



Siméon Denis Poisson  
(1781-1840)<sub>France</sub>

“Life is good for only two things:  
discovering mathematics and teaching  
mathematics.”

## Poisson Process

- **Wikipedia:** A **Poisson process**, named after the French mathematician **Siméon-Denis Poisson** (1781 – 1840), is the **stochastic process** in which events (e.g. arrivals) occur continuously and **independently of one another**.
- **Formal Definition:** The Poisson Process is a **counting function**  $\{N(\tau), \tau \geq 0\}$  where  $N(\tau)$  is the number of events that have occurred up to time  $t$ , i.e. in the interval  $[0, \tau]$ .
- **Fact:** The number of events between time  $a$  and time  $b$  is given as  $N(b) - N(a)$  and has a **Poisson distribution**.
- The Poisson process is a **continuous-time process**: Time is **continuous**
  - Its discrete-time counterpart is the *Bernoulli process*
    - Bernoulli process is a discrete-time stochastic process consisting of a sequence of independent random variables taking values over two symbols.

## Examples of using Poisson Process

- The number of web page requests arriving at a server may be characterized by a Poisson process except for unusual circumstances such as coordinated denial of service attacks.
- The number of telephone calls arriving at a switchboard, or at an automatic phone-switching system, may be characterized by a Poisson process.
- The number of raindrops falling over a wide spatial area may be characterized by a spatial Poisson process.
- The arrival of "customers" is commonly modelled as a Poisson process in the study of simple queueing systems.
- The execution of trades on a stock exchange, as viewed on a tick by tick basis, is a Poisson process.

3

## Applications of Poisson<sub>Review</sub>

- ▶ **Context:** number of events occurring in a fixed period of time
  - ▶ Events occur with a known average rate and are **independent**
- ▶ Poisson distribution is characterized by the **average rate**  $\lambda$ 
  - ▶ The average number of arrival in the fixed time period.
- ▶ **Examples**
  - ▶ The number of cars passing a fixed point in a 5 minute interval.  
*Average rate:  $\lambda = 3$  cars/5 minutes*
  - ▶ The number of calls received by a switchboard during a given period of time. *Average rate:  $\lambda = 3$  call/minutes*
  - ▶ The number of message coming to a router per second
  - ▶ The number of travelers arriving to the airport for flight registration

## Poisson Distribution Review

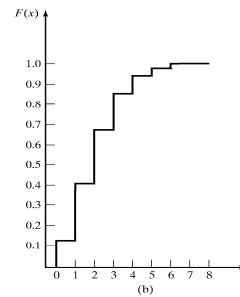
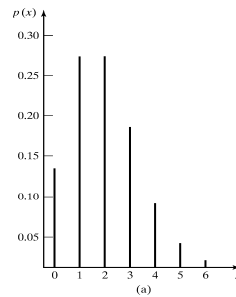
- The Poisson distribution with the average rate parameter  $\lambda$

$$\text{PMF: } p(k) = P(X = k) = \begin{cases} \frac{\lambda^k}{k!} \exp(-\lambda) & \text{for } k = 0, 1, 2, \dots \\ 0, & \text{otherwise} \end{cases}$$

$$\text{CDF: } F(k) = P(X \leq k) = \sum_{i=0}^k \frac{\lambda^i}{i!} \cdot \exp(-\lambda)$$

$$\text{Expected value: } E[X] = \lambda$$

$$\text{Variance: } V[X] = \lambda$$



## Poisson R.V. Review

- Ex. A production line with .4 percent of its items are defective,  $n=500$  items are taken for a quality control. What is the probability that 0, 1, 3 items of them are defective
- That is  $X \sim b(500, 0.004)$  approx. To Poisson

$$P(X = k) = \frac{(\lambda)^k}{k!} e^{-\lambda}$$

$$\lambda = 500 * .004 = 2$$

$$P(x = 0) = e^{-2}$$

$$P(x = 1) = 2e^{-2}$$

$$P(x = 3) = \frac{4}{3} e^{-2}$$

## Example: Poisson Review

- The number of cars that enter the parking follows a Poisson distribution with a mean rate equal to  $\lambda = 20$  cars/hour
  - The probability of having **exactly 15 cars** entering the parking in **one hour**:

$$p(15) = P(X = 15) = \frac{20^{15}}{15!} \cdot \exp(-20) = 0.051649$$

## Example: Poisson Review

- Example: A computer repair person is “beeped” each time there is a call for service. The number of beeps per hour  $\sim$  Poisson( $\alpha = 2$  per hour).

