

Last lecture

Continuous RV (Ch 3)

- Motivation
- Cumulative Distribution Function (Ch 3.1)
- Examples
- CDF to PMF and probabilistic density function (PDF)

Continuous RV & Probability Density Function (Ch 3.2)

- Definition
- Facts

Uniform Distribution (Ch 3.3)

Agenda

Uniform Distribution (Ch 3.3)

Exponential Distribution (Ch 3.4)

- Memoryless property
- Connect to $\text{Geo}(p)$

Poisson process (Ch 3.5)

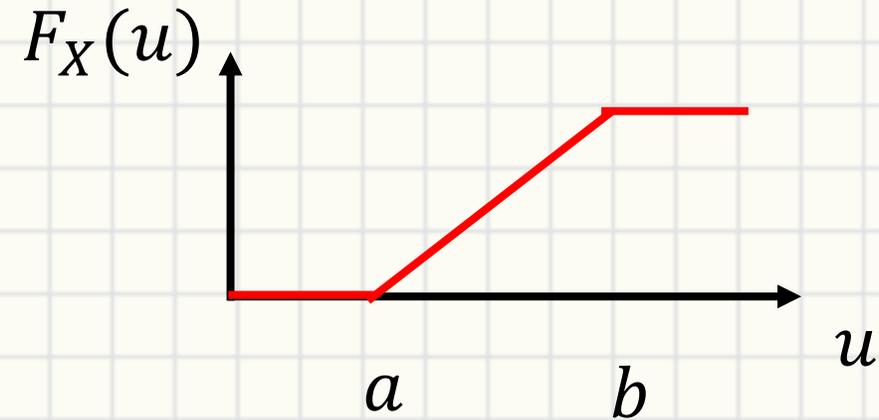
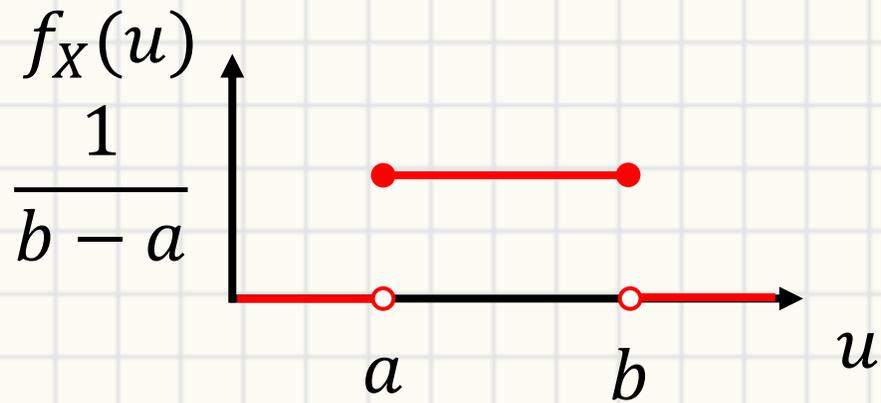
- Motivation
- Bernoulli process to Poisson process
- Definition
- Properties

Erlang Distribution (Ch 3.5.3)

Uniform Distribution

Uniform Distribution

$$f_X(u) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq u \leq b \\ 0 & \text{else} \end{cases}$$



Properties

$$f_X(u) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq u \leq b \\ 0 & \text{else} \end{cases}$$

- $E[X] = \int_{-\infty}^{\infty} u f_X(u) du = \int_a^b \frac{u}{b-a} du = \frac{b-a}{2}$
- $E[X^2] = \int_{-\infty}^{\infty} u^2 f_X(u) du = \int_a^b \frac{u^2}{b-a} du = \frac{b^2+ab+a^2}{3}$
- $Var(X) = \frac{(a-b)^2}{12}$
- Special case, when $(a, b) = (0, 1)$
 - k^{th} moment $E[X^k] = \frac{1}{k+1}$
 - $Var(X) = \frac{1}{12}$

