

# Probability and Statistics

## 4 – Continuous Random Variables

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HV (due next lecture on friday) : Exc. 6, Q1-3

# Exponential Distributions

## Definition (4.39)

A random variable  $X$  has a *exponential distribution*,  $X \sim \exp(\lambda)$ , for some parameter  $\lambda > 0$ , if it has a pdf defined by:

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases}$$

$$\left( \text{Note: } \int_0^{\infty} \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^{\infty} = 1 \right)$$

# Exponential Distributions

## Lemma (4.40)

If  $X \sim \exp(\lambda)$ , then:

$$(i) F_X(x) = \begin{cases} 1 - e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

$$(ii) E(X^n) = \frac{n!}{\lambda^n} \quad \text{for } n \in \mathbb{N}$$

$$(iii) \underline{E(X) = \frac{1}{\lambda}}, \quad \text{Var}(X) = \frac{1}{\lambda^2}, \quad \text{skew}(X) = 2, \quad \text{kurt}(X) = 9$$

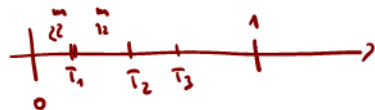
$$(iv) \phi_X(t) = \frac{\lambda}{\lambda - t} \quad \text{for } t < \lambda$$

$$(v) F_X^{-1}(p) = -\frac{\ln(1-p)}{\lambda} \quad \text{for } p \in (0, 1)$$

(vi) first quartile, median and third quartile are:

$$\ln(4/3)/\lambda \approx 0.288/\lambda, \quad \ln(2)/\lambda \approx 0.693/\lambda, \quad \ln(4)/\lambda \approx 1.386/\lambda$$

$m = E(X)$ ,  $\lambda := \frac{1}{m}$  [success] rate  
 $t = x$  (time)



# Memorylessness of Exponential Distributions

## Theorem (4.42)

*X has an exponential distribution, if and only if X has the memoryless property:*

$$\Pr(X > t + h \mid X > t) = \Pr(X > h) \quad \text{for all } t, h \geq 0$$

$$\Pr(T \leq t + \Delta t \mid T > t) = \Pr(T \leq \Delta t) = F_T(\Delta t)$$

# Exponential Distributions

## Theorem (4.44)

Let  $T$  be a continuous random variable with a continuous pdf  $f_T$  with constant success rate  $\lambda$ , i. e. for all  $t > 0$ , we have:

$$\lambda = \lim_{\Delta t \rightarrow 0} \frac{\Pr(T \leq t + \Delta t \mid T > t)}{\Delta t}$$

Then  $T \sim \exp(\lambda)$ .

|| (4.42)

Proof: Assume  $T \sim \exp(\lambda)$

$$\begin{aligned} \lim_{\Delta t \rightarrow 0} \frac{F_T(\Delta t)}{\Delta t} &= \lim_{\Delta t \rightarrow 0} \frac{1 - e^{-\lambda \Delta t}}{\Delta t} = \frac{d}{dt} (1 - e^{-\lambda t}) \Big|_{t=0} \\ &= \lambda e^{-\lambda t} \Big|_{t=0} \\ &= \lambda \end{aligned}$$

# Exponential Distributions

$$\lambda = \lim_{\Delta t \rightarrow 0} \frac{\Pr(T \leq t + \Delta t \mid T > t)}{\Delta t} = \lim_{\Delta t \rightarrow 0} \frac{\Pr(t \leq T \leq t + \Delta t)}{\Pr(T > t) \cdot \Delta t}$$

$$= \lim_{\Delta t \rightarrow 0} \frac{\int_t^{t+\Delta t} f_T(x) dx}{(1 - F_T(t)) \cdot \Delta t} = \frac{1}{1 - F_T(t)} f_T(t)$$

$$\Rightarrow f_T(t) = \lambda (1 - F_T(t)) \quad \Rightarrow \quad \frac{d}{dt} f_T(t) = -\lambda f_T(t) \quad \text{differential equation for } f_T(t)$$

$$\Rightarrow \left. \begin{array}{l} f_T(t) = c \cdot e^{-\lambda t} \\ \int_{-\infty}^{\infty} f_T(t) dt = 1 \end{array} \right\} \Rightarrow c = \lambda, \quad f_T \text{ is the p.d.f. of } \exp(\lambda) - \text{dist.}$$

# Minimum of independent $X_i \sim \exp(\lambda_i)$

## Theorem (4.41)

Let  $X_1, \dots, X_n$  be independent random variables with  $X_i \sim \exp(\lambda_i)$  for  $i = 1, \dots, n$ . Then

$$X = \min\{X_1, \dots, X_n\}$$

is a random variable with  $X \sim \exp(\lambda_1 + \dots + \lambda_n)$ .

