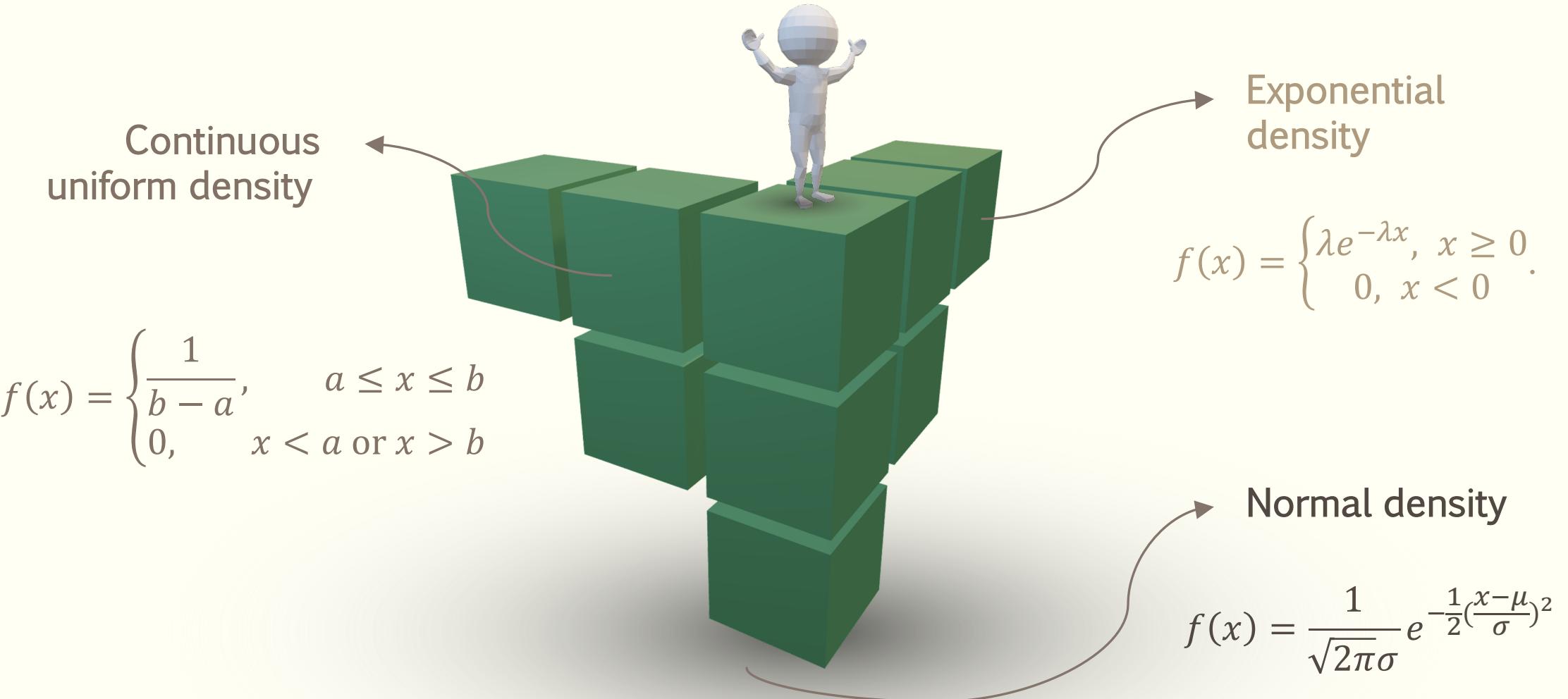


# IMPORTANT DENSITIES

- Continuous uniform density
- Exponential density
- Normal density

# Important Densities

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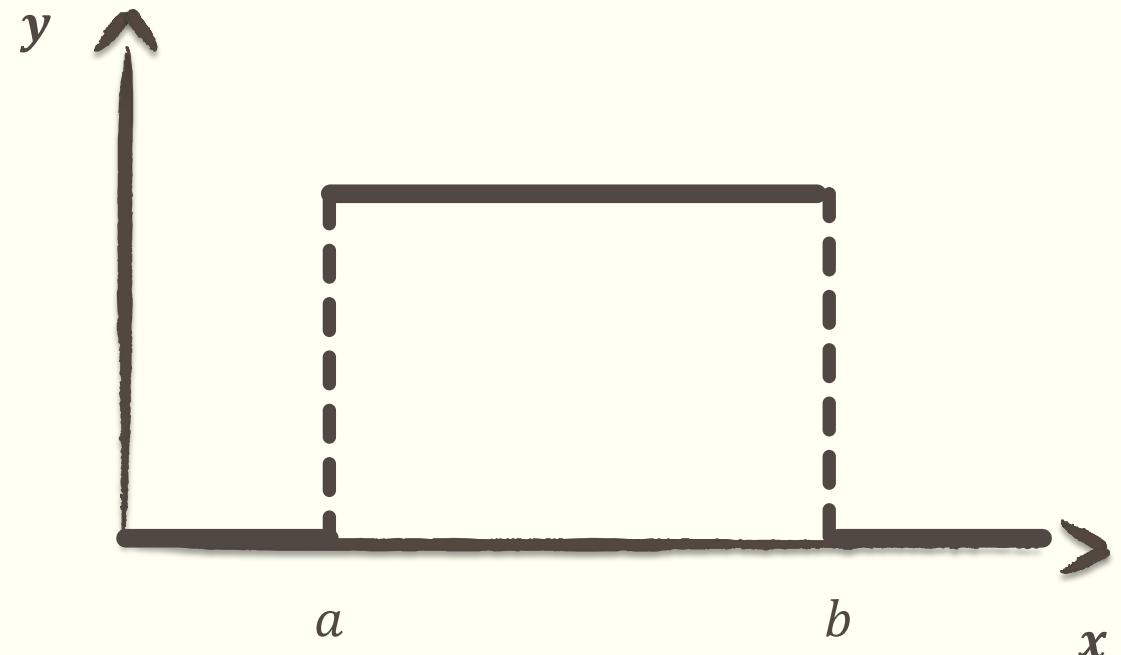
# Continuous Uniform Distribution

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- The distribution describes an experiment where there is an arbitrary outcome that lies between certain bounds.
- The bounds are defined by the parameters,  $a$  and  $b$ , which are the minimum and maximum values.
- The probability density function of the continuous uniform distribution is

$$f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & x < a \text{ or } x > b \end{cases}$$

