

IEOR 4102 Lec 10

Intro to point processes,
renewal processes,
Elementary Renewal Theorem

Poisson processes

A simple point process

is a sequence of points $\{t_n : n \geq 1\}$

$$0 < t_1 < t_2 < \dots \quad \lim_{n \rightarrow \infty} t_n = \infty \quad \left(t_0 \stackrel{\text{def}}{=} 0 \right)$$

$\Psi = \{t_n : n \geq 1\}$ interarrival times

$$X_1 = t_1$$

$$X_2 = t_2 - t_1$$

$$\vdots \quad X_n = t_n - t_{n-1}, \quad n \geq 1$$

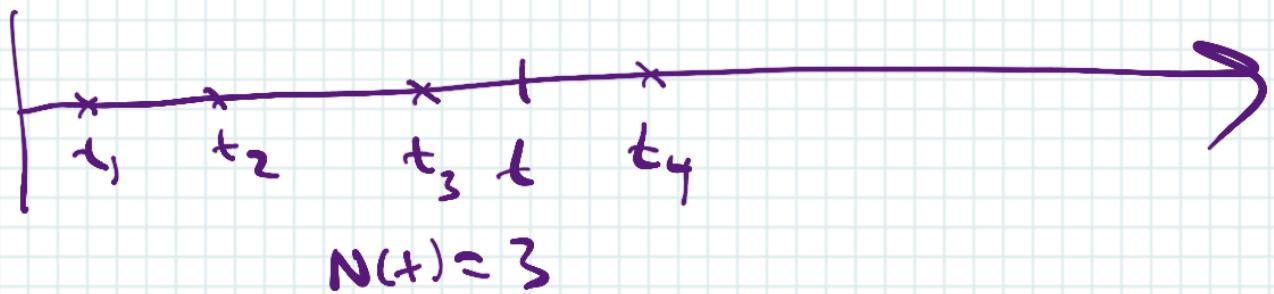
$$t_n = X_1 + \dots + X_n$$

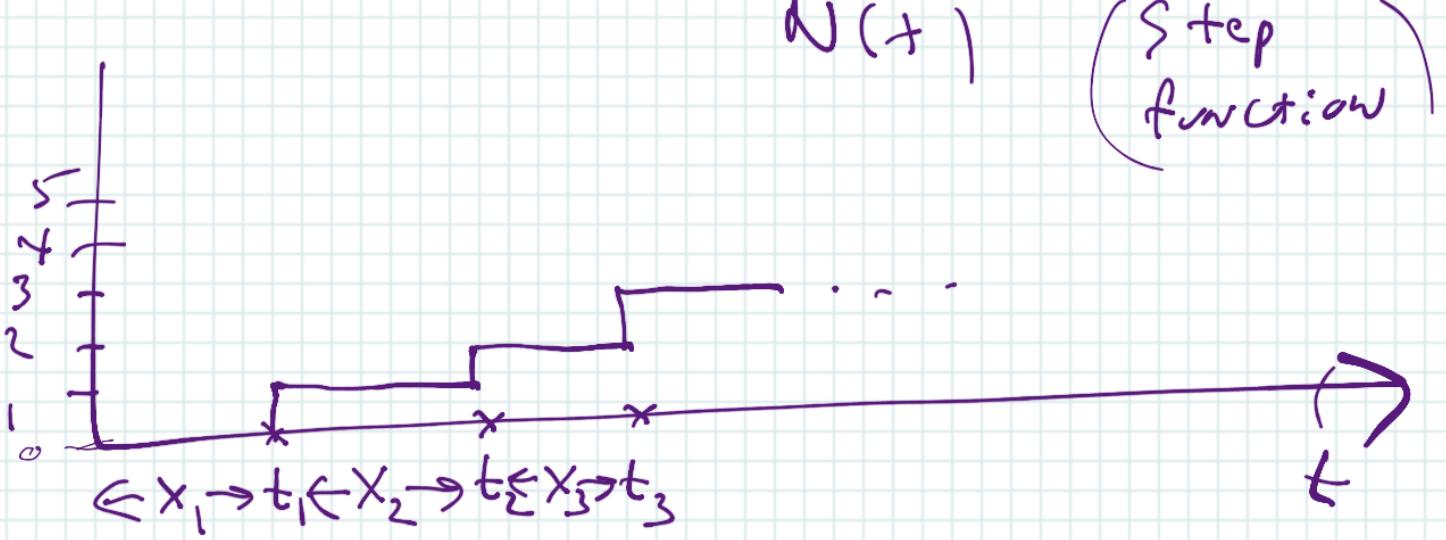
The counting process
 $\{N(t) : t \geq 0\}$

