

CHAPTER 3
STANDARD POISSON PROCESS.

Motivations:

- theoretical importance :
 - "simple" continuous-time stochastic process.
 - "universal" stationary process: $\mathbb{R}_+ \rightarrow \mathbb{N}$ indep. increments and jump of size 1.
- applications
 - queuing processes.
 - compound poisson process.

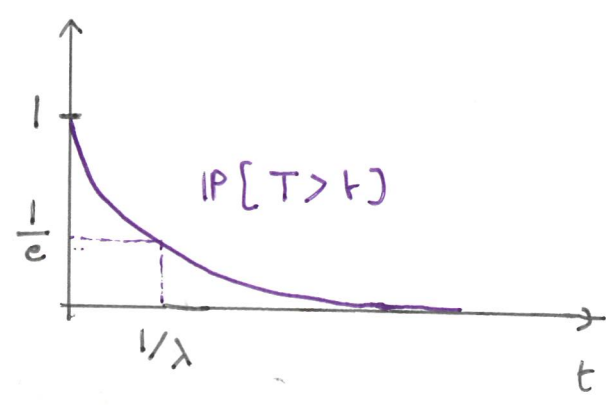
Framework: $(\Omega, \mathcal{F}, \mathbb{P})$ probability space
 • time space $\mathbb{R}_+ = [0, \infty)$.

1 EXPONENTIAL RANDOM VARIABLES

Def: Let $\lambda > 0$. A real n.v. T is exponential with parameter λ (we write $T \sim \text{Exp}(\lambda)$) if it has density

$$f(t) = \lambda e^{-\lambda t} \quad \forall t \geq 0.$$

Rk: $T \sim \text{Exp}(\lambda) \Leftrightarrow \forall t \geq 0 \quad P[T \geq t] = e^{-\lambda t}.$

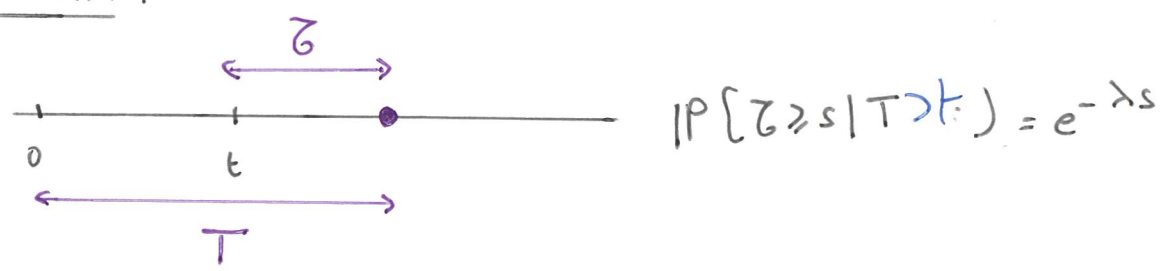


Prop. (memory less property)

Let $T \sim \text{Exp}(\lambda)$.

$$\forall s, t \geq 0 \quad \mathbb{P}[T \geq s+t \mid T \geq t] = \mathbb{P}[T \geq s]$$

Interpretation.



"condition on $T \geq t$, $Z := T - t$ is $\text{Exp}(\lambda)$."

Prop. (min of indep. exponentials.)

Let $n \geq 0$, T_1, \dots, T_n indep. $T_i \sim \text{Exp}(\lambda_i)$ $\lambda_i > 0$.

- $\min(T_1, \dots, T_n) \sim \text{Exp}(\lambda_1 + \dots + \lambda_n)$

- $\mathbb{P}[T_1 = \min(T_1, \dots, T_n)] = \frac{\lambda_1}{\lambda_1 + \dots + \lambda_n}$

Reminder: If X is a real n.v. with density f and Y is a n.v. with values in some measurable space (E, \mathcal{F}) independent of X . Then for all $\varphi: \mathbb{R} \times E \rightarrow \mathbb{R}$ meas. bounded, we have.

$$E[\varphi(X, Y)] = \int_{\mathbb{R}} E[\varphi(x, Y)] f(x) dx$$

heuristic $E[\varphi(X, Y)] = \int \underbrace{E[\varphi(X, Y) | X=x]}_{\substack{= E[\varphi(x, Y)] \\ \text{indep.}}} \underbrace{P[X=x]}_{f(x)} dx$

Proof: For every $t \geq 0$

$$P[\min(T_1, \dots, T_n) \geq t] \stackrel{\text{indep.}}{=} \prod_{i=1}^n P[T_i \geq t] = \exp(-(\lambda_1 + \dots + \lambda_n)t)$$

• $P[T_1 = \min(T_1, \dots, T_n)]$

reminder $\rightarrow \int_0^{\infty} P[t = \min(t, T_2, \dots, T_n)] \lambda_1 e^{-\lambda_1 t} dt$

$$= \int_0^{\infty} P[\min(T_2, \dots, T_n) \geq t] \lambda_1 e^{-\lambda_1 t} dt$$

$$= \lambda_1 \int_0^{\infty} e^{-(\lambda_1 + \dots + \lambda_n)t} dt = \frac{\lambda_1}{\lambda_1 + \dots + \lambda_n}$$

Prop. (Sum of exponentials.)

Let $\lambda > 0$, $n \geq 1$. Let T_1, \dots, T_n iid $\text{Exp}(\lambda)$.

Then $S_n := T_1 + \dots + T_n$ is $\Gamma(n, \lambda)$ -distributed.

ie. S_n is a continuous n.v. with density

$$f_{S_n}(t) = \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!}$$

Rk: $\Gamma(1, \lambda) = \text{Exp}(\lambda)$.

