

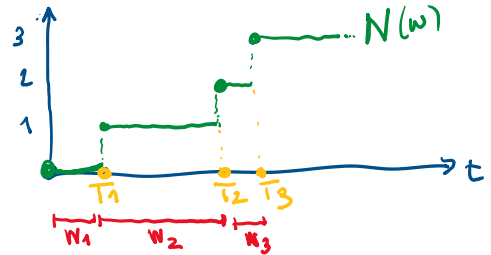
ADVANCED STOCHASTIC PROCESSES - 14: LECTURE

SUMMARY: "Equivalent representations of PP on \mathbb{R}^+ "

We have seen that a PP $(\lambda), (N_t)_{t \geq 0}$, can be thought in different ways:

1. As a trajectory $N = (N_t)_{t \geq 0}$ represented ($\forall \omega \in \Omega$) as an integer-valued step function;

2. A sequence of points $(T_m)_{m \geq 0}$ in \mathbb{R}^+ with $T_m := \inf\{t \geq 0 : N_t = m\}$



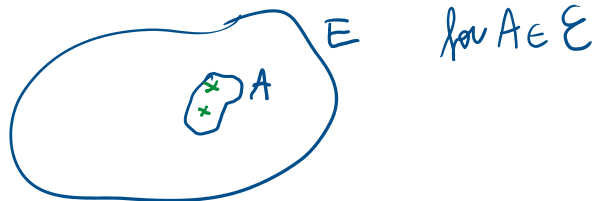
3. A sequence of time-intervals $(W_m)_{m \in \mathbb{N}}$ s.t. $W_m = T_m - T_{m-1}$

While repr. 1. and 3. are specific to time-evolution in \mathbb{R}^+ , the repr. 3. admits generalizations \rightarrow main idea is to construct random set of points on more general spaces

$(\mathbb{R}^+ \rightarrow \text{e.g. } \mathbb{R}, \mathbb{R}^d \text{ or general metric space } (E, \mathcal{E}) \text{ which is complete, separable, locally compact})$
 \rightarrow Polish space

MAIN IDEA:

$N_t = \# \text{ arrivals in } [0, t]$
 $= \# \text{ points in } [0, t]$ $\xrightarrow{\text{generalized to}}$ $N(A) = \# \text{ points in } A$



Def [Poisson Point Process]

Let Λ be a σ -finite measure on (E, \mathcal{E}) (stably, $\Lambda \in \mathcal{U}(E, \mathcal{E})$)

A Poisson Point Process with intensity Λ (PPP/ Λ) is a function

$N: \Omega \times E \rightarrow \overline{\mathbb{N}}_0$ such that $\forall k \in \mathbb{N}, \forall m_1, \dots, m_k \in \mathbb{N}_0,$
 $(\omega, A) \rightarrow N_\omega(A) \quad \forall \text{ disjoint } A_1, \dots, A_k \in \mathcal{E}:$

$$(\omega, A) \rightarrow N_\omega(A) \quad \forall \text{ disjoint } A_1, \dots, A_k \in \mathcal{E}:$$

$$P(N(A_i) = m_i, \forall i=1, \dots, k) = \prod_{i=1}^k \frac{\Lambda(A_i)^{m_i}}{m_i!} e^{-\Lambda(A_i)}$$

Interpretation: $N_\omega(A) = \#$ points in A (for the realization ω)

Remarks:

- If $\mathcal{E} = \mathbb{R}^d$ (or subsets of it) and $\Lambda = \lambda \cdot \text{leb}_{\mathbb{R}^d}$, with $\lambda > 0$, we obtain the precise generalisation of homogeneous PPP(λ) on \mathbb{R}^d .

In this case: $\Lambda(A) = \lambda \text{Vol}(A)$

- The definition implicitly uses that $\forall A \in \mathcal{E}$
 $N(A): \Omega \rightarrow \overline{\mathbb{N}}$ is a r.v. with law $\text{Poi}(\Lambda(A))$

Poisson Point Process versus Poisson Random Measure

A countable set of points $(x_i)_{i \in \mathbb{N}} \in E$ can be put in correspondence with a counting measure μ on (E, \mathcal{E}) , by setting

$$\mu(A) = \sum_{i \in \mathbb{N}} \delta_{x_i}(A) = \# \text{ of points in } A$$

In this sense, the PPP(Λ) can be considered as a random counting measure on (E, \mathcal{E}) . let us formalize this idea.

Def [random measure]: Consider (Ω, \mathcal{F}, P) prob. space and (E, \mathcal{E}) .

