

# Chapter 3. Continuous Distribution

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Math 3215 Summer 2023

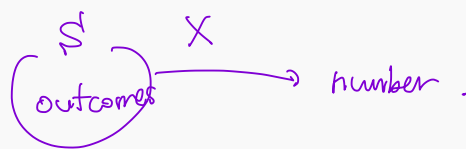
Georgia Institute of Technology

# **Section 1.**

## **Random Variables of the Continuous Type**

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## Continuous Random Variables

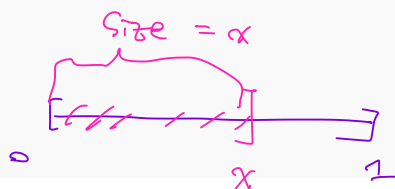


Let the random variable  $X$  denote the outcome when a point is selected at random from an interval  $[0, 1]$ .

If the experiment is performed in a fair manner, it is reasonable to assume that the probability that the point is selected from an interval  $[\frac{1}{3}, \frac{1}{2}]$  is

The cdf of  $X$  is

$$F(x) = x$$



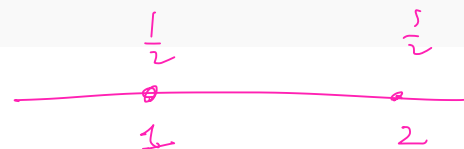
$$P = \text{size of } [\frac{1}{3}, \frac{1}{2}] = \frac{1}{2} - \frac{1}{3} = \frac{1}{6}$$

A horizontal line segment from 0 to 1. A bracket above the segment from  $\frac{1}{3}$  to  $\frac{1}{2}$  is labeled 'P'. The segment from  $\frac{1}{3}$  to  $\frac{1}{2}$  is shaded with diagonal lines.

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

$$P(\frac{1}{6} \leq X \leq \frac{1}{3}) = \frac{1}{6}$$

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$$\frac{1}{2} = P(X = 1) \quad , \quad P(X = 2) = \frac{1}{2}$$

## Continuous Random Variables

### Definition

We say a random variable  $X$  on a sample space  $S$  is a **continuous random variable** if there exists a function  $f(x)$  such that

- $f(x) \geq 0$  for all  $x$ ,
- $\int_{S(X)} f(x) dx = 1$ , and
- For any interval  $(a, b) \subset \mathbb{R}$ ,

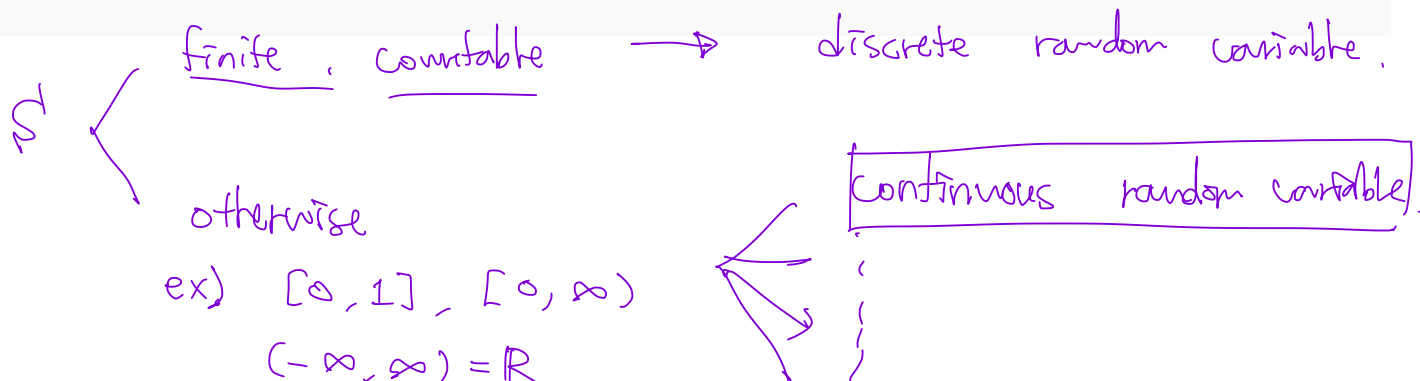
↑ similar to pmf.

$$\mathbb{P}(X \in (a, b)) = \mathbb{P}(a < X < b) = \int_a^b f(x) dx.$$

The function  $f(x)$  is called the **probability density function (pdf)** of  $X$ .

density

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$X$  is conti.

there exists a density  $f(x)$  need not to be conti.

$$P(a < X < b) = \int_a^b f(x) dx$$

## Continuous Random Variables

The cdf of  $X$  is  $F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$

The expectation (mean) of  $X$  is  $E[X] = \int_{-\infty}^{\infty} x f(x) dx$

The variance of  $X$  is  $Var(X) = E[(X - E[X])^2] = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx = \sigma^2$   
 $\mu = E[X]$

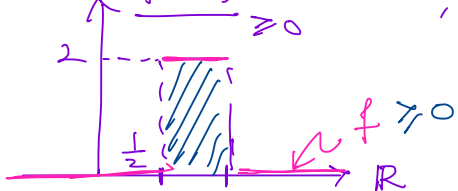
The standard deviation of  $X$  is  $std(X) = \sqrt{Var(X)} = \sigma = \sigma_X$

The moment generating function of  $X$  is

$$M(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx$$

For pmf  $\underbrace{f(k)}_{\geq 0}$ ,  $\sum f(k) = 1 \Rightarrow f(k) \leq 1$

For pdf  $\underbrace{f(x)}_{\geq 0}$ ,  $\int_{-\infty}^{\infty} f(x) dx = 1$  but  $f(x) > 1$  for some  $x$

Ex:   $\int f dx = 1$  but  $f(x) > 1$  for  $x \in [1/2, 1]$

## Continuous Random Variables

### Properties

The pmf of a discrete random variable is bounded by 1. But for pdf,  $f(x)$  can be greater than 1.

For cdf  $F$ , we have  $F'(x) = f(x)$  where  $F$  is differentiable at  $x$ .

