

*the  
Poisson  
process*

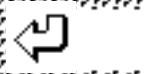
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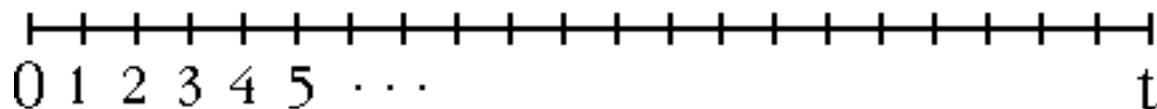
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The Poisson Process  
as a limiting case of  
the Bernoulli Process



Consider the following situation:



A time interval of length  $t$  seconds is divided into one-second intervals, with the probability of a vehicle arriving at an intersection during a one-second interval being a small number  $p$ . (Assume that the probability that more than one vehicle arrives is negligible.)

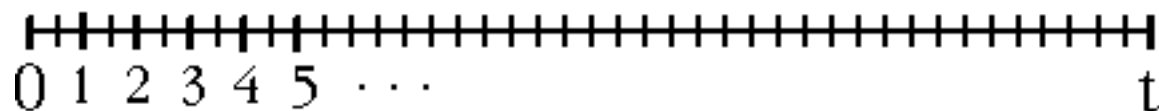
Consider the Bernoulli process  $\{X_k; k=1,2,\dots\}$  where  $X_k = 1$  if a vehicle arrives during the  $k^{\text{th}}$  second, and the associated counting process  $\{N_t\}$  which counts the number of arrivals during the interval  $[0,t]$ .

Then  $N_t$  has the binomial distribution:

$$P\{N_t = x\} = \binom{t}{x} p^x (1-p)^{t-x}$$

with expected value  $v = tp$ .

Consider what happens as we divide  $[0,t]$  into  $n$  smaller time intervals, but in such a way that the expected number of arrivals in  $[0,t]$  remains constant,  $v$ .



That is, the probability of an arrival in each of these small intervals must be  $v/n$ , and

$$P\{N_t = x\} = \binom{n}{x} \left(\frac{v}{n}\right)^x \left(1 - \frac{v}{n}\right)^{n-x}$$

$$\begin{aligned}
 P\{N_t = x\} &= \binom{n}{x} \left(\frac{v}{n}\right)^x \left(1 - \frac{v}{n}\right)^{n-x} \\
 &= \left(\frac{n!}{x!(n-x)!}\right) \left(\frac{v^x}{n^x}\right) \left(1 - \frac{v}{n}\right)^n \left(1 - \frac{v}{n}\right)^{-x} \\
 &= \frac{v^x}{x!} \left(1 - \frac{v}{n}\right)^n \left(\frac{n!}{(n-x)!}\right) \frac{1}{n^x \left(1 - \frac{v}{n}\right)^x}
 \end{aligned}$$

Consider the limit of this distribution as  $n \rightarrow +\infty$

$$\begin{aligned}
 P\{N_t = x\} &= \frac{v^x}{x!} \underbrace{\left(1 - \frac{v}{n}\right)^n}_{\downarrow e^{-v}} \underbrace{\left(\frac{n!}{(n-x)!}\right)}_{\parallel} \frac{1}{\underbrace{n^x \left(1 - \frac{v}{n}\right)^x}_{\downarrow \left[n\left(1 - \frac{v}{n}\right)\right]^x}} \\
 &= \frac{v^x}{x!} e^{-v} \frac{n(n-1)(n-2) \dots (n-x+1)}{\left[n\left(1 - \frac{v}{n}\right)\right]^x} \\
 &\rightarrow \frac{v^x}{x!} e^{-v} = 1
 \end{aligned}$$



