}(t) (\lambda_j + \mu_j)h + \lambda_{j-1}(h) P_{i, j-1}(t)$$

$$+ \mu_{j+1}(h) P_{i, j+1}(t) + o(h)$$

Dividing by  $h$  and taking limit  $h \rightarrow 0$ .

$$\lim_{h \rightarrow 0} \left[ \frac{P_{ij}(t+h) - P_{ij}(t)}{h} \right] = -(\lambda_j + \mu_j) P_{ij}(t) + \lambda_{j-1} P_{i,j-1}(t) + \mu_{j+1} P_{i,j+1}(t) + \lim_{h \rightarrow 0} \left[ \frac{o(h)}{h} \right].$$

$$P_{ij}'(t) = -(\lambda_j + \mu_j) P_{ij}(t) + \lambda_{j-1} P_{i,j-1}(t) + \mu_{j+1} P_{i,j+1}(t)$$

at  $t=0$ ,

$$P_{ij}'(0) = -(\lambda_j + \mu_j) P_{ij}(0) + \lambda_{j-1} P_{i,j-1}(0) + \mu_{j+1} P_{i,j+1}(0)$$

Let  $P_j(t) = P_{ij}(t)$ , suppressing the suffix 'i'.

Then we get,

$$P_j'(t) = -(\lambda_j + \mu_j) P_j(t) + \lambda_{j-1} P_{j-1}(t) + \mu_{j+1} P_{j+1}(t)$$

At  $j=0$ ,

$$P_0'(t) = -(\lambda_0 + \mu_0) P_0(t) + \lambda_{-1} P_{-1}(t) + \mu_1 P_1(t)$$

Since initial condition,  $P_j(0) = \begin{cases} 0 & j=1, 2, \dots \\ 1 & j=0 \end{cases}$

$$P_0'(t) = -\lambda_0 P_0(t) + \mu_1 P_1(t) \quad \therefore \mu_0 = 0$$

The forward Kolmogorov partial diff. eqn. for B & D process.

$$P_j'(t) = -(\lambda_j + \mu_j) P_j(t) + \lambda_{j-1} P_{j-1}(t) + \mu_{j+1} P_{j+1}(t)$$

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

The soln,  $P_j(t)$  can be obtained by solving the forward Kolmogorov partial diff. eqn. subject to the initial condition.



## Uses of B and D process

B and D process serves as a good model in the study of different disciplines of queuing theory.

Let  $x(t)$  be the no. of persons waiting or being served for service at a counter at time 't'. This number may increase due to an arrival & decrease due to service completion.

If  $\lambda_j$  denotes the arrival rate,  $\mu_j$  denote the service rate and  $j$  persons on the queue, then the process  $x(t)$  can be modelled as a B and D process with assumptions given in (1).

### Notes

For different values of  $\lambda_j, \mu_j$  differential models are formed. If  $\lambda_j = j\lambda$  and  $\mu_j = 0$ , then the B and D process becomes linear birth process.

If  $\lambda_j = \lambda$  and  $\mu_j = 0$ , then the B and D process becomes immigration process. If  $\lambda_j = 0, \mu_j = j\mu$ , then the B and D process is called as emigration process. If  $\mu_j = j\mu$  and  $\lambda_j = 0$ ,

then B and D process is called linear death process. If  $\lambda_j = \lambda_j$  and  $\mu_j = 0$ , then B & D process becomes pure birth process. If  $\lambda_j = 0$  and  $\mu_j = \mu_j$ , then B and D process becomes pure death process. The pure birth process becomes state-furry process if  $\lambda_j = j\lambda$ .

**Pure - birth process:**

Let  $x(t)$  denotes the popl size at time 't'.  $x(t)$  can increase by birth but multiple birth are not included in the process.

Further  $x(t)$  cannot decrease by death as well as migration.

(ie)  $\mu_k = 0$ , then the forward kolmogorov differential eqn's can be derived as follows.,

$$\text{Let } x(0) = 0, P_n(t) = P\{x(t) = n / x(0) = 0\}$$

$$P_n(t+h) = P_0(h) P_n(t) + P_1(h) P_{n-1}(t) + \sum_{i=2}^n P_i(h) P_{n-i}(t)$$

$$= [1 - \lambda_n h + o(h)] P_n(t) + \lambda_{n-1} h P_{n-1}(t) + o(h) + \dots$$

$$\lim_{h \rightarrow 0} \frac{P_n(t+h) - P_n(t)}{h} = -P_n(t) \lambda_n + P_{n-1}(t) \lambda_{n-1} + \lim_{h \rightarrow 0} \frac{o(h)}{h}$$

$$P_n'(t) = -\lambda_n P_n(t) + \lambda_{n-1} P_{n-1}(t) + 0 \quad \text{--- (1)}$$



At  $n=0$ ,

$$P_0'(t) = -\lambda_0 P_0(t) \quad \text{--- (2)}$$

Equ (1) & (2) are called Kolmogorov partial diff. equ. of the pure birth process.