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$$

$$\frac{d}{dx} F(x) = \frac{d}{dx} \int_{-\infty}^x f(t) dt = f(x)$$

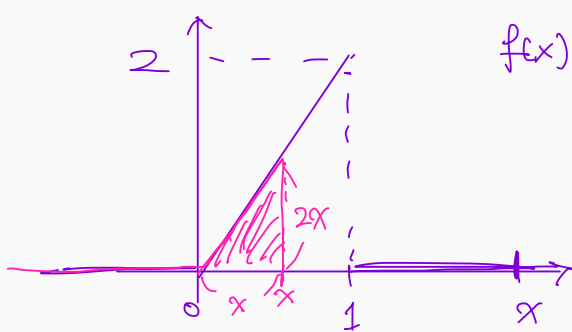
Fundamental Thm of Calculus.

## Continuous Random Variables

### Example

Let  $X$  be a continuous random variable with a pdf  $f(x) = 2x$  for  $0 < x < 1$ .

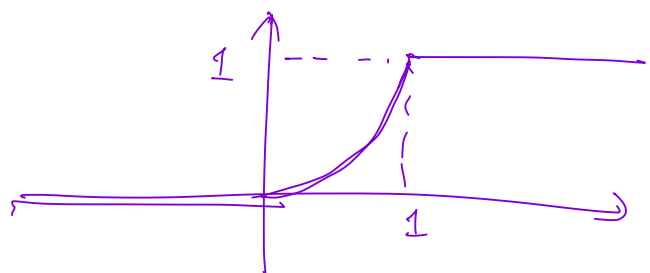
Find the cdf and the expectation.



$$f(x) = \begin{cases} 2x & \text{if } 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt = \begin{cases} 0 & , x \leq 0 \\ \frac{1}{2} \cdot x \cdot (2x) = x^2 & , 0 < x < 1 \\ 1 & , x \geq 1 \end{cases}$$

$$F(x) = \begin{cases} 0 & x \leq 0 \\ x^2 & 0 < x < 1 \\ 1 & x \geq 1 \end{cases}$$



## Continuous Random Variables

### Example

Let  $X$  be a continuous random variable with a pdf  $\underbrace{f}_g(x) = 2x$  for  $0 < x < 1$ .

Find the cdf and the expectation.

$$\begin{aligned} \mathbb{E}[X] &= \int_{-\infty}^{\infty} x \cdot \underline{f(x)} \, dx = \int_0^1 x \cdot 2x \, dx \\ &= \int_0^1 2 \boxed{x^2} \, dx = \left[ 2 \cdot \frac{1}{3} \cdot x^3 \right]_0^1 = \frac{2}{3}. \end{aligned}$$



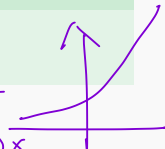
## Continuous Random Variables

### Example

Let  $X$  have the pdf  $f(x) = xe^{-x}$ . Find the mgf.

$$f(x) = \begin{cases} xe^{-x}, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

$$M(t) = \mathbb{E}[e^{tx}] = \int_0^{\infty} e^{tx} f(x) dx = \int_0^{\infty} \underbrace{(x - e^{-(t-1)x})}_{\substack{\text{IBP} \\ \downarrow \\ 1}} dx$$

$t > 1$  

$$\stackrel{\text{IBP}}{\uparrow} \left[ x \cdot \frac{1}{(t-1)} e^{(t-1)x} \right]_0^{\infty} - \int_0^{\infty} \frac{1}{(t-1)} e^{(t-1)x} dx$$

$t < 1$

$$\int \underbrace{u(x)}_{u'(x)} \cdot \underbrace{v'(x)}_{V(x)} dx = u(x) V(x) - \int u'(x) \cdot V(x) dx$$

$$= \left[ -\frac{1}{(t-1)^2} e^{(t-1)x} \right]_0^{\infty} = \frac{1}{(t-1)^2}, \quad t < 1.$$

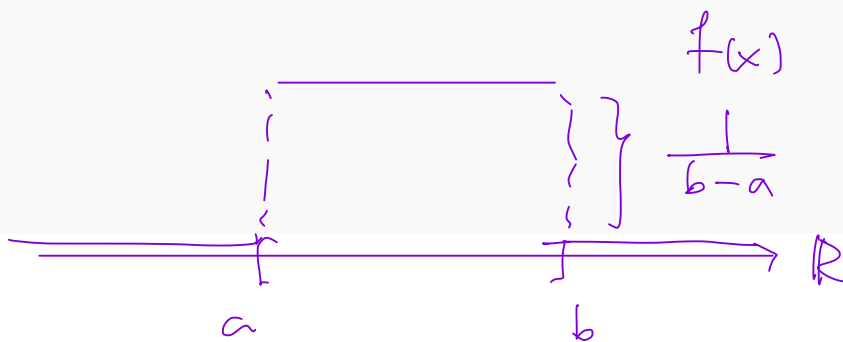
## Uniform Random Variables

### Definition

$X$  is a uniform random variable if its pdf is constant on its support.

If its support is  $[a, b]$ , then the pdf is

We denote by  $X \sim U(a, b)$ .



## Uniform Random Variables

$$f(x) = \begin{cases} \frac{1}{b-a} & , a \leq x \leq b \\ 0 & , \text{o.w.} \end{cases}$$

### Theorem

If  $X \sim U(a, b)$ , then

$$\mathbb{E}[X] = \frac{a+b}{2}$$

$$\text{Var}[X] = \frac{1}{12} (a-b)^2$$

$$M(t) = \text{Exercise.}$$

$$\begin{aligned} \mathbb{E}[X] &= \int_a^b \frac{1}{b-a} \cdot \underbrace{x} \, dx = \frac{1}{b-a} \cdot \left[ \frac{x^2}{2} \right]_a^b = \frac{b^2 - a^2}{2 \cdot (b-a)} \\ &= \frac{a+b}{2} \end{aligned}$$

$$b^2 - a^2 = (b-a) \cdot (b+a)$$

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$$\begin{aligned} \mathbb{E}[X^2] &= \int_a^b \frac{1}{b-a} x^2 \, dx = \frac{1}{(b-a)} \cdot \frac{1}{3} \cdot \underbrace{(b^3 - a^3)}_{(b-a) \cdot (a^2 + ab + b^2)} \\ &= \frac{1}{3} (a^2 + ab + b^2) \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= \frac{1}{3} (a^2 + ab + b^2) - \frac{1}{4} (a^2 + 2ab + b^2) \\ &= \frac{1}{12} (a^2 - 2ab + b^2) = \frac{(a-b)^2}{12} \end{aligned}$$

