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

$$\lim_{\delta \to 0} \frac{1}{\delta} \sum_{k=2}^{\infty} \frac{(\lambda\delta)^k}{k!} = \lim_{\delta \to 0} \left(\frac{\lambda^2}{2}\delta - \frac{\lambda^3}{3!}\delta^2 + \frac{\lambda^4}{4!}\delta^3 + \cdots\right) = 0$$

$$\therefore \sum_{k=2}^{\infty} \frac{(\lambda\delta)^k}{k!} \text{ is } o(\delta)$$
Hence,
$$P(A(t+\delta)-A(t)=0) = 1 - \lambda\delta + o(\delta)$$
 (3.12)

$$P(A(t+\delta)-A(t)=1) = e^{-\lambda \delta} \lambda \delta = \lambda \delta - (\lambda \delta)^{2} + (\lambda \delta)^{3} - (\lambda \delta)^{4} + \dots$$

$$P(A(t+\delta)-A(t)=1) = \lambda \delta + o(\delta)$$
(3.13)

(b) Let [ti,tit] and [tz,tzt] be two disjoint intervals

let N be the number of arrivals in [ti,tit] U[tz,tzt]

Let N, be the number of arrivals in [ti,tit]

N, be the number of arrivals in [tz,tzt]

$$\Rightarrow N_1 = A(t_1 + T_1) - A(t_1)$$

$$N_2 = A(t_2 + T_2) - A(t_2)$$

Since [ti, ti+ []] and [ti) tz+[] are disjoint, then N= Ni+ Nz

$$P(N=n) = \sum_{k=0}^{n} P(N_1=k) P(N_2=n-k)$$

$$= \sum_{k=0}^{n} e^{-\lambda T_1} \frac{(\lambda T_1)^k}{k!} e^{-\lambda T_2} \frac{(\lambda T_2)^{n-k}}{(n-k)!}$$

$$= \frac{e^{-\lambda (T_1+T_2)}}{n!} \frac{(\lambda T_2)^n}{k!} \sum_{k=0}^{n} \frac{n!}{k! \frac{(n-k)!}{T_2}} \frac{T_1}{k!}$$

$$= \frac{e^{-\lambda (T_1+T_2)}}{n!} \frac{(\lambda T_2)^n}{n!} \left[\frac{1+\frac{T_1}{T_2}}{n!} \right]^n$$

$$= \frac{e^{-\lambda (T_1+T_2)}}{n!} \left[\lambda (T_1+T_2) \right]^n$$

(c)
$$A(t) = A_1(t) + A_2(t) + \cdots + A_k(t)$$

$$A_1, A_2, \dots, A_k \text{ are independent}$$

$$A(t'+t) - A(t) = \sum_{i=1}^{k} \left[A_i(t+t) - A_i(t) \right]$$

$$\sim Poisson R.V. \text{ with mean } (\lambda_i t)$$

As we saw in part (b), the sum of two independent Poisson R.V.s with means λ_a and λ_b is a Poisson R.V. with mean $\lambda_a t \lambda_b$ By induction, it follows that the sum of k independent Poisson R.V.s with mean $\lambda_1, \dots, \lambda_k$ is also Poisson with mean $\lambda_1, \dots, \lambda_k$

Note: one can also prove that the sum of k independent Poisson R.V.s is also a Poisson R.V. using characteristic functions

$$\begin{array}{lll}
x_{i} & \sim & P(x_{i}=k) = \frac{\lambda_{i}^{k} e^{-\lambda_{i}}}{k!} & k=0,1,\dots \\
E[e^{js\times i}] & = \sum_{k=0}^{\infty} e^{jsk} \frac{\lambda_{i}^{k} e^{-\lambda_{i}}}{k!} = e^{\lambda_{i}} \sum_{k=0}^{\infty} \frac{(\lambda_{i}e^{js})^{k}}{k!} = e^{\lambda_{i}} e^{\lambda_{i}}e^{js} \\
& = exp(-\lambda_{i}(1-e^{js})) \\
\times & = x_{1} + x_{2} + \dots + x_{k}, \quad \{x_{i}\} \text{ independent} \\
E[e^{js\times}] & = E[\frac{1}{x_{i}} e^{js\times i}] = \prod_{i=1}^{k} exp(-\lambda_{i}(1-e^{js})) \\
& = \exp((\sum_{i=1}^{k} \lambda_{i})(1-e^{js})) \\
& = \exp((\sum_{i=1}^{k} \lambda_{i})(1-e^{js}))
\end{array}$$

$$\begin{array}{ll}
& e^{is} \\
& = e^{is} e^{is} \\
& = e^{is} e^{is} \\
& = e^{is} e^{is} e^{is} e^{is} e^{is} \\
& = e^{is} e^{is} e^{is} e^{is} e^{is} e^{is} e^{is} \\
& = e^{is} e^{i$$

show, $P(\text{first arrival in A comes from } A_1) = \frac{\lambda_1}{\lambda_1}$

P(I arrival from A in [o,t] I arrival from A in [o,t])

A,8 A are not indep.

= P(1 arrival from A1 in [oit], o arrival from A2 in [oit])

P(1 arrival from A in [oit])

A18 Az are indep

= P(I arrival from A1 in [ort]) P(o arrival from A2 in [ort])
P(I arrival from A in [ort])

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

= M(larrival in [ti,s)) P(no arrival in [s,tz))

P(larrival in [ti,tz])

 $= \frac{\lambda(5-t_2) e^{-\lambda(5-t_1)} \times e^{-\lambda(t_2-t_1)}}{\lambda(t_2-t_1) e^{-\lambda(t_2-t_1)}}$

 $= \frac{s-t_1}{t_2-t_1}$