Proof: By induction.

• $n=1$ ok.

• Assume that the result holds for some $n \geq 1$.

Let T_1, \dots, T_{n+1} iid $\text{Exp}(\lambda)$. $S_n = T_1 + \dots + T_n$

and T_{n+1} are indep., hence $S_{n+1} = S_n + T_{n+1}$

admits a density given by the convolution

$$f_{S_{n+1}}(t) = \int_0^t f_{S_n}(s) f_{T_{n+1}}(t-s) ds$$

$$\stackrel{\text{(IH)}}{=} \int_0^t \lambda^2 e^{-\lambda s} \cdot \frac{(\lambda s)^{n-1}}{(n-1)!} e^{-\lambda(t-s)} ds$$

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

2. DEFINITION OF POISSON PROCESSES

Setup: $\lambda > 0$.

$(T_i)_{i \geq 1}$ iid $\text{Exp}(\lambda)$

$S_n := T_1 + \dots + T_n$

Def. The stochastic process $(N_t)_{t \geq 0}$ defined by

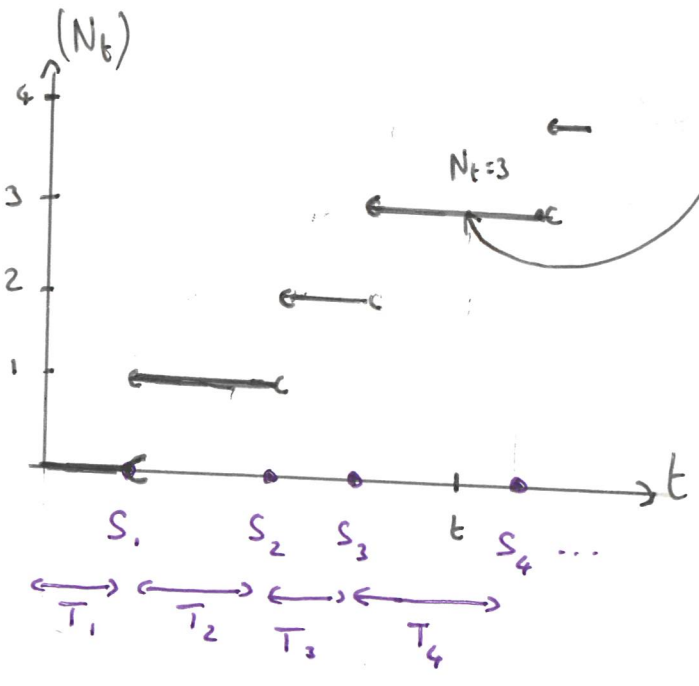
$$\forall t \geq 0 \quad N_t = \sum_{i=1}^{\infty} \mathbb{1}_{S_i \leq t}$$

is called Poisson process with intensity λ (PP(λ))

The n.v. T_1, T_2, \dots are the inter-arrival times

and S_1, S_2, \dots are the jump times of the process.

Interpretation



$N_t =$ "number of clock rings before time t "

Elementary properties

- the mapping $t \rightarrow N_t$ is a.s. right-continuous nondecreasing, with values in \mathbb{N} .

- For fixed $t \geq 0$ $N_t \sim \text{Poisson}(\lambda t)$

ie.
$$P[N_t = n] = \frac{(\lambda t)^n}{n!} e^{-\lambda t} \quad (\forall n \in \mathbb{N})$$

Proof: The first item follows from the definition. For the second item, first notice $P[N_t = 0] = P[T_1 > t] = e^{-\lambda t}$ and then

$$\forall n \geq 1 \quad P[N_t = n] = P[S_n \leq t, S_n + T_{n+1} > t]$$

S_n, T_{n+1} indep. \rightarrow

$$= \int_0^t P[s \leq t, s + T_{n+1} > t] \lambda e^{-\lambda s} \frac{(\lambda s)^{n-1}}{(n-1)!} ds$$

$$= \int_0^t \lambda e^{-\lambda(t-s)} e^{-\lambda s} \frac{(\lambda s)^{n-1}}{(n-1)!} ds$$

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

Rk: other def. of Poisson processes exist in the literature. The definition above ensures that Poisson processes exist. (because a sequence of iid exists)

The memoryless property of the exponentials implies a Markov property for the Poisson process.

Notation: Fix $t > 0$. Consider the process $N^{(t)} = (N_s^{(t)})_{s \geq 0}$ defined by $\forall s \geq 0$ $N_s^{(t)} = N_{t+s} - N_t$

