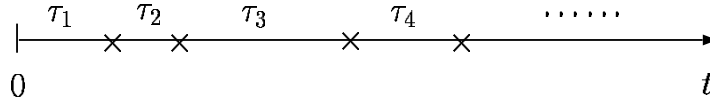


RENEWAL PROCESSES  
AND  
POISSON PROCESSES

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## Renewal Processes



A **renewal process** is a point process characterized by the fact that the successive inter-arrival times  $\tau_1, \tau_2, \tau_3$  etc... are distributed with the same cdf  $F(t)$  with density  $f(t)$  and expected value  $E[\tau] = m_1$ :

$$F(t) = Pr\{\tau \leq t\} \quad ; \quad f(t) = \frac{dF(t)}{dt}$$

$$E[\tau] = m_1$$

Let us define the corresponding Laplace transforms:

$$f^*(s) = \mathcal{L}[f(t)] \quad ; \quad F^*(s) = \mathcal{L}[F(t)]$$

The  $\tau_i$  form a sequence of independent identically distributed random variables.

Let  $s_k$  denote the time up to the  $k$  arrival:

$$s_k = \sum_{i=1}^k \tau_i \tag{1}$$

Let  $F_k(t)$  denote the cdf of  $s_k$ , and  $f_k(t)$  its density.

$$F_k(t) = Pr\{s_k \leq t\} \quad ; \quad f_k(t) = \frac{dF_k(t)}{dt}$$

In terms of Laplace transforms, we obtain:

$$F_k^*(s) = \frac{1}{s} [f^*(s)]^k \quad ; \quad f_k^*(s) = [f^*(s)]^k$$

## Renewal Processes - number of arrivals

Let  $N(t)$  denote the number of arrivals in the interval  $0 - t$ .

$$N(t) < k \quad \text{if and only if} \quad s_k > t$$

from which:

$$Pr\{N(t) < k\} = Pr\{s_k > t\} = 1 - F_k(t) \quad (2)$$

$$\begin{aligned} Pr\{N(t) = k\} &= Pr\{N(t) < k + 1\} - Pr\{N(t) < k\} \\ &= F_k(t) - F_{k+1}(t) \end{aligned} \quad (3)$$

Expected number of arrivals in  $(0 - t)$ ,  $H(t)$ :

$$\begin{aligned} H(t) = E[N(t)] &= \sum_{k=0}^{\infty} k Pr\{N(t) = k\} \\ &= \sum_{k=0}^{\infty} k [F_k(t) - F_{k+1}(t)] = \sum_{k=1}^{\infty} F_k(t) \end{aligned} \quad (4)$$

In Laplace transforms:

$$H^*(s) = \sum_{k=1}^{\infty} F_k^*(s) = \frac{1}{s} \sum_{k=1}^{\infty} f^{*k}(s)$$

## The Renewal Equation

The renewal density  $h(t)$  is defined as:

$$h(t) = \lim_{\Delta t \rightarrow 0} \frac{\text{Pr}\{\text{one or more events occur in } (t - t + \Delta t)\}}{\Delta t}$$

The probability that the  $k$ -th event occurs in  $(t - t + \Delta t)$  is:

$$\text{Pr}\{k\text{-th event occurs in } (t - t + \Delta t)\} = f_k(t) + O(\Delta t)$$

Hence:

$$h(t) = \sum_{k=1}^{\infty} f_k(t) = \frac{dH(t)}{dt}$$

In Laplace transform:

$$\begin{aligned} h^*(s) &= \sum_{k=1}^{\infty} f_k^*(s) \\ &= f^*(s) + [f^*(s)]^2 + \dots + [f^*(s)]^k + \dots \\ &= \frac{f^*(s)}{1 - f^*(s)} \end{aligned}$$

From which we obtain the **renewal equation** in LT and time domain:

$$h^*(s) = f^*(s) + h^*(s) \cdot f^*(s)$$

$$h(t) = f(t) + \int_0^t h(t-u) \cdot f(u) du$$

## The Fundamental Renewal Equation

In terms of the renewal function  $H(t)$ , the renewal equation can be derived in the following way.