From (2),

$$P_0'(t) + \lambda_0 P_0(t) = 0$$

$$P_0(t) e^{\int \lambda_0 dt} = k \Rightarrow P_0(t) e^{\lambda_0 t} = k$$

At  $t=0$ ,

$$P_0(0) e^{\lambda_0(0)} = k$$

$$\Rightarrow k = 1 \quad \left\{ \because P_0(0) = 1 \right\}$$

$$\therefore P_0(t) = e^{-\lambda_0 t} \quad \text{--- (3)}$$

Put  $n=1$  in equ (1), we get

$$P_1'(t) = -\lambda_1 P_1(t) + \lambda_0 P_0(t)$$

$$= -\lambda_1 P_1(t) + \lambda_0 e^{-\lambda_0 t}$$

[using (3)]

$$P_1'(t) + \lambda_1 P_1(t) = \lambda_0 e^{-\lambda_0 t}$$

$$P_1(t) e^{\int \lambda_1 dt} = \int \lambda_0 e^{-\lambda_0 t} e^{\int \lambda_1 dt} dt + C$$

$$P_1(t) e^{\lambda_1 t} = \int \lambda_0 e^{-\lambda_0 t} e^{\lambda_1 t} dt + C$$

$$= \lambda_0 \int e^{-(\lambda_0 - \lambda_1)t} dt + C$$

$$= \lambda_0 \frac{e^{-(\lambda_0 - \lambda_1)t}}{-(\lambda_0 - \lambda_1)}$$

$$P_0(t) e^{\int \lambda_0 dt} = k$$

diff. both sides

$$P_0'(t) e^{\lambda_0 t} + P_0(t) \lambda_0 e^{\lambda_0 t} = 0$$

$$e^{\lambda_0 t} [P_0'(t) + P_0(t) \lambda_0] = 0$$

$$P_0'(t) + \lambda_0 P_0(t) = 0$$

$$P_1(t) e^{\lambda_1 t} = \frac{\lambda_0}{\lambda_1 - \lambda_0} e^{-(\lambda_0 - \lambda_1)t} + c.$$

At  $t = 0$

$$P_1(0) e^{\lambda_1(0)} = \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right) e^{-(\lambda_0 - \lambda_1)(0)} + c.$$

$$0.11 = \frac{\lambda_0}{\lambda_1 - \lambda_0} + c$$

$$c = \left( -\frac{\lambda_0}{\lambda_1 - \lambda_0} \right)$$

$$P_1(t) e^{\lambda_1 t} = \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right) e^{-(\lambda_0 - \lambda_1)t} - \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right)$$

$$P_1(t) = \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right) e^{-\lambda_0 t + \lambda_1 t - \lambda_1 t} - \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right) e^{-\lambda_1 t}$$

$$= \left( \frac{\lambda_0}{\lambda_1 - \lambda_0} \right) \left[ e^{-\lambda_0 t} - e^{-\lambda_1 t} \right] \quad (4)$$

$$P_k(t) = \lambda_{k-1} e^{-\lambda_k t} \int_0^t e^{\lambda_k x} P_{k-1}(x) dx \quad (5)$$

for  $k = 1, 2, \dots$

which is called that soln. of pure birth process.  $P_k(t)$  can be evaluate for  $k = 1, 2, \dots$  by iteration.



Linear Birth process: (Yule-Furry Process)

when  $\lambda_n = n\lambda$ ,  $x(0) = 1$ , then the pure birth process is called as Yule-Furry process (Linear Birth process).