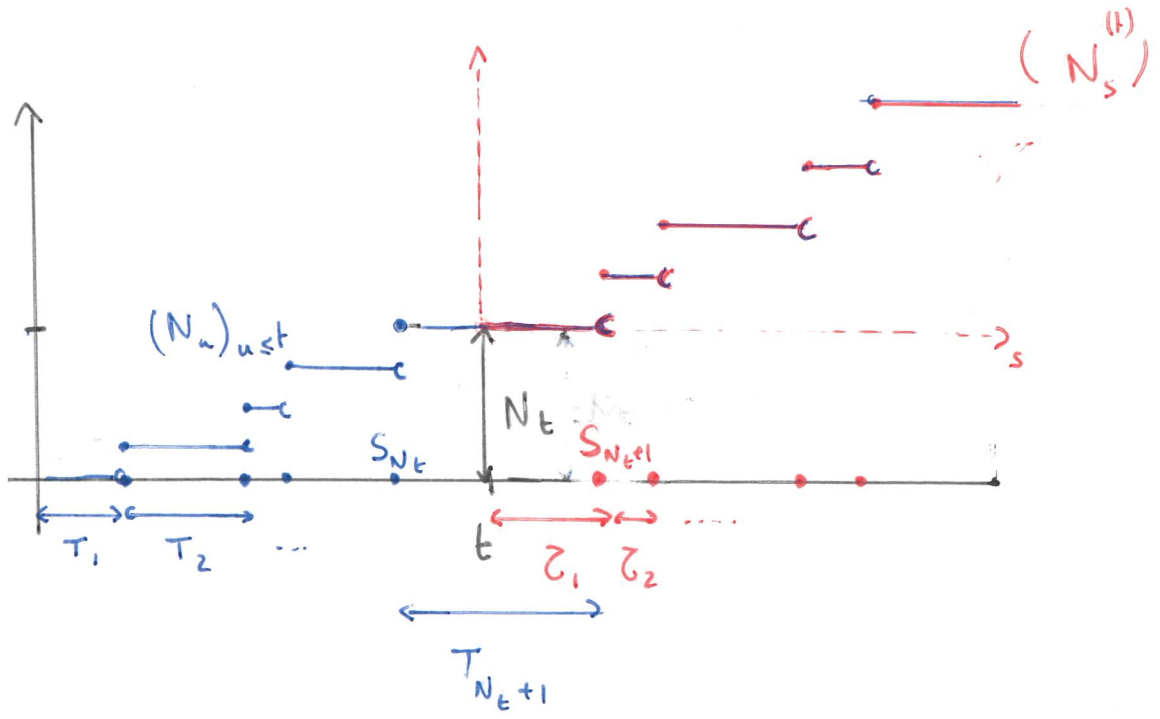
Thm [Markov property for the Poisson processes]
Fix $t > 0$. The process $N^{(t)}$ is a PP(λ) independent of $(N_u)_{0 \leq u \leq t}$.

ie: There exist some $(\tau_i)_{i \geq 1}$ iid $\text{Exp}(\lambda)$ p.t.

$$N_s^{(t)} = \sum_{i=1}^{\infty} \mathbb{1}_{\tau_1 + \dots + \tau_i \leq t}$$

and for every $u_1, \dots, u_k \leq t$ $s_1, \dots, s_l \geq 0$

$(N_{u_1}, \dots, N_{u_k})$ and $(N_{s_1}^{(t)}, \dots, N_{s_l}^{(t)})$ indep.



Proof: Define

$$\tau_1 = S_{N_t+1} - t \quad \text{and for } i \geq 2 \quad \tau_i = T_{N_t+i}$$

Step 1: $\tau_1 \sim \text{Exp}(\lambda)$ independent of N_t

$$\forall s \geq 0, n \geq 1 \quad \mathbb{P}[N_t = n, \tau_1 > s] = \mathbb{P}[S_n \leq t, S_{n+1} > t, S_{n+1} > t+s]$$

$$= \mathbb{P}[S_n \leq t, S_n + T_{n+1} > t+s]$$

$$= \int_0^t \underbrace{\mathbb{P}[u + T_{n+1} > t+s]}_{\substack{\text{memoryless property} \\ \text{of } T_{n+1}}} f_{S_n}(u) du$$

$$= \mathbb{P}[T_{n+1} > (t-u)+s]$$

$$\stackrel{\text{memoryless property of } T_{n+1}}{=} e^{-\lambda s} \times \mathbb{P}[T_{n+1} > t-u]$$

$$= e^{-\lambda s} \int_0^t \mathbb{P}[u + T_{n+1} > t] f_{S_n}(u) du$$

$$= e^{-\lambda s} \mathbb{P}[S_n \leq t, S_{n+1} > t]$$

$$= e^{-\lambda s} \mathbb{P}[N_t = n]$$

(the case $n=0$ can be deduced from above, or proved directly by the memoryless property of T_1 .)

Step 2: $(Z_i)_{i \geq 1}$ are iid $\text{Exp}(\lambda)$, indep. of N_t

$$\forall i \geq 1 \quad \forall s_1, \dots, s_i \geq 0 \quad \forall n \in \mathbb{N}$$

$$\begin{aligned} & \mathbb{P}\{N_t = n, Z_1 > s_1, \dots, Z_i > s_i\} \\ &= \mathbb{P}\left[\underbrace{N_t = n, Z_1 > s_1}_{\in \sigma(T_1, \dots, T_{n+1})}, \underbrace{T_{n+2} > s_2, \dots, T_{n+i} > s_i}_{\in \sigma(T_{n+2}, \dots, T_{n+i})} \right] \end{aligned}$$

$$\stackrel{\text{indep.}}{=} \mathbb{P}\{N_t = n, Z_1 > s_1\} \cdot e^{-\lambda(s_2 + \dots + s_i)}$$

$$\stackrel{\text{Step 1}}{=} \mathbb{P}\{N_t = n\} e^{-\lambda(s_1 + \dots + s_i)}$$

Since $N_s^{(t)} = \sum_{i=1}^{\infty} \mathbb{1}_{Z_1 + \dots + Z_i \leq s}$ (for every $s \geq 0$), this

concludes that $N^{(t)}$ is a $\text{pp}(\lambda)$ indep. of N_t .

Step 3: $N^{(t)}$ is independent of $(N_u)_{u \leq t}$.