In Laplace transform:

$$H^*(s) = \frac{h^*(s)}{s} = \frac{1}{s} \frac{f^*(s)}{1 - f^*(s)}$$

In a similar way, we can write:

$$\begin{aligned} H^*(s) &= \sum_{k=1}^{\infty} F_k^*(s) = \frac{1}{s} \sum_{k=1}^{\infty} f^{*k}(s) \\ &= \frac{1}{s} \{f^*(s) + f^{*2}(s) + \dots + f^{*k}(s) + \dots\} = \frac{1}{s} \frac{f^*(s)}{1 - f^*(s)} \end{aligned}$$

From the above we derive:

$$H^*(s) = \frac{f^*(s)}{s} + H^*(s) f^*(s)$$

In time domain:

$$H(t) = F(t) + \int_0^t H(t-u) \cdot f(u) du$$

This is known as the fundamental **renewal equation**.

## Poisson Process

A **Poisson process** is a renewal process in which the inter-arrival times are exponentially distributed with parameter  $\lambda$ .

$$f(t) = \lambda e^{-\lambda t} \quad \Longrightarrow \quad f^*(s) = \frac{\lambda}{s + \lambda}$$

The cdf and density of the time up to the  $k$ -th arrival  $s_k$  are in LT:

$$f_k^*(s) = \left( \frac{\lambda}{s + \lambda} \right)^k \quad ; \quad F_k^*(s) = \frac{\lambda^k}{s(s + \lambda)^k}$$

$$f_1(t) = \mathcal{L}^{-1} \left[ \frac{\lambda}{s + \lambda} \right] = \lambda e^{-\lambda t}$$

$$f_2(t) = \mathcal{L}^{-1} \left[ \frac{\lambda^2}{(s + \lambda)^2} \right] = \lambda^2 t e^{-\lambda t}$$

... ..

$$f_k(t) = \mathcal{L}^{-1} \left[ \frac{\lambda^k}{(s + \lambda)^k} \right] = \frac{\lambda (\lambda t)^{k-1}}{(k-1)!} e^{-\lambda t}$$

$$F_k(t) = \int_0^t f_k(u) du = 1 - \sum_{i=0}^{k-1} \frac{(\lambda t)^i}{i!} e^{-\lambda t}$$

## Poisson Process

Let us define:

$$P_k(t) = \Pr\{N(t) = k\} = F_k(t) - F_{k+1}(t)$$

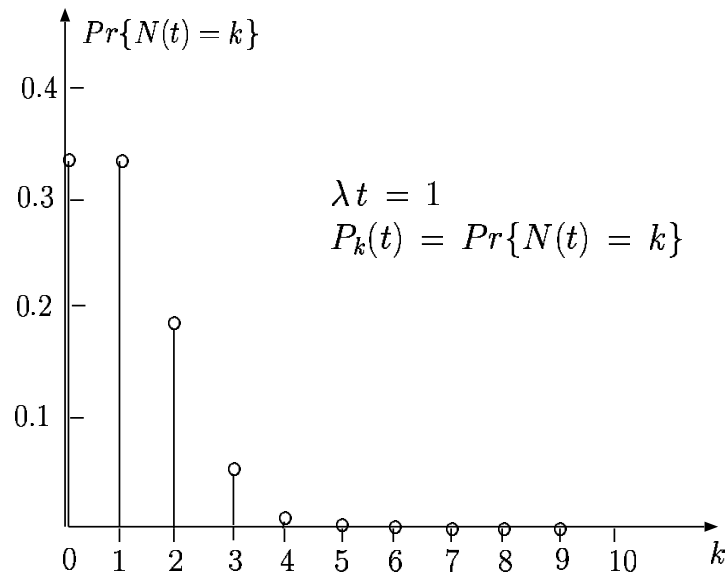
Taking Laplace transforms:

$$P_k^*(s) = \frac{\lambda^k}{s(s + \lambda)^k} - \frac{\lambda^{k+1}}{s(s + \lambda)^{k+1}} = \frac{\lambda^k}{(s + \lambda)^{k+1}}$$

Inverting again in the time domain, we obtain the **Poisson distribution**:

$$P_k(t) = \Pr\{N(t) = k\} = \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$