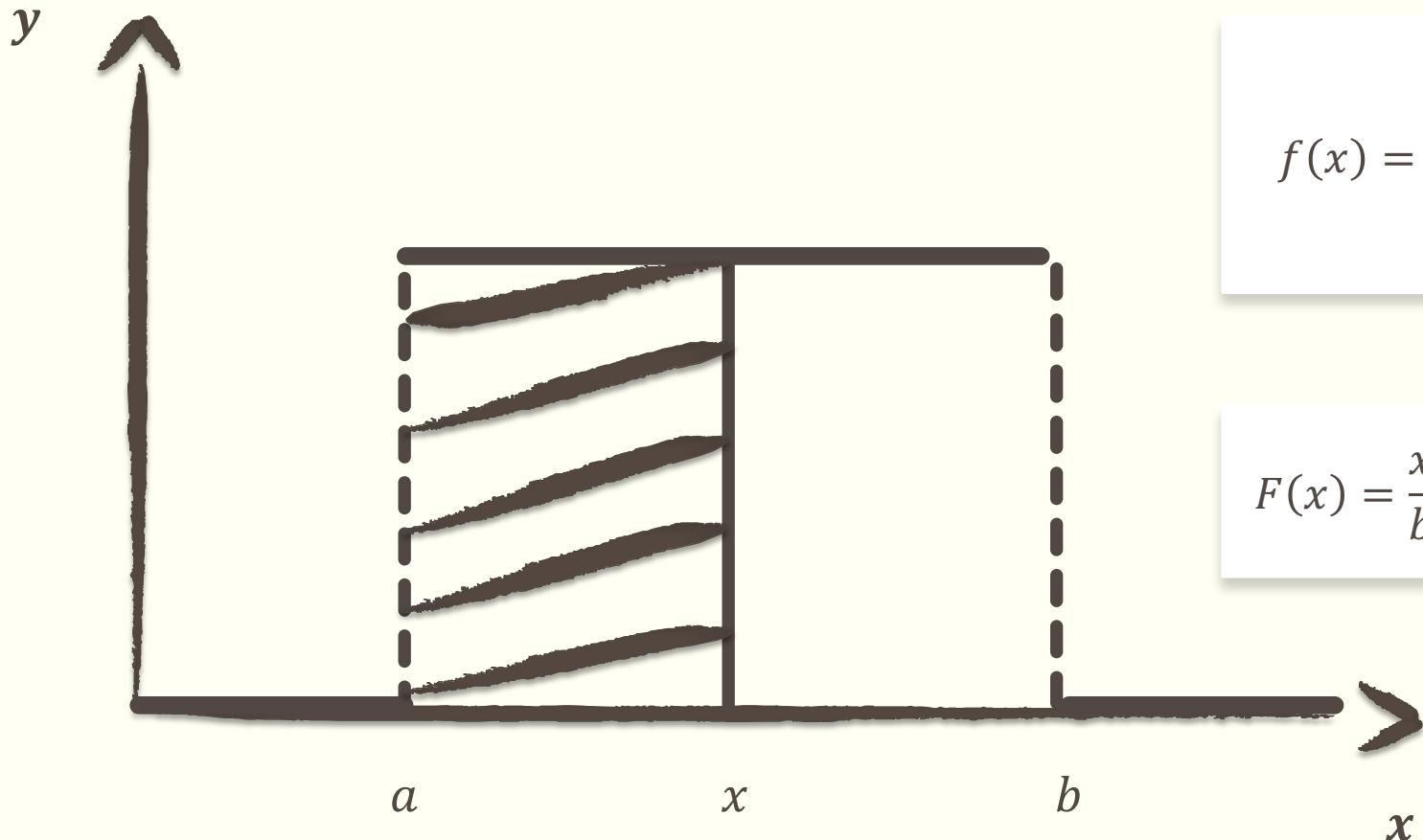


# Calculate the Cumulative Distribution Function of the Uniform Distribution

Let  $X$  be a continuous real-valued random variable with density function  $f(x)$ . Then the function defined by

$$F_X(x) = \int_{-\infty}^x f(s)ds,$$

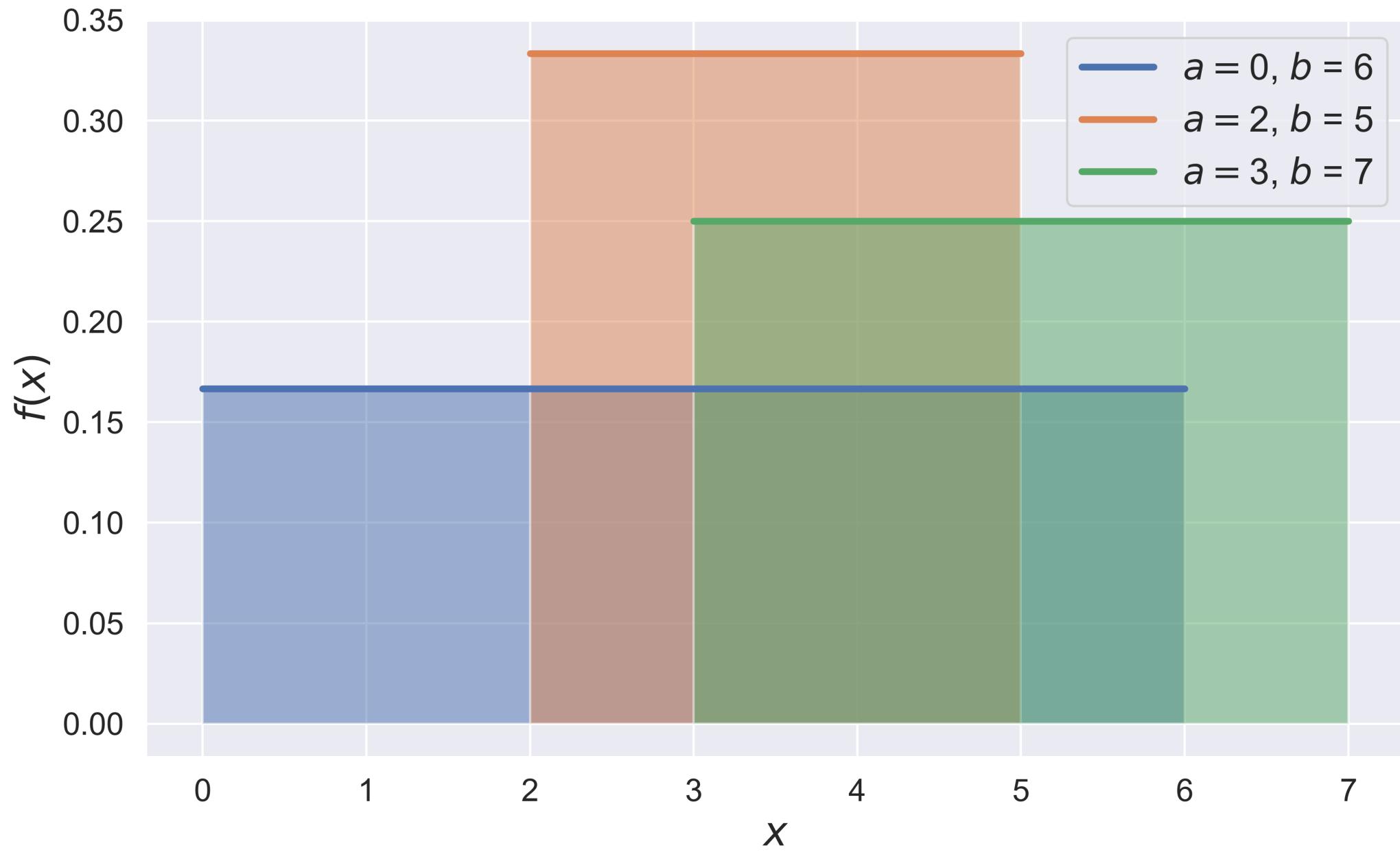
is the cumulative distribution function of  $X$ .



$$f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & x < a \text{ or } x > b \end{cases}$$

$$F(x) = \frac{x-a}{b-a}, \quad a \leq x \leq b$$

# Continuous uniform distribution



## Example 1

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A real number  $U$  is chosen at random from  $[0, 1]$  with uniform probability, and then this number is squared.

Let  $X$  represent the result,  $X = U^3$ .

- What is the cumulative distribution function of  $X$ ?
- What is the density function of  $X$ ?

$$F_U(u) = P(U \leq u) = \dots$$

00

01

$$F_X(x) = P(X \leq x) = \dots$$

range or  $X$  is ...

02

$$\frac{d}{dx} F_X(x) = f(x) = \dots$$

**01**

$$\begin{aligned}F_X(x) &= P(X \leq x) = P(U \leq \sqrt[3]{x}) \\&= \sqrt[3]{x} \\0 \leq X &\leq 1\end{aligned}$$

**02**

$$\frac{d}{dx} F_X(x) = f(x) = \frac{d}{dx} \sqrt[3]{x} = \frac{1}{3\sqrt[3]{x^2}}$$

## Example 2

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---

Two real numbers  $X$  and  $Y$  are chosen at random and uniformly from  $[0, 1]$ . Let  $Z = X + Y$ .