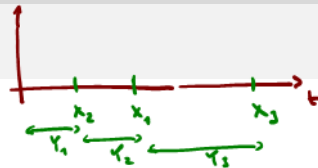
$$\begin{aligned} F_X(x) &= P_r(X \leq x) = 1 - P_r(X > x) = 1 - P_r(X_1 > x, X_2 > x, \dots, X_n > x) \\ &= 1 - P_r(X_1 > x) \cdot P_r(X_2 > x) \cdot \dots \cdot P_r(X_n > x) \\ &= 1 - e^{-\lambda_1 x} \cdot e^{-\lambda_2 x} \cdot \dots \cdot e^{-\lambda_n x} = 1 - e^{-(\lambda_1 + \dots + \lambda_n)x} \end{aligned}$$

$$\begin{aligned} P_r(X_i > x) &= 1 - F_{X_i}(x) \\ &= 1 - (1 - e^{-\lambda_i x}) = e^{-\lambda_i x} \end{aligned}$$

↑  
cdf of  $\exp(\lambda_1 + \dots + \lambda_n)$ -dist.  
□

# Exponential Distributions

$n=3$



$i_1 = 2$   
 $i_2 = 1$   
 $i_3 = 3$

## Theorem (4.43)

Let  $\lambda > 0$  and let  $X_1, \dots, X_n$  be independent random variables with  $X_i \sim \exp(\lambda)$  for  $i = 1, \dots, n$ . Furthermore, let  $X_{i_1}, \dots, X_{i_n}$  be a reordering of  $X_1, \dots, X_n$ , such that:

$$X_{i_1} \leq \dots \leq X_{i_n}$$

Then

$$Y_k := X_{i_k} - X_{i_{k-1}} \quad (X_{i_0} := 0)$$

is random variable with  $Y_k \sim \exp((n+1-k)\lambda)$ .

$$[Y_n \sim \exp(\lambda)]$$

$k=1$ :  $Y_1 \sim \exp(n\lambda)$ : true by (4.41)

$$Y_2: X_{i_2}, \dots, X_{i_n} \quad \Pr(X_{i_j} > X_{i_1} + t) = e^{-\lambda t}$$

because of the memoryless property

$\rightarrow$  Result follows by induction



# Reliability theory

## Definition (4.45)

Let  $T$  be a random variable, describing the lifetime or time-to-failure of a device or system and let  $f_T$  and  $F_T$  be its density functions.

(i) The reliability function is defined by:

$$R(t) := \Pr(T > t) = 1 - F_T(t)$$

(ii) The mean time to failure (MTTF) is defined to be the expected lifetime:

$$\text{MTTF} := E(T)$$

(iii) The failure rate is defined to be

$$\lambda(t) := \lim_{\Delta t \rightarrow 0} \frac{\Pr(T \leq t + \Delta t \mid T > t)}{\Delta t}$$



# Reliability theory

Lemma (4.46)

$$\text{MTTF} = \int_0^{\infty} R(t) dt$$

$$R(t) = 1 - \bar{F}_T(t)$$

$$\text{MTTF} = E(T) = \int_0^{\infty} t \cdot f_T(t) dt = -t \cdot \overset{R(t)}{(1 - \bar{F}_T(t))} \Big|_0^{\infty} + \int_0^{\infty} 1 \cdot R(t) dt$$

$\downarrow$   $\downarrow t \rightarrow \infty$   
 $\infty \cdot 0 = 0$

add. assumption :

# Reliability theory

Lemma (4.47)

(i)

$$\lambda(t) \stackrel{\checkmark}{=} \frac{f_T(t)}{\underbrace{1 - F_T(t)}_{R(t)}} \stackrel{\uparrow \checkmark}{=} \frac{f_T(t)}{\int_t^\infty f_T(\tau) d\tau} \stackrel{\checkmark}{=} -\frac{R'(t)}{R(t)}$$

(ii)

$$F_T(t) = 1 - e^{-\int_0^t \lambda(\tau) d\tau}, \quad f_T(t) = \lambda(t) e^{-\int_0^t \lambda(\tau) d\tau}$$

(iii) If  $\lambda(t) = \lambda$  is constant, then  $T \sim \exp(\lambda)$ .

$$(i) \quad \lambda(t) = \lim_{\Delta t \rightarrow 0} \frac{\Pr(t < T \leq t + \Delta t)}{\Pr(T > t) \cdot \Delta t} = \frac{1}{\Pr(T > t)} \cdot \frac{F_T(t + \Delta t) - F_T(t)}{\Delta t} = \frac{f_T(t)}{\Pr(T > t)} = \frac{f_T(t)}{1 - F_T(t)}$$

(ii), (iii): HW