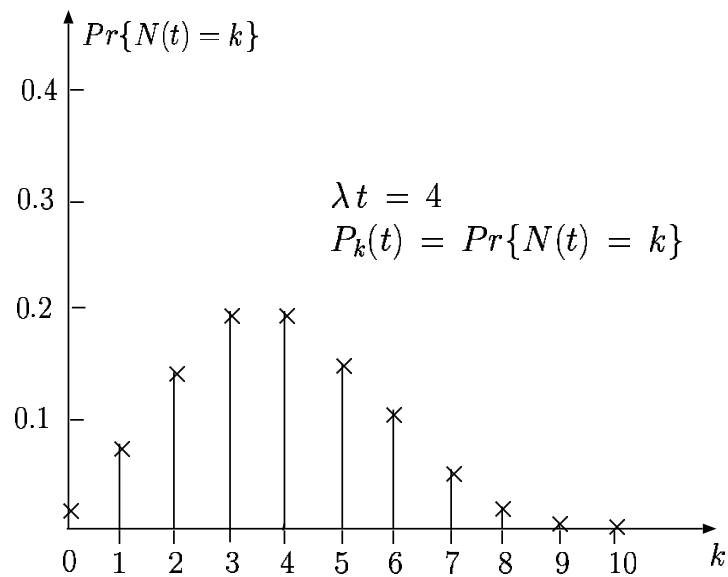
# Poisson Distribution



$k$	$P_k$
0	0.368
1	0.368
2	0.184
3	0.061
4	0.015
5	0.003
6	$5.1 \cdot 10^{-4}$
...	...
10	$1.01 \cdot 10^{-7}$



# Poisson Distribution



$k$	$P_k$
0	0.0183
1	0.073
2	0.146
3	0.195
4	0.195
5	0.156
6	0.104
7	0.059
8	0.029
9	0.013
10	0.005

## Poisson Process

### Expected number of events

$$\begin{aligned}
 H(t) &= E[N(t)] = \sum_{k=0}^{\infty} k \Pr\{N(t) = k\} = \sum_{k=0}^{\infty} k P_k(t) \\
 &= e^{-\lambda t} \sum_{k=1}^{\infty} \frac{(\lambda t)^k}{(k-1)!} \\
 &= \lambda t \cdot e^{-\lambda t} (1 + \lambda t + \frac{(\lambda t)^2}{2!} + \dots) = \lambda t
 \end{aligned}$$

An alternative derivation in Laplace transforms:

$$H^*(s) = E^*[N(s)] = \frac{1}{s} \frac{f^*(s)}{1 - f^*(s)}$$

Since in the Poisson process:

$$\begin{aligned}
 f^*(s) = \frac{\lambda}{s + \lambda} &\implies E^*[N(s)] = \frac{\lambda}{s^2} \\
 &\implies h^*(s) = \frac{\lambda}{s}
 \end{aligned} \tag{5}$$

We obtain:

$$H(t) = E[N(t)] = \lambda t$$

$$h(t) = \lambda$$

## Buffer design in a Poisson Process

Jobs arrive according to a Poisson process of rate  $\lambda$  and must be stored in a buffer during an interval  $T$ .

The design problem consists in evaluating the number of slots in a buffer such that the probability of refusing an incoming job in the interval  $T$  is less than a prescribed (small) risk  $\alpha$  ( $0 \leq \alpha \leq 1$ ).

Let  $N(T)$  be the number of arrivals in the interval  $T$  and let  $K$  be the number of slots to be determined.

The design problem can be formulated as:

$$Pr\{N(T) > K\} \leq \alpha$$

$$Pr\{N(T) \leq K\} > 1 - \alpha$$

$$Pr\{N(T) \leq K\} = \sum_{j=0}^K \frac{(\lambda T)^j}{j!} e^{-\lambda T} > 1 - \alpha$$

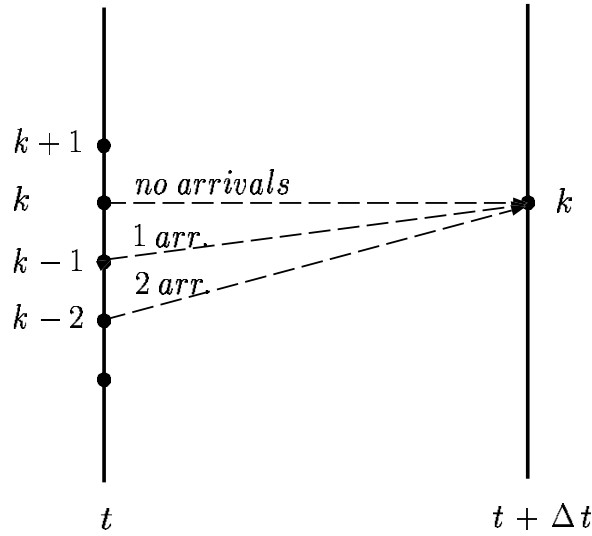
The value of  $K$  is the smallest integer that satisfies the above equation.