Please derive expressions for the cumulative distribution and the density function of  $Z$ .

01

$F_Z(z) = P(Z \leq z) = \dots$   
range of  $Z$  is ...

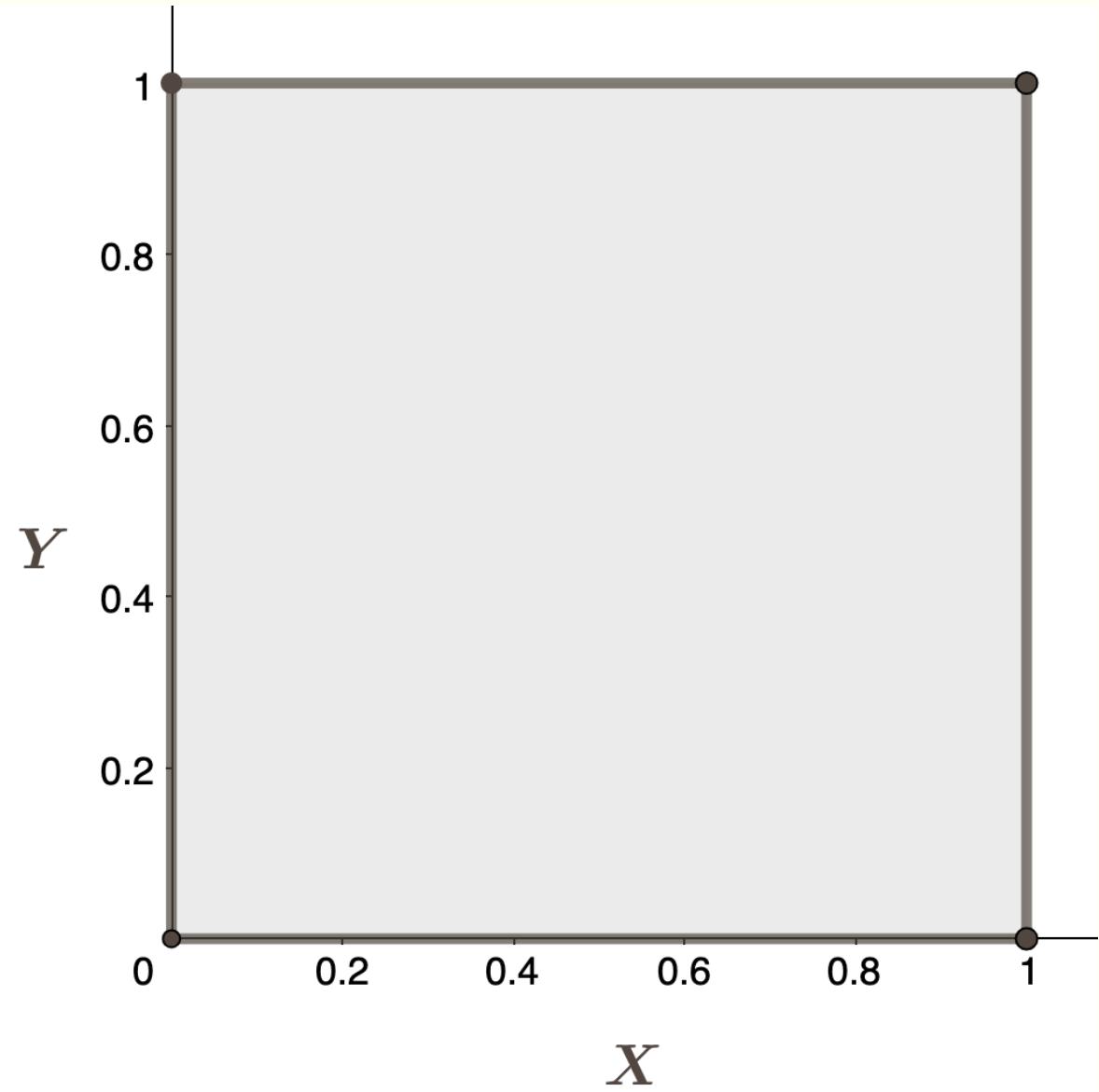
$$\frac{d}{dz} F_Z(z) = f(z) = \dots$$

02

- Two real numbers  $X$  and  $Y$  are chosen at random and uniformly from  $[0, 1]$ . Let  $Z = X + Y$ .
- Please derive expressions for the cumulative distribution and the density function of  $Z$ .

**01**

$$\begin{aligned} F_Z(z) &= P(Z \leq z) \\ &= P(X + Y \leq z) \end{aligned}$$



Choose a number  $U$  from the unit interval  $[0, 1]$  with uniform distribution.  
Consider the following random variables:

- $X = U + 2$
- $Y = 2U$
- $Z = |U - \frac{1}{2}|$

Which one of them follows a uniform distribution as well?

01

$F_Z(z) = P(Z \leq z) = \dots$   
range or  $Z$  is ...

$$\frac{d}{dz} F_Z(z) = f(z) = \dots$$

02

Choose a number  $U$  from the unit interval  $[0, 1]$  with uniform distribution.  
Consider the following random variables:

- $X = U + 2$
- $Y = 2U$
- $Z = \left|U - \frac{1}{2}\right|$

Which one of them follows a uniform distribution as well?

01

$$F_Z(z) = P(Z \leq z) = P\left(\left|U - \frac{1}{2}\right| \leq z\right) = 2z$$

range of  $Z$  is  $[0, \frac{1}{2}]$

$$\frac{d}{dz} F_Z(z) = f(z) = \frac{d}{dz}(2z) = 2$$

02

# Uniform distribution

Consider the following random variable:

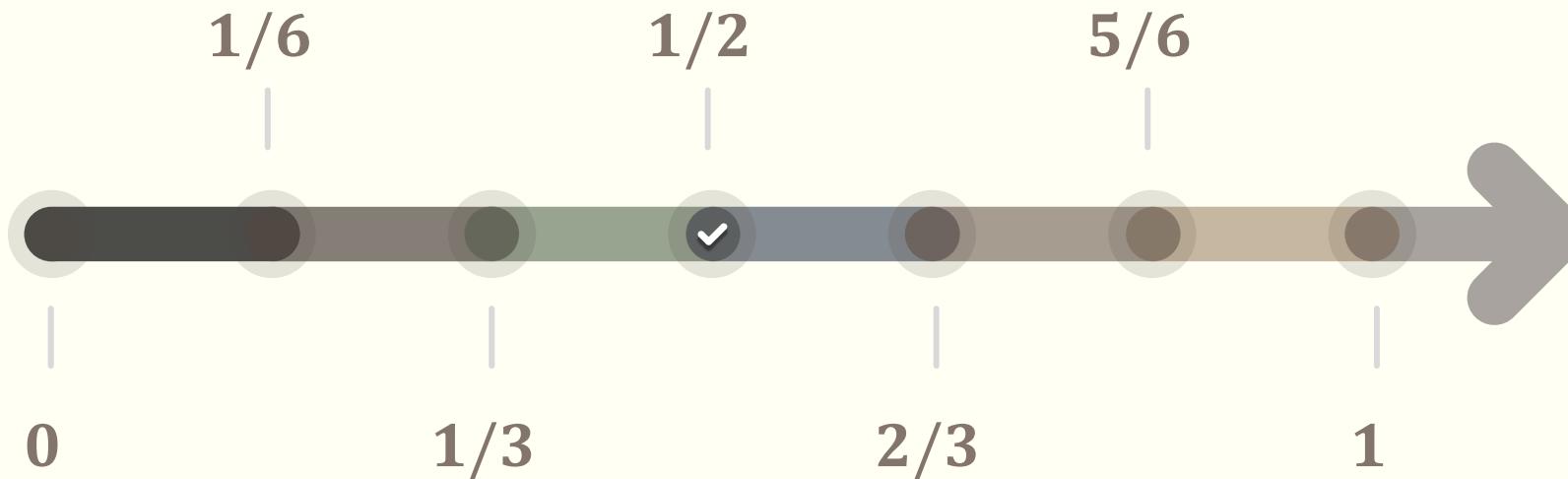
- $Z = |U - \frac{1}{2}|$

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

=

$$P\left(Z = \frac{1}{6}\right) = P\left(U = \frac{1}{3}\right) + P\left(U = \frac{2}{3}\right)$$

=



$$P(Z = 0) = \frac{1}{2} P(Z = \frac{1}{6})$$

It cannot be uniform?

