

# MS-C2111 Stochastic Processes



## Lecture 7

### *Random point patterns and counting processes*

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# Contents

Random point patterns

Exponential distributions are exponential races

Simulating independently scattered point patterns

# Contents

Random point patterns

Exponential distributions are exponential races

Simulating independently scattered point patterns

## Study object: Unpredictable events

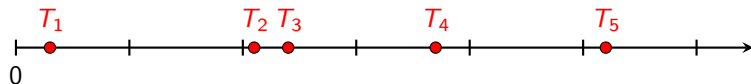
Time instants which cannot be accurately predicted:

- Occurrences of earthquakes
- Service requests in data centers
- Press releases of publicly traded companies

## Study goals

- Learn to model unpredictable time instants as random point patterns
- Derive the distribution of point counts of an independently scattered pattern
- Learn to analyze independently scattered point patterns using a Poisson process

# Random point patterns



## Random point pattern

= Countable set of points  $\{T_1, T_2, \dots\}$  on the time line  $\mathbb{R}_+$  such that the location of each point is a random variable

- Countable  $\implies$  The points can be listed using  $1, 2, \dots$

## Distribution of a random point pattern

= The joint probability distribution of *all* random variables  $T_1, T_2, \dots$

- Usually implicitly defined via a generating mechanism
- What is a natural mechanism?

# Independent scattering

Maximally random time instants: Information about  $[0, t]$  is irrelevant for predicting what happens in  $(t, \infty)$

A random point pattern  $X$  with counting measure  $N(A) = |X \cap A|$  is

- **Independently scattered** if

$$A_1, \dots, A_n \text{ disjoint} \implies N(A_1), \dots, N(A_n) \text{ independent}$$

- **Homogeneous** if

$$N(A + t) =_{\text{st}} N(A) \quad \text{for all } t \geq 0,$$

where  $A + t = \{a + t : a \in A\}$  is the  $t$ -shifted version of  $A$

# Thought experiment

How to construct an independently scattered point pattern using  $U_1, U_2, \dots =_{\text{st}} \text{Unif}(0, 1)$  as building blocks?

## Observations

In an independently scattered point pattern:

- The number of points on every nonempty interval is unlimited
- The distance between two points can be arbitrarily small and arbitrarily large

# Distribution of point counts

## Theorem

Let  $X$  be a homogeneous and independently scattered random point pattern on  $(0, \infty)$  with intensity

$$\lambda = \mathbb{E}N((0, 1]).$$

Then the number of points of  $X$  in the interval  $(0, t]$  follows a *Poisson distribution* with parameter  $\lambda t$ :

$$\mathbb{P}\left(N((0, t]) = k\right) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \quad k = 0, 1, \dots$$



## Distribution of point counts: Proof (1/2)

Let  $v(t) = \mathbb{P}(N(0, t] = 0)$  be the probability that  $(0, t]$  contains no points of  $X$ .

$$\begin{aligned}v(s + t) &= \mathbb{P}(N(0, s + t] = 0) \\&= \mathbb{P}(N(0, s] = 0, N(s, s + t] = 0) \\&= \mathbb{P}(N(0, s] = 0) \mathbb{P}(N(s, s + t] = 0) \\&= \mathbb{P}(N(0, s] = 0) \mathbb{P}(N(0, t] = 0) \\&= v(s)v(t)\end{aligned}$$

$t \mapsto v(t)$  is decreasing  $\implies$

$$v(t) = e^{-\alpha t} \quad \text{for some } \alpha > 0.$$

## Distribution of point counts: Proof (2/2)

Divide  $(0, t]$  into small subintervals  $I_{n,j} = (\frac{j-1}{n}t, \frac{j}{n}t]$  and denote

$$Z_n = \sum_{j=1}^n \theta_j, \quad \theta_j = 1(N(I_{n,j}) \geq 1).$$

Independent scattering & homogeneity  $\implies Z_n \stackrel{\text{st}}{=} \text{Bin}(n, q_n)$   
where

$$q_n = 1 - v(t/n) = 1 - e^{-\alpha t/n}.$$

$$\text{l'H\^opital} \implies nq_n = \frac{1 - e^{-\alpha t/n}}{1/n} \rightarrow \alpha t, \quad n \rightarrow \infty.$$

Law of small numbers (+ a density argument):

$$\mathbb{P}(N(0, t] = k) \approx \mathbb{P}(Z_n = k) \rightarrow e^{-\alpha t} \frac{(\alpha t)^k}{k!}.$$

Moreover,  $\alpha = \mathbb{E}N(0, 1] = \lambda$ .