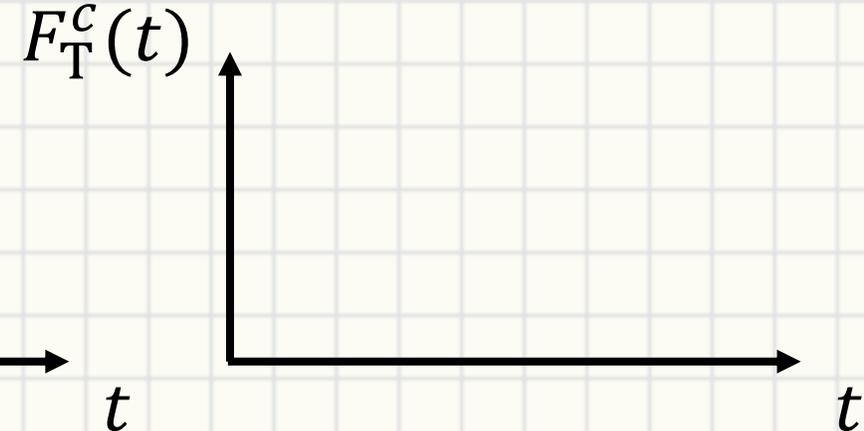
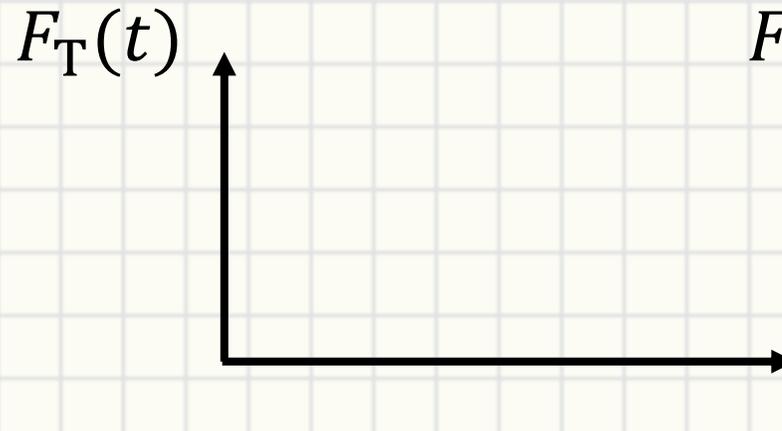
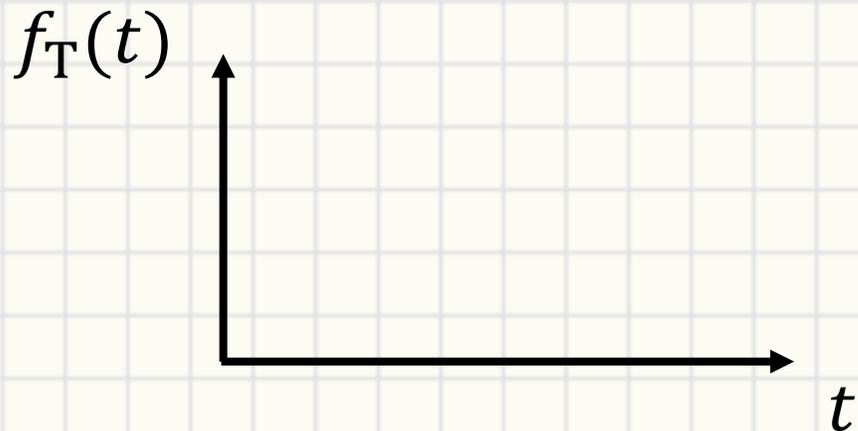
Exponential Distribution

Exponential Distribution

Motivation – System life for failure rate λ

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

$$F_T(t) = \begin{cases} 1 - e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$



Properties

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

- $$\begin{aligned} E[T^n] &= \int_0^{\infty} t^n \lambda e^{-\lambda t} dt \\ &= -t^n e^{-\lambda t} \Big|_0^{\infty} + \int_0^{\infty} n t^{n-1} e^{-\lambda t} dt \\ &= 0 + \frac{n}{\lambda} \int_0^{\infty} t^{n-1} \lambda e^{-\lambda t} dt = \frac{n}{\lambda} E[T^{n-1}] \end{aligned}$$

- $$E[T] = \frac{1}{\lambda} \quad E[T^2] = \frac{2}{\lambda^2} \quad E[T^n] = \frac{n!}{\lambda^n}$$

- $$\text{Var}(T) =$$

Examples

$$F_T(t) = \begin{cases} 1 - e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