$$P\{N_t = x\} = \frac{v^x}{x!} e^{-v}$$

If the arrival rate is  $\lambda$  /second, then  $v = \lambda t$   
and

$$P\{N_t = x\} = \frac{(\lambda t)^x}{x!} e^{-\lambda t}$$

for  $x=0, 1, 2, 3, \dots$

Poisson  
Distribution



**Poisson  
Distribution**

$$P\{N_t = x\} = \frac{(\lambda t)^x}{x!} e^{-\lambda t}$$

for  $x=0, 1, 2, 3, \dots$

**Mean Value**

$$E(N_t) = \lambda t$$

**Variance**

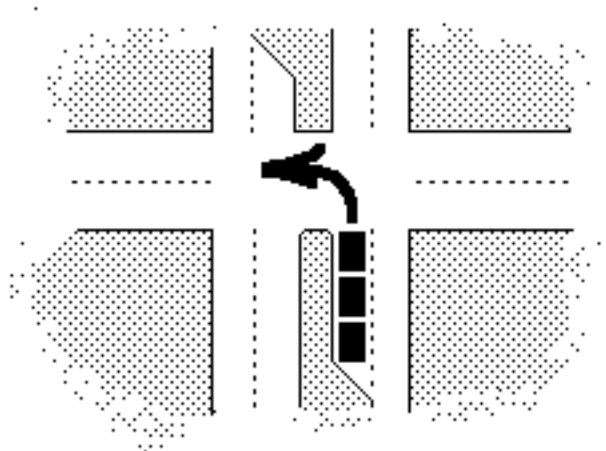
$$\text{Var}(N_t) = \lambda t$$

*mean and variance are equal!*



**Example** A left-turn lane at an intersection has a capacity of 3 autos. 30% of autos arriving at the intersection wish to turn left. The *expected number* of autos arriving during a red signal is 6.

What is the probability that the capacity of the left-turn lane is exceeded during a red signal?



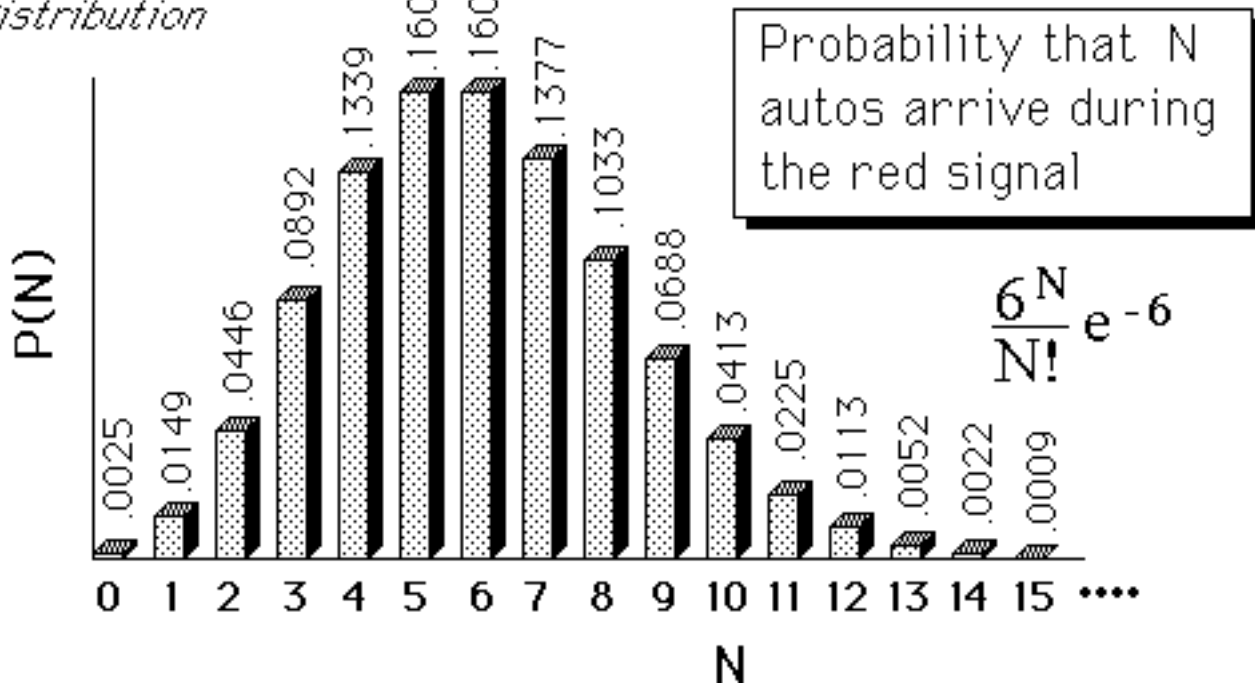
Given that  $\mathbf{N}$  autos arrive, the number  $\mathbf{X}$  of left-turning autos has the *binomial* distribution.