# Law of small numbers

## Theorem

If  $Z_n =_{\text{st}} \text{Bin}(n, q_n)$  and  $nq_n \rightarrow \alpha \in (0, \infty)$  as  $n \rightarrow \infty$ , then

$$\mathbb{P}(Z_n = k) \rightarrow e^{-\alpha} \frac{\alpha^k}{k!}, \quad n \rightarrow \infty.$$

In other words:  $\text{Bin}(n, q_n) \rightarrow \text{Poi}(\alpha)$  as  $n \rightarrow \infty$ .

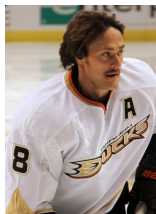
## Proof.

The probability that  $Z_n = k$  can be written as

$$\binom{n}{k} q_n^k (1 - q_n)^{n-k} = \underbrace{\frac{n!}{n^k (n-k)!}}_{\rightarrow 1} \underbrace{\frac{1}{(1 - q_n)^k}}_{\rightarrow 1} \underbrace{\frac{(nq_n)^k}{k!}}_{\rightarrow \frac{\alpha^k}{k!}} \underbrace{\left(1 - \frac{nq_n}{n}\right)^n}_{\rightarrow e^{-\alpha}}.$$



## Example: Teemu Selänne



- Scored 1430 points in 1387 games during NHL regular seasons 1993–2013
- $\lambda = 1.03$  points per game on average
- $\# \text{Points}/\text{game} =_{\text{st}} \text{Poi}(\lambda)$  if goal occurrence instants independently scattered

Points	Pr (predicted)	#Games (predicted)	#Games (realized)
0	0.35665	495	505
1	0.36771	510	506
2	0.18955	263	251
3	0.06514	90	87
4	0.01679	23	29
5	0.00346	5	9
6	0.00059	1	0
7	0.00009	0	0
> 7	0.00001	0	0

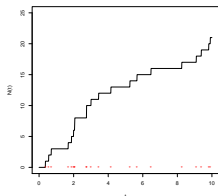
# Poisson process

$N : \mathbb{R}_+ \rightarrow \mathbb{Z}_+$  is a **Poisson process** with intensity  $\lambda$  if

- $N(t) - N(s) =_{\text{st}} \text{Poi}(\lambda(t - s))$  for all  $(s, t] \subset \mathbb{R}_+$
- $N$  has independent increments:

$$N(t_1) - N(s_1), \dots, N(t_n) - N(s_n)$$

are independent for disjoint  
 $(s_1, t_1], \dots, (s_n, t_n] \subset \mathbb{R}_+$



The previous theorem rephrased:

## Theorem

*The counting process  $N(t) = N(0, t]$  of a homogeneous independently scattered point pattern is a Poisson process with intensity  $\lambda = \mathbb{E}N(0, 1]$ .*

# Contents

Random point patterns

Exponential distributions are exponential races

Simulating independently scattered point patterns

# Exponential distribution

Random number  $X \geq 0$  is  $\text{Exp}(\lambda)$ -distributed with rate parameter  $\lambda > 0$  if

$$\mathbb{P}(X \leq t) = 1 - e^{-\lambda t}, \quad t \geq 0.$$

- $X$  has a density  $f(t) = \lambda e^{-\lambda t}$ ,  $t \geq 0$
- $\mathbb{E}X = \frac{1}{\lambda}$

# Memoryless property

## Theorem

$X =_{\text{st}} \text{Exp}(\lambda)$  satisfies

$$\mathbb{P}(X > s + t \mid X > s) = \mathbb{P}(X > t) \quad \text{for all } s, t \geq 0.$$

**Proof:**

$$\begin{aligned}\mathbb{P}(X > s + t \mid X > s) &= \frac{\mathbb{P}(X > s + t, X > s)}{\mathbb{P}(X > s)} \\ &= \frac{\mathbb{P}(X > s + t)}{\mathbb{P}(X > s)} \\ &= \frac{e^{-\lambda(s+t)}}{e^{-\lambda s}} = e^{-\lambda t} = \mathbb{P}(X > t).\end{aligned}$$

Stochastic public traffic: Having waited for a bus for  $s$  minutes, the remaining waiting time is  $\text{Exp}(\lambda)$ -distributed.