The partial diff. eqn. of the Yule-Furry process can be obtained by subst.  $\lambda_n = n\lambda$  in eqn (1)

$$\begin{aligned} \text{(ii)} \quad P_n'(t) &= -\lambda_n P_n(t) + \lambda_{n-1} P_{n-1}(t) \\ &= -n\lambda P_n(t) + (n-1)\lambda P_{n-1}(t) \\ &= -\lambda [n P_n(t) - (n-1) P_{n-1}(t)] \end{aligned} \quad \text{--- (2)}$$

Suppose the initial condition is

$$P_i(0) = \begin{cases} 0 & ; i \neq 1 \\ 1 & ; i = 1 \end{cases}$$

(iii) The process started with only one member at time  $t=0$ .  
when  $n=1$ ,

$$P_1'(t) = -\lambda P_1(t)$$

$$P_1'(t) + \lambda P_1(t) = 0$$

$$P_1(t) e^{\int \lambda dt} = k$$

$$P_1(t) e^{\lambda t} = k$$

At  $t=0$ ,

$$P_1(0) e^{\lambda(0)} = k \Rightarrow k = 1 \quad \left\{ \text{let } P_1(0) = 1 \right\}$$

$$\therefore P_1(t) e^{\lambda t} = 1$$

$$P_1(t) = e^{-\lambda t} \quad (**)$$

when  $n=2$ ,

$$P_2'(t) = -2\lambda P_2(t) + \lambda P_1(t)$$

$$P_2'(t) + 2\lambda P_2(t) = \lambda P_1(t)$$

$$= \lambda e^{-\lambda t} \quad (\text{from } *)$$

$$P_2(t) e^{2\lambda t} = \int \lambda e^{-\lambda t} e^{2\lambda t} dt + C$$

$$P_2(t) e^{2\lambda t} = \lambda \int e^{-\lambda t} e^{2\lambda t} dt + C$$

$$= \lambda \int e^{\lambda t} dt + C$$

$$= \lambda \frac{e^{\lambda t}}{\lambda} + C$$

At  $t=0$ ,

$$P_2(0) e^{2\lambda(0)} = 1 + C \quad \left\{ \text{Let } P_2(0) = 0 \right\}$$

$$0 = 1 + C$$

$$\Rightarrow C = -1$$

$$P_2(t) e^{2\lambda t} = e^{\lambda t} - 1$$

$$P_2(t) = \frac{e^{\lambda t} - 1}{2\lambda t}$$

$$= \left( e^{\lambda t} - 1 \right) e^{-2\lambda t} = e^{\lambda t - 2\lambda t} - e^{-2\lambda t}$$

$$= e^{-\lambda t} - e^{-2\lambda t}$$

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



iii)ly, 
$$P_n(t) = e^{-\lambda t} \left[ 1 - e^{-\lambda t} \right]^{n-1}; \text{ for } n \geq 1$$

where 
$$p = e^{-\lambda t}; \quad q = 1 - e^{-\lambda t}$$

$$P_n(t) = p q^{n-1}$$

which is called as Geometric dist.  
 The prob. generating fn. of the Geometric dist. is

$$P(s, t) = \sum_{n=1}^{\infty} P_n(t) s^n$$

$$= \sum_{n=1}^{\infty} \left[ e^{-\lambda t} (1 - e^{-\lambda t})^{n-1} \right] s^n$$

$$= \frac{s e^{-\lambda t}}{1 - s(1 - e^{-\lambda t})}$$

$$P(s, t) = \frac{s p}{1 - s q}$$

$$E(n) = 1/p$$

$$V(n) = q/p^2$$

which is the pdf of Yule-furry process.

Mean of the process =  $\frac{1}{e^{-\lambda t}} = e^{\lambda t}$

Variance of the process =  $\frac{1 - e^{-\lambda t}}{(e^{-\lambda t})^2}$

$$= (1 - e^{-\lambda t}) e^{2\lambda t}$$

$$= e^{2\lambda t} - e^{\lambda t}$$

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





The process  $\{x(t)\}$  stays that in the state 'j' for a random length of time, whose dist. is exponential with mean  $1/\mu_j$ , then it moves to the state  $(j-1)$ ;  $j=1, 2, \dots, N$ . This transition continues until the state '0' is reached. Since the  $1^{\text{st}}$  row of the matrix  $Q$  is  $(0, 0, \dots, 0)$ . Then the ~~state~~ state '0' is called Absorbing state.