- The probability of three beeps in the next hour:

$$P(X = 3) = \frac{(\lambda)^k}{k!} e^{-\lambda} = \frac{(2)^3}{3!} e^{-2} = 0.18$$

Or

$$P(X = 3) = F(3) - F(2) = 0.857 - 0.677 = 0.18$$

- The probability of two or more beeps in a 1-hour period:

$$P(X \geq 2) = 1 - p(0) - p(1) = 1 - F(1) = 0.594$$

## Stationary Poisson Process

- Also known as **homogeneous Poisson process** is characterized by a **CONSTANT rate parameter  $\lambda$** , also known as **intensity**, such that the number of events in time interval  $(t, t + \tau]$  follows a Poisson distribution with associated parameter  $\lambda \cdot \tau$ .
- Formally, A counting process  $\{N(t), t \geq 0\}$  is a (Stationary) Poisson process with mean rate  $\lambda$  if:

for  $t \geq 0$  and  $n = 0, 1, 2, \dots$

$$\text{PMF: } p[N(t + \tau) - N(t) = n] = p[N(\tau) = n] = \frac{(\lambda \cdot \tau)^n}{n!} \exp(-\lambda \cdot \tau)$$

- $N(t + \tau) - N(t)$  describes the number of events in time interval

- The mean and the variance are equal

$$E[N(\tau)] = V[N(\tau)] = \lambda \cdot \tau$$

## Stationary Poisson Process

- **Properties of Poisson process**

- Arrivals occur one at a time (not simultaneous)
- $\{N(\tau), \tau \geq 0\}$  has **stationary increments**, which means
 
$$N(t) - N(s) = N(t - s)$$

The number of arrivals in time  $s$  to  $t$  is also Poisson-distributed with mean

- $\lambda \cdot (t - s)$  has **independent increments**
- $$\{N(\tau), \tau \geq 0\}$$

## Generating a Poisson Process

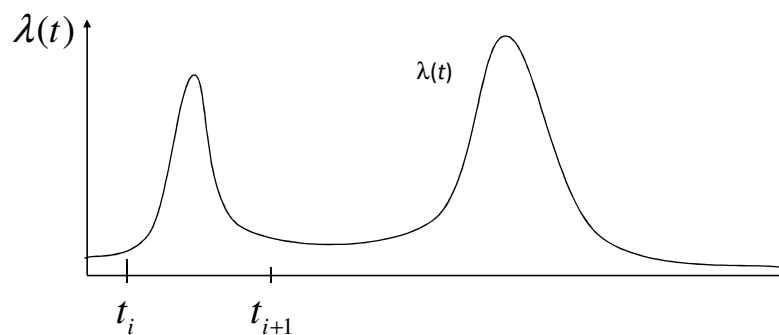
- Stationary with rate  $\lambda > 0$
- Time between events  $A_i = t_i - t_{i-1}$  are IID exponential
- Algorithm

1. Generate  $U \sim U(0,1)$
2. Return  $t_i = t_{i-1} - \left(\frac{1}{\lambda}\right) \ln U$

11

## Nonstationary Case

- Can we simply generalize?



12

## Thinning Algorithm

1. Set  $t = t_{i-1}$
2. Generate  $U_1, U_2$  I.I.D.  $U(0,1)$
3. Replace  $t$  by

$$t - \left( \frac{1}{\lambda^*} \right) \ln U_1, \text{ where } \lambda^* = \max_t \{\lambda(t)\}$$

4. If  $U_2 \leq \lambda(t) / \lambda^*$  return  $t_i = t$ . Otherwise, go back to step 2.

13

## (Homogenous) Poisson Process

- The **homogeneous Poisson process** is characterized by a **CONSTANT rate parameter  $\lambda$** , also known as **intensity**, such that the number of events in time interval  $(t, t + \tau]$  follows a Poisson distribution with associated parameter  $\lambda \cdot \tau$ .
- Formally, A counting process  $\{N(\tau), \tau \geq 0\}$  is a (homogenous) Poisson process with mean rate  $\lambda$  if:

for  $t \geq 0$  and  $n = 0, 1, 2, \dots$

$$\text{PMF: } p[N(t + \tau) - N(t) = n] = p[N(\tau) = n] = \frac{(\lambda \cdot \tau)^n}{n!} \exp(-\lambda \cdot \tau)$$

