

6.2 Poisson Process

Summarize Bernoulli Process

- $X_1, X_2, \dots$ $\left\{ \begin{array}{l} \text{(i.i.d)} \\ \text{independent} \\ \text{identically} \\ \text{distributed} \\ \text{Bernoulli r.v.s} \end{array} \right.$
- $Y_k = \text{time to } k^{\text{th}} \text{ success (arrival)}$
- $T_k = Y_k - Y_{k-1} = k^{\text{th}} \text{ interarrival time}$
- $N_t = X_1 + \dots + X_t = \# \text{ of arrivals in } \{1, 2, \dots, t\}$

We know!

$$P_{Y/K}(t) = \binom{t-1}{k-1} \cdot p^k (1-p)^{t-k} \quad k=k, k+1, \dots$$

$$P_{I/K}(t) = (1-p)^{t-1} p \quad t=1, 2, \dots$$

$$P_{N_t}(k) = \binom{t}{k} p^k (1-p)^{t-k} \quad k=0, 1, 2, \dots$$

For a continuous time arrival process
we define

$$P(k, \tau) = P(\# \text{ of arrivals in a time interval of len. } \tau \text{ equal to } k)$$

Defn

Such an arrival Process is a
Poisson Process with rate $\lambda > 0$
iff

(a) time homogeneity

$P(k, \tau)$ is the same for all
intervals of length τ .

(b) independence

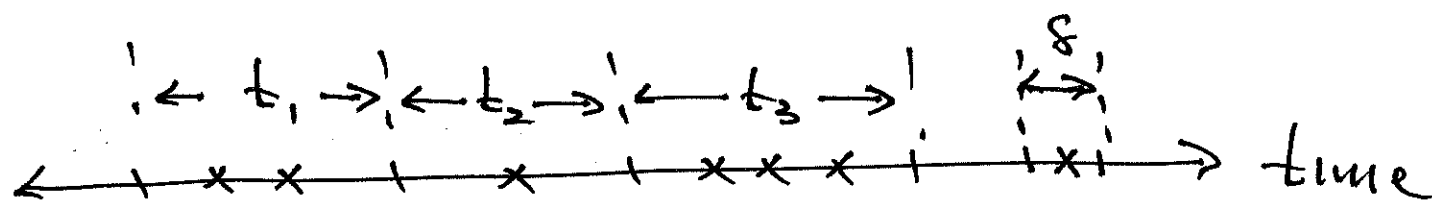
The # of arrivals in disjoint
time intervals are independent.

(c) Small interval Probabilities

let $\delta > 0$ be small. Then

$$P(k, \delta) \approx \begin{cases} 1 - \lambda\delta & \text{if } k=0 \\ \lambda\delta & \text{if } k=1 \\ 0 & \text{if } k=2, 3, \dots \end{cases}$$

we call λ the arrival rate



more precisely

$$P(k, \delta) = \begin{cases} 1 - \lambda\delta & k=0 \\ \lambda\delta & k=1 \\ 0 & k=2, 3, \dots \end{cases} + o(\delta)$$

where $o(\delta)$ denotes a function with the property

$$\lim_{\delta \rightarrow 0} \frac{o(\delta)}{\delta} = 0$$

For instance, the function $c\delta^2$ has this property.