## Poisson Process

### An alternative derivation

A counting process  $N(t)$  is a stochastic point process identified by a sequence of random points in time. The **state** of the process at time  $t$  is defined by the number of arrivals  $N(t)$  in the interval  $(0 - t)$ .

Let  $E_i(t)$  be the event  $N(t) = i$ , i.e.  $i$  arrivals in  $(0 - t)$ .



According to the figure, the event  $E_i(t + \Delta t)$  can be decomposed into the sequence of independent and mutually exclusive events:

$$\begin{aligned}
 E_i(t + \Delta t) &= \{E_i(t), \text{ no arrivals in } \Delta t\} \\
 &+ \{E_{i-1}(t), 1 \text{ arrival in } \Delta t\} \\
 &+ \{E_{i-2}(t), 2 \text{ arrivals in } \Delta t\} \\
 &+ \dots
 \end{aligned}$$

## Poisson Process

### An alternative derivation

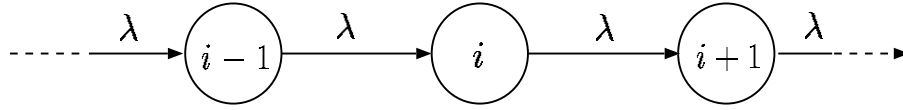
By the theorem of the total probability, we can write:

$$\begin{aligned} Pr\{N(t + \Delta t) = i\} &= Pr\{\text{no arrivals in } \Delta t \mid N(t) = i\} \cdot \\ &\quad \cdot Pr\{N(t) = i\} \\ &+ Pr\{1 \text{ arrival in } \Delta t \mid N(t) = i - 1\} \cdot \\ &\quad \cdot Pr\{N(t) = i - 1\} \\ &+ Pr\{2 \text{ arrivals in } \Delta t \mid N(t) = i - 2\} \cdot \\ &\quad \cdot Pr\{N(t) = i - 2\} \\ &+ \dots \end{aligned}$$

## Poisson Process

### An alternative derivation

A stochastic point process  $N(t)$  is a Poisson process, if the probability of having one event in any interval  $dt$  is constant and equal to  $\lambda$ .



By the theorem of the total probability, we can write for  $i > 0$ :

$$Pr\{N(t + \Delta t) = i | N(t) = i\} = 1 - \lambda \Delta t + O(\Delta t)$$

$$Pr\{N(t + \Delta t) = i | N(t) = i - 1\} = \lambda \Delta t + O(\Delta t)$$

Where:

$$\lim_{\Delta t \rightarrow 0} \frac{O(\Delta t)}{\Delta t} = 0$$

For  $i = 0$ , we can write:

$$Pr\{N(t + \Delta t) = 0 | N(t) = 0\} = 1 - \lambda \Delta t + O(\Delta t)$$

## Poisson Process

### An alternative derivation

Let us define:

$$P_i(t) = \Pr\{N(t) = i\}$$

According to the above relations we can write:

$$\begin{cases} P_0(t + \Delta t) = (1 - \lambda \Delta t) P_0(t) & i = 0 \\ P_i(t + \Delta t) = (1 - \lambda \Delta t) P_i(t) + \lambda \Delta t P_{i-1}(t) & i > 0 \end{cases}$$

$$\begin{cases} \frac{P_0(t + \Delta t) - P_0(t)}{\Delta t} = -\lambda P_0(t) & i = 0 \\ \frac{P_i(t + \Delta t) - P_i(t)}{\Delta t} = -\lambda P_i(t) + \lambda P_{i-1}(t) & i > 0 \end{cases}$$