$N(t) = \# \text{ pts in the time interval } [0, t]$

$$N(0) = 0$$

$$N(t) = \max \{ n \geq 1 : t_n \leq t \}$$





$$\{N(+) = 0\} = \{t_1 > t\} = \{x_1 > t\}$$

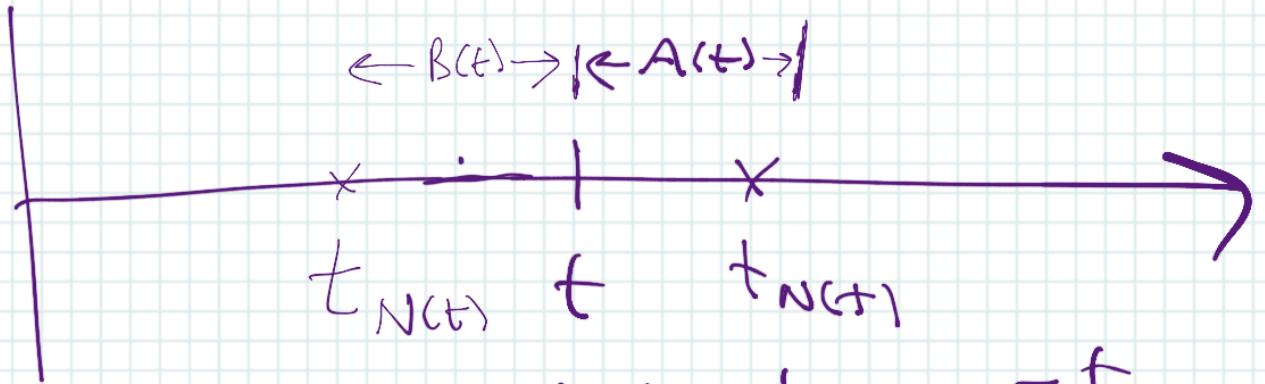
$$\{N(+) = n\} = \{t_n \leq t, t_{n+1} > t\}$$

$$\{N(+) \geq n\} = \{t_n \leq t\}$$

Modeling

- 1) times at which your phone calls come in
- 2) times at which jobs are sent to a printer
- 3) times that subways arrive to a station
- 4) times at which a claim comes in for an insurance company

etc.



$$A(t) = t_{N(+)} + t$$

$$B(t) = t - t_{N(+)}$$

When $\{t_n\}$ ($\{x_n\}$)

are random variables

We call $\Psi = \{t_n\}$ a
random point process

$$P(N(t) = 0) = P(t_1 > t) = P(X_1 > t)$$

$$\boxed{P(N(t) = n)} = P(t_n \leq t, t_{n+1} > t)$$

$$= P(X_1 + \dots + X_n \leq t, X_1 + \dots + X_{n+1} > t)$$

$$E(N(t)) = ? \quad \text{Var}(N(t)) = ?$$

A renewal process is
a random point process
 $\Psi = \{t_n\}$
in which the interarrival
times
 $\{X_n\}$ are iid with interarrival time

dist. $F(x) = P(X \leq x)$ $x \geq 0$

$$\left(0 < E(X) < \infty \right) \text{ we assume} \quad \lambda \stackrel{\text{def}}{=} \frac{1}{E(X)}$$

"rate"

