

DEF: **Moments**. Let $n \in \mathbb{N}$,

$$m_n(x) = \mathbb{E}[x^n] \quad \text{"moment"},$$

$$\mu_n(x) = \mathbb{E}[(x - \mathbb{E}[x])^n] \quad \text{"centered moment"}.$$

PROP: • If $\mathbb{E}[|x|^n] < +\infty$, then $\phi_x \in C^n(\mathbb{I})$,

I open set containing 0, and :

$$m_k = \frac{1}{i^k} \cdot \frac{\partial^k}{\partial z^k} \phi_x(0), \quad k=1, \dots, n.$$

• If ϕ_x has n continuous derivatives in 0,
then :

$$m_k = \frac{1}{i^k} \frac{\partial^k}{\partial z^k} \phi_x(0), \quad k=1, 2, \dots, n.$$

DEF: **Moment generating function.**

$$M_x(u) = \mathbb{E}[e^{u \cdot x}]$$

$$m_n = \frac{\partial^n}{\partial u^n} M_x(0).$$

Rmk: $M_x(u) = \phi_x(-iu)$: easy to go $M_x \leftrightarrow \phi_x$.

DEF: **Characteristic exponent**.

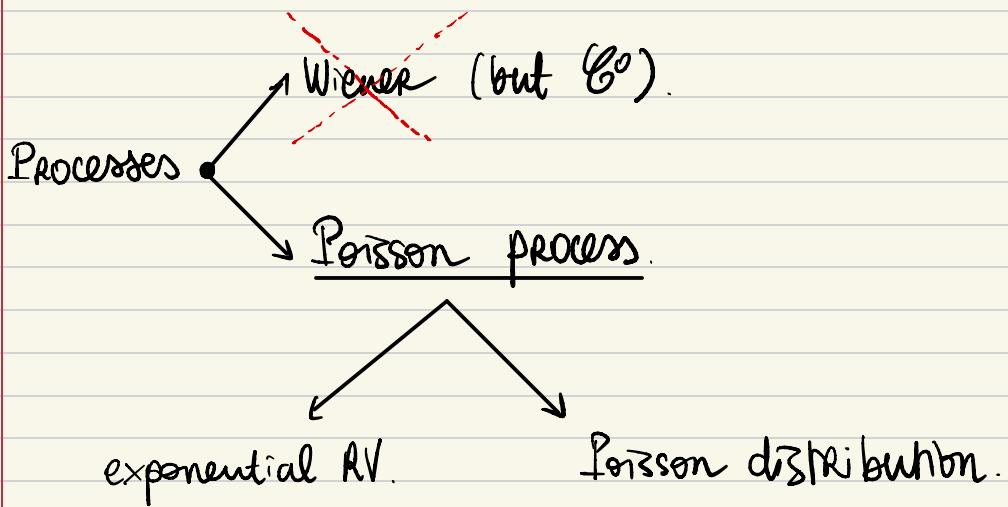
$$? \Psi_X \text{ s.t } \phi_X(u) = e^{\Psi_X(u)}.$$

$$\phi_X(0) = E[e^{i \cdot 0 \cdot X}] = E[e^0] = 1$$

Assume that ϕ_X is C^0 in 0. Then:

$$\boxed{\Psi_X(u) = \log(\phi_X(u))} \text{ in a neighborhood of 0}$$

$$(\Psi_X(0) = 0)$$



DEF: Exponential R.V. Y R.V. $\sim \text{Exp}(\lambda)$ iff:

$$Pr(Y \leq y) = \underbrace{1 - e^{-\lambda y}}_{\text{cdf}}, \forall y \in \mathbb{R}^+.$$

$$\text{pdf} = \lambda e^{-\lambda y} \mathbb{1}_{\{y \geq 0\}}$$

THM: Absence of memory. Let $T \geq 0$ a RV s.t

$$\mathbb{P}(T > t+s | T > t) = \mathbb{P}(T > s), \forall t, s > 0.$$

This is equivalent to: $T \sim \text{Exp}$.

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

$$\begin{aligned}\mathbb{P}(T > t+s | T > t) &= \frac{\mathbb{P}(T > t+s)}{\mathbb{P}(T > t)} = \frac{1 - \mathbb{P}(T \leq t+s)}{1 - \mathbb{P}(T \leq t)} \\ &= \frac{\lambda e^{-\lambda(t+s)}}{\lambda - (\lambda - e^{-\lambda t})} = \frac{e^{-\lambda(t+s)}}{e^{-\lambda t}} = e^{-\lambda s} \\ &= 1 - (1 - e^{-\lambda s}) = 1 - \mathbb{P}(T \leq s) = \mathbb{P}(T > s)\end{aligned}$$