- $N(t + \tau) - N(t)$  describes the number of events in time interval  $(t, t + \tau]$
- The mean and the variance are equal  $E[N(\tau)] = V[N(\tau)] = \lambda \cdot \tau$

14

## (Homogenous) Poisson Process

- **Properties of Poisson process**

- Arrivals occur one at a time (not simultaneous)
- $\{N(\tau), \tau \geq 0\}$  has **stationary increments**, which means
 
$$N(t) - N(s) = N(t - s)$$

The number of arrivals in time  $s$  to  $t$  is also Poisson-distributed with mean  $\lambda \cdot (t - s)$

- $\{N(\tau), \tau \geq 0\}$  has **independent increments**

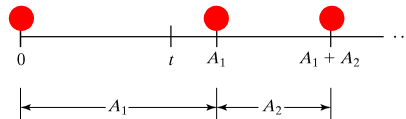
15

## Inter-Arrival Times of a Poisson Process

CDF of Exponential distribution

$$F(x; \lambda) = \begin{cases} 1 - e^{-\lambda x}, & x \geq 0, \\ 0, & x < 0. \end{cases}$$

- **Inter-arrival time:** time between two consecutive arrivals
  - The inter-arrival times of a Poisson process are random.
  - What is its distribution?**
- Consider the **inter-arrival times** of a Poisson process  $(A_1, A_2, \dots)$ , where  $A_i$  is the elapsed time between arrival  $i$  and arrival  $i+1$



- The first arrival occurs after time  $t$  **MEANS** that there are no arrivals in the interval  $[0, t]$ , As a consequence:

$$p(A_1 > t) = p(N(t) = 0) = \exp(-\lambda \cdot t)$$

$$p(A_1 \leq t) = 1 - p(A_1 > t) = 1 - \exp(-\lambda \cdot t)$$

The Inter-arrival times of a Poisson process are exponentially distributed and independent with mean  $1/\lambda$

16

## Slide 16

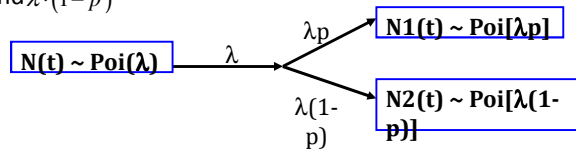
---

**B1** Poi is not an abbreviation of Poisson that I have ever seen  
Brian, 07/01/2005

# Splitting and Pooling

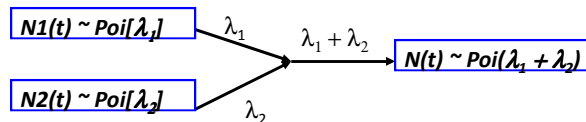
- Splitting**

- A Poisson process can be split into two Poisson processes: The first with a probability  $p$  and the second with probability  $1-p$ .
- $N(t) = N_1(t) + N_2(t)$  where  $N_1(t)$  and  $N_2(t)$  both Poisson processes with rates  $\lambda \cdot p$  and  $\lambda \cdot (1-p)$



- Pooling**

- The summation of two Poisson processes is a Poisson process
- $N_1(t) + N_2(t) = N(t)$  where  $N_1(t)$  and  $N_2(t)$  are Poisson processes with rates  $\lambda_1 + \lambda_2$



## Non Homogenous (Non-stationary) Poisson Process (NSPP)

- The **non homogeneous Poisson process** is characterized by a **VARIABLE rate parameter  $\lambda(t)$** , the arrival rate at time  $t$ . In general, the rate parameter may change over time.



- The stationary increments, property is not satisfied

$$\exists s, t : N(t) - N(s) \neq N(t-s)$$

- The **expected number of events** (e.g. arrival) between time  $s$  and time  $t$  is

$$\lambda_{s,t} = \int_s^t \lambda(u) \cdot du$$