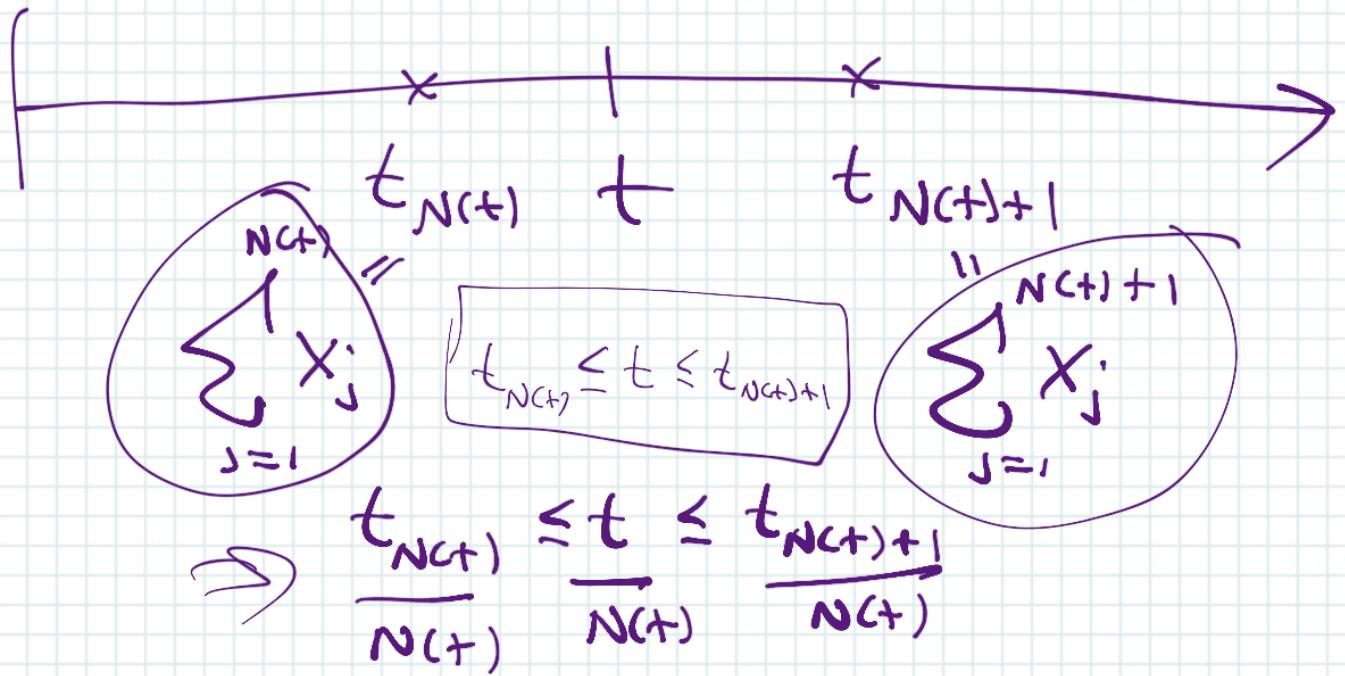
Elementary renewal Thm:

For a renewal process

a) $\lim_{t \rightarrow \infty} \frac{N(t)}{t} = \lambda$ wp |
long-run rate
at which points
occur

b) $\lim_{t \rightarrow \infty} \frac{E(N(t))}{t} = \lambda$

Proof of (a) $t_n = \underbrace{x_1 + \dots + x_n}_{\text{iid}}$



$$L(+)=\frac{1}{N(+)} \sum_{j=1}^{N(+)} X_j \leq \left(\frac{t}{N(+)} \right) \leq \frac{1}{N(+)} \sum_{j=1}^{N(+)+1} X_j = R(+)$$

$$\lim_{t \rightarrow \infty} L(+) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^n X_j = E(X) = \frac{1}{\lambda}$$

wp | SLLN

$$\lim_{t \rightarrow \infty} R(+) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^{n+1} X_j = E(X) = \frac{1}{\lambda}$$

$$\Rightarrow \lim_{t \rightarrow \infty} \frac{t}{N(+)} = \frac{1}{\lambda} \text{ wp } \boxed{\Rightarrow \lim_{t \rightarrow \infty} \frac{N(+)}{t} = \lambda \text{ wp }}$$

$$P(N+J=5) = ?$$

We would need
more information
about $F(x)$

We now will consider
the very special case of
a renewal process for which

$$F(x) = P(X \leq x) = 1 - e^{-\lambda x} \quad x \geq 0$$

$$E(X) = \frac{1}{\lambda} \quad \text{exp}(\lambda) \text{ dist.}$$

Called a Poisson Process at rate λ

$$t_n = X_1 + \dots + X_n$$

has an Erlang (n, λ) dist,

X_i is called the
 $1 \leq i \leq n$ "i-th phase"