In order to achieve Step 3, one can first use an induction (on k) and Step 2 to show that for every $k \geq 1$,

$$\forall p \geq k \quad \forall 0 = t_0 < t_1 < \dots < t_p \quad \forall n_1, \dots, n_p \in \mathbb{N}$$

$$\begin{aligned} \mathbb{P}\{N_{t_1} = n_1, \dots, N_{t_p} = n_p\} &= \mathbb{P}\{N_{t_1} = n_1, \dots, N_{t_k} = n_k\} \cdot \\ & \quad \mathbb{P}\{N_{t_{k+1}-t_k} = n_{k+1} - n_k, \dots, N_{t_p-t_k} = n_p - n_k\} \end{aligned}$$

(exercise)

Now let $u_1 < \dots < u_k < t$ $0 \leq s_1 < \dots < s_p$
 $n_1, \dots, n_k \in \mathbb{N}$ $m_1, \dots, m_p \in \mathbb{N}$

$\forall n \in \mathbb{N}$ the formula above implies

$$P[N_{u_1} = n_1, \dots, N_{u_p} = n_p, N_t = n, N_{s_1}^{(t)} = m_1, \dots, N_{s_p}^{(t)} = m_p]$$

$$= P[N_{u_1} = n_1, \dots, N_{u_p} = n_p, N_t = n] \underbrace{P[N_{s_1} = m_1, \dots, N_{s_p} = m_p]}_{= P[N_{s_1}^{(t)} = m_1, \dots, N_{s_p}^{(t)} = m_p]}$$

By summing over all $n \in \mathbb{N}$, we obtain the desired independence property.

4 INDEPENDENT AND STATIONARY INCREMENTS.

Def. A stochastic process $(X_t)_{t \geq 0}$ is said to have independent and stationary increments if

$\forall k \geq 1 \forall 0 = t_0 < \dots < t_k$ $X_{t_1} - X_{t_0}, \dots, X_{t_k} - X_{t_{k-1}}$ are independent

and

$\forall s < t \forall h \geq 0$ $X_t - X_s \stackrel{\text{Law}}{=} X_{t+h} - X_{s+h}$

TRM (stationary and independent increments.)

(i) $\forall k \quad \forall 0 = t_0 < t_1 < \dots < t_k$

$N_{t_1} - N_{t_0}, \dots, N_{t_k} - N_{t_{k-1}}$ are independent

(ii) $\forall t, s \geq 0$

$N_{t+s} - N_t \sim \text{Pois.}(\lambda s)$

In particular $(N_t)_{t \geq 0}$ has stationary and independent increments

Interpretation:



$N_t - N_s =$ number of arrivals in the interval $[s, t]$

(i) \Leftrightarrow the arrivals in disjoint intervals are independent.

(ii) \Leftrightarrow the law of the number of arrivals depends only on the length of the interval

Rk: (i) and (ii) is equivalent to $\forall k \geq 1$.

$\forall 0 = t_0 < \dots < t_k \quad \forall n_1, \dots, n_k \in \mathbb{N}$

(*) $\mathbb{P} [N_{t_1} - N_{t_0} = n_1, \dots, N_{t_k} - N_{t_{k-1}} = n_k] = \prod_{i=1}^k \left(\frac{(\lambda(t_i - t_{i-1}))^{n_i}}{n_i!} e^{-\lambda(t_i - t_{i-1})} \right)$

Proof: We prove (*) by induction on k .

• The case $k = 1$ corresponds to $N_t \sim \text{Poiss.}(\lambda t)$.

• Now let $k \geq 1$ and assume that (*) holds.

Let $t_0 < \dots < t_{k+1}$ $n_1, \dots, n_{k+1} \in \mathbb{N}$

$$\mathbb{P} \left[\underbrace{N_{t_1} - N_{t_0} = n_1, \dots, N_{t_k} - N_{t_{k-1}} = n_k}_{\in \sigma((N_u)_{u \leq t_k})}, \underbrace{N_{t_{k+1}} - N_{t_k} = n_{k+1}}_{= N_{t_{k+1}-t_k}^{(t_k)}} \right]$$

$$\stackrel{MP}{=} \underbrace{\mathbb{P} [N_{t_1} - N_{t_0} = n_1, \dots, N_{t_k} - N_{t_{k-1}} = n_k]}_{\substack{\text{induction} \\ = \prod_{i \leq k} \frac{[\lambda(t_i - t_{i-1})]^{n_i}}{n_i!} e^{-\lambda(t_i - t_{i-1})}}} \cdot \underbrace{\mathbb{P} [N_{t_{k+1}} - N_{t_k} = n_{k+1}]}_{= \frac{(\lambda(t_{k+1} - t_k))^{n_{k+1}}}{n_{k+1}!} e^{-\lambda(t_{k+1} - t_k)}}$$

$$= \prod_{i \leq k+1} \frac{[\lambda(t_i - t_{i-1})]^{n_i}}{n_i!} e^{-\lambda(t_i - t_{i-1})}$$

■

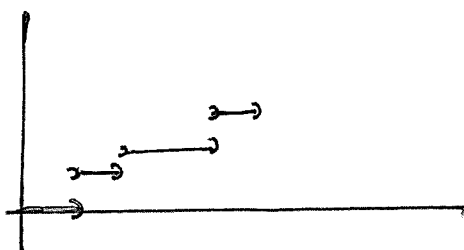
5 FINITE MARGINALS CHARACTERIZATION

Motivation: If N is a p.p. (λ), we know the law of the vector $(N_{t_1}, \dots, N_{t_k})$ for every fixed $t_1 < \dots < t_k$. (called the finite marginal laws: they characterize the law of the stochastic process $(N_t)_{t \geq 0}$)

Conversely, if a stochastic process $(N_t)_{t \geq 0}$ has the same finite marginals as a p.p. (λ), is it a p.p. (λ)?