The number  $\mathbf{N}$  of autos arriving during the red signal has the *Poisson* distribution.

$$P\{X>3\} = \sum_{N=4}^{\infty} \underbrace{P\{X>3 | N \text{ arrivals}\}}_{\text{computed using binomial distr.}} \underbrace{P\{N \text{ arrivals}\}}_{\text{computed using Poisson distr.}}$$

$$\begin{aligned} P\{X>3 | N \text{ arrivals}\} &= \sum_{x=4}^N \binom{N}{x} (0.3)^x (0.7)^{N-x} && \boxed{\text{binomial distr.}} \\ &= 1 - \sum_{x=0}^3 \binom{N}{x} (0.3)^x (0.7)^{N-x} \\ P\{N \text{ arrivals}\} &= \frac{6^N}{N!} e^{-6} && \boxed{\text{Poisson distr.}} \end{aligned}$$

*Poisson  
Distribution*





## Time between arrivals

Suppose that the number of arrivals in an interval has the Poisson distribution with arrival rate  $\lambda$  /second.

Let  $T_1 =$  time of the first arrival.

What is the distribution of  $T_1$  ?





$$P\{T_1 > t\} = P\{\text{NO arrivals occur in interval } [0, t]\}$$

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

↖ Poisson distribution  $P\{N_t=0\}$

$$\text{CDF: } P\{T_1 \leq t\} = F(t) = 1 - e^{-\lambda t}$$

$$\text{Density function: } f(t) = \frac{d}{dt}F(t) = \lambda e^{-\lambda t}$$

Exponential  
Distribution

## Exponential Distribution

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

Mean Value

$$E(T_1) = \frac{1}{\lambda}$$

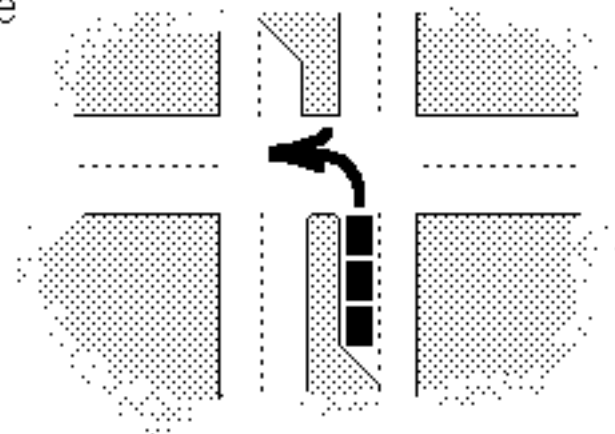
Variance

$$\text{Var}(T_1) = \frac{1}{\lambda^2}$$

**Example**

Suppose that the arrival rate for northbound autos is 6 per 30 second red signal, i.e.,  $0.2/\text{second}$

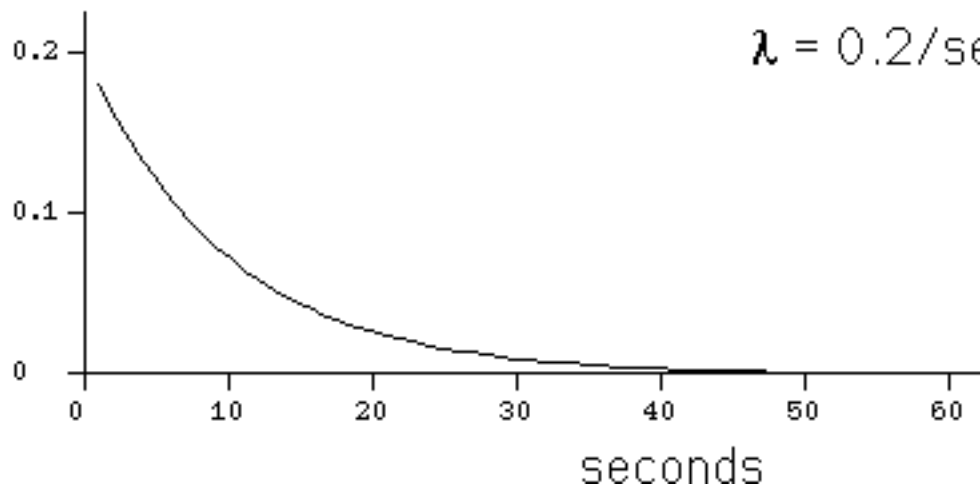
What is the distribution of the arrival time of the first auto?  
(This will also be the distribution of the time between arrivals!)