Observe that

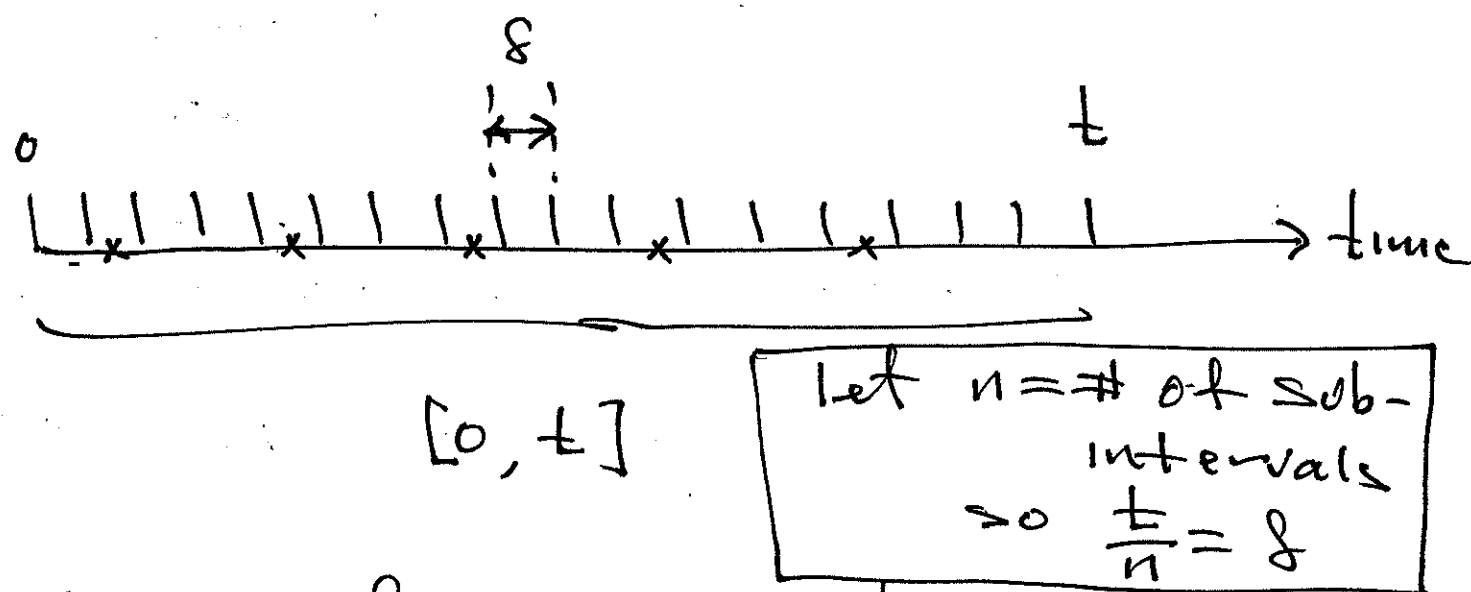
$$E[\text{\# arrivals in } [t, t+\delta]]$$

$$= (1-\lambda\delta) \cdot 0 + \lambda\delta \cdot 1$$

$$= \lambda\delta$$

$\therefore \lambda$ is expected # arrivals per unit time.

To reason about a large time interval, we divide into small intervals



choose δ so small that prob. of more than one arrival in one small interval is negligible

we now have (approximately) a Bernoulli process with

$$p = \lambda \delta = \frac{\lambda t}{n}$$

note $\delta \rightarrow 0$ iff $n \rightarrow \infty$. so

$$P(k \text{ arrivals in } [0, t])$$

$$= \binom{n}{k} \left(\frac{\lambda t}{n}\right)^k \left(1 - \frac{\lambda t}{n}\right)^{n-k} \quad \left\{ \begin{array}{l} \text{Binomial} \\ \text{dist.} \end{array} \right.$$

In limit as $n \rightarrow \infty$ ($\delta \rightarrow 0$) this becomes Poisson dist. with Param. λt . Thus

$$P(k \text{ arrivals in } [0, t])$$

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

Let N_t denote the #
of arrivals of the Poisson
Process in $[0, t]$, then

$$P_{N_t}(k) = e^{-\lambda t} \cdot \frac{(\lambda t)^k}{k!} \quad (k=0, 1, \dots)$$