# Uniform distribution

Consider the following random variable:

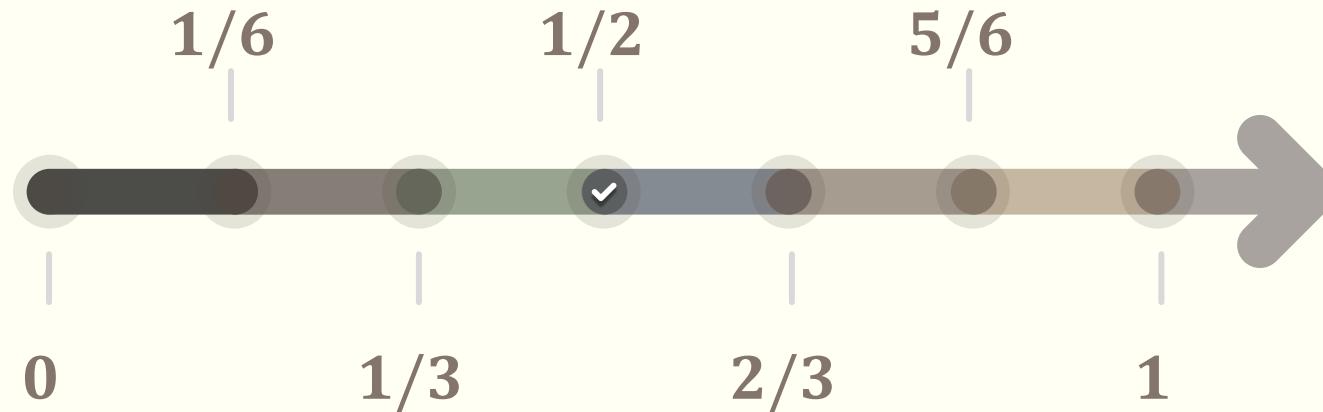
- $Z = |U - \frac{1}{2}|$

$$P(Z = 0) = P\left(U = \frac{1}{2}\right) = 0$$

=

$$P\left(Z = \frac{1}{6}\right) = P\left(U = \frac{1}{3}\right) + P\left(U = \frac{2}{3}\right) = 0$$

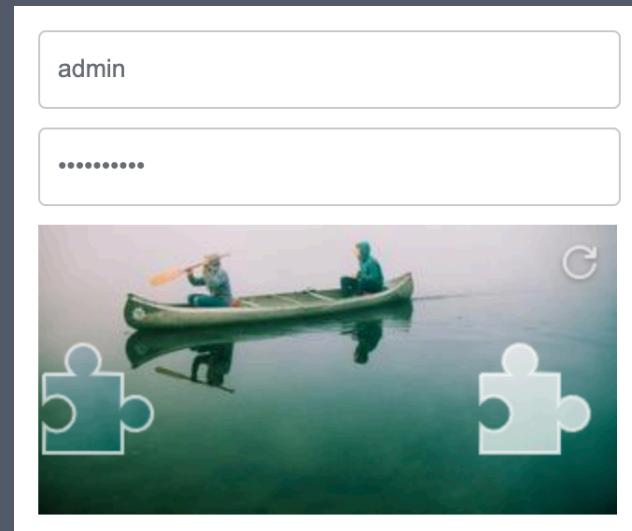
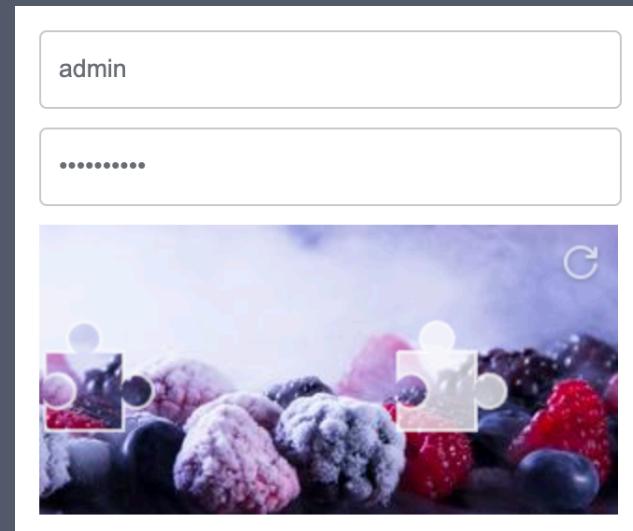
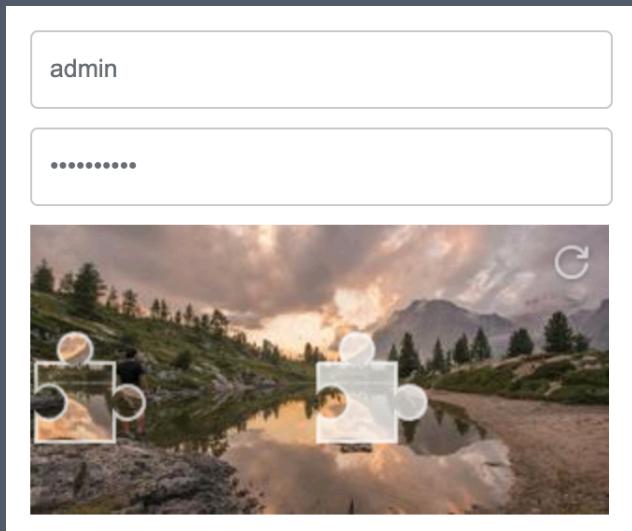
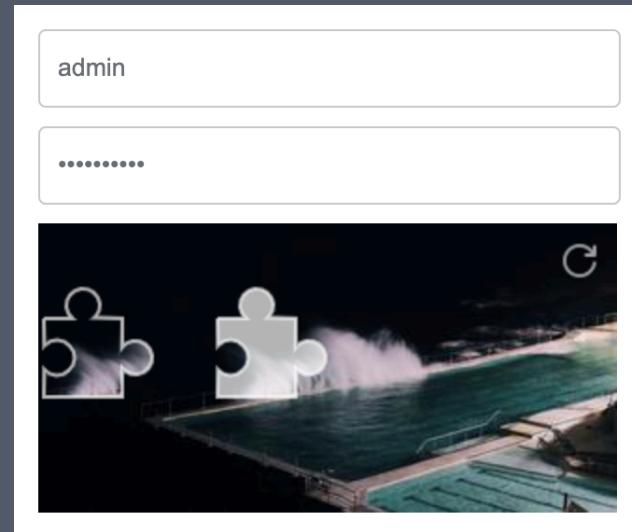
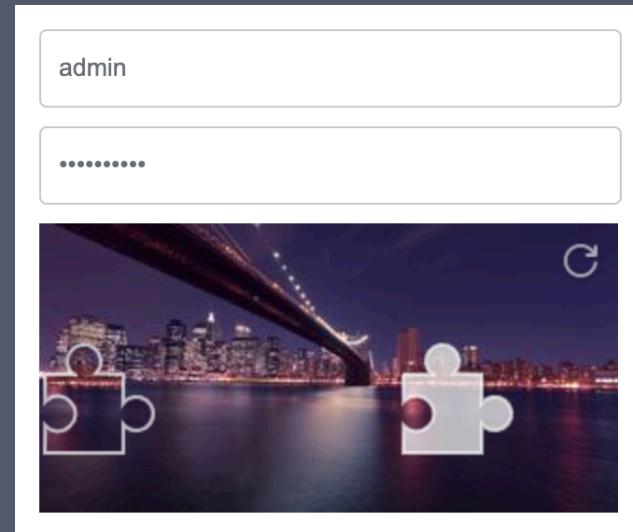
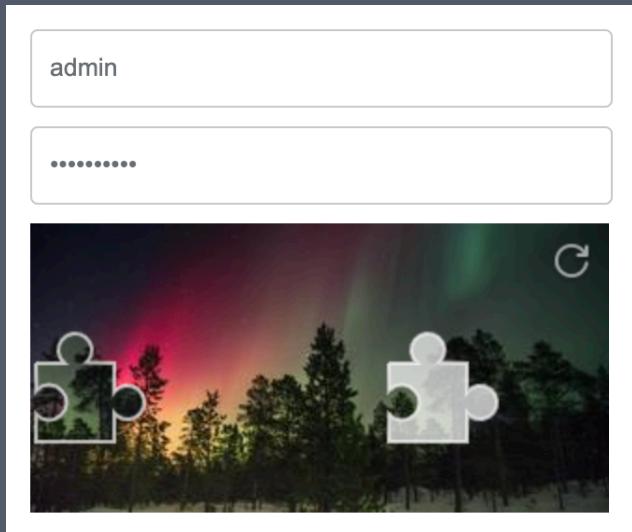
=