The Forward Kolmogorov partial diff. eqns are obtained as

$$\begin{aligned}
 P_j'(t) &= -\mu_j P_j(t) + \mu_{j+1} P_{j+1}(t) \quad ; j=1, 2, \dots, N-1 \\
 P_N'(t) &= -\mu_N P_N(t) \quad ; j=N \\
 P_0'(t) &= \mu_1 P_1(t) \quad ; j=0
 \end{aligned}$$

The above set of diff. eqns are solved subject to the initial conditions

$$P_j(0) = \begin{cases} 1 & \text{if } j=N \\ 0 & \text{if } j \neq N \end{cases}$$

It may be solved by using iterative method, the soln is obtained by

$$P_N(t) = e^{-\mu_N(t)}$$

## Linear Birth and Death Process:

Let us consider the Kolmogorov forward partial diff. eqn. for the Birth and Death process as

$$P_j'(t) = -(\lambda_j + \mu_j) P_j(t) + \lambda_{j-1} P_{j-1}(t) + \mu_{j+1} P_{j+1}(t);$$

$$\forall t \geq 1 \quad \text{--- (1)}$$

at  $j=0$

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

at  $j=n$ , the eqn (1) becomes

$$P_n'(t) = -(\lambda_n + \mu_n) P_n(t) + \lambda_{n-1} P_{n-1}(t) + \mu_{n+1} P_{n+1}(t)$$

when  $\lambda_n = n\lambda$ ,  $\mu_n = n\mu$

$$P_n'(t) = -(n\lambda + n\mu) P_n(t) + \lambda(n-1) P_{n-1}(t) + \mu(n+1) P_{n+1}(t)$$

$$\forall n \geq 1 \quad \text{--- (2)}$$

at  $n=0$

$$P_0'(t) = -(0\lambda + 0\mu) P_0(t) + \lambda(0-1) P_{0-1}(t) + \mu(0+1) P_{0+1}(t)$$

$$\Rightarrow P_0'(t) = \mu P_1(t)$$

By using method of prob. generating fn. eqn (2) can be obtained

$$P^*(s, t) = \sum_{n=0}^{\infty} P_n(t) s^n \quad \left\{ \text{PGF of } P_n(t) \right\}$$

$$\frac{\partial P^*}{\partial s} = \sum_{n=0}^{\infty} n s^{n-1} P_n(t)$$



By multiplying (2) by  $s^n$  and summing over all values of  $n = 1, 2, \dots$  we get

$$\begin{aligned} \frac{\partial P}{\partial t} &= -(\lambda + \mu) \sum_{n=1}^{\infty} n P_n(t) s^n + \lambda \sum_{n=1}^{\infty} (n-1) P_{n-1}(t) s^n \\ &\quad + \mu \sum_{n=1}^{\infty} (n+1) P_{n+1}(t) s^n \\ &= -(\lambda + \mu) s \left[ \sum_{n=1}^{\infty} n P_n(t) s^{n-1} \right] + \lambda s^2 \left[ \sum_{n=1}^{\infty} (n-1) P_{n-1}(t) s^{n-2} \right] \\ &\quad + \mu \left[ \sum_{n=1}^{\infty} (n+1) P_{n+1}(t) s^n \right] \end{aligned}$$

$$= -(\lambda + \mu) s \frac{\partial P^*}{\partial s} + \lambda s^2 \frac{\partial P^*}{\partial s} + \mu \frac{\partial P^*}{\partial s}$$

$$\frac{\partial P}{\partial t} = \left[ \mu - (\lambda + \mu) s + \lambda s^2 \right] \frac{\partial P^*}{\partial s} \quad \text{--- (3)}$$

where  $P^*$  is a p.g.f.

The eq (3) is solved subject to the initial condition

$$P^*(s, 0) = s \quad \text{--- (4)}$$

The final soln for  $P^*(s, t)$  is obtained as  $P^{**}(s, t)$  and it satisfies the eq (3).

### Ex: M/M/s (Queue)

In this model, the no. of service channels 's' and queuing discipline FIFO arrivals at queue follows poisson process with rate  $\lambda$ .

Interarrival of the customer joining the queue is exponential with parameter  $\lambda/\mu$ . Service time dist. of all the customers and for all servers is same and it is distributed as exponential with mean  $1/\mu$  and the service rate is  $\mu$  per server.

If  $x(t) \leq s$ , all  $x(t)$  persons are being served, and departure rate  $\mu x(t)$ .

If  $x(t) > s$ , the departure rate is  $s\mu$ .

An account of the assumption of the exponential dist. for inter arrival time and service time are mutually independent.