• Let $g(t) = \mathbb{P}(T > t)$. We assume that

$$\mathbb{P}(T > t+s | T > t) = \mathbb{P}(T > s) \quad \forall t, s > 0.$$

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$$\frac{\mathbb{P}(T > t+s)}{\mathbb{P}(T > t)}$$

So:

$$\begin{aligned}g(t+s) &= \mathbb{P}(T > t+s) = \mathbb{P}(T > t) \mathbb{P}(T > s) \\ &= g(t)g(s)\end{aligned}$$

$g(t) = \mathbb{P}(T > t)$ is $\begin{cases} \text{RIGHT \&} \\ \text{NON-INCREASING} \end{cases}$ and

$g(t+s) = g(t)g(s) \quad \forall t, s > 0$. So: $g(t) = e^{-\lambda t}$ for any t . So $T \sim \text{Exp}(\lambda)$. ■

DEF: Poisson distribution. N r.v with values in \mathbb{N} .
 $N \sim \text{Pois}(\lambda)$ iff: $P(N=n) = e^{-\lambda} \frac{\lambda^n}{n!}, \forall n \in \mathbb{N}$.

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$$M(u) = e^{\lambda(e^u - 1)}$$

PROP: $(\tau_i)_{i \geq 1}$ R.V iid $\sim \text{Exp}(\lambda)$,

$$N_t = \inf \{n \geq 0 : \sum_{i=1}^{n+1} \tau_i > t\}, \quad \forall t > 0.$$

We have that $N_t \sim \text{Pois}(\lambda t)$, that is:

$$\forall n \in \mathbb{N}, \quad P(N_t = n) = e^{-\lambda t} \frac{(\lambda t)^n}{n!}. \quad \text{pdf of a sum of iid RV}$$

Proof: Let: $T_n = \sum_{i=1}^n \tau_i$. this comes from basic theory.
 $P_n(t) = \lambda e^{-\lambda t} \frac{(\lambda t)^{n-1}}{(n-1)!} \underset{[0, +\infty)}{\uparrow}(t)$ } This is the pdf of T_n .

$$\begin{aligned} P(T_{n+1} > t) &= 1 - P(T_{n+1} \leq t) \xleftarrow{\text{cdf of } T_{n+1}} \\ &= 1 - \int_0^t \lambda e^{-\lambda s} \frac{(\lambda s)^n}{n!} ds \end{aligned}$$

$P_{n+1}(s)$ is pdf of T_{n+1} .

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

f' g

$$f = e^{-\lambda s}, \quad g^1 = \frac{(\lambda s)^{n-1}}{(n-1)!} \times \lambda$$

$$\begin{aligned} \text{So: } P(T_{n+1} > t) &= 1 + \left[e^{-\lambda s} \frac{(\lambda s)^n}{n!} \right]_0^t - \underbrace{\int_0^t e^{-\lambda s} \frac{(\lambda s)^{n-1}}{(n-1)!} ds}_{P_n(s)} \\ &= 1 + e^{-\lambda t} \frac{(\lambda t)^n}{n!} - P(T_n < t) \\ &= e^{-\lambda t} \frac{(\lambda t)^n}{n!} + P(T_n > t) \end{aligned}$$

It means:

$$\begin{aligned} P(T_{n+1} > t) - P(T_n > t) &= e^{-\lambda t} \frac{(\lambda t)^n}{n!} \\ P(N_t = n) &\quad \text{we recognize a Poisson distribution.} \end{aligned}$$

$N_t \leq n$

So finally: $N_t \sim \text{Poisson}(\lambda t)$. ■

PROP: • $Y_1 \sim \text{Poisson}(\lambda_1)$, $Y_2 \sim \text{Poisson}(\lambda_2)$, & $Y_1 \perp Y_2$.

Then: $Y_1 + Y_2 \sim \text{Poisson}(\lambda_1 + \lambda_2)$.

• $Y \sim \text{Poisson}(\lambda)$. Then: $Y = \sum_{i=1}^n Y_i$ where Y_1, Y_2, \dots, Y_n iid $\sim \text{Exp}(\lambda/n)$, $\forall n$ [INFINITE DIVISIBILITY]

Poisson Process:

DEF: $(\tau_i)_{i \geq 1}$ sequence of iid RV $\sim \text{Exp}(\lambda)$.

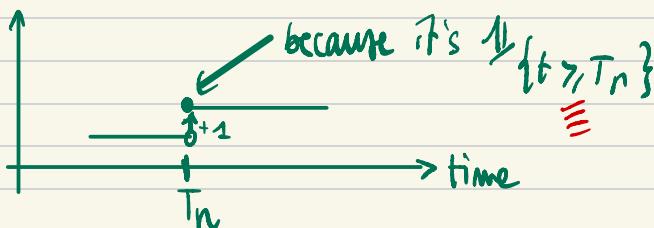
$T_n = \sum_{i=1}^n \tau_i$. $N_t = \sum_{n=1}^{+\infty} \mathbb{1}_{\{t \geq T_n\}}$ is a Poisson Process with intensity λ .