$$P(Z = 0) = P\left(Z = \frac{1}{6}\right) = 0$$

For continuous random variable, the probability at a single point cannot be used to decide the distribution!

# Sliding Verification Code



# Sliding Verification Code

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admin

.....



```
canvasInit() {
    // # generate random number
    const random = (min, max) => {
        return Math.floor(Math.random() * (max - min + 1) + min);
    };
    //x: 254, y: 109
    let mx = random(127, 244),
        bx = random(10, 128),
        y = random(10, 99);

    this.slider = { mx, bx };

    this.draw(mx, bx, y);
},
```

## Javascript: Math.random()

The `Math.random()` function returns a floating-point, pseudo-random number in the range 0 to less than 1 (inclusive of 0, but not 1) with approximately **uniform distribution** over that range, which you can then scale to your desired range.

# Exponential Distribution

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- The exponential distribution is the probability distribution of the **time** between events in a **Poisson point process**. That is, a process in which events occur continuously and independently at a constant average rate  $\lambda$ .
- The density function of an exponential distribution is

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

- The cumulative distribution function of an exponential distribution is

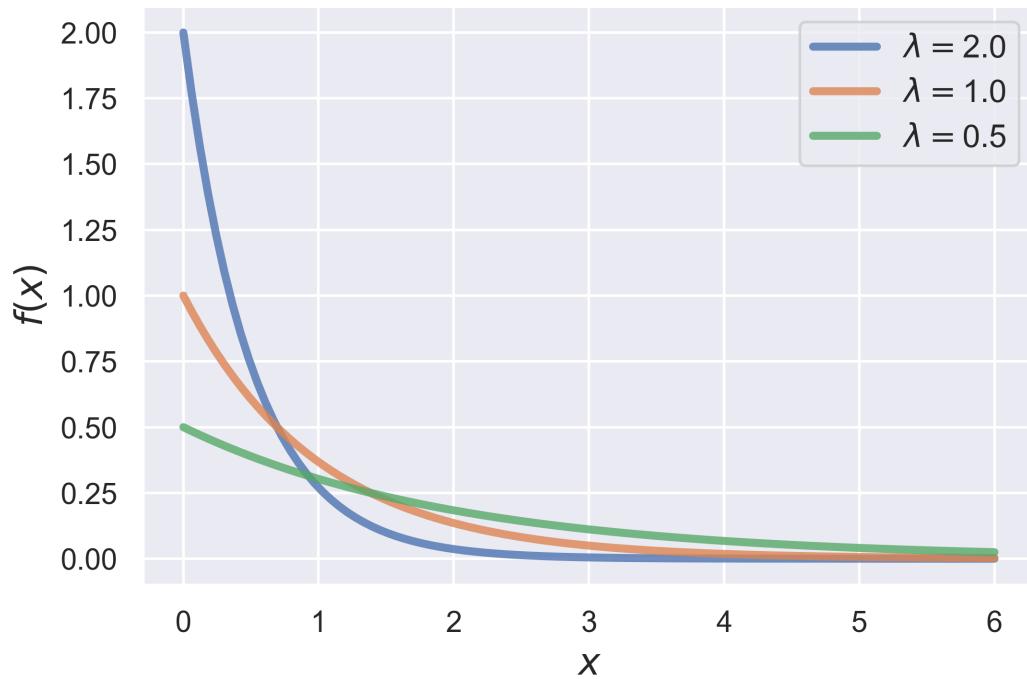
$$F(x) = \begin{cases} 1 - e^{-\lambda x}, & x \geq 0 \\ 0, & x < 0 \end{cases}.$$

# Exponential Distribution

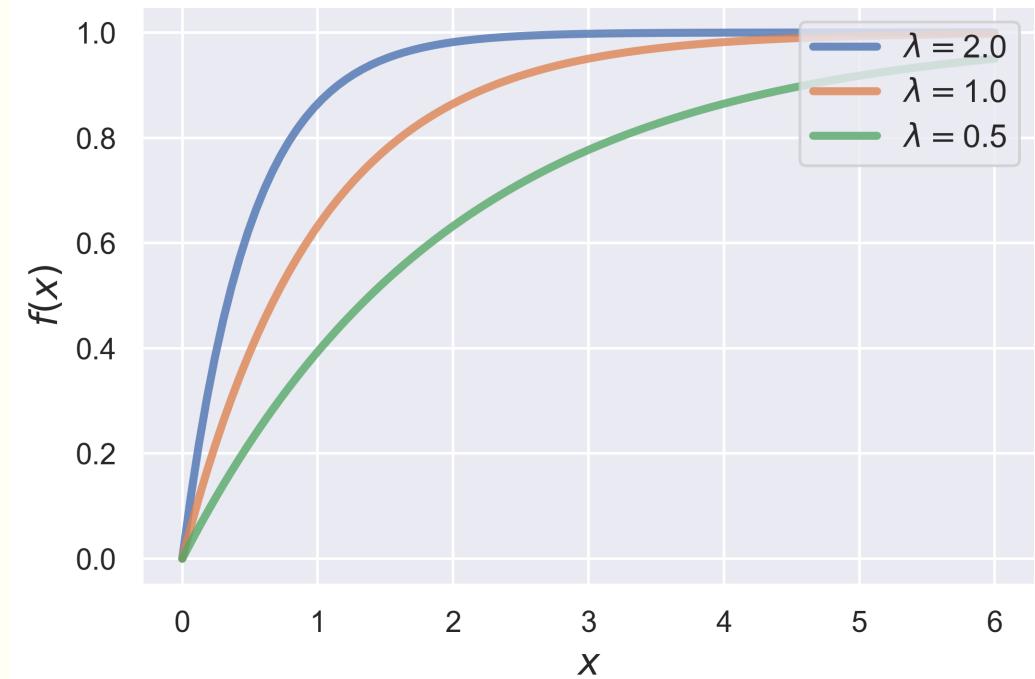
$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0 \\ 0, & x < 0 \end{cases}.$$

$$F(x) = \begin{cases} 1 - e^{-\lambda x}, & x \geq 0 \\ 0, & x < 0 \end{cases}.$$