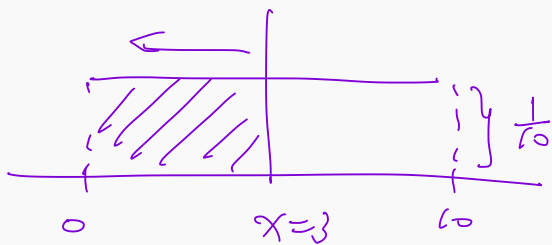
## Uniform Random Variables

$$X \sim \text{Unif}(0, 10)$$

### Example

If  $X$  is uniformly distributed over  $(0, 10)$ , calculate  $\mathbb{P}(X < 3)$ ,  $\mathbb{P}(X > 6)$ , and  $\mathbb{P}(3 < X < 8)$ .

$$\mathbb{P}(X < 3) = 3 \cdot \frac{1}{10} = 0.3$$



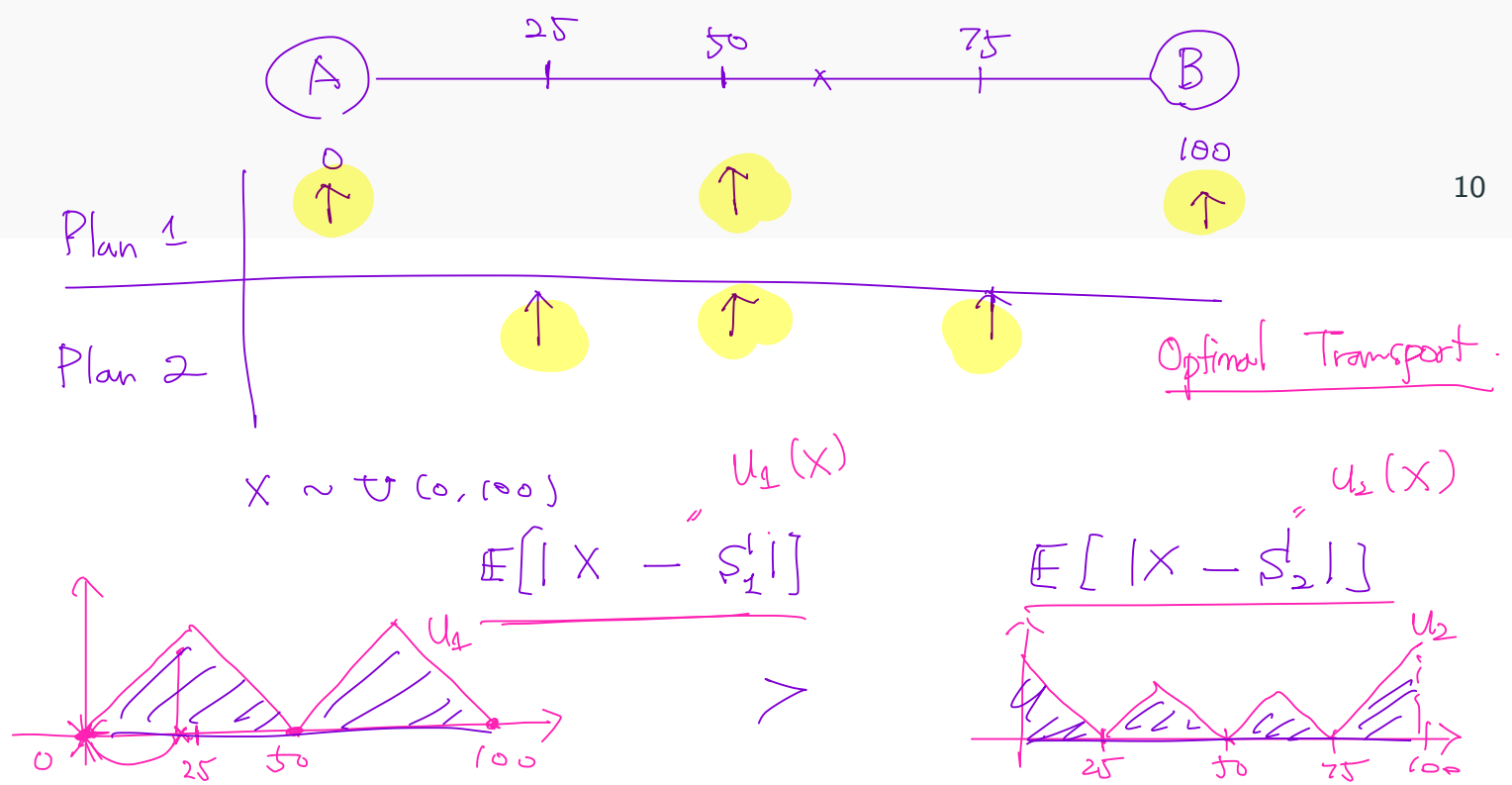
$$\mathbb{P}(X > 6) = \frac{4}{10}$$

$$\mathbb{P}(3 < X < 8) = \frac{5}{10}$$

## Uniform Random Variables

### Example

A bus travels between the two cities A and B, which are 100 miles apart. If the bus has a breakdown, the distance from the breakdown to city A has a  $U(0, 100)$  distribution. There are bus service stations in city A, in B, and in the center of the route between A and B. It is suggested that it would be more efficient to have the three stations located 25, 50, and 75 miles, respectively, from A. Do you agree? Why?