## First jump instant of a Poisson process

Let  $T_1 = \min\{t \geq 0 : N(t) = 1\}$  be the first jump instant of a Poisson process  $N$  with intensity  $\lambda$ . What is the distribution of  $T_1$ ?

$$\mathbb{P}(T_1 > t) = \mathbb{P}(N(t) = 0) = e^{-\lambda t} \frac{\lambda^0}{0!} = e^{-\lambda t}$$

Because the cumulative distribution function

$\mathbb{P}(T_1 \leq t) = 1 - e^{-\lambda t}$ ,  $t \geq 0$  determines the distribution, we conclude that  $T_1$  is  $\text{Exp}(\lambda)$ -distributed.

More generally, the distance

$$\Delta = \min\{t \geq s : N(t) = N(s) + 1\}$$

from any point  $s$  to the next jump instant of the Poisson process is  $\text{Exp}(\lambda)$ -distributed.

## Winning time in an exponential race

Winning time of a race is  $U = \min\{X_1, \dots, X_n\}$ , where the competitors' times  $X_i \stackrel{\text{st}}{=} \text{Exp}(\lambda_i)$  are independent.

What is the distribution of  $U$ ?

$$\begin{aligned}\mathbb{P}(U > t) &= \mathbb{P}(X_1 > t, \dots, X_n > t) \\ &= \mathbb{P}(X_1 > t) \cdots \mathbb{P}(X_n > t) \\ &= e^{-\lambda_1 t} \cdots e^{-\lambda_n t} \\ &= e^{-(\sum_{i=1}^n \lambda_i) t}\end{aligned}$$

$$\implies U \stackrel{\text{st}}{=} \text{Exp}(\sum_{i=1}^n \lambda_i)$$

The minimum of independent exponentially distributed random numbers is exponentially distributed.

## Winning probability in an exponential race

What is the probability that 1 wins?

Competitor 1 wins if  $X_1 < \min\{X_2, \dots, X_n\} =: \tilde{U}$ .

Because  $X_1$  and  $\tilde{U}$  are independent,

$$\mathbb{P}(X_1 < \tilde{U}) = \int_0^\infty \mathbb{P}(t < \tilde{U}) \lambda_1 e^{-\lambda_1 t} dt.$$

Because  $\tilde{U} =_{\text{st}} \text{Exp}(\sum_{i=2}^n \lambda_i)$ , competitor 1 wins with probability

$$\begin{aligned} \mathbb{P}(X_1 < \tilde{U}) &= \int_0^\infty e^{-(\sum_{i=2}^n \lambda_i)t} \lambda_1 e^{-\lambda_1 t} dt \\ &= \lambda_1 \int_0^\infty e^{-(\sum_{i=1}^n \lambda_i)t} dt = \frac{\lambda_1}{\sum_{i=1}^n \lambda_i} \end{aligned}$$

# Exponential race — Summary

## Theorem

Let  $U = \min\{X_1, \dots, X_n\}$  where  $X_i =_{\text{st}} \text{Exp}(\lambda_i)$  are independent.  
Then:

- $U =_{\text{st}} \text{Exp}(\sum_{i=1}^n \lambda_i)$
- $\mathbb{P}(X_i = U) = \frac{\lambda_i}{\lambda_1 + \dots + \lambda_n}$
- $U$  and  $I = \arg \min_i \{X_1, \dots, X_n\}$  are independent.

# Contents

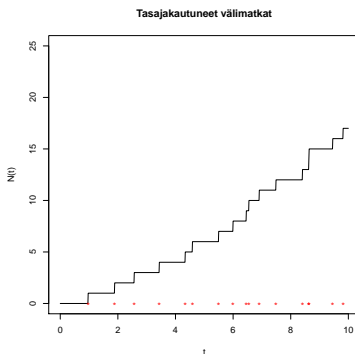
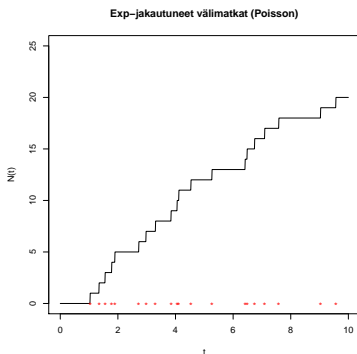
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# Simulating independently scattered point patterns