A random measure m on (E, \mathcal{E}) is a function $m: \Omega \times \mathcal{E} \rightarrow \overline{\mathbb{R}}_+$
 $(\omega, A) \rightarrow m_\omega(A)$

such that:

- * $m(A): \Omega \rightarrow \overline{\mathbb{R}}_+$ is a r.v. $\forall A \in \mathcal{E}$
- * $m_\omega: \mathcal{E} \rightarrow \overline{\mathbb{R}}_+$ is a measure $\forall \omega \in \Omega$

Moreover, it is a random counting measure if

Moreover, it is a random counting measure if

$$m_\omega := \sum_{i \in \mathbb{N}} \delta_{X_i(\omega)}, \text{ for } (X_i)_{i \in \mathbb{N}} \text{ set of random point}$$

Notation:

If $f: E \rightarrow \mathbb{R}_+$ and m is a random measure on (E, \mathcal{E}) , then:

$$m_\omega(f) := \int_E m_\omega(dx) f(x) \stackrel{\text{if counting measure}}{=} \sum_{i \in \mathbb{N}} f(X_i(\omega)) \quad \left(\begin{array}{l} \text{average of } f \\ \text{w.r.t. } m_\omega \end{array} \right)$$

$$\rightarrow m(f): \Omega \rightarrow \bar{\mathbb{R}}_+ \text{ is a r.v. with mean } \int_\Omega P(d\omega) m_\omega(f) \stackrel{\text{if counting meas.}}{=} \sum_{i \in \mathbb{N}} \mathbb{E}(f(X_i))$$

In particular, taking $f = \mathbb{1}_A$ for $A \in \mathcal{E}$, $m_\omega(\mathbb{1}_A) = m_\omega(A)$,

$$\text{we define } \mu(A) := \mathbb{E}(m(A)) = \int_\Omega P(d\omega) m_\omega(A), \quad \forall A \in \mathcal{E}$$

and notice that $\mu: \mathcal{E} \rightarrow \bar{\mathbb{R}}_+$ is a measure on (E, \mathcal{E})

called mean of the random measure m (called mean or intensity measure)

Example 1: Let $(X_k)_{k \in \mathbb{N}}$ i.i.d r.v. with common law ν on (E, \mathcal{E}) , and let $Y \sim \text{Poi}(\lambda)$, independent of the X_k 's.

* Define the random measure $m(A) = \sum_{k=1}^Y \delta_{X_k}(A)$, $\forall A \in \mathcal{E}$

$$\left[\begin{array}{l} \bullet \quad m(A): \Omega \rightarrow \mathbb{R}_+ \quad \text{is a r.v. } \forall A \in \mathcal{E} \\ \quad \quad \omega \rightarrow m_\omega(A) \\ \bullet \quad m_\omega = \mathcal{E} \rightarrow \mathbb{R}_+ \quad \text{is a measure, } \forall \omega \in \Omega \\ \quad \quad A \rightarrow m_\omega(A) \end{array} \right]$$

We compute its intensity measure μ :

$$\mu(A) = \mathbb{E}(m(A)) = \mathbb{E}(\mathbb{E}(m(A)|Y)) = \sum_{j=1}^{\infty} P(Y=j) \cdot j \cdot \overset{\nu(A)}{P(X_k \in A)}$$

$$= \nu(A) \cdot \mathbb{E}(Y) = \lambda \cdot \nu(A)$$

$$\rightarrow \text{hence } \mu: \mathcal{E} \rightarrow \mathbb{R}_+$$

$$\rightarrow \text{hence } \mu: \mathcal{E} \rightarrow \mathbb{R}_+ \\ A \rightarrow \lambda \cdot \nu(A)$$

The main tool to characterize the law of $m_{(\cdot)}$ is the Laplace functional, as provided by the next Proposition (without proof!)

Proposition: The law of $m_{(\cdot)}$ is uniquely determined by the
$$\mathcal{L}_m(f) := \mathbb{E}(e^{-m(f)})$$
, $\forall f: E \rightarrow \mathbb{R}^+$, $f \in \mathcal{E}$ where $f \mapsto \mathcal{L}_m(f)$ is called Laplace functional.

Continuity property: if $(f_n) \uparrow f \Rightarrow \mathcal{L}_m(f_n) \xrightarrow{n \rightarrow \infty} \mathcal{L}_m(f)$

Note: 1. the Laplace functional of a random measure m , plays the role of a generating function for r.v.
2. the same characterization holds true for the characteristic functional
$$\psi_m(f) = \mathbb{E}(e^{im(f)})$$
, for $f: E \rightarrow \mathbb{R}$, $f \in \mathcal{E}$ and bounded.