We have

$$E[N_t] = \lambda t$$

$$\text{Var}(N_t) = \lambda t$$

Let T be time to 1st arrival. We derive the distribution at T . First note

$$\{T > t\} = \{N_t = 0\}$$

Thus

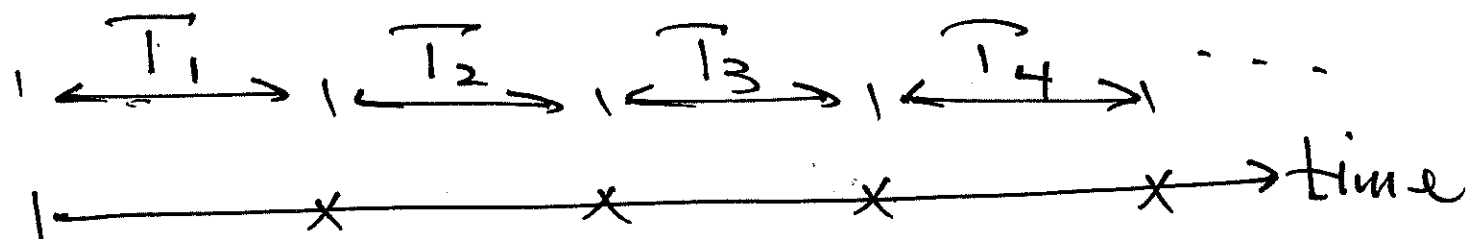
$$\begin{aligned} F_T(t) &= P(T \leq t) \\ &= 1 - P(T > t) \\ &= 1 - P(N_t = 0) \\ &= 1 - P_{N_t}(0) \\ &= 1 - e^{-\lambda t} \end{aligned}$$

Now diff. w.r.t t to get

$$f_T(t) = \lambda e^{-\lambda t} \quad (t \geq 0)$$

Thus T is exponential with parameter λ .

Now let T_k be the k^{th} inter-arrival time



Recall, the exponential r.v. is memoryless

Thus each T_k is exponential with Param. λ

$$f_{T_k}(t) = \lambda e^{-\lambda t} \quad (t \geq 0)$$

Now define

$$Y_k = T_1 + T_2 + \dots + T_k \quad (k=1, 2, \dots)$$

i.e. $Y_k = \text{time to } k^{\text{th}} \text{ arrival.}$

The dist. of Y_k is derived in the text (P. 316-317), called Erlang PDF of order k

$$f_{Y_k}(t) = \frac{\lambda^k \cdot t^{k-1} \cdot e^{-\lambda t}}{(k-1)!} \quad (t \geq 0)$$

Exercise

Prove that the sum of k independent exponential r.v.'s (of Param. λ) is Erlang of order k (Param λ)

Hint use induction on k

$$\text{Base: } Y_1 = T_1$$

$$\text{Induction: } Y_k = Y_{k-1} + T_k$$

lets summarize analogy

Bernoulli

Poisson

time: $t \in \{1, 2, 3, \dots\}$

$t \in [0, \infty)$

rate: p/trial

$\lambda/\text{unit time}$

N_t : Binomial(t, p) Poisson(λt)

T_k : Geometric(p) Exponential(λ)

Y_k : Pascal(p, k) Erlang(λ, k)

Ex.

You receive emails according to a Poisson process at rate

$$\lambda = 2 \text{ msg./hour}$$

you currently have no messages.

- a.) What is probability that you have 3 new messages when you check 30 min from now?

$$P(N_{\frac{1}{2}} = 3) = P_{N_{\frac{1}{2}}}^{(3)}$$

$$= e^{-2 \cdot \frac{1}{2}} \cdot \frac{(2 \cdot \frac{1}{2})^3}{3!} = e^{-1} \cdot \frac{1}{6} = \boxed{.0613}$$

b.) what is Prob. at no messages?

$$P(N_{\frac{1}{2}} = 0) = P_{N_{\frac{1}{2}}}(0)$$

$$= e^{-2 \cdot \frac{1}{2}} \cdot \frac{(2 \cdot \frac{1}{2})^0}{0!}$$

$$= e^{-1} = \boxed{.3679}$$

c.) what is expected time to the 5th new message?

$$E[Y_5] = E[T_1 + \dots + T_5]$$

$$= 5 \cdot \frac{1}{\lambda} = \frac{5}{2} = \boxed{2.5 \text{ hours}}$$

The Sum of Poisson R.V.s

let X, Y be ind. Poisson
with Param. λ_1, λ_2 resp. we
will show that

$$Z = X + Y$$

is also Poisson with Param.
 $\lambda_1 + \lambda_2$.

Proof later ...