Erlang(n, λ) is a special case
of a Gamma (n, λ) dist.

$$\begin{aligned} E(t_n) &= E(X_1 + \dots + X_n) \\ &= n E(X) = \frac{n}{\lambda}, \quad n \geq 1 \end{aligned}$$

why is this called a
"Poisson" process?

Because: for every fixed $t > 0$,
 $N(t)$ has a Poisson dist.
 with mean $E(N(t)) = \lambda t = \text{Var}(N(t))$

$$P(N(t) = k) = \frac{-\lambda^k (\lambda^t)^k}{k!}, \quad k \geq 0$$

$$P(N(t) = 0) = e^{-\lambda t}$$

=

$$P(t_1 > t)$$

$$= P(X_1 > t)$$

for all $s \geq 0 \quad t \geq 0$

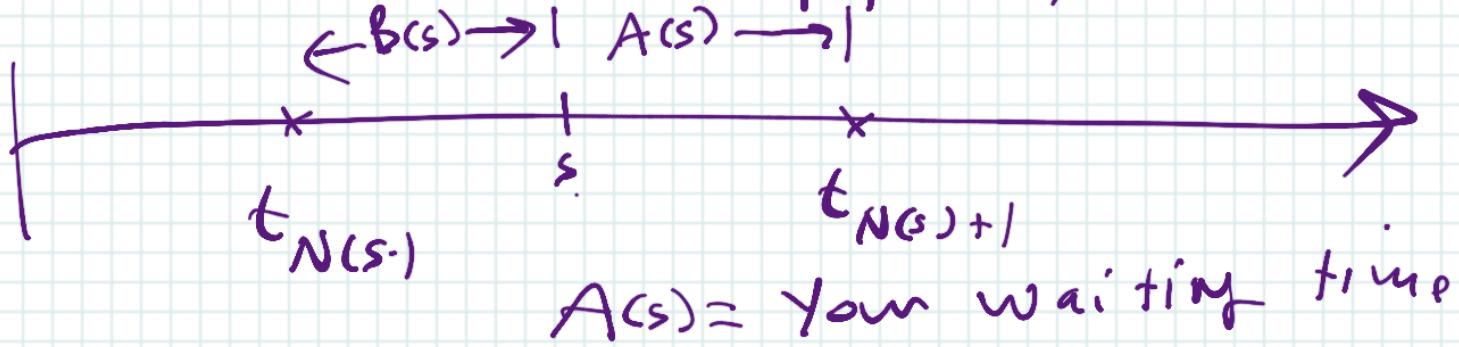
$$N(s+t) - N(s) = \begin{matrix} \# \text{ arrivals in } (s, s+t] \\ \text{length interval} \end{matrix}$$

$\stackrel{s \text{ ist}}{=} N(t)$

$$\begin{aligned} & \Pr(N(s+t) - N(s) = k) \\ &= \Pr(N(s+t) = k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \quad k \geq 0 \end{aligned}$$

imagine subways arriving

according to a PP(λ)



$A(s)$ = your waiting time

By the memoryless
Property of the exponential
dist.