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

Exponential  
Distribution

$$\lambda = 0.2/\text{sec.}$$

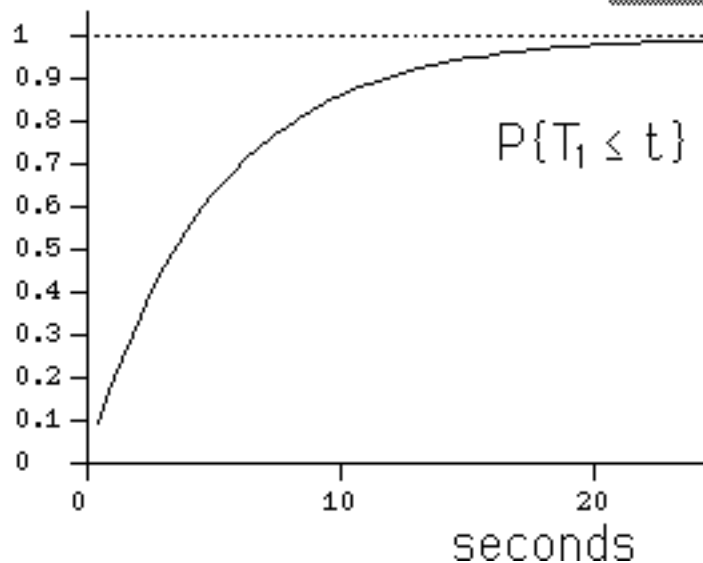


$\lambda = 0.2/\text{sec.}$

t	F(t)
1	0.18127
2	0.32968
3	0.45119
4	0.55067
5	0.63212
6	0.69881
7	0.75340
8	0.79810
9	0.83470
10	0.86466
11	0.88920
12	0.90928
13	0.92573
14	0.93919
15	0.95021
16	0.95924
17	0.96663
18	0.97268
19	0.97763
20	0.98168

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

Exponential  
Distribution



**Memoryless  
Property**

**Exponential  
Distribution**

Suppose that it is known that, at time  $t_0$ , the first arrival has not yet occurred, i.e.,  $T_1 > t_0$ .

What is the conditional distribution of  $T_1$  ?

That is, what is  $P\{T_1 \leq t \mid T_1 > t_0\}$  for  $t > t_0$  ?



Memoryless  
Property

Exponential  
Distribution

$$\begin{aligned} P\{T_1 \leq t \mid T_1 > t_0\} &= \frac{P\{T_1 \leq t \cap T_1 > t_0\}}{P\{T_1 > t_0\}} = \frac{P\{t_0 \leq T_1 \leq t\}}{P\{T_1 > t_0\}} \\ &= \frac{F(t) - F(t_0)}{1 - F(t_0)} = \frac{(1 - e^{-\lambda t}) - (1 - e^{-\lambda t_0})}{e^{-\lambda t_0}} \\ &= \frac{e^{-\lambda t_0} - e^{-\lambda t}}{e^{-\lambda t_0}} = 1 - e^{-\lambda(t-t_0)} \end{aligned}$$

**Memoryless  
Property****Exponential  
Distribution**

$$P\{T_1 \leq t \mid T_1 > t_0\} = 1 - e^{-\lambda(t-t_0)} = P\{T_1 \leq t - t_0\}$$

If the time  $\tau$  is reckoned from time  $t_0$ ,  
i.e.,  $\tau = t - t_0$ , then

$$P\{T_1 \leq t \mid T_1 > t_0\} = P\{T_1 \leq t - t_0\} = P\{T_1 \leq \tau\}$$

In other words, the failure of an arrival to occur before time  $t_0$  does not alter one's prediction of the length of time (from  $t_0$ ) before the next arrival.



## Time of $k^{\text{th}}$ Arrival

Let  $T_k$  = time of  $k^{\text{th}}$  arrival,

$\tau_k = T_k - T_{k-1}$  = time between arrivals  $k-1$  and  $k$ .

Suppose that  $\tau_k$  ( $k=1,2,3,\dots$ ) have identical and independent exponential distributions with rate  $\lambda$ .