Example 1 (continuation): let $m(A) = \sum_{k=1}^Y \mathbb{1}_A(X_k)$,

for $(X_k)_{k \in \mathbb{N}}$ i.i.d. $\sim \nu$ and indep. of $Y \sim \text{Poi}(\lambda)$, and $f \geq 0$, \mathcal{E} -meas.

$$\begin{aligned} \mathcal{L}_m(f) &= \mathbb{E}\left(e^{-m(f)}\right) = \mathbb{E}\left(\prod_{k=1}^Y e^{-f(X_k)}\right) = \mathbb{E}\left(\mathbb{E}(\cdot \mid Y)\right) \\ &= \sum_{j=0}^{\infty} e^{-\lambda} \frac{\lambda^j}{j!} \mathbb{E}\left(e^{-f(X_k)}\right)^j = e^{-\lambda} \sum_{j=0}^{\infty} \frac{\lambda^j}{j!} \nu\left(e^{-f}\right)^j \\ &= e^{-\lambda(1-\nu(e^{-f}))} = e^{-\lambda\nu(1-e^{-f})} \end{aligned}$$

$\int_E e^{-f(x)} \nu(dx) = \nu(e^{-f})$

• Random measure m is characterized by its Laplace functional

$$\mathcal{L}_m(f) = \mathbb{E}(e^{-m(f)}), \text{ for } f: E \rightarrow \mathbb{R}^+, f \in \mathcal{E}$$

Theorem [Characterization of PPP(Λ)]

Let N be a random measure on (E, \mathcal{E}) . Then N is a PPP(Λ)

$$\iff \mathcal{L}_N(f) = e^{-\Lambda(1-e^{-f})}, \quad \forall f \in \mathcal{E}, f \geq 0 \quad (*)$$

$$\text{where } \Lambda(1-e^{-f}) = \int_E (1-e^{-f(x)}) \Lambda(dx)$$

Proof: \Rightarrow :

1. First we verify (*) for function $f = a \cdot \mathbb{1}_A$, with $A \in \mathcal{E}$, $a > 0$:

$$N(a \mathbb{1}_A) = \int_E a \mathbb{1}_A(x) N(dx) = a N(A), \text{ hence}$$

$$\mathbb{E}(e^{-a N(A)}) = \sum_{j=0}^{\infty} e^{-a \cdot j} \cdot \frac{(\Lambda(A))^j}{j!} e^{-\Lambda(A)} = e^{-\Lambda(A)(1-e^{-a})}$$

check! \downarrow

$$= e^{-\Lambda(1-e^{-a \mathbb{1}_A})} \quad \checkmark$$

2. Then we verify (*) for simple functions $f = \sum_{i=1}^m a_i \cdot \mathbb{1}_{A_i}$, where $A_1, \dots, A_m \in \mathcal{E}$ are disjoint, $a_1, \dots, a_m > 0$:

$$N(f) = \sum_{i=1}^m a_i N(A_i), \text{ where } N(A_i) \text{ are indep. by def. of PPP}(\Lambda).$$

$$\text{Hence: } \mathbb{E}(e^{-N(f)}) = \prod_{i=1}^m \mathbb{E}(e^{-a_i N(A_i)}) = e^{-\sum_{i=1}^m \Lambda(A_i)(1-e^{-a_i})}$$

$$= e^{-\Lambda(1-e^{-f})} \quad \checkmark$$

3. We conclude the proof for general functions $f \in \mathcal{E}$, $f \geq 0$.

Consider $(f_m)_{m \in \mathbb{N}}$ simple functions s.t. $f_m \uparrow f$.

By the continuity of the Laplace functionals:

$$\mathbb{E}(e^{-N(f)}) = \lim_{m \rightarrow \infty} \mathbb{E}(e^{-N(f_m)}) = \lim_{m \rightarrow \infty} e^{-\Lambda(1-e^{-f_m})} \stackrel{\substack{\text{monotone convergence} \\ \text{theorem}}}{=} e^{-\Lambda(1-e^{-f})}$$

increasing to $1-e^{-f}$ \checkmark

\Leftarrow) follows from Proposition on the Laplace functional, together with the preceding computation. #

with the preceding computation.

#

Remark: Consider Example 1 with

$$m(A) = \sum_{k=1}^Y \mathbb{1}_A(X_k) \quad \text{for } Y \sim \text{Poi}(\lambda) \quad \left. \begin{array}{l} (X_k)_{k \in \mathbb{N}} \text{ i.i.d.} \sim \nu \\ \text{indep.} \end{array} \right\}$$

We have shown that

$$L_m(f) = e^{-\lambda \nu(1 - e^{-f})} \quad (\lambda \in \mathbb{R}^+, \nu \in \mathcal{P}(E, \mathcal{E}))$$

By the Theorem above, $m \sim \text{PPP}(\Lambda)$, with $\Lambda = \lambda \nu$.

This provides a general idea to construct $\text{PPP}(\Lambda)$ as shown by the next result.