Density function



Cumulative distribution function



## Exponential Distribution

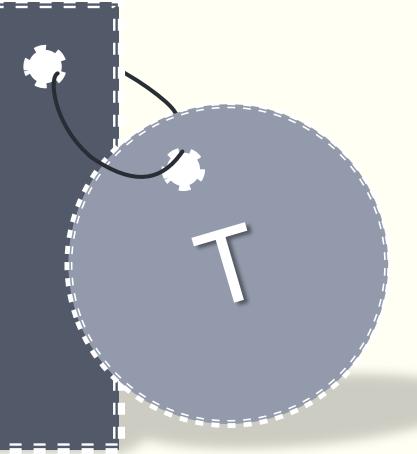
$$P(T > t) = e^{-\lambda t}$$

$$\begin{aligned} P(T > r + s | T > r) &= \frac{P(T > r + s \cap T > r)}{P(T > r)} = \frac{P(T > r + s)}{P(T > r)} \\ &= \frac{e^{\lambda-(r+s)}}{e^{-\lambda r}} = e^{-\lambda s} = P(T > s) \end{aligned}$$



## Memoryless Property

$$P(T > r + s | T > r) = P(T > s)$$

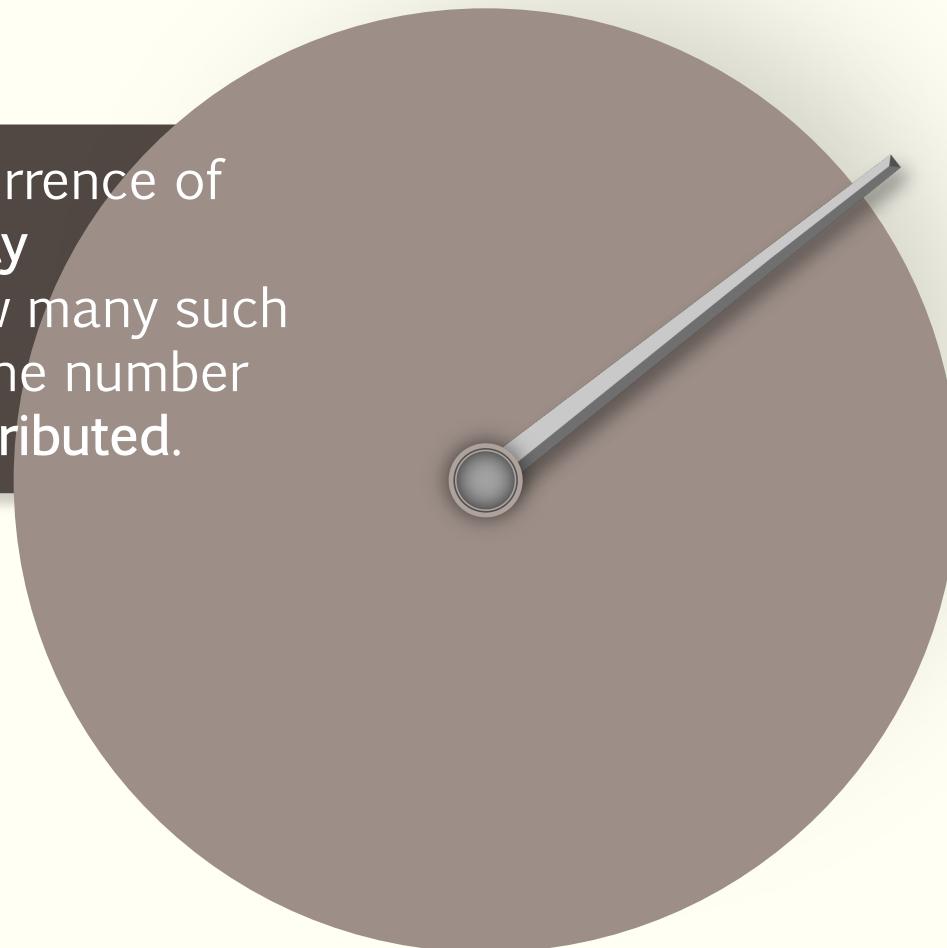


# Poisson Process

## Binomial distribution

$$b(n, p, k) = \binom{n}{k} p^k q^{n-k}$$

Poisson process: the occurrence of each event is **exponentially distributed**. By time t, how many such events have happened? The number of events are **Poisson-distributed**.



## Poisson Distribution

$$P(X = k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

## Exponential Distribution

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

# Poisson Process

- A Poisson Process is a model for a series of discrete event where the **average time** between events is known, but the exact timing of events is random.
- The arrival of an event is independent of the event before (waiting time between events is **memoryless**).
- A Poisson process meets the following criteria (in reality many phenomena modeled as Poisson processes do not meet these exactly):
  - Events are **independent** of each other. The occurrence of one event does not affect the probability another event will occur.
  - The average rate (events per time period) is constant.
  - Two events cannot occur at the same time.



## Retail

Events are independent of each other.

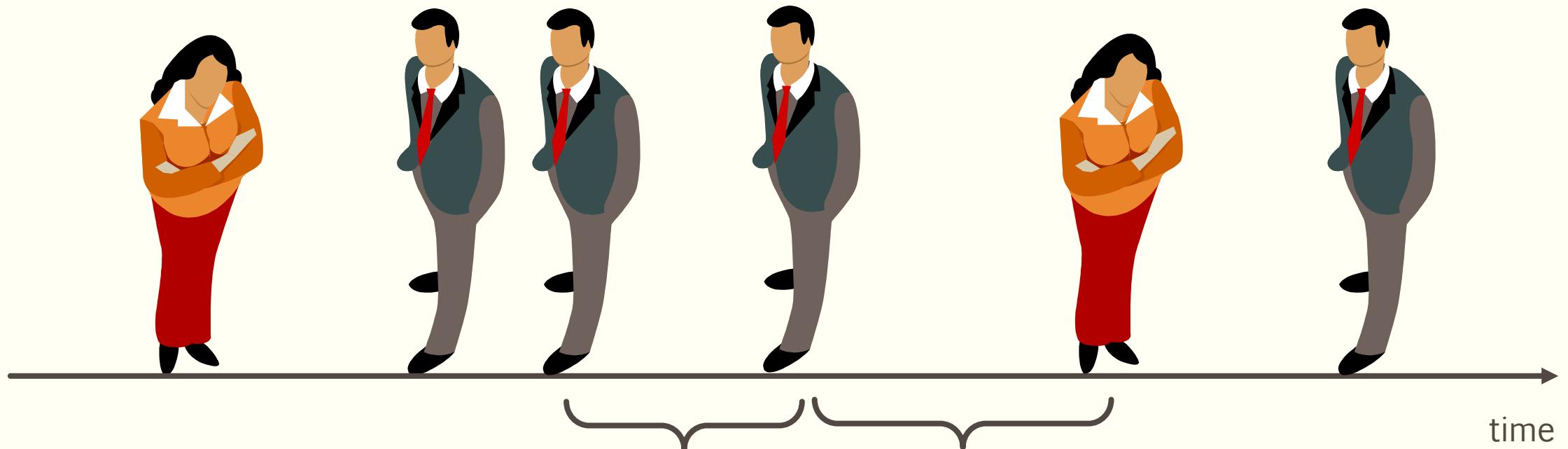
The average rate is constant.

Two events cannot occur at the same time.

## Customers

# Poisson Process

- A Poisson Process is a model for a series of discrete event where the **average time** between events is known, but the exact timing of events is random.
- The arrival of an event is independent of the event before (waiting time between events is **memoryless**).