Rmk: This definition is exactly equivalent to the previous one with the inf.

If we fix time, N_t RV $\sim \text{Poisson}(\lambda t)$.

$(N_t)_{t \geq 0}$ Poisson Process w/ intensity λ .

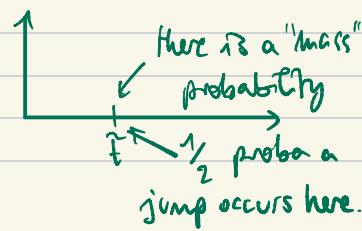
Rmk: $(N_t)_{t \geq 0}$ is a CADLAG process:



$\rightarrow N_{t^-} = N_t$ with probability 1. Let's write things about that: assume $\exists \tilde{t}$ s.t

$$\mathbb{P}(N_{\tilde{t}^-} \neq N_{\tilde{t}}) = \frac{1}{2} (\neq 1)$$

We don't want such things.



Fortunately, with Poisson Process we don't have mess !

PROP:

$$\bullet \mathbb{P}(N_t = n) = e^{-\lambda t} \frac{(\lambda t)^n}{n!} \quad \forall t, \forall n.$$

$$\bullet \mathbb{E}[e^{iuN_t}] = e^{\lambda t(e^{iu}-1)}$$

$\bullet (N_t)_{t \geq 0}$ has independent increments:

$$\forall t_1 < t_2 < \dots < t_n, N_{t_n} - N_{t_{n-1}} \perp\!\!\!\perp N_{t_{n-1}} - N_{t_{n-2}}, \dots$$

$$N_{t_2} - N_{t_1}, N_{t_1} \\ (N_0 = 0)$$

$$\bullet \mathbb{E}[N_t] = \lambda t, \quad \forall t \geq 0.$$

⚠ We don't have martingale property with (N_t) .

So we need to define the "compensated Poisson Process".

DEF:

Compensated Poisson Process :

$$\hat{N}_t = N_t - \lambda t.$$

$$\hat{N}_0 = 0 \quad \text{and} \quad E[\hat{N}_t] = E[N_t] - \lambda t = 0.$$

Clearly, $(\hat{N}_t)_t$ is not a Poisson Process since it doesn't take only \mathbb{N} values.

PROP: For the compensated Poisson Process :

$$\phi(z) = e^{\lambda t(e^{iz} - 1 - iz)}$$

$$\begin{aligned}
 \phi(z) &= E[e^{iz\hat{N}_t}] = E[e^{iz(N_t - \lambda t)}] \\
 &= E[e^{izN_t} e^{-iz\lambda t}] = e^{-iz\lambda t} \phi_{N_t}(z) \\
 &= e^{-iz\lambda t} e^{\lambda t(e^{iz} - 1)} = e^{\lambda t(e^{iz} - 1 - iz)} \blacksquare
 \end{aligned}$$

PROP:

$(\hat{N}_t)_{t \geq 0}$ is a martingale.

$$\forall t > s, \quad E[\hat{N}_t | \hat{N}_s] = \hat{N}_s.$$

Proof:

$$E[\hat{N}_t | \hat{N}_s] = E[\tilde{N}_t - \tilde{N}_s + \tilde{N}_s | \hat{N}_s]$$

$$= E[\tilde{N}_s | \hat{N}_s] + E[\tilde{N}_t - \tilde{N}_s | \hat{N}_s]$$

$$= \tilde{N}_s + \underbrace{E[\tilde{N}_t - \tilde{N}_s]}_{=0} = \tilde{N}_s. \blacksquare$$

THM: Let $(X_t)_{t \geq 0}$ be a counting process with independent & stationary increments. Then $(X_t)_{t \geq 0}$ is a Poisson Process.

Rank: "Stationary increments" means that

$$\forall t > s, h > 0, X_{t+h} - X_{s+h} \sim X_t - X_s.$$

The previous theorem tells us that:

Poisson Process is one only "counting process" choice

If we want to work w/ independent & stationary increments : this will be the framework of this course.

Rank: T_1, \dots, T_n sequence of random jump times.

$$N_t = \#\{i \geq 1 : T_i \in [0, t]\}, \quad N_t - N_s = \#\{i \geq 1 : T_i \in]s, t]\}. \quad \left. \right\} \text{intuitive way of thinking about it.}$$

DEF: Let's introduce a probability space :

Ω , \mathcal{F} event and :