This process  $\{x(t), t \geq 0\}$  is a M.P. of Birth & Death trial with  $\lambda_n = \lambda$  and

$$\mu_n = \begin{cases} n\mu & ; n \leq s \\ s\mu & ; n > s \end{cases} \quad \text{--- (1)}$$

using (1) in the expression

$$P_n = \left( \frac{\lambda_0, \lambda_1, \dots, \lambda_{n-1}}{\mu_1, \mu_2, \dots, \mu_n} \right) P_0, \text{ we get}$$

$$P_n = \begin{cases} \left( \frac{\lambda}{\mu} \right)^n \frac{1}{n!} P_0 & ; \text{ if } n \leq s \\ \left( \frac{\lambda}{\mu} \right)^n \frac{1}{s! s^{n-s}} P_0 & ; \text{ if } n > s \end{cases}$$



$P_0$  can be obtained by using  $\sum P_n = 1$

such that

$$P_0^{-1} = 1 + \sum_{n=1}^{\infty} \left(\frac{\lambda}{\mu}\right)^n \frac{1}{n!} + \sum_{n=s+1}^{\infty} \left(\frac{\lambda}{\mu}\right)^n \frac{1}{s!} \frac{1}{s^{n-s}}$$

$\frac{\lambda}{s\mu} < 1$  is necessary condition for the existence of the steady state for the queuing model  $M/M/s$ .

If  $\left(\frac{\lambda}{s\mu}\right) = 1$ , the queue size increased infinitely and hence there is no stability in the queuing system.

Ex: when  $s=1$  (ie)  $M/M/1$  - FIFO:

S.T a single server queuing system with  $\rho < 1$  has a stationary dist. of the queue size

$$P_n = (1-\rho)\rho^n$$

Solu

when  $s=1$ , in  $M/M/s$  queuing system is obtained,  $M/M/1$  queuing system is obtained.

In this case the process  $x(t)$  will be a B & D process with  $\lambda_n = \lambda$ ,  $\mu_n = \mu \forall n=1, 2, \dots$

and  $\mu_0 = 0$ ,  $\lambda_0 = \lambda$  for  $n=0$

$$\text{Let } P_n(t) = P \left\{ x(t) = n / x(0) = . \right\}$$

Then the Kolmogorov forward partial diff. eqn of B & D process be

$$P_n'(t) = -(\lambda + \mu)P_n(t) + \lambda P_{n-1}(t) + \mu P_{n+1}(t), \quad n \geq 1 \quad \text{--- (1)}$$

$$P_0'(t) = -\lambda P_0(t) + \mu P_1(t) \quad \text{--- (2)} \quad \left\{ \begin{array}{l} \lambda = \lambda_0 \\ \mu = \mu_0 = 0 \end{array} \right.$$

Let us assume the existence of steady state as  $t \rightarrow \infty$

$$P_n(t) \rightarrow \lim_n P_n, \text{ inde. of } t.$$

The steady state prob.

$P\{x=n\} = P_n$ , can be obtained by using

$$\left[ \begin{array}{l} P_n'(t) = 0 \text{ and } P_n(t) = P_n \text{ in (1) \& (2) we get} \\ 0 = -(\lambda + \mu)P_n + \lambda P_{n-1} + \mu P_{n+1}, \quad n \geq 1 \quad \text{--- (3)} \\ 0 = -\lambda P_0 + \mu P_1 \quad \text{--- (4), } n=0 \end{array} \right.$$

The eqn (3) & (4) are called as balance eqn of equilibrium eqn.

From (4) we get

$$\mu P_1 = \lambda P_0$$

$$P_1 = \frac{\lambda}{\mu} P_0$$

$$P_1 = \rho P_0 \quad \left\{ \because \rho = \frac{\lambda}{\mu} \right\}$$



$$\Rightarrow P_0 = P P_1$$

$$\vdots$$

$$P_n = P P_{n-1}$$

$$\left. \begin{aligned} P_0 &= P P_1 \\ &= P(P P_0) \\ &= P^2 P_0 \end{aligned} \right\}$$

$$= P(P^{n-1} P_0)$$

$$\left. \begin{aligned} \therefore P_{n-1} &= P^{n-1} P_0 \end{aligned} \right\}$$

$$P_n = P^n P_0 \quad \text{--- (5)}$$

from eq (3)

$$0 = -(\lambda P_n + \mu P_n) + \lambda P_{n-1} + \mu P_{n+1}$$

$$0 = -\lambda P_n - \mu P_n + \lambda P_{n-1} + \mu P_{n+1}$$

$$\mu P_{n+1} - \lambda P_n = \mu P_n - \lambda P_{n-1}$$

$\div \mu$  on both sides,

$$P_{n+1} - \left(\frac{\lambda}{\mu}\right) P_n = P_n - \left(\frac{\lambda}{\mu}\right) P_{n-1}$$