\Leftrightarrow No: For T_1, T_2, \dots iid $\text{Exp}(\lambda)$, consider the

$$\text{process } \tilde{N}_t = \sum_{i \geq 1} \mathbb{1}_{S_i < t} \quad (S_i = T_1 + \dots + T_i)$$



$(\tilde{N}_t)_{t \geq 0}$ is not a p.p. because it has not right-continuous trajectories a.s. But it has the same finite marginals as the p.p. defined by $N_t = \sum_{i \geq 1} \mathbb{1}_{S_i \leq t}$.

• One can even construct a "worse" counter-example, (inspired from W. Werner lecture notes on Brownian motion). Let N be a p.p. (λ). Let $(X_i)_{i \in \mathbb{N}}$ iid $\text{Exp}(1)$, indep. of N . Notice that $\mathcal{X} = \{X_1, X_2, \dots\}$ is dense in \mathbb{R}_+ .

Consider the process defined by $\forall t \tilde{N}_t = N_t + \mathbb{1}_{t \in \mathbb{Z}}$

The process \tilde{N}_t is not a $pp(\lambda)$ (it is nowhere right continuous a.s.) but it has the same finite marginals.

↳ Yes: In this section, we will see that if we add a regularity condition on $t \rightarrow N_t$. Then it is characterized by its finite marginal laws.

Def: Let $N := (N_t)_{t \geq 0}$ be a continuous-time stochastic process. We say that N is a counting process if the following holds a.s.:

$N_0 = 0$ and $t \mapsto N_t$ is non decreasing, right-continuous, with values in \mathbb{N} .

In this case we can define the successive jump times by setting $S_1 = \min \{t : N_t > 0\}$ and by induction for $i \geq 1$

$$S_{i+1} = \min \{t \geq S_i : N_t > N_{S_i}\}.$$

Rk: N is a $pp(\lambda)$ if and only if N is a counting process with jumps of size 1 [i.e. $(\forall t \limsup_{s \rightarrow 0} N_t - N_{t-s} \leq 1)$ a.s.] and $S_1, S_2 - S_1, S_3 - S_2$ are iid $Exp(\lambda)$.

Thm: Let $\lambda > 0$. Let $N = (N_t)_{t \geq 0}$ be a counting process.

The following are equivalent.

(i) N is a PP(λ)

(ii) $\forall k \geq 1 \quad \forall t_0 = 0 < t_1 < \dots < t_k \quad \forall n_1, \dots, n_k \in \mathbb{N}$

$$P[N_{t_1} - N_{t_0} = n_1, \dots, N_{t_k} - N_{t_{k-1}} = n_k] = \prod_{i=1}^k e^{-\lambda(t_i - t_{i-1})} \frac{[\lambda(t_i - t_{i-1})]^{n_i}}{n_i!}$$

Proof: (i) \Rightarrow (ii) already seen.

(ii) \Rightarrow (i) We first prove that $(N_t)_{t \geq 0}$ has jumps of size 1 on every segment $[0, A]$, $A > 0$.



Let $E_n = \{ \forall i \leq n, N_{\frac{iA}{n}} - N_{\frac{(i-1)A}{n}} \leq 1 \}$ for $n \geq 1$.

We have $P[E_n] = \prod_{i=1}^n \left(e^{-\frac{\lambda A}{n}} + e^{-\frac{\lambda A}{n}} \frac{\lambda A}{n} \right) = e^{-A} \left(1 + \frac{\lambda A}{n} \right)^n \rightarrow 1$.

Let $E = \bigcup_{n \geq 1} E_n$. We have $P[E] = 1$ (because $P[E] \geq P[E_n]$ for all $n \geq 1$) and furthermore $\forall \omega \in E$

$$\forall t \leq A \quad \limsup_{s \rightarrow 0} N_t(\omega) - N_{t-s}(\omega) \leq 1.$$

This concludes that N has jumps of size 1.

Fix $k \geq 1$. We prove that $T_1 = S_1, T_2 = S_2 - S_1, \dots, T_k = S_k - S_{k-1}$ are iid $\text{Exp}(\lambda)$

We begin with the computation of the law of (S_1, \dots, S_k) .

Let $U = \{(s_1, \dots, s_k) \in \mathbb{R}^k : 0 \leq s_1 \leq \dots \leq s_k\}$. We show

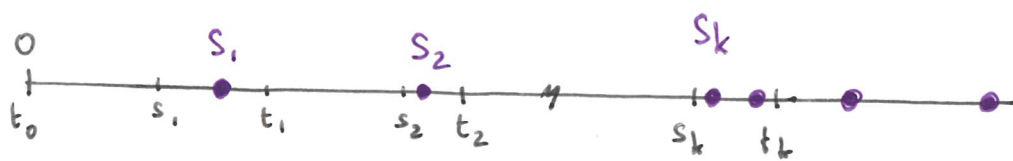
that

$$\forall H \in \mathcal{B}(U) \quad \mathbb{P}[(s_1, \dots, s_k) \in H] = \int_H \lambda^k e^{-\lambda y_k} dy_1 \dots dy_k.$$

By Dynkin's lemma, it suffices to prove it for

$$H = [s_1, t_1) \times \dots \times [s_k, t_k),$$

with $s_1 < t_1 < \dots < s_k < t_k$. (convention: $t_0 = 0$)



$$\mathbb{P}[\forall i \leq k \quad S_i \in [s_i, t_i)]$$

$$= \mathbb{P}\left[\bigcap_{i \leq k} \{N_{s_i} - N_{t_{i-1}} = 0\} \cap \bigcap_{i < k} \{N_{t_i} - N_{s_i} = 1\} \cap \{N_{t_k} - N_{s_k} \geq 1\} \right]$$

$$= \prod_{i \leq k} e^{-\lambda(s_i - t_{i-1})} \prod_{i < k} \lambda(t_i - s_i) e^{-\lambda(t_i - s_i)} \times (1 - e^{-\lambda(t_k - s_k)})$$

$$= \prod_{i < k} \lambda(t_i - s_i) \times e^{-\lambda s_k} \times (1 - e^{-\lambda(t_k - s_k)})$$

$$= \prod_{i < k} \int_{s_i}^{t_i} \lambda dy_i \times \int_{s_k}^{t_k} \lambda e^{-\lambda y_k} dy_k$$

Hence (S_1, \dots, S_k) has density $f(y_1, \dots, y_k) = \lambda^k e^{-\lambda y_k} \mathbb{1}_{y_1 < \dots < y_k}$.