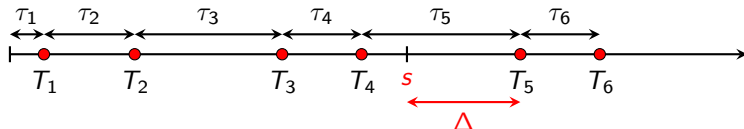
$$T_n = \tau_1 + \cdots \tau_n$$



## Theorem

*If the interpoint distances  $\tau_1, \tau_2, \dots$  are independent and  $\text{Exp}(\lambda)$ -distributed, then the point pattern  $X = \{T_1, T_2, \dots\}$  is homogeneous and independently scattered, and the corresponding counting process  $N(t)$  is a Poisson process with intensity  $\lambda$ .*

## Sketch of proof



$$\Delta = T_{N(s)+1} - s$$

Condition on  $A_s = \{N((0, s]) = 4, T_1 = t_1, \dots, T_4 = t_4\}$

$$\begin{aligned}\mathbb{P}(\Delta > t \mid A_s) &= \mathbb{P}(T_5 - s > t \mid A_s) \\ &= \mathbb{P}(t_4 + \tau_5 - s > t \mid A_s) \\ &= \mathbb{P}(\tau_5 > s - t_4 + t \mid \tau_5 > s - t_4, A_s) \\ &= \mathbb{P}(\tau_5 > s - t_4 + t \mid \tau_5 > s - t_4) \\ &= \mathbb{P}(\tau_5 > t) \\ &= e^{-\lambda t}\end{aligned}$$

$\Delta =_{\text{st}} \text{Exp}(\lambda)$  and independent of  $N(s)$

$\implies$  The points in  $[s, \infty)$  are independent of the points in  $[0, s]$   
(We conditioned on a zero-probability event.)





What is the distribution of the  $n$ -th time instant  $T_n$ ?

## Gamma distribution

A random number  $X \geq 0$  is  $\text{Gam}(n, \lambda)$ -distributed with shape parameter  $n$  and rate parameter  $\lambda$  if it has a density function

$$f(x) = \begin{cases} \lambda e^{-\lambda x} \frac{(\lambda x)^{n-1}}{(n-1)!}, & x \geq 0, \\ 0, & x < 0. \end{cases}$$

### Lemma

*If  $\tau_1, \dots, \tau_n$  are independent and  $\text{Exp}(\lambda)$ -distributed, then  $T_n = \sum_{i=1}^n \tau_i$  is  $\text{Gam}(n, \lambda)$ -distributed.*

### Proof.

Classroom problem 8.1.



## Distribution of $N(t)$

Let  $T_n = \tau_1 + \cdots + \tau_n$  and let  $N(t)$  be the counting process of  $\{T_1, T_2, \dots\}$ .

$$\begin{aligned}\mathbb{P}(N(t) = n) &= \mathbb{P}(N(t) \geq n) - \mathbb{P}(N(t) \geq n+1) \\&= \mathbb{P}(T_n \leq t) - \mathbb{P}(T_{n+1} \leq t) \\&= \int_0^t \lambda e^{-\lambda s} \frac{(\lambda s)^{n-1}}{(n-1)!} ds - \int_0^t \lambda e^{-\lambda s} \frac{(\lambda s)^n}{(n)!} ds \\&= e^{-\lambda t} \frac{(\lambda t)^n}{n!} + \int_0^t \lambda e^{-\lambda s} \frac{(\lambda s)^n}{(n)!} ds - \int_0^t \lambda e^{-\lambda s} \frac{(\lambda s)^n}{(n)!} ds \\&= e^{-\lambda t} \frac{(\lambda t)^n}{n!}.\end{aligned}$$

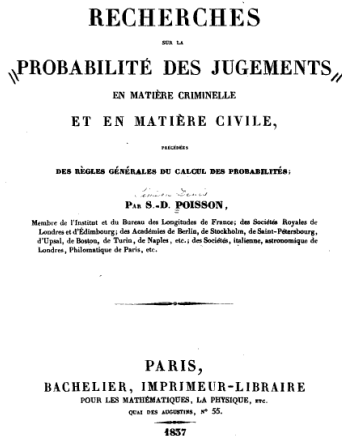
Hence  $N(t) =_{\text{st}} \text{Poisson}(\lambda t)$ .

# Summary

- Random point pattern on  $\mathbb{R}$  = model for unpredictable time instants
- Independently scattered point patterns have Poisson distributed point counts and exponentially distributed interpoint distances.
- Independently scattered point patterns can be constructed using Exp-distributed interpoint distances



Siméon Denis Poisson (1781–1840)



# References

## Photos

1. Image courtesy of think4photop at FreeDigitalPhotos.net
2. Image courtesy of Hockeybroad/Cheryl Adams at Wikipedia.