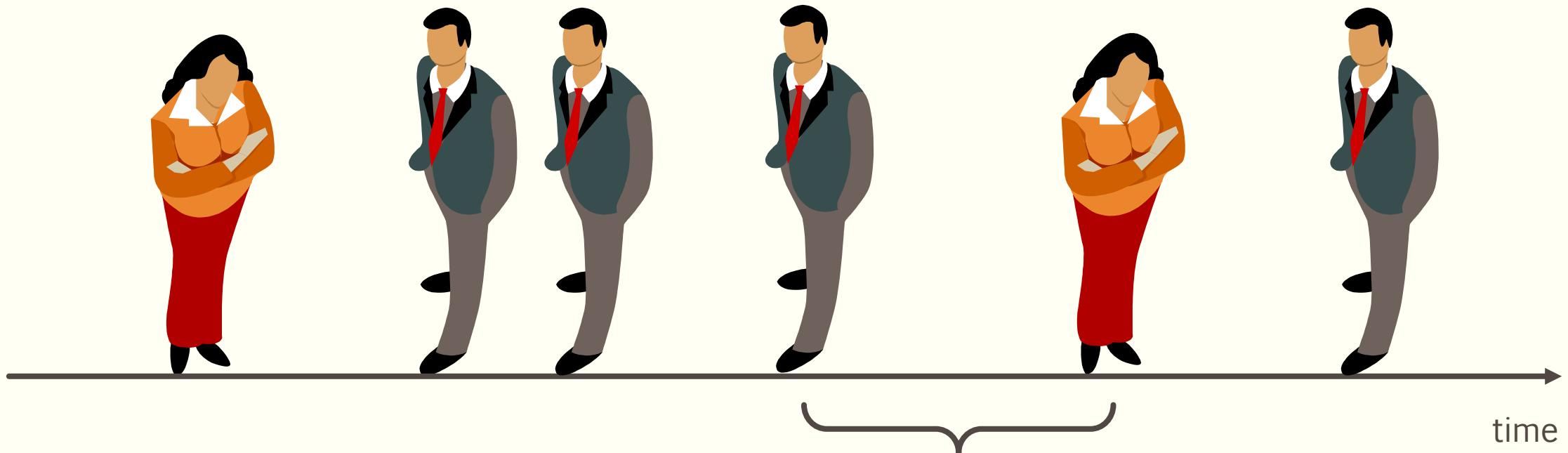


# Waiting time

- An intriguing part of a Poisson process involves figuring out how long we have to wait until the next event (this is sometimes called the interarrival time).
- The waiting time follows an exponential distribution:

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

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



# Number of events

- By time  $t$ , how many such events have happened? The number of events are Poisson-distributed as

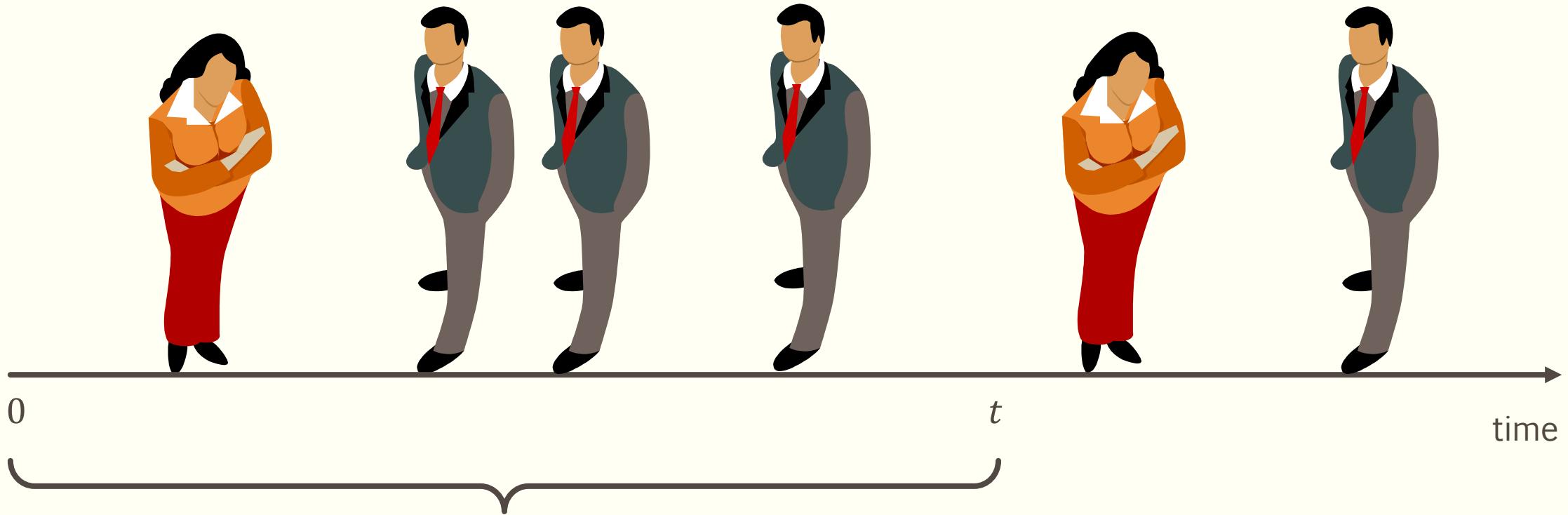
$$P(N = n) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$$

$$np = \lambda t, t = 1, n \rightarrow \infty, p \rightarrow 0$$

=

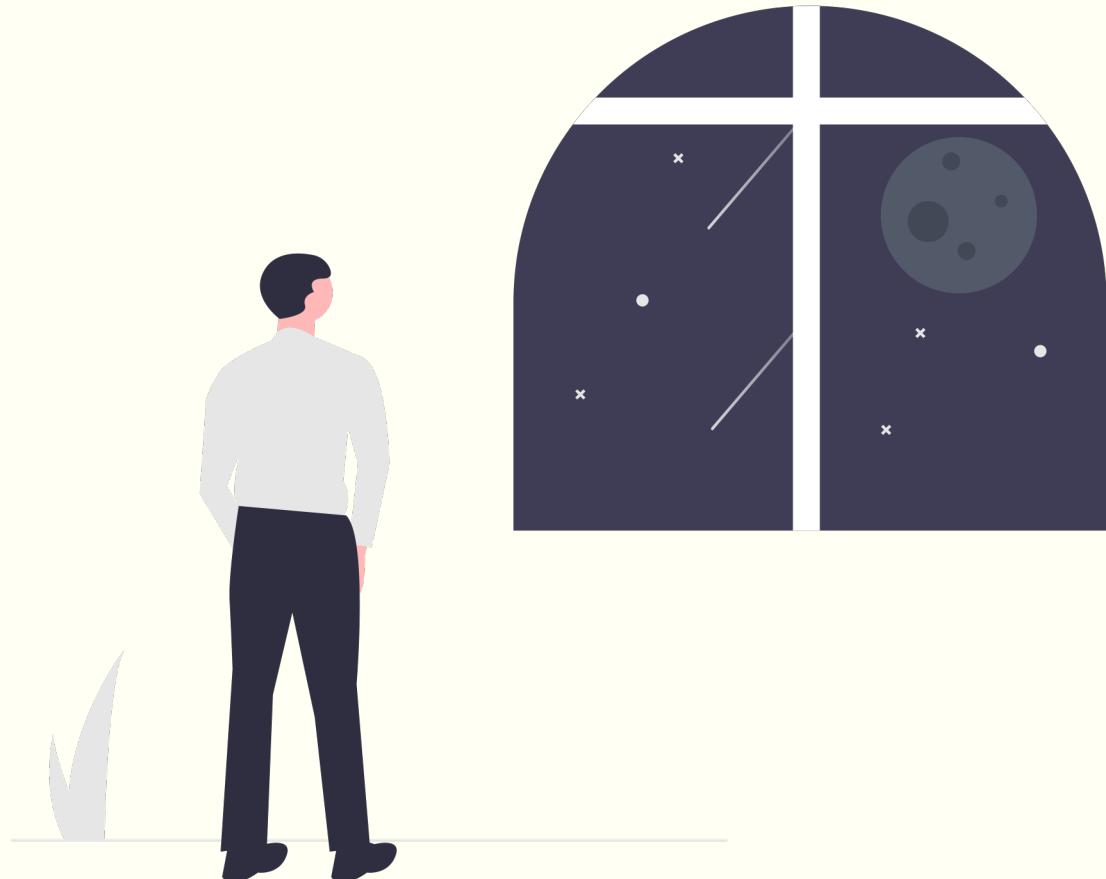
## Poisson Distribution

$$P(X = k) = \frac{\lambda^k}{k!} e^{-\lambda}$$



# EXAMPLE

The probability of seeing a shooting star in 1 hour is 91%. What is the probability of seeing a shooting star in 30 minutes?