Let $T \sim \text{Exp}(\lambda = \ln 2)$, find $P\{T \geq t\}$ and $P(T \leq 1 | T \leq 2)$

Memoryless Property

$$F_T(t) = \begin{cases} 1 - e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

$$P\{T \geq t\} = e^{-\lambda t}$$

- $P\{T \geq s + t | T \geq s\} =$
- If T is the system lifetime

Connecting *Exp* with *Geo*

$$F_T(t) = \begin{cases} 1 - e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

Summary - $F_L\left(\left\lfloor \frac{c}{h} \right\rfloor\right) \rightarrow F_T(c)$ when $h \rightarrow 0$

- $L \sim Geo(p = \lambda h)$
- $T \sim Exp(\lambda = \lambda)$

A lightbulb of average lifetime 1000hrs

- Failed hour = $L \sim Geo(p = \frac{1}{1000})$
- Let's assume it will only fail at start of each ticks h hours (e.g., sec, $h = 1/3600$)
- Failed ticks = $L_h \sim Geo(p_h = \frac{1}{1000} \times h)$

Connecting *Exp* with *Geo*

$$F_T(t) = \begin{cases} 1 - e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

A lightbulb of average lifetime 1000hrs

- Failed hour = $L \sim Geo(p = \frac{1}{1000})$
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- Failed ticks = $L_h \sim Geo(p = \frac{1}{1000} \times h)$
- $$P\{L_h h > c\} = P\left\{L_h > \frac{c}{h}\right\} = F_{L_h}^c\left(\left\lceil \frac{c}{h} \right\rceil\right) = (1 - p)^{\left\lceil \frac{c}{h} \right\rceil}$$
$$= (1 - \lambda h)^{\left\lceil \frac{c}{h} \right\rceil} = \lim_{h \rightarrow 0} -e^{-\lambda c} = F_T^c(c)$$

Slido – Waiting for the bus

Say a bus comes to the stop every 10 minutes in average

- Expected waiting time follows $T \sim \text{Exp}(\lambda = \frac{1}{10})$
- Alice knows the time last bus leave at $t = 0$, what's her best strategy to arrive t^* at the stop minimizing her waiting time T' ?

(A) $t^* = 0$

(B) $t^* = 5$

(C) $t^* = 10$

(D) Doesn't matter



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Poisson Process

Motivation

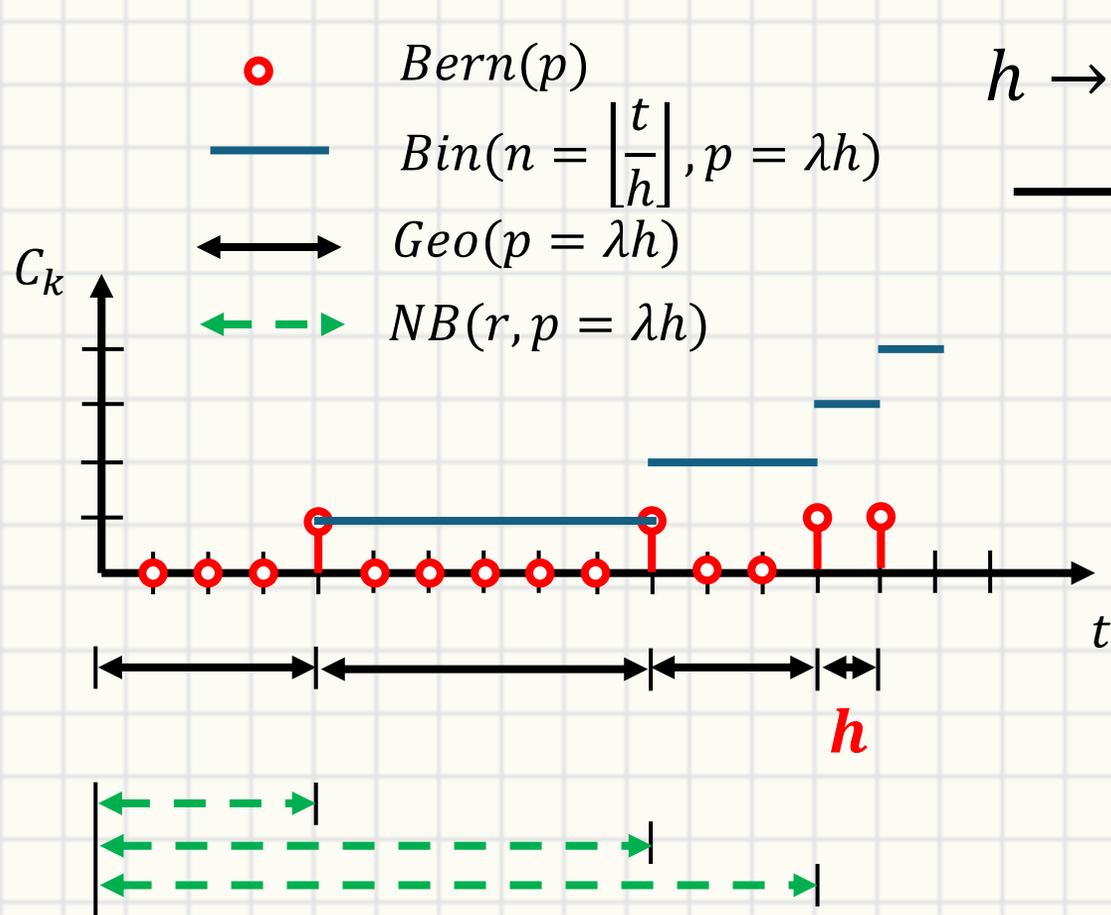
- Model the process of customer coming inside a coffee shop
- The process of incoming calls for customer service
 - If a customer (call) only comes at the $t = n \cdot \Delta t$ of every Δt minutes -> Bernoulli process
 - But what if we want to model the time
 - Define a small h between Bernoulli trials