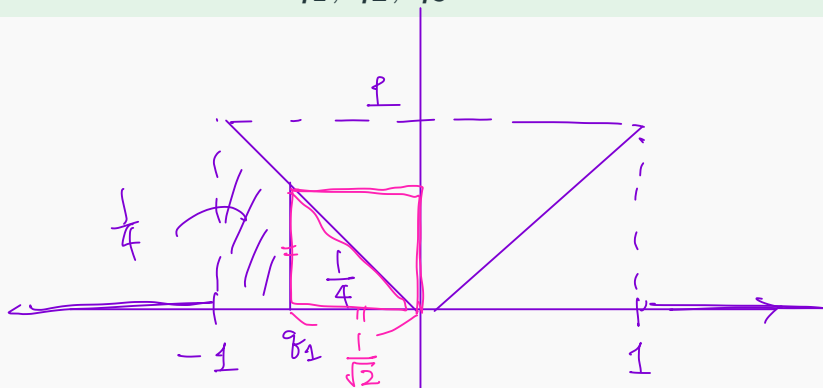




## Percentile

### Example

Let  $X$  be a continuous random variable with pdf  $f(x) = |x|$  for  $-1 < x < 1$ . Find  $q_1, q_2, q_3$ .

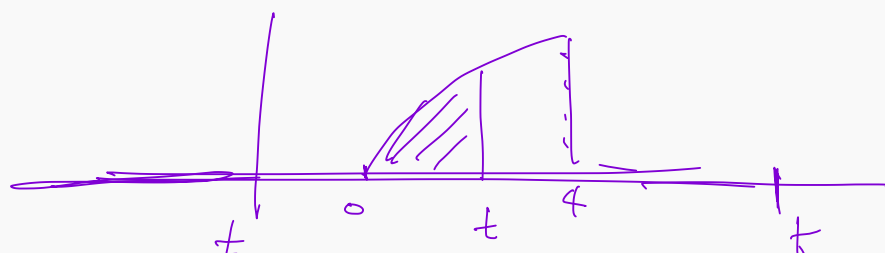


$$q_2 = 50^{\text{th}} \text{ percentile} = \text{median} = \pi_{0.5} = 0$$

$$q_1 = 25^{\text{th}} \text{ percentile} = 1^{\text{st}} \text{ quartile} = \pi_{0.25} = -\frac{1}{\sqrt{2}}$$

$$q_3 = \frac{1}{\sqrt{2}}$$

## Exercise



Let  $f(x) = c\sqrt{x}$  for  $0 \leq x \leq 4$  be the pdf of a random variable  $X$ .

Find  $c$ , the cdf of  $X$ , and  $\mathbb{E}[X]$ .

$$1 = \int_0^4 c \sqrt{x} \, dx = c \cdot \left[ \frac{2}{3} \cdot x^{\frac{3}{2}} \right]_0^4 = c \cdot \frac{2}{3} \cdot 8 \quad \therefore c = \frac{3}{16}.$$

$$F(t) = \mathbb{P}(X \leq t) = \int_0^t c \cdot \sqrt{x} \, dx = \left[ c \cdot \frac{2}{3} \cdot x^{\frac{3}{2}} \right]_0^t$$

$$\underline{0 \leq t \leq 4} \quad = \frac{3}{16} \cdot \frac{2}{3} \cdot t^{\frac{3}{2}} = \frac{1}{8} t^{\frac{3}{2}}$$

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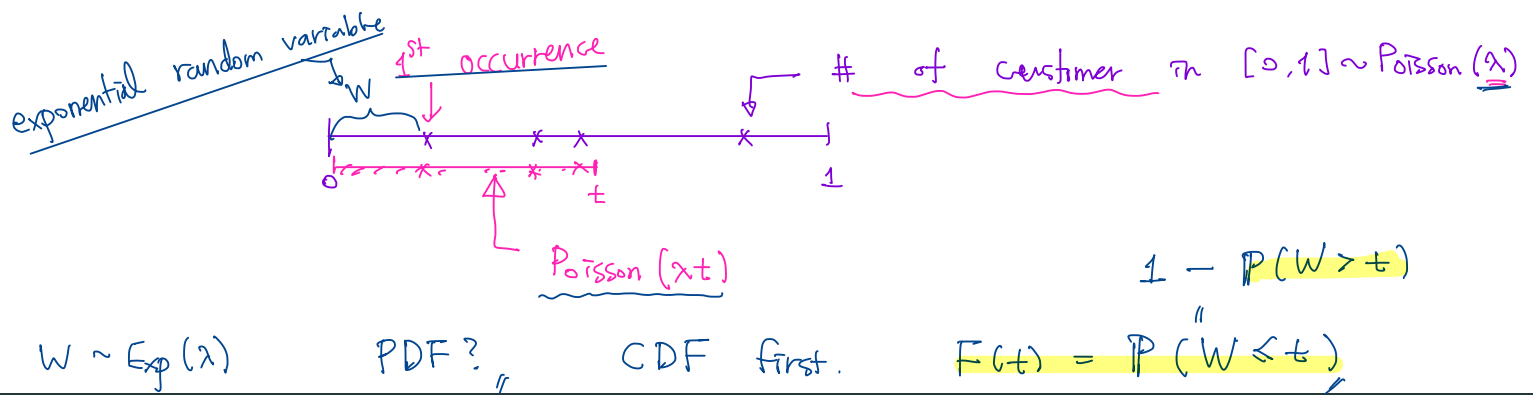
$$F(t) = \begin{cases} 0 & t < 0 \\ \frac{1}{8} t^{\frac{3}{2}} & 0 \leq t \leq 4 \\ 1 & t \geq 4 \end{cases}$$



## **Section 2.**

# **The Exponential, Gamma, and Chi-Square Distributions**

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## Exponential random variables

Consider a Poisson random variable  $X$  with parameter  $\lambda$ .

This represents the number of occurrences in a given interval, say  $[0, 1]$ .

If  $\lambda = 5$ , that means the expected number of occurrences in  $[0, 1]$  is 5.