Theorem: For any $\Lambda \in \mathcal{M}(E, \mathcal{E})$, there exists a $\text{PPP}(\Lambda)$.

Proof

1. Assume that $\Lambda(E) < \infty$ and define $\nu \in \mathcal{P}(E, \mathcal{E})$ s.t.

$$\nu(A) := \frac{\Lambda(A)}{\Lambda(E)}, \quad \forall A \in \mathcal{E}$$

Then consider $(X_k)_{k \in \mathbb{N}}$ i.i.d. $\sim \nu$ and indep. of $Y \sim \text{Poi}(\Lambda(E))$

and define

$$m(A) := \sum_{k=1}^Y \mathbb{1}_A(X_k), \quad \forall A \in \mathcal{E}$$

We have already shown that m is a random measure

with Laplace functional

$$L_m(f) = e^{-\Lambda(E) \cdot \nu(1 - e^{-f})} \stackrel{\nu(\cdot) = \frac{\Lambda(\cdot)}{\Lambda(E)}}{=} e^{-\Lambda(1 - e^{-f})}$$

By the previous Theorem [Characterization of $\text{PPP}(\Lambda)$], it turns out that $m \sim \text{PPP}(\Lambda)$ as wanted.

2. If $\Lambda(E) = \infty$ use com. (no. the) but that Λ is σ -finite

1. $m \sim \dots$ as usual.

2. If $\Lambda(E) = \infty$, we can use the fact that Λ is σ -finite, and hence that exist $(E_m)_{m \geq 1}$ s.t. $E_m \uparrow E$ with $\Lambda(E_m) < \infty$.

$$\text{Define: } \begin{cases} \Lambda^1(A) = \Lambda(E_1 \cap A) \\ \Lambda^n(A) := \Lambda((E_n \setminus E_{n-1}) \cap A) \end{cases} \quad \forall n \geq 2, \quad \forall A \in \mathcal{E}$$

Then Λ^n is finite, and by step 1. we can construct $m^n \sim \text{PPP}(\Lambda_n)$ independently, and set $m := \sum_{n=1}^{\infty} m^n \implies m \sim \text{PPP}(\Lambda)$ #

Corollary [Properties of PPP(Λ)]

1. If $f \in L^1(\Lambda) \implies \mathbb{E}(N(f)) = \Lambda(f)$

\implies for $f = \mathbb{1}_A$, we get $\mu(A) = \mathbb{E}(N(A)) = \Lambda(A)$ (mean intensity)

2. If $f \in L^2(\Lambda) \cap L^1(\Lambda)$, $\implies \text{Var}(N(f)) = \Lambda(f^2)$

\implies for $f = \mathbb{1}_A$, we get $\text{Var}(N(A)) = \Lambda(A)$

* not obvious if Λ is not a finite measure

Verify! The proof is an application of the previous Theorem.

Example

1. Particles in boxes:

• Let E countable, $\mathcal{E} = \mathcal{P}(E)$, and $\Lambda \in \mathcal{M}(E)$

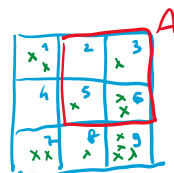
\hookrightarrow collection of boxes of \mathbb{R}^d

• Consider independent r.v. $(W_x)_{x \in E}$ s.t. $W_x \sim \text{Poi}(\Lambda(\{x\}))$

\hookrightarrow number of particles in the box x

Then $N(A) := \sum_{x \in E} W_x \mathbb{1}_A(x)$ $\forall A \in \mathcal{E}$
" $\delta_x(A)$ "

is a PPP(Λ) describing the particles in boxes \rightarrow

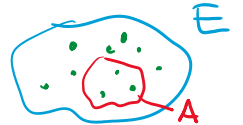


is a PPP(λ) describing the particles in boxes \rightarrow

2. Stones in a field:

- Let $Y \sim \text{Poi}(\lambda) \rightarrow$ number of stones
- Let $(X_m)_{m \in \mathbb{N}}$ iid on (E, \mathcal{E}) , with law ν , independent of Y
 \hookrightarrow each stone is thrown, and takes a position in E , independently of the other, described by X_m (for stone m)

Then $N(A) = \sum_{m=1}^Y \mathbb{1}_A(X_m)$, $\forall A \in \mathcal{E}$



is a PPP($\lambda \cdot \nu$) describing the configuration of the thrown stones in E . Indeed, this corresponds to the example 1 above where we obtained (with updated notation)

$$\mathcal{L}_\nu(A) = e^{-\lambda(1-\nu(A))} = e^{-\lambda\nu(1-e^{-\int A})}$$

\uparrow verify! $\rightarrow \lambda \in \mathbb{R}^+, \nu \in \mathcal{M}(E)$