Bernoulli Process

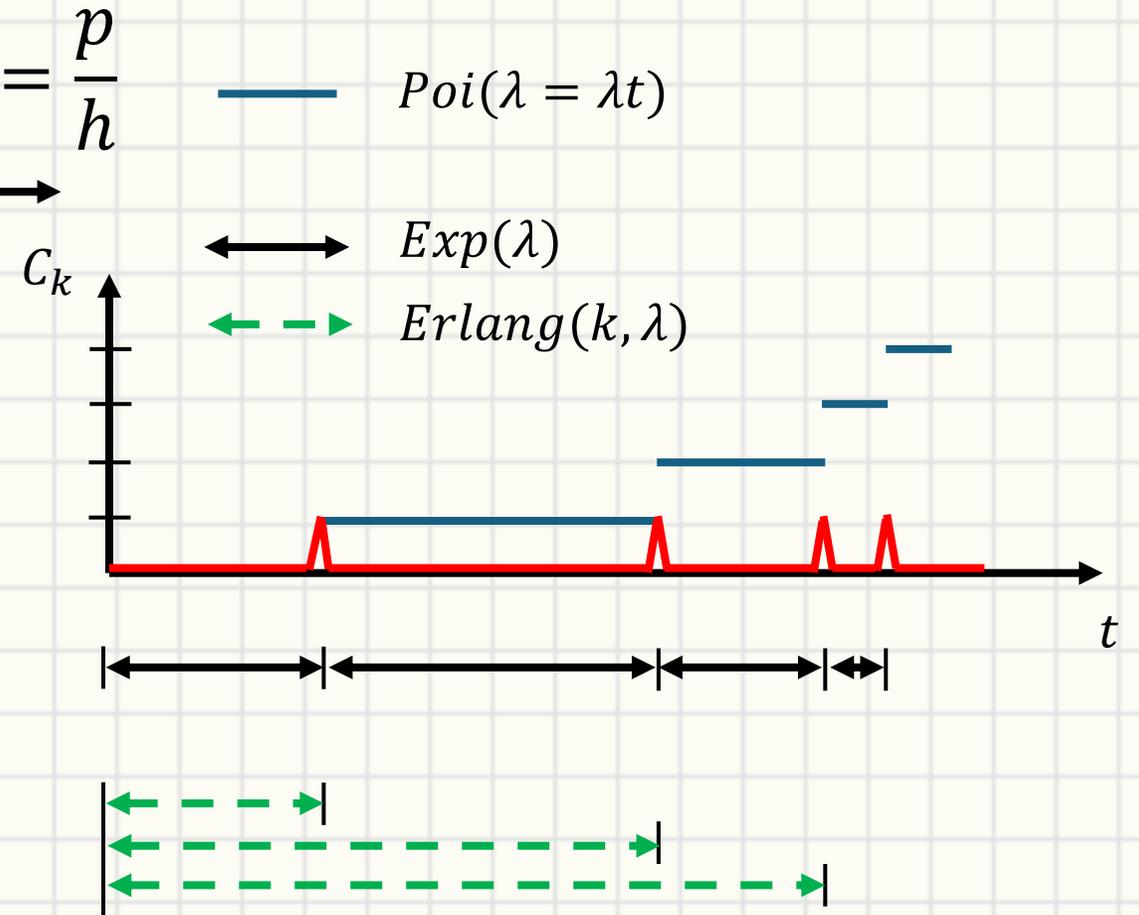
$$h \rightarrow 0, \lambda = \frac{p}{h}$$