# Poisson distribution

- By time  $t$ , how many such events have happened? The number of events are Poisson-distributed as

$$P(N = n) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$$

$t = 1$  hour

$$P(X \geq 1) = 1 - e^{-\lambda t} = 91\%$$

$t = 30$  minutes

$$P(X \geq 1) = 1 - e^{-\lambda t} = \dots$$



$t = 1$  hour

$$P(X \geq 1) = 1 - e^{-\lambda t} = 91\%$$

$$e^{-\lambda t} = 9\% = 0.09$$

$t = 30$  minutes

$$P(X \geq 1) = 1 - e^{-\lambda t} = 70\%$$

$$e^{-\lambda t} = 0.3$$



# Binomial distribution

- success: a shooting star
- failure: no shooting star
- probability of a success:  $p$

$t = 1 \text{ hour}$

$$P(X = 0) = (1 - p)^2 = 0.09$$

$t = 30 \text{ minutes}$

$$P(X = 0) = 1 - p = 0.3$$

30 minutes

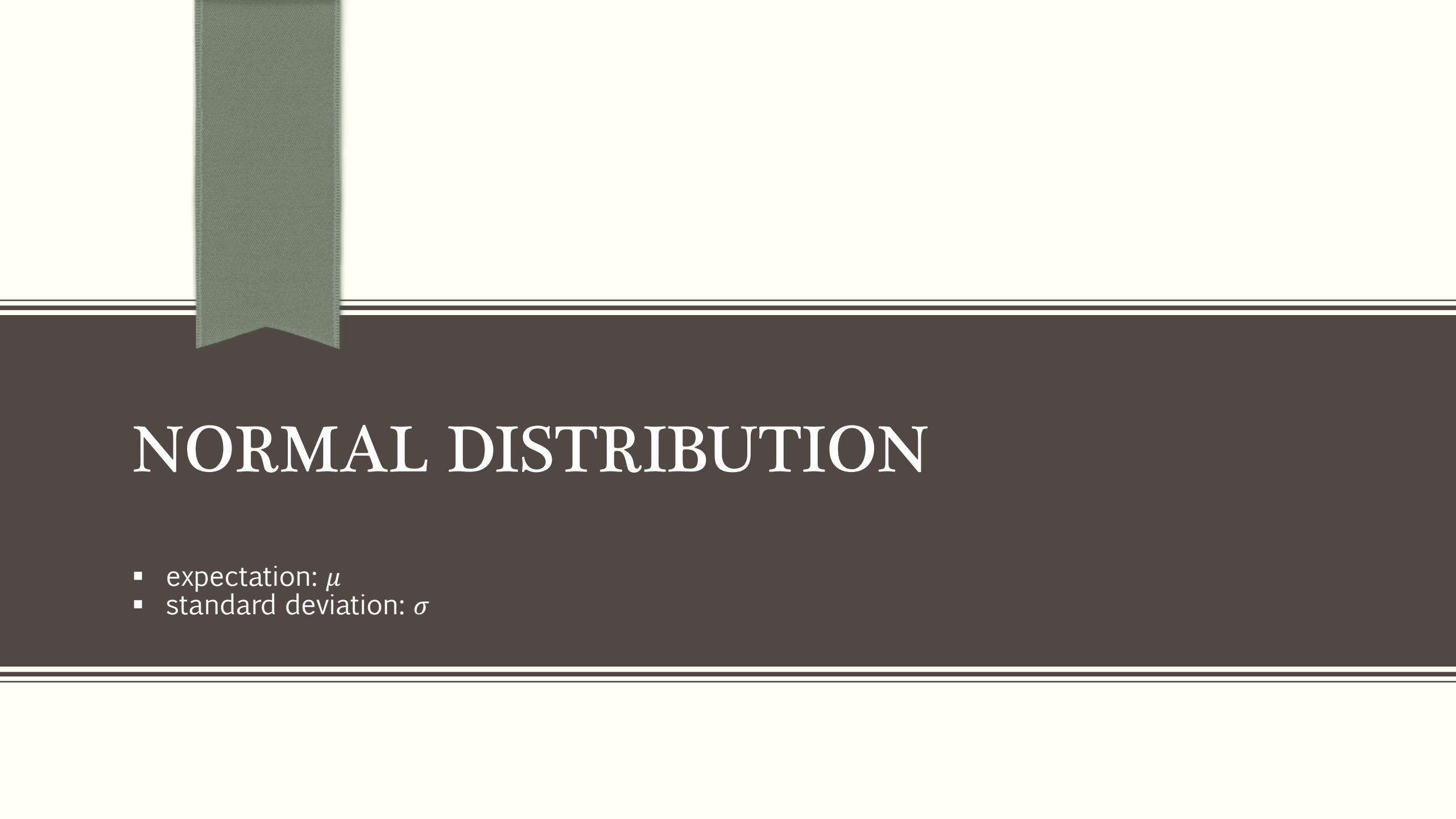


30 minutes



X

✓



# NORMAL DISTRIBUTION

- expectation:  $\mu$
- standard deviation:  $\sigma$

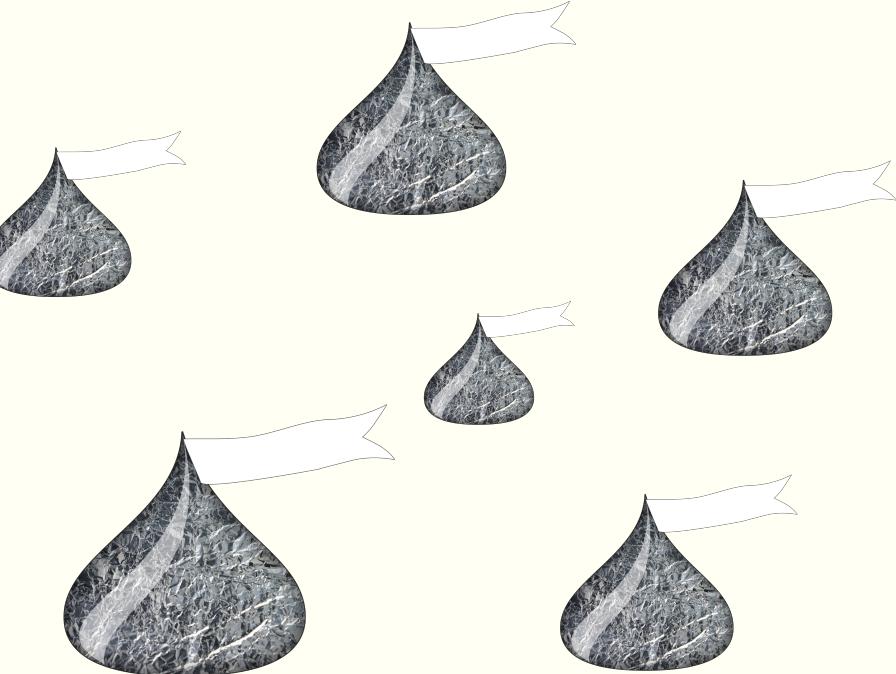
# How Much Does a Hershey Kiss Weight?

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