Define the map $h(t_1, \dots, t_k) = (t_1, t_1+t_2, \dots, t_1+\dots+t_k)$. This way, we have $(T_1, \dots, T_k) = h^{-1}((S_1, \dots, S_k))$. By change of variables (and using that the jacobian of h is 1),

(T_1, \dots, T_k) admits the density

$$\begin{aligned}
 (f \circ h)(t_1, \dots, t_k) &= \lambda^k e^{-\lambda(t_1+\dots+t_k)} \mathbb{1}_{t_1 < t_1+t_2 < \dots < t_1+\dots+t_k} \\
 &= \prod_{i=1}^k (\lambda e^{-\lambda t_i} \mathbb{1}_{t_i > 0}),
 \end{aligned}$$

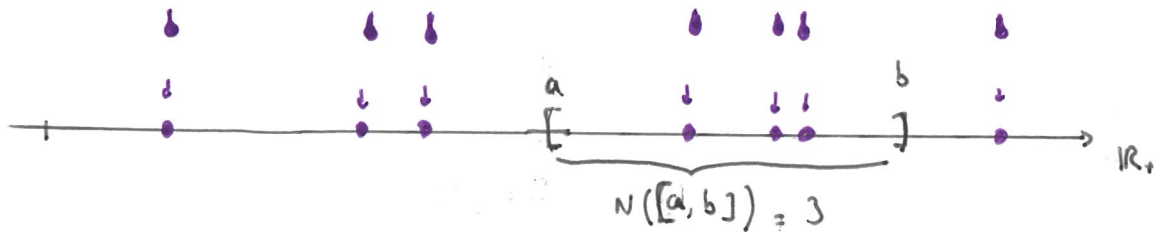
which establishes that T_1, \dots, T_k are iid $\text{Exp}(\lambda)$

Since k is arbitrary, we conclude that the interarrival times T_1, T_2, \dots are iid $\text{Exp}(\lambda)$. ■

6 MICROSCOPIC CHARACTERIZATION.

Motivation:

Droplets of water falling on the half-line \mathbb{R}_+ .



↳ random points on \mathbb{R}_+ = position at which the droplets have fallen.

Define for every interval $I \subset \mathbb{R}_+$

$$N(I) = \#\{\text{random points on } [a, b]\}$$

Hypotheses 1. If I_1, \dots, I_k are disjoint intervals

$N(I_1), \dots, N(I_k)$ are independent.

• If $I' = (a+h, b+h)$ is a translate of $I = (a, b)$, $h \geq 0$

$$N(I') \stackrel{\text{law}}{=} N(I)$$

• \forall bounded interval $I \subset \mathbb{R}_+$

$$N(I) \in \mathbb{N} \text{ a.s.}$$

The hypotheses imply that the stochastic process $N_t := N([0, t])$

is a counting process with independent and stationary increments

We know that such process exists \rightarrow Poisson process.

Q: Is it the only one?

Answer : yes ! But we need an additional condition
fixing the density λ .

Thm: Let N be a counting process, $\lambda > 0$. The following are equivalent.

- (i) N is a $pp(\lambda)$
- (ii) N has independent and stationary increments and

$IP[N_t = 1] = \lambda t + o(t)$	$[ie. \lim_{t \rightarrow 0} \frac{P[N_t = 1]}{\lambda t} = 1]$
$IP[N_t \geq 2] = o(t)$	$[ie. \lim_{t \rightarrow 0} \frac{IP[N_t \geq 2]}{t} = 0]$

Lemma (Poisson approximation)

Let $(p_n)_{n \geq 1}$ be a sequence of parameters $p_n \in [0, 1]$ and $\lambda \in (0, \infty)$ s.t.

$$\lim_{n \rightarrow \infty} n p_n = \lambda.$$

For every n , let $X_n \sim \text{Bin}(n, p_n)$. Then

$$X_n \xrightarrow[n \rightarrow \infty]{} \text{Poisson}(\lambda) \text{ in distribution.}$$

Proof: See [Probability Theory, p. 47]

Proof of Thm 1

$$(i) \Rightarrow (ii) \quad \mathbb{P}[N_t = 1] = \lambda t e^{-\lambda t} = \lambda t + o(t)$$

$$\mathbb{P}[N_t \geq 2] = 1 - e^{-\lambda t} - \lambda t e^{-\lambda t} = o(t)$$

(ii) \Rightarrow (i) We already have that (N_t) has independent increments. It suffices to prove that

$$\forall s \leq t \quad N_t - N_s \sim \text{Poisson}(\lambda(t-s)).$$

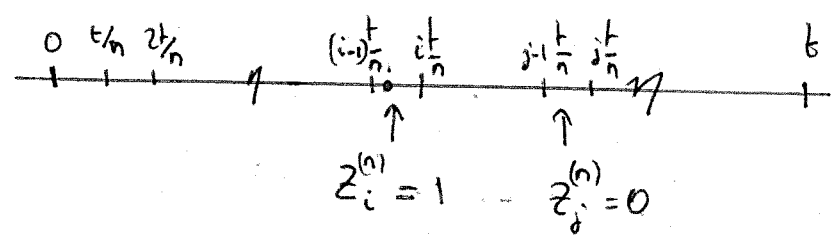
Since N_t has stationary increments, it suffices to prove

$$\forall t \quad N_t \sim \text{Poisson}(\lambda t).$$

Fix $t \in (0, \infty)$.