Then  $T_k$  is the *sum* of  $k$  random variables with exponential distributions.

It is said to have a  *$k$ -Erlang* distribution.



## Erlang Distribution

*time of  $k^{\text{th}}$  arrival in  
a Poisson process*

## Density function

$(k > 0, \lambda > 0, t \geq 0)$

$$\begin{aligned} f(t) &= \frac{\lambda (\lambda t)^{k-1} e^{-\lambda t}}{(k-1)!} \\ &= \frac{\lambda (\lambda t)^{k-1} e^{-\lambda t}}{\Gamma(k)} \end{aligned}$$

where the *Gamma*  
function is defined by  
(for  $k > 0$ , not necessarily  
integer!)

$$\begin{aligned} \Gamma(k) &= \int_0^{\infty} e^{-u} u^{k-1} du \\ &= (k-1)! \quad \text{if } k \text{ integer} \end{aligned}$$

## Erlang Distribution

CDF

$$F(t) = \frac{\Gamma(k, \lambda t)}{\Gamma(k)}$$

where  $\Gamma(k, x)$  is the "*incomplete Gamma function*"  
defined by

$$\Gamma(k, x) = \int_0^x e^{-u} u^{k-1} du$$

*tabulated*

Alternate computation,  
when  $k$  is integer

**CDF**

$$\begin{aligned}F(t) &= P\{T_k \leq t\} = P\{N_t \geq k\} \\ &= 1 - P\{N_t < k\} \\ &= 1 - P\{N_t \leq k-1\}\end{aligned}$$

where  $N_t = \#$  arrivals at time  $t$   
has the *Poisson* distribution:

$$F(t) = 1 - \sum_{x=0}^{k-1} \frac{(\lambda t)^x}{x!} e^{-\lambda t}$$

## Erlang Distribution

Mean Value

$$\mu = \frac{k}{\lambda}$$

Variance

$$\sigma^2 = \frac{k}{\lambda^2}$$

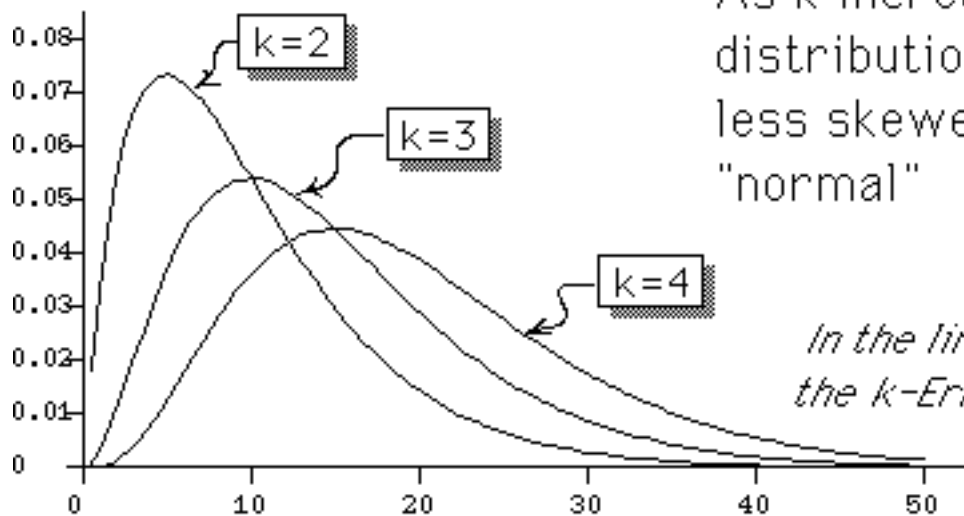
*(These expressions result from the fact that the random variable is the sum of  $k$  i.i.d. random variables.)*

More generally, when  $k$  is not an integer, the probability distribution is called the *Gamma* distribution.

## Erlang Distribution

*example:*

$$\lambda = 0.2$$



As  $k$  increases, the distribution becomes less skewed, and more "normal"

*In the limit, as  $k \rightarrow \infty$ , the  $k$ -Erlang distribution converges to the Normal distribution!*

**Bernouilli process**

Binomial distn.

# of events

Geometric distn.

time until  
1st event

Pascal distn.

time until  
 $k^{\text{th}}$  event**Poisson process**

Poisson distn.

Exponential distn.

Erlang distn.