Let  $W$  be the waiting time for the first occurrence. Then,

$$P(W > t) = P(\text{no occurrences in } [0, t]) = P(Y = 0)$$

for  $t > 0$ .

$$= e^{-\lambda t} \cdot \frac{(\lambda t)^0}{0!} = e^{-\lambda t}$$

$Y = \#$  of events in  $[0, t] \sim \text{Poisson}(\lambda t)$

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$$F(t) = 1 - e^{-\lambda t} \quad \text{PDF} = f(t) = \lambda e^{-\lambda t}, \quad t \geq 0.$$

$$F(t) = \int_{-\infty}^t f(s) ds$$

PDF

$$F'(t) = f(t)$$

$W =$  waiting time of 1st occurrence  
 $\sim \text{Exp}(\lambda)$



## Exponential random variables

### Definition

We say  $X$  is an exponential random variable with parameter  $\lambda$  (or mean  $\theta$  where  $\lambda = \frac{1}{\theta}$ ) if its pdf is

$$f(x) = \lambda e^{-\lambda x}$$

for  $x \geq 0$  and otherwise 0. Here,  $\lambda$  is the parameter and  $\theta$  is the mean.

$$f(t) = \lambda e^{-\lambda t}, \quad t \geq 0 \quad X \sim \text{Exp}(\lambda)$$

$$\begin{aligned} \mathbb{E}[X] &= \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_0^{\infty} \lambda x e^{-\lambda x} dx, \quad \begin{aligned} \lambda x &= t \\ \lambda dx &= dt \\ dx &= \frac{1}{\lambda} dt \end{aligned} \\ &= \frac{1}{\lambda} \int_0^{\infty} \underbrace{t}_{\substack{\uparrow \\ 1}} \cdot \underbrace{e^{-t}}_{\substack{\uparrow \\ -e^{-t}}} dt \quad \stackrel{\text{IBP}}{=} \frac{1}{\lambda} \left( \underbrace{\left[ -t e^{-t} \right]_0^{\infty}}_{=0} + \underbrace{\int_0^{\infty} e^{-t} dt}_{= \left[ -e^{-t} \right]_0^{\infty} = 1} \right) \\ &= \frac{1}{\lambda} = \theta. \end{aligned}$$

## Exponential random variables

### Theorem

Suppose that  $X$  is an exponential random variable with parameter  $\lambda = \frac{1}{\theta}$ .

$$\mathbb{E}[X] = \frac{1}{\lambda} = \theta$$

$$\text{Var}[X] = \frac{1}{\lambda^2} = \theta^2 = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$$

$$M(t) = \frac{\lambda}{\lambda - t} = \frac{1}{1 - \theta t}$$

$$M(t) = \mathbb{E}[e^{tx}] = \int_0^{\infty} e^{tx} \cdot \lambda e^{-\lambda x} dx$$

$$= \lambda \int_0^{\infty} e^{-\underbrace{(\lambda - t)}_{\geq 0} x} dx$$

$$\stackrel{\substack{\text{if } \lambda - t > 0 \\ \uparrow}}{=} \frac{\lambda}{\lambda - t} \quad \stackrel{\substack{\uparrow \\ \theta = \frac{1}{\lambda}}}{=} \frac{1}{1 - \theta t}.$$

## Exponential random variables

$$f(x) = \lambda e^{-\lambda x} = \frac{1}{20} \cdot e^{-\frac{x}{20}}$$

### Example

Let  $X$  have an exponential distribution with a mean  $\theta = 20$ .

$$\lambda = \frac{1}{\theta} = \frac{1}{20}$$

Find  $\mathbb{P}(X < 18)$ .

$$\begin{aligned} \mathbb{P}(X < 18) &= \int_{-\infty}^{18} f(x) dx = \int_0^{18} \frac{1}{20} e^{-\frac{x}{20}} dx \\ &= \left[ -e^{-\frac{x}{20}} \right]_0^{18} = 1 - e^{-\frac{18}{20}} \end{aligned}$$

$$\textcircled{1} \quad F(x) = \mathbb{P}(X \leq x) = \mathbb{P}(X < x) = 1 - e^{-\lambda x}$$

$$\textcircled{2} \quad \mathbb{P}(X > x) = e^{-\lambda x}$$

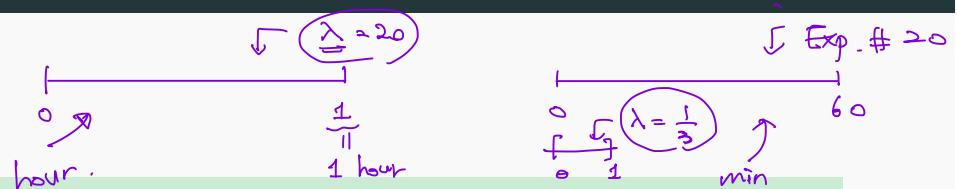
$$\mathbb{P}(X > t+s \mid X > t) = \frac{\mathbb{P}(X > t+s)}{\mathbb{P}(X > t)} = \frac{e^{-\lambda(t+s)}}{e^{-\lambda t}}$$

$$= e^{-\lambda s} = \mathbb{P}(X > s)$$

memoryless property.

$$P(W > t) = e^{-\frac{1}{3} \cdot 5} = e^{-\frac{5}{3}}$$

## Exponential random variables



### Example

Customers arrive in a certain shop according to an approximate Poisson process at a mean rate of 20 per hour.

What is the probability that the shopkeeper will have to wait more than five minutes for the arrival of the first customer?

$X = \# \text{ of customer in 1 hour} \sim \text{Poisson}(\lambda)$ ,  $\lambda = 20$

$\Downarrow$

$W = \text{waiting time} \sim \text{Exp}(20)$

$$P(W > \frac{1}{12}) = e^{-20 \cdot \frac{1}{12}} = e^{-\frac{5}{3}}$$

$$P(W > t) = e^{-\lambda t}$$

Binomial = # of success in  $n$  trials      Poisson = # of customers in  $[0, t]$

Geometric = # of trials until 1<sup>st</sup> success      Exp. = Waiting time until 1<sup>st</sup> customer

Neg. Bin. = # of trials until  $r^{\text{th}}$  success      Gamma = Waiting time until  $r^{\text{th}}$  customers

## Gamma random variables

Consider a Poisson random variable  $X$  with  $\lambda$ .