$$A(s) \sim \exp(\lambda)$$

and is independent of $B(s)$
independent of the past

$$\{N(u) : 0 \leq u < s\}$$

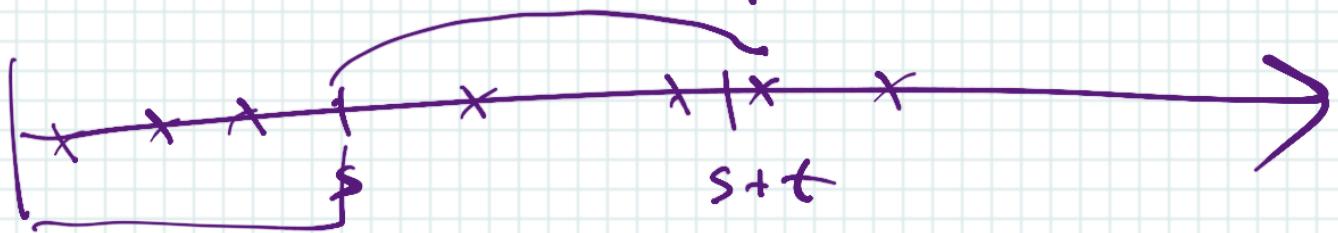
$$P(A(s) > t) = e^{-\lambda t}, \quad t \geq 0$$

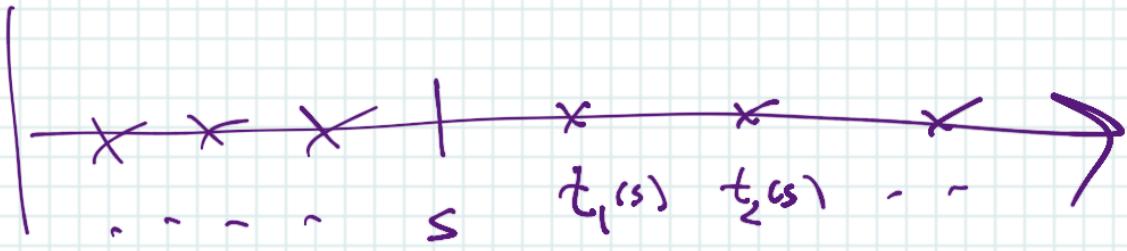
For any s

$\Rightarrow N(s+t) - N(s)$ is independent of $N(s)$

$\stackrel{\text{dist}}{=} N(t)$ for all s

the $PP(\lambda)$ process from time s onward is yet again $PP(\lambda)$ independent of the past





$$\begin{aligned} t_1(s) &= t_{N(s)+1} - s \\ t_2(s) &= t_{N(s)+2} - s \end{aligned} \quad \left. \right)$$

we consider
the PP(λ)
starting
from time s

$\{t_n(s) : n \geq 1\} \sim \text{PP}(\lambda)$
independent of $\{N(u) : 0 \leq u \leq s\}$

$$\underline{N(s+t)} = \underbrace{N(s)}_{\mathcal{T}} + \underbrace{(N(s++) - N(s))}_{\text{independent}} \stackrel{\text{dist.}}{=} \underline{N(+)}$$

Present State

Markov property holds

Continuous-time Markov
chain

$(N(t))$ for a $PP(\lambda)$

has both stationary
and independent increments

1) Stationary increments Definition

For all $s \geq 0, t \geq 0$

$N(s+t) - N(s)$ has a dist.
that only depend on the length t
of the interval $\{s, s+t\}$

$\stackrel{\text{dist}}{=}$

$N(t)$

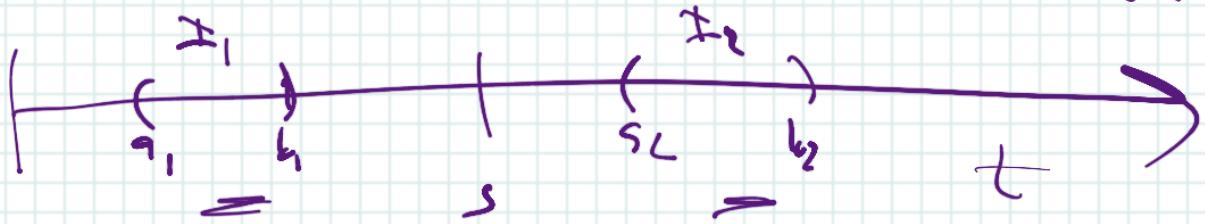
2) Independent increments Definition:

$I = (a, b)$ $a \leq a < b$ an interval.