Let $n \geq 1$. By independence and stationarity of the increments, the variables $Z_i = \mathbb{1}_{N_{i \frac{t}{n}} - N_{(i-1) \frac{t}{n}} \geq 1}$

are i.i.d Bernoulli (p_n) random variables,
 where $p_n := P[N_{\frac{t}{n}} \geq 1] = \lambda \frac{t}{n} + o\left(\frac{t}{n}\right)$



Hence $X_n := \sum_{i=1}^n z_i^{(n)}$ is a Binomial (n, p_n) random variable. Since $np_n \xrightarrow{n \rightarrow \infty} \lambda t$, the lemma implies for any $k \in \mathbb{N}$

$$P[X_n = k] \xrightarrow{n \rightarrow \infty} \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$

We have: for every $n \geq 1$

$$\begin{aligned} P[N_t \neq X_n] &= P\left[\bigcup_{1 \leq i \leq n} \{N_{i\frac{t}{n}} - N_{(i-1)\frac{t}{n}} \geq 2\}\right] \\ &\stackrel{\text{union bound}}{\leq} \sum_{i=1}^n P\left[N_{i\frac{t}{n}} - N_{(i-1)\frac{t}{n}} \geq 2\right] \\ &\stackrel{\text{stationarity}}{=} n \times P\left[N_{\frac{t}{n}} \geq 2\right] \end{aligned}$$

Since $P[N_{\frac{t}{n}} \geq 2] = o\left(\frac{t}{n}\right)$, we get that

$$\lim_{n \rightarrow \infty} P[N_t \neq X_n] = 0.$$

Fix $k \in \mathbb{N}$. For every $n \geq 1$

$$|P[N_t = k] - P[X_n = k]| \leq E[|1_{N_t = k} - 1_{X_n = k}|] \leq P[N_t \neq X_n].$$

$$\text{Hence } P[N_t = k] = \lim_{n \rightarrow \infty} P[X_n = k] = \frac{(\lambda t)^k}{k!} e^{-\lambda t}.$$

4 PROPERTIES OF THE POISSON PROCESSES

4.1 Law of large numbers

Thm: Let $(N_t)_{t \geq 0}$ be a Poisson process on (Ω, \mathcal{F}, P) of rate λ .

Then we have

$$\lim_{t \rightarrow \infty} \frac{N_t}{t} = \lambda \quad \text{a.s.}$$

Proof: Let $X_i = N_i - N_{i-1}$, $i \geq 1$.

The X_i are pairwise independent, identically distributed random variables (actually, they are i.i.d.)

$X_i \sim \text{Poisson}(\lambda)$. Hence

$$E[|X_i|] = E[X_i] = \lambda < \infty.$$

By the strong law of large numbers (see [PROBABILITY, p. 36])

$$\frac{X_1 + \dots + X_i}{i} \xrightarrow{i \rightarrow \infty} \lambda \quad \text{a.s.}$$

Since for every $i \leq t < i+1$

$$\frac{X_1 + \dots + X_i}{i+1} \leq \frac{N_t}{t} \leq \frac{X_1 + \dots + X_{i+1}}{i}$$

We also have $\frac{N_t}{t} \xrightarrow{t \rightarrow \infty} \lambda$ a.s. ■

4.2 Thinning

Let $(N_t)_{t \geq 0}$ be a Poisson process with rate $\lambda > 0$, jump times (s_k) .

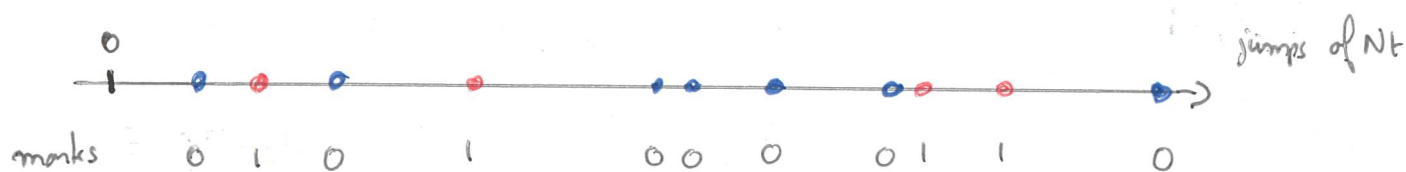
Let $(X_n)_{n \geq 1}$ i.i.d Bernoulli variable with parameter $p \in (0, 1)$

"such a sequence is called a marking of $(N_t)_{t \geq 0}$ ".

Define the thinned processes.

$$N_t^1 = \sum_{k \geq 1} \mathbb{1}_{\{s_k \leq t, X_k = 1\}}$$

$$N_t^0 = \sum_{k \geq 1} \mathbb{1}_{\{s_k \leq t, X_k = 0\}}$$



Applications: • Clients of type 0 and type 1 in a queue.

• Valid / non valid claims at insurance Company.

Rk: $N_t = N_t^0 + N_t^1$ a.s.