Let  $W$  be the waiting time until  $\alpha$ -th occurrences, then its cdf is

$$F(t) = \mathbb{P}(W \leq t) = 1 - \mathbb{P}(W > t) = 1 - \sum_{k=0}^{\alpha-1} \frac{(\lambda t)^k e^{-\lambda t}}{k!}.$$

Thus, the pdf is

$$f(x) = \frac{\lambda (\lambda x)^{\alpha-1}}{(\alpha-1)!} e^{-\lambda x}.$$

differentiate  
in  $t$

This random variable is called a gamma random variable with  $\lambda$  and  $\alpha$  where  $\lambda = \frac{1}{\theta} > 0$ .

This can be extended to non-integer  $\alpha > 0$ .

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$W$  = waiting time until  $\alpha^{\text{th}}$  customers

# of customers in  $[0, t] \sim \text{Poisson}(\lambda t)$

$\gamma$

$$\mathbb{P}(W > t) = \mathbb{P}(\gamma \leq \alpha-1) = \sum_{k=0}^{\alpha-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!}$$

$$\alpha=3. \quad F(t) = 1 - \underbrace{e^{-\lambda t}} - \underbrace{\lambda t e^{-\lambda t}} - \frac{1}{2}(\lambda t)^2 e^{-\lambda t}$$

$$f(t) = F'(t) = \cancel{\lambda e^{-\lambda t}} - \cancel{\lambda e^{-\lambda t}} + \cancel{\lambda^2 t e^{-\lambda t}} - \cancel{\lambda^2 t e^{-\lambda t}} + \frac{1}{2} \cdot \lambda^3 \cdot t^2 \cdot e^{-\lambda t}$$

## Gamma functions

The gamma function is defined by

$$\Gamma(t) = \int_0^{\infty} y^{t-1} e^{-y} dy$$

$\Gamma(1) = \int_0^{\infty} y^{1-1} e^{-y} dy$   
 $= \int_0^{\infty} e^{-y} dy$   
 $= [-e^{-y}]_0^{\infty} = 1$   
 $\Gamma(2) = \int_0^{\infty} y^{2-1} e^{-y} dy$  (IBP)

for  $t > 0$ .

By integration by parts, we have

$$\begin{aligned}
 &= \left[ -y^{t-1} e^{-y} \right]_0^{\infty} + \int_0^{\infty} (t-1) y^{t-2} e^{-y} dy \\
 &= (t-1) \cdot \int_0^{\infty} y^{(t-1)-1} e^{-y} dy \\
 &= (t-1) \cdot \Gamma(t-1)
 \end{aligned}$$



Def  $X \sim \text{Gamma}(\lambda, \underline{\alpha})$  if

$$f(x) = \frac{\lambda \cdot (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} \cdot e^{-\lambda x} \quad \text{for } x \geq 0$$

## Gamma functions

$$\Gamma(t) = \Gamma(t-1) \cdot (t-1) \quad \Rightarrow \quad \Gamma(n) = (n-1)!$$

In particular,  $\Gamma(1) = 1$

$$\Gamma(2) = (2-1) \cdot \Gamma(2-1) = \Gamma(1) = 1$$

$$\Gamma(3) = (3-1) \cdot \Gamma(3-1) = 2 \cdot \Gamma(2) = 2$$

$$\Gamma(n) = (n-1) \cdot \Gamma(n-1) = (n-1) \cdot (n-2) \cdot \Gamma(n-2) = (n-1) \cdot (n-2) \cdots \Gamma(1)$$

$$= (n-1)!$$

for integers  $n$ .

$$X \sim \text{Gamma}(\lambda, \alpha)$$

$$f = \frac{\lambda \cdot (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} e^{-\lambda x}, \quad x \geq 0.$$

## Gamma random variables

$$\Gamma(t) = \int_0^{\infty} y^{t-1} e^{-y} dy \quad // \quad \underline{\Gamma(t) = (t-1)\Gamma(t-1)}$$

### Theorem

$$\mathbb{E}[X] = \frac{\alpha}{\lambda}$$

When computing, use definition of Gamma function.

$$\text{Var}[X] = \frac{\alpha}{\lambda^2}$$

$$M(t) = \frac{1}{(1-\theta t)^\alpha} \text{ for } t \leq \frac{1}{\theta}.$$

$$\theta = \frac{1}{\lambda}$$

## Gamma random variables

$$f(x) = \frac{(\lambda) \cdot ((\lambda x)^{2-1})}{\Gamma(2)} e^{-\lambda x} = \frac{1}{9} \cdot x e^{-\frac{x}{3}}$$

### Example

Suppose the number of customers per hour arriving at a shop follows a Poisson random variable with mean 20.

$\lambda$  = mean of Poisson.

That is, if a minute is our unit, then  $\lambda = \frac{1}{3}$ .

$\theta$  = mean of Exp.

What is the probability that the second customer arrives more than five minutes after the shop opens for the day?

$W$  = waiting time for 2<sup>nd</sup> customers  $\sim \text{Gamma}(\frac{1}{3}, 2)$

$$P(W > \underline{\underline{5}}) = \int_{\underline{\underline{5}}}^{\infty} \left( \frac{1}{9} x \right) e^{-\left(\frac{x}{3}\right)} dx = \frac{1}{3} \int_{\frac{5}{3}}^{\infty} y e^{-\frac{y}{3}} dy$$

$$\frac{x}{3} = y, \quad dx = 3 dy$$

$$= \int_{\frac{5}{3}}^{\infty} \underbrace{y}_{1} \underbrace{e^{-y}}_{-e^{-y}} dy = \left[ -y e^{-y} \right]_{\frac{5}{3}}^{\infty} + \int_{\frac{5}{3}}^{\infty} e^{-y} dy$$

$$= \frac{5}{3} e^{-\frac{5}{3}} + e^{-\frac{5}{3}} = \frac{8}{3} e^{-\frac{5}{3}}$$

## Chi-square distribution

Let  $X$  have a gamma distribution with  $\lambda = \frac{1}{2}$ ,  $\theta = 2$  and  $\alpha = r/2$ , where  $r$  is a positive integer.