Poisson Process

- Assume each trial takes h time to complete



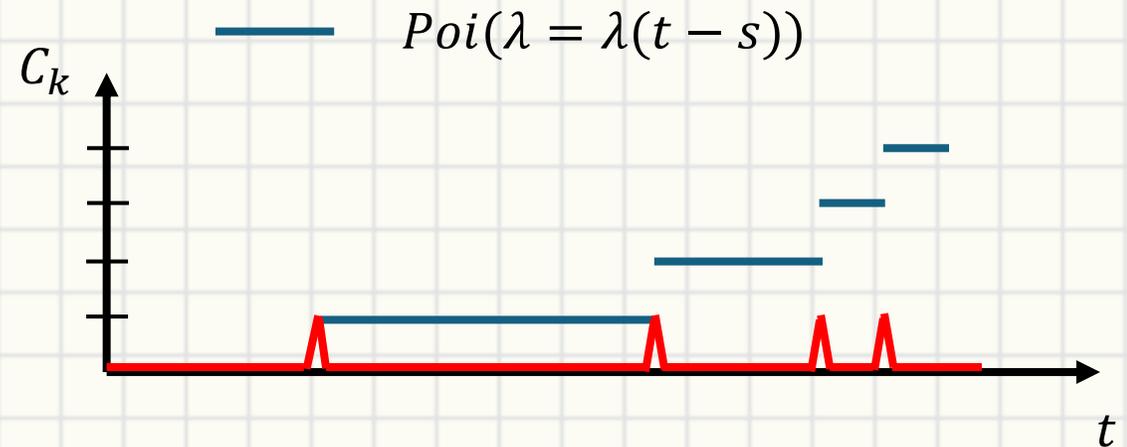
$$h \rightarrow 0, \lambda = \frac{p}{h}$$



Definition

A Poisson process with rate λ is a random counting process $N = (N_t: t \geq 0)$ s.t.

- $N_t - N_s$ follows Poisson distribution $Poi(\lambda = \lambda(t - s))$
- For $0 \leq t_1 \leq t_2 \dots \leq t_k$, $N_{t_k} - N_{t_{k-1}}$ the increments are independent with each other



Example

Consider a Poisson process with rate $\lambda > 0$ in time interval $[0, T]$

- X is the total number of count during $[0, T]$
- X_1 is the count during $[0, \tau]$, $0 < \tau < T$
- X_2 is the count during $[\tau, T]$

- Solve $P\{X = n\}$, $P\{X_1 = i\}$ and $P\{X_2 = j\}$

Example

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- Let $n = i + j$
- Solve $P(X = n | X_1 = i)$

Example

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- Let $n = i + j$
- Solve $P(X_1 = i | X = n)$

Example

Calls arrive to a support center at rate $\lambda = 2$ calls per minute.

Let N_t denotes the number of calls until time t (mins)

- $N_t \sim$
- $P_{N_t}(k) =$
- $P\{ \text{No calls arrive in the first 3.5 minutes} \}$
- $P\{ \text{The third call arrives after time } t = 3. \}$

Erlang Distribution

Definition

Let T_r denotes the time of r^{th} count of a Poisson process

- $T_r = \sum_{i=1}^r U_i, U_i \sim Exp(\lambda)$
- $F_{T_r}^C(t) = P\{T_r > t\}$: “At most $r - 1$ count by time t ”
- $F_{T_r}^C(t) =$
- $f_{T_r}(t) = -\frac{dF_{T_r}(t)}{dt} =$