Theorem: (thinning of Poisson processes)

$(N_t^0)_{t \geq 0}$ and $(N_t^1)_{t \geq 0}$ are independent Poisson processes with rate $\lambda_0 = (1-p)\lambda$ and $\lambda_1 = p\lambda$ resp.

Proof: later with punctual Poisson processes on general spaces

$\xrightarrow{\text{just families of random variables}}$
 \underline{Rk} Let $(X_t)_{t \geq 0}$ $(Y_t)_{t \geq 0}$ be two stochastic processes on the same probability space. They are independent iff

$\forall t_1, \dots, t_k \quad \forall s_1, \dots, s_l \quad (X_{t_1}, \dots, X_{t_k})$ indep. of $(Y_{s_1}, \dots, Y_{s_l})$.

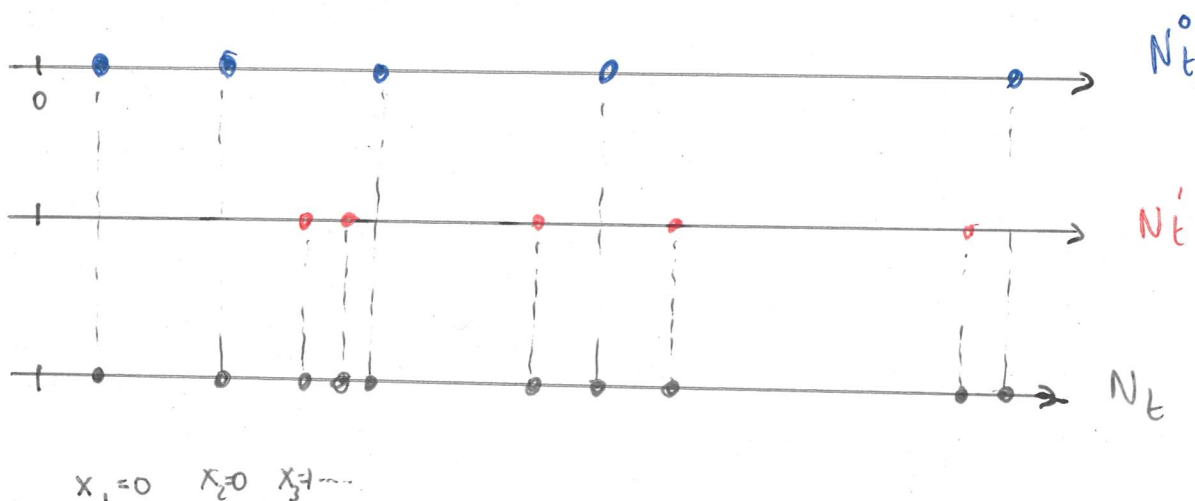
$\swarrow \quad \searrow$
 "random vectors"

4.3 Superposition.

Let $(N_t^0)_{t \geq 0}$ and $(N_t^1)_{t \geq 0}$ be two independent Poisson processes with respective rates $\lambda_0 > 0$, $\lambda_1 > 0$.

Define

$$N_t = N_t^0 + N_t^1, \quad t \geq 0. \quad \text{"superposition process"}$$



N_t is a counting process and we define

$$X_k = \begin{cases} \text{the } k\text{-th jump of } N_t \text{ is} \\ \text{a jumping time of } N_t^1 \end{cases}$$

Thm: (superposition of Poisson processes)

$(N_t)_{t \geq 0}$ is a Poisson process with rate $\lambda = \lambda_0 + \lambda_1$
 and $(X_k)_{k \geq 1}$ is a marking of $(N_t)_{t \geq 0}$ with

$$\forall k \quad P[X_k = 1] = \frac{\lambda_1}{\lambda_0 + \lambda_1}$$

Proof: $(N_t)_t$ is a counting process (it follows from the definition -)

We consider (independently of N^0, N^1)

- $(\tilde{N}_t)_{t \geq 0}$ Poisson process intensity $\lambda = \lambda_0 + \lambda_1$
- $(\tilde{X}_k)_{k \geq 1}$ iid Bernoulli $\left(\frac{\lambda_1}{\lambda_0 + \lambda_1}\right)$.

By the theorem in the previous section, the thinned processes $(\tilde{N}_t^0), (\tilde{N}_t^1)$ are independent processes with respective rates $\lambda_0, \lambda_1, \dots$

For every $t_1 < \dots < t_k$, $f: \mathbb{R}^k \rightarrow \mathbb{R}$ bounded.

$$\begin{aligned} E[f(N_{t_1}, \dots, N_{t_k})] &= E[f(N_{t_1}^0 + N_{t_1}^1, \dots, N_{t_k}^0 + N_{t_k}^1)] \\ &= E[f(\tilde{N}_{t_1}^0 + \tilde{N}_{t_1}^1, \dots, \tilde{N}_{t_k}^0 + \tilde{N}_{t_k}^1)] \\ &= E[f(\tilde{N}_{t_1}, \dots, \tilde{N}_{t_k})] \end{aligned}$$

Therefore N is a pp(λ). Similarly, for every $t_1 < \dots < t_k$
 for every $p \geq 1$ and every $f: \mathbb{R}^k \times \{0,1\}^p \rightarrow \mathbb{R}$ meas. bounded

$$E[f(N_{t_1}, \dots, N_{t_k}, X_1, \dots, X_p)] = E[f(\tilde{N}_{t_1}, \dots, \tilde{N}_{t_k}, \tilde{X}_1, \dots, \tilde{X}_p)]$$

Hence X_1, \dots, X_p are iid Bernoulli($\frac{\lambda_i}{\lambda_0 + \lambda_i}$) indep. of
 $(N_{t_1}, \dots, N_{t_k})$. ■