Let $N(I) = N(b) - N(a) = \# \text{ pts that fall in } I = (a, b)$

If I_1 and I_2 are non-overlapping intervals

then $N(I_1)$ and $N(I_2)$ are independent vs.



For a PP(λ)

$N(t)$ has a Poisson dist.
with mean λt .

and by stationary increments we thus
have $N(s+t) - N(t) \xrightarrow{\text{dist.}} N(t)$

$$P(N(s+t) - N(t) = k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \quad k \geq 0$$

Examples

(cars arrive to the GW Bridge
 $\sim \text{PP}(x)$ $\lambda = 1000/\text{hr}$

a) $E(\# \text{arrives between } 2 \text{ pm} \text{ and } 3 \text{ pm})$
 $= E(N(3) - N(2)) = E(N(1))$ Stationary increments
 $= \lambda \cdot 1 = \lambda$

b) $E(N(3) - N(2) \mid N(10) - N(9) = 700) = 1000$
 $\uparrow = E(N(3) - N(2))$ Independent increments
 $= E(N(1)) = 1000$

c) $P(\text{No arrivals in a given 15 minute interval})$

$$15 \text{ min} = \frac{1}{4} \text{ hr}$$

$$P(N(\frac{1}{4})=0) = e^{-\lambda \cdot \frac{1}{4}} = e^{-250}$$

=

$$P(t_1 > \frac{1}{4}) \\ = P(X_1 > \frac{1}{4}) = e^{-\lambda \cdot \frac{1}{4}} = e^{-250}$$

Fact: The PP(λ) process

is the only simple point process with both stationary and independent increments

So if you have a simple point process and it has both stationary & independent increments then it is a Poisson process with $\lambda = \underline{E(N(1))}$

Proof involves proving that

(X_n) are iid exponential rvs.

Let's show that X_1 is exponential

via showing it has the
memoryless property

$$\text{P}(X_1 > s+t) \stackrel{?}{=} \text{P}(X_1 > s) \text{P}(X_1 > t)$$

$$\text{P}(N(s+t) = 0) = \text{P}(N(s) = 0, N(s+t) - N(s) = 0)$$

$$= \text{P}(N(s) = 0) \text{P}(N(s+t) - N(s) = 0)$$

$$= \text{P}(N(s) = 0) \text{P}(N(t) = 0)$$

$$= \text{P}(X_1 > s) \text{P}(X_1 > t) \quad \checkmark$$

independent increments

stationary increments

We continue using
independent & stationary
increments sequentially to show
(induction) that X_{n+1} has the same exponential
dist. as X_n
and is independent of X_1, \dots, X_n

$\Rightarrow \{X_n\}$ iid $\text{exp}(\lambda)$ some
 $\lambda > 0.$

Subway Example

Consider 94th / Broadway Station Subway

to go downtown.

(N₁(t)) 1) Locals (L) ~ PP (λ_1) $\lambda_1 = 7/\text{hr}$

(N₂(t)) 2) Express (E) ~ PP (λ_2) $\lambda_2 = 3/\text{hr}$
(E) independent

You arrive

q) You decide to take 1st express

$$\begin{aligned} \text{P(Wait > 10 min)} &= \text{P}(N_2(\frac{1}{6}) = 0) \\ &= e^{-\lambda_2 \cdot \frac{1}{6}} = e^{\frac{1}{2}} \approx 0.61 \end{aligned}$$

b) You decide to take which ever arrives last

$$E(\text{Waiting time}) = E(\min(X_{1(1)}, X_{1(2)}))$$

$\stackrel{=}{\rightarrow} E(Z)$

$X_{1(1)} \sim \text{Exp}(\lambda_1)$ = time until first L

$X_{1(2)} \sim \text{Exp}(\lambda_2)$ = time until first E

$\frac{1}{\lambda_1 + \lambda_2} = \frac{1}{10}$

$\lambda_1 + \lambda_2 = 6 \text{ minutes}$

$$E(\text{Waiting time} \mid \text{local was first})$$

$$E(Z \mid X_{1(1)} < X_{1(2)}) = E(Z) = 6 \text{ min}$$

also.