Taking the limit  $\Delta t \rightarrow 0$ , a Poisson process is characterized by the following set of linear differential equations:

$$\begin{cases} \frac{d P_0(t)}{d t} = -\lambda P_0(t) & i = 0 \\ \frac{d P_i(t)}{d t} = -\lambda P_i(t) + \lambda P_{i-1}(t) & i > 0 \end{cases}$$

with initial conditions:

$$\begin{cases} P_0(0) = 1 & i = 0 \\ P_i(0) = 0 & i > 0 \end{cases}$$

## Poisson Process

### An alternative derivation

Taking Laplace transforms of the time domain differential equations:

$$\begin{cases} s P_0^*(s) - 1 = -\lambda P_0^*(s) & i = 0 \\ s P_i^*(s) = -\lambda P_i^*(s) + \lambda P_{i-1}^*(s) & i > 0 \end{cases}$$

$$P_0^*(s) = \frac{1}{s + \lambda}$$

$$P_1^*(s) = \frac{\lambda}{s + \lambda} \cdot P_0^*(s) = \frac{\lambda}{(s + \lambda)^2}$$

$$\dots \quad \dots$$

$$P_i^*(s) = \frac{\lambda}{s + \lambda} \cdot P_{i-1}^*(s) = \frac{\lambda^i}{(s + \lambda)^{i+1}}$$

$$\dots \quad \dots$$

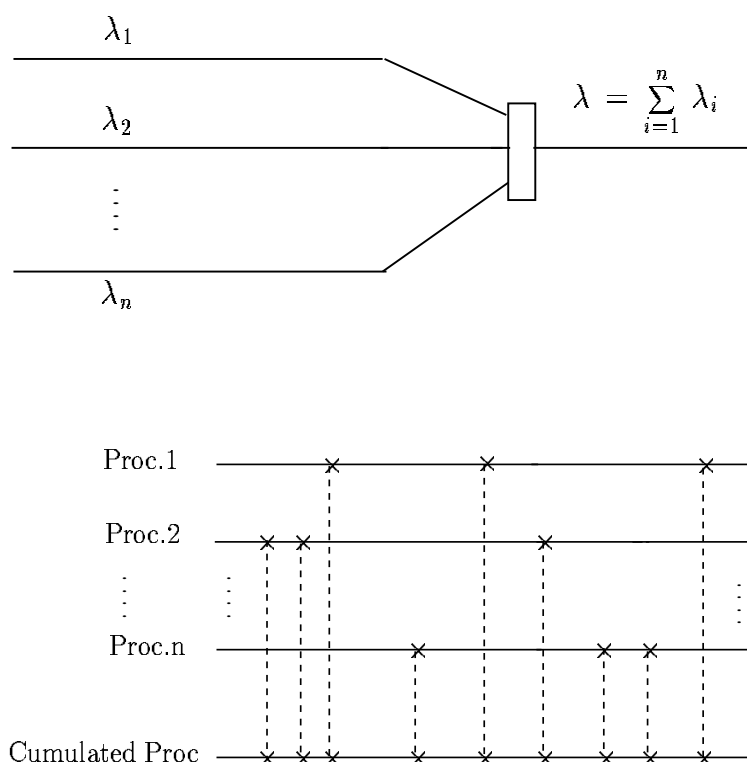
From the above equations we derive again the Poisson distribution:

$$P_k(t) = Pr\{N(t) = k\} = \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$



## Superposition of Poisson Processes

A superposition of Poisson processes is obtained by cumulating the occurrences of  $n$  independent sources of Poisson processes with parameters  $\lambda_1, \lambda_2, \dots, \lambda_n$ , respectively.



The cumulated process is still a Poisson process:

$$Prob\{N(t + \Delta t) = k + 1 \mid N(t) = k\} =$$

$$\lambda_1 \Delta t + \lambda_2 \Delta t + \dots + \lambda_n \Delta t + O(t)$$

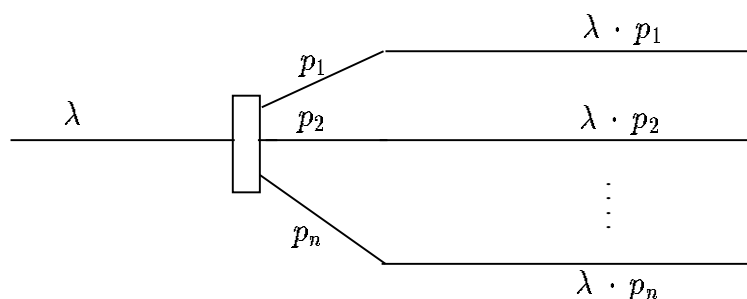
The parameter of the cumulated Poisson process is given by:

$$\lambda = \sum_{i=1}^n \lambda_i$$

## Decomposition of Poisson Processes

A Poisson process with parameter  $\lambda$  is split into  $n$  independent branches according to a probability distribution  $p_1, p_2, \dots, p_n$  with:

$$\sum_{i=1}^n p_i = 1$$

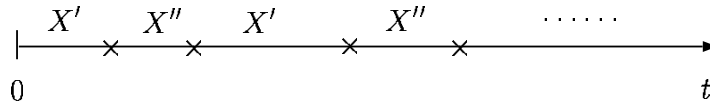


By decomposing the original Poisson process of parameter  $\lambda$ ,  $n$  Poisson processes are generated:  $N_1(t), N_2(t), \dots, N_n(t)$ , with parameters:  $\lambda p_1, \lambda p_2, \dots, \lambda p_n$ .

$$Pr\{N_i(t + \Delta t) = k + 1 \mid N_i(t) = k\} = p_i \lambda \Delta t + O(\Delta t)$$

## Alternating Renewal Processes

The process is constituted by a sequence of *Type I* variables  $X'$  with density  $f_1(x)$  followed by a *Type II* variables  $X''$  with density  $f_2(x)$ . The process starts with probability 1 with a *Type I* variable.



If we look at the sequence formed by the occurrence of the *Type II* variables, the process is an ordinary renewal process with inter-arrival time  $(X' + X'')$ .

The mean number of *Type II* occurrences satisfies (in LT):

$$H_2^*(s) = \frac{f_1^*(s) \cdot f_2^*(s)}{s \{ 1 - f_1^*(s) \cdot f_2^*(s) \}}$$

For the *Type I* occurrences we have a modified renewal process, for which the expected number of renewals is:

$$H_1^*(s) = \frac{f_1^*(s)}{s \{ 1 - f_1^*(s) \cdot f_2^*(s) \}}$$

In both cases:

$$h_i^*(s) = s H_i^*(s) \quad i = 1, 2$$

## Alternating Renewal Processes

Let be:

$\pi_1(t)$  - Probability *Type I* variable occurs at time  $t$

$\pi_2(t)$  - Probability *Type II* variable occurs at time  $t$

*Type I* is in use at time  $t$  if:

a) - No *Type I* event occurs in  $(0 - t)$ ;

b) - A *Type II* event occurs in  $u - u + du$  ( $u < t$ ), and no *Type I* events occur in  $(t - u)$ :

$$\pi_1(t) = [1 - F_1(t)] + \int_0^t h_2(u) [1 - F_1(t - u)] du$$

In Laplace transform:

$$\begin{aligned} \pi_1^*(s) &= \frac{1 - f_1^*(s)}{s} + \frac{f_1^*(s) \cdot f_2^*(s)}{1 - f_1^*(s) \cdot f_2^*(s)} \frac{1 - f_1^*(s)}{s} \\ \pi_1^*(s) &= \frac{1 - f_1^*(s)}{s \{ 1 - f_1^*(s) \cdot f_2^*(s) \}} \end{aligned}$$

Note also that:

$$\pi_1^*(s) = H_2^*(s) - H_1^*(s) + \frac{1}{s}$$

$$\pi_1(t) = H_2(t) - H_1(t) + 1$$

## Alternating Poisson Process

*Type I* variable is exponential with rate  $\lambda$ ;

*Type II* variable is exponential with rate  $\mu$ .

$$\begin{aligned}\pi_1^*(s) &= \frac{s + \mu}{s(s + \lambda + \mu)} \\ &= \frac{\mu}{\lambda + \mu} \cdot \frac{1}{s} + \frac{\lambda}{\lambda + \mu} \cdot \frac{1}{s + \lambda + \mu}\end{aligned}$$

$$\pi_1(t) = \frac{\mu}{\lambda + \mu} + \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

$$\pi_2^*(s) = \frac{\lambda}{s(s + \lambda + \mu)}$$

$$\pi_2(t) = \frac{\lambda}{\lambda + \mu} - \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

$$\lim_{t \rightarrow \infty} \pi_1(t) = \frac{\mu}{\lambda + \mu} \quad ; \quad \lim_{t \rightarrow \infty} \pi_2(t) = \frac{\lambda}{\lambda + \mu}$$