$$P_{n+1} - P P_n = P_n - P P_{n-1}$$

for  $n = n-1$ ,

$$P_{n-1+1} - P P_{n-1} = P_{n-1} - P P_{n-1-1}$$

$$P_n - P P_{n-1} = P_{n-1} - P P_{n-2}$$

for  $n = n-2$

$$P_{n-2+1} - P P_{n-2} = P_{n-2} - P P_{n-2-1}$$

$$P_{n-1} - P P_{n-2} = P_{n-2} - P P_{n-3}$$

$$\begin{aligned}
 \therefore P_n - \rho P_{n-1} &= P_{n-1} - \rho P_{n-2} \\
 &= P_{n-2} - \rho P_{n-3} \\
 &\vdots \\
 &= P_1 - \rho P_0 \\
 &= 0 \quad \because n \geq 0
 \end{aligned}$$

Since  $P_n = \rho P_{n-1}$   
 $= \rho^n P_0$  and  $\sum_{n=0}^{\infty} P_n = 1$

$$\begin{aligned}
 \sum P_0 \rho^n &= P_0 (\sum \rho^n) \\
 &= P_0 (1 + \rho + \rho^2 + \dots + \rho^n + \dots) \\
 &= P_0 (1 - \rho)^{-1} = 1 \quad \left\{ \because P_0 = 1 - \rho \right.
 \end{aligned}$$

$$\Rightarrow P_0 = 1 - \rho$$

$$P_n = (1 - \rho) \rho^n, \quad n \geq 0$$

$\rho = \frac{\lambda}{\mu} < 1$  The steady state solution of the process (simple queuing model) exists.

The distn.  $P_n = (1 - \rho) \rho^n$  is of a geometric distn. Hence mean & variance of the queue size can be obtained as

$$E(x) = \frac{\rho}{1 - \rho} = \frac{\lambda}{\mu - \lambda}$$



$$V(x) = \frac{\rho}{(1-\rho)^2} = \frac{\lambda\mu}{(\mu-\lambda)^2}$$

The Prob. that the system is empty is obtained as

$$P\{x=0\} = P_0 = 1-\rho$$

The Prob. that the system is not empty is obtained as

$$P\{x \geq 1\} = 1 - P\{x < 1\}$$

$$= 1 - P\{x=0\}$$

$$= 1 - P_0$$

$$= 1 - (1-\rho)$$

$$P\{x \geq 1\} = \rho$$

Ex: M/M/∞; FIFO

It is obtained M/M/S; FIFO as a special case of M/M/S with  $s = \infty$ . In real life the no. of servers can be increase so that at any time the no. of servers equal to the no. of customers.

There will be no waiting. This model is more appropriate in the large communication system with the assumptions used in M/M/S system in

addition to the above assumption.  
 The process  $x(t)$  will be a M.P of  
 B & D trial with  $\lambda_n = \lambda$ ,  $\mu_n = n\mu \forall n=0,1,2,\dots$   
 eqn (1) is substituted in

$$P_n = \left[ \begin{array}{c} \lambda_0 \lambda_1 \dots \lambda_{n-1} \\ \mu_1 \mu_2 \dots \mu_n \end{array} \right] P_0$$

Hence we get,

$$P_n = \left( \frac{\lambda}{\mu} \right)^n \frac{1}{n!} P_0 ; \forall n=1,2,\dots \quad \text{--- (2)}$$

$$\sum P_n = 1 \Rightarrow P_0 \sum_n \left( \frac{\lambda}{\mu} \right)^n \frac{1}{n!} = 1$$

$$P_0 \sum \frac{p^n}{n!} = 1 ; \text{ where } p = \frac{\lambda}{\mu}$$

$$P_0 (e^p) = 1 \Rightarrow P_0 = e^{-p} \quad \left\{ \begin{array}{l} e^x = 1 + \frac{x}{1!} + \frac{x^2}{2!} + \dots \end{array} \right.$$

$$\therefore P_n = \frac{e^{-p} p^n}{n!}$$

Thus the stable distn. of  $M/M/\infty$   
 exist whatever be the value of  $p$  and  
 it is of the Poisson distn. with  
 parameter  $p$ . The steady state soln of  
 this model exist if the series  $\sum \frac{p^n}{n!}$   
 convergent.