The pdf of  $X$  is

$$f(x) = \frac{1}{\Gamma(\frac{r}{2})2^{\frac{r}{2}}} x^{\frac{r}{2}-1} e^{-\frac{x}{2}}$$

← many application in stat.

for  $x > 0$ .

We say that  $X$  has a  $\chi^2$  chi-square distribution with  $r$  degrees of freedom and we use the notation  $X \sim \chi^2(r)$ .

## Exercise

Let  $X$  have an exponential distribution with  $\lambda = \frac{1}{\theta}$  mean  $\theta$ .

Compute  $\mathbb{P}(X > 15 | X > 10)$  and  $\mathbb{P}(X > 5)$ .

$$\textcircled{1} \quad \mathbb{P}(X > t) = e^{-\lambda t} \quad \Rightarrow \quad \mathbb{P}(X > 5) = e^{-\lambda \cdot 5} = e^{-5/\theta}$$

$$\textcircled{2} \quad \mathbb{P}(X > t+s | X > t) = \mathbb{P}(X > s)$$

$$\mathbb{P}(X > 10+5 | X > 10) = \mathbb{P}(X > 5) = e^{-5/\theta}.$$

## Section 3.

# The Normal Distribution

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Central Limit Theorem.

## Gaussian random variables

$$X \sim \text{Exp}(\lambda) \quad f(x) = \lambda e^{-\lambda x}, \quad x \geq 0$$

### Definition

We say  $X$  is a Gaussian random variable or has a normal distribution if its pdf is given by

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right), \quad \forall x \in \mathbb{R}$$

Here  $\mu$  is the mean and  $\sigma$  is the standard deviation. We use the notation  $X \sim N(\mu, \sigma^2)$ .

↑ mean    ↑ variance

If  $\mu = 0$ ,  $\sigma^2 = 1$ .

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$X \sim N(0, 1)$  the standard normal (Gaussian).

$$X \sim N(\mu, \sigma^2) \quad f(x) = \frac{1}{\sqrt{2\pi} \cdot \sigma} e^{-\frac{|x-\mu|^2}{2\sigma^2}}$$

## Gaussian random variables

### Theorem

$$\int_{\mathbb{R}} f(x) dx = 1$$

$$\mathbb{E}[X] = \mu$$

$$\text{Var}[X] = \sigma^2$$

$$M(t) = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right)$$

For  $\mu=0$  ,  $\sigma^2=1$  ,

$$\int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{|x|^2}{2}} dx = 1 \quad (\text{Gaussian Integral})$$

$$Z \sim N(0, 1)$$



$$\textcircled{1} \quad Z \sim N(0,1) \Rightarrow X = \sigma Z + \mu \sim N(\mu, \sigma^2)$$

$$\textcircled{2} \quad X \sim N(\mu, \sigma^2) \Rightarrow Z = \frac{X - \mu}{\sigma} \sim N(0,1)$$

## Standard normal distribution

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

In particular, if  $\mu = 0$  and  $\sigma^2 = 1$ , then  $Z \sim N(0,1)$  is called the standard normal random variable.

### Example

Let  $Z \sim N(0,1)$ .

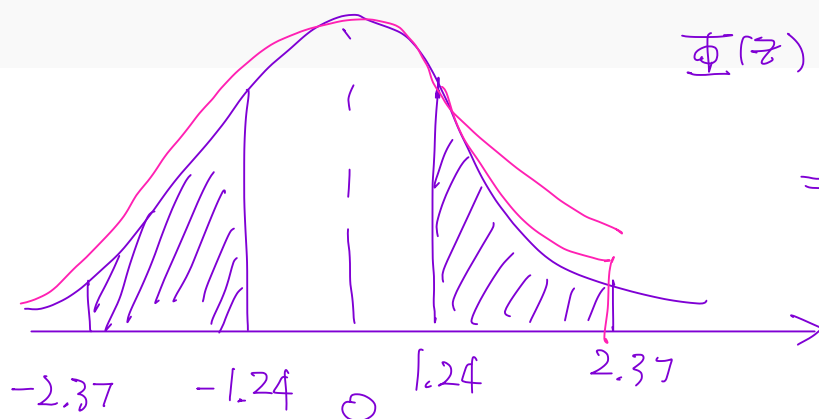
Find  $\mathbb{P}(Z \leq 1.24)$ ,  $\mathbb{P}(1.24 \leq Z \leq 2.37)$ , and  $\mathbb{P}(-2.37 \leq Z \leq -1.24)$ .

$$\mathbb{P}(Z \leq 1.24) = \underline{\hspace{2cm}}$$

$$\mathbb{P}(-2.37 \leq Z \leq -1.24) = \Phi(2.37) - \Phi(1.24)$$

$$\Phi(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$= \mathbb{P}(Z \leq z)$$



## Standard normal distribution

### Theorem

If  $X \sim N(\mu, \sigma^2)$ , then  $Z = \frac{X - \mu}{\sigma}$  is the standard normal.

## Standard normal distribution

$$\begin{aligned} X &= 4Z + 3 \\ Z &= \frac{X - 3}{4} \sim N(0, 1) \end{aligned}$$

### Example

Let  $X \sim N(\underline{3}, 16)$ .  $\mu = \underline{3}$ ,  $\sigma^2 = 16$ ,  $\sigma = \underline{4}$ .

Find  $\mathbb{P}(4 \leq X \leq 8)$ ,  $\mathbb{P}(0 \leq X \leq 5)$ , and  $\mathbb{P}(-2 \leq X \leq 1)$ .

$$\begin{aligned} \mathbb{P}(4 \leq X \leq 8) &\stackrel{\substack{\uparrow \\ \text{in terms of } Z}}{=} \mathbb{P}(\underline{4} \leq 4Z + \underline{3} \leq \underline{8}) \\ &= \mathbb{P}(1 \leq 4Z \leq 5) \\ &= \mathbb{P}(0.25 \leq Z \leq 1.25) \\ &= \Phi(1.25) - \Phi(0.25) \end{aligned}$$

## Standard normal distribution

### Example

Let  $X \sim N(25, 36)$ .  $\mu = 25$ ,  $\sigma^2 = 36$ ,  $\sigma = 6$

Find a constant  $c$  such that  $\mathbb{P}(|X - 25| \leq c) = 0.9544$ .

$c = 12$

$$Z \sim N(0, 1)$$

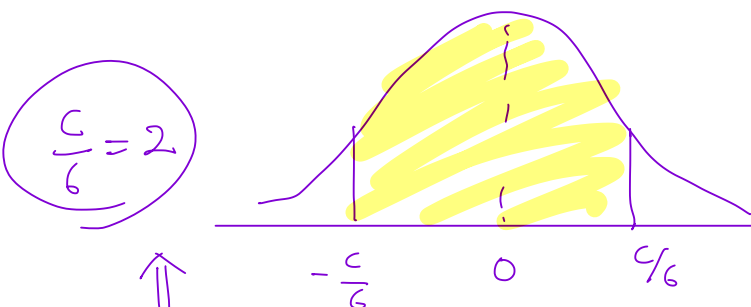
$$X = \sigma Z + \mu = 6Z + 25$$

$$\mathbb{P}(|X - 25| \leq c)$$

$$= \mathbb{P}(|6Z| \leq c) = \mathbb{P}(|Z| \leq c/6)$$

$$= \mathbb{P}\left(-\frac{c}{6} \leq Z \leq \frac{c}{6}\right) \stackrel{\uparrow}{=} \Phi\left(\frac{c}{6}\right) - \Phi\left(-\frac{c}{6}\right)$$

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$$\Phi(z) = \mathbb{P}(Z \leq z)$$

$$= 2 \cdot \mathbb{P}\left(0 \leq Z \leq \frac{c}{6}\right)$$

$$= 2 \cdot \left(\mathbb{P}\left(Z \leq \frac{c}{6}\right) - \mathbb{P}(Z \leq 0)\right)$$

$$\Phi\left(\frac{c}{6}\right) = \frac{1.9544}{2} = 0.9772 = \Phi\left(\frac{c}{6}\right) - \underbrace{\Phi(0)}_{1/2}$$

$$0.9544 = 2 \cdot \Phi\left(\frac{c}{6}\right) - 1$$

## Standard normal distribution

### Theorem

If  $Z$  is the standard normal, then  $Z^2$  is  $\chi^2(1)$ .

## **Section 4.**

### **Additional Models**

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## Weibull distribution

Recall the postulates of an approximate Poisson:

- The numbers of occurrences in nonoverlapping subintervals are independent.
- The probability of two or more occurrences in a sufficiently short subinterval is essentially zero.
- The probability of exactly one occurrence in a sufficiently short subinterval of length  $h$  is approximately  $\lambda h$ .

## Weibull distribution

One can think the event occurrence as a failure and so  $\lambda$  can be understood as the failure rate.

Poisson distribution and its waiting time (exponential distribution) has a constant failure rate.

Sometimes, it is more natural to choose  $\lambda$  as a function of  $t$  in the last assumption.

Then the waiting time  $W$  for the first occurrence satisfies

$$\mathbb{P}(W > t) = \exp \left( - \int_0^t \lambda(w) dw \right).$$



## Weibull distribution

### Definition

If  $\lambda(t) = \alpha \frac{t^{\alpha-1}}{\beta^\alpha}$ , then the waiting time  $W$  for the first occurrence has the density

$$g(t) = \lambda(t) \exp \left( - \int_0^t \lambda(w) dw \right) = \alpha \frac{t^{\alpha-1}}{\beta^\alpha} \exp \left( - \left( \frac{t}{\beta} \right)^\alpha \right).$$

$W$  is called the Weibull random variable.

## Weibull distribution

### Example

If  $\lambda(t) = 2t$ , then the waiting time  $W$  has the density

and it is a Weibull random variable with  $\alpha =$  and  $\beta =$  .

If  $W_1, W_2$  are independent Weibull with  $\alpha$  and  $\beta$  above, is the minimum of  $W_1, W_2$  Weibull?

## Weibull distribution

### Theorem

The mean of  $W$  is  $\mu = \beta\Gamma(1 + \frac{1}{\alpha})$ .

The variance is  $\sigma^2 = \beta^2 (\Gamma(1 + \frac{2}{\alpha}) - \Gamma(1 + \frac{1}{\alpha})^2)$ .

## Mixed type random variables

### Example

Suppose  $X$  has a cdf

$$F(x) = \begin{cases} 0, & x < 0 \\ \frac{x^2}{4}, & 0 \leq x < 1 \\ \frac{1}{2}, & 1 \leq x < 2 \\ \frac{x}{3}, & 2 \leq x < 3 \\ 1, & x \geq 3. \end{cases}$$

Find  $\mathbb{P}(0 < X < 1)$ ,  $\mathbb{P}(0 < X \leq 1)$ , and  $\mathbb{P}(X = 1)$ .

## Mixed type random variables

### Example

Consider the following game: A fair coin is tossed.

If the outcome is heads, the player receives \$2.

If the outcome is tails, the player spins a balanced spinner that has a scale from 0 to 1.

The player then receives that fraction of a dollar associated with the point selected by the spinner.

Let  $X$  be the amount received. Draw the graph of the cdf  $F(x)$ .

## Exercise

The cdf of  $X$  is given by

$$F(x) = \begin{cases} 0, & x < -1 \\ \frac{x}{4} + \frac{1}{2}, & -1 \leq x < 1 \\ 1, & x \geq 1. \end{cases}$$

Find  $\mathbb{P}(X < 0)$ ,  $\mathbb{P}(X < -1)$ , and  $\mathbb{P}(-1 \leq X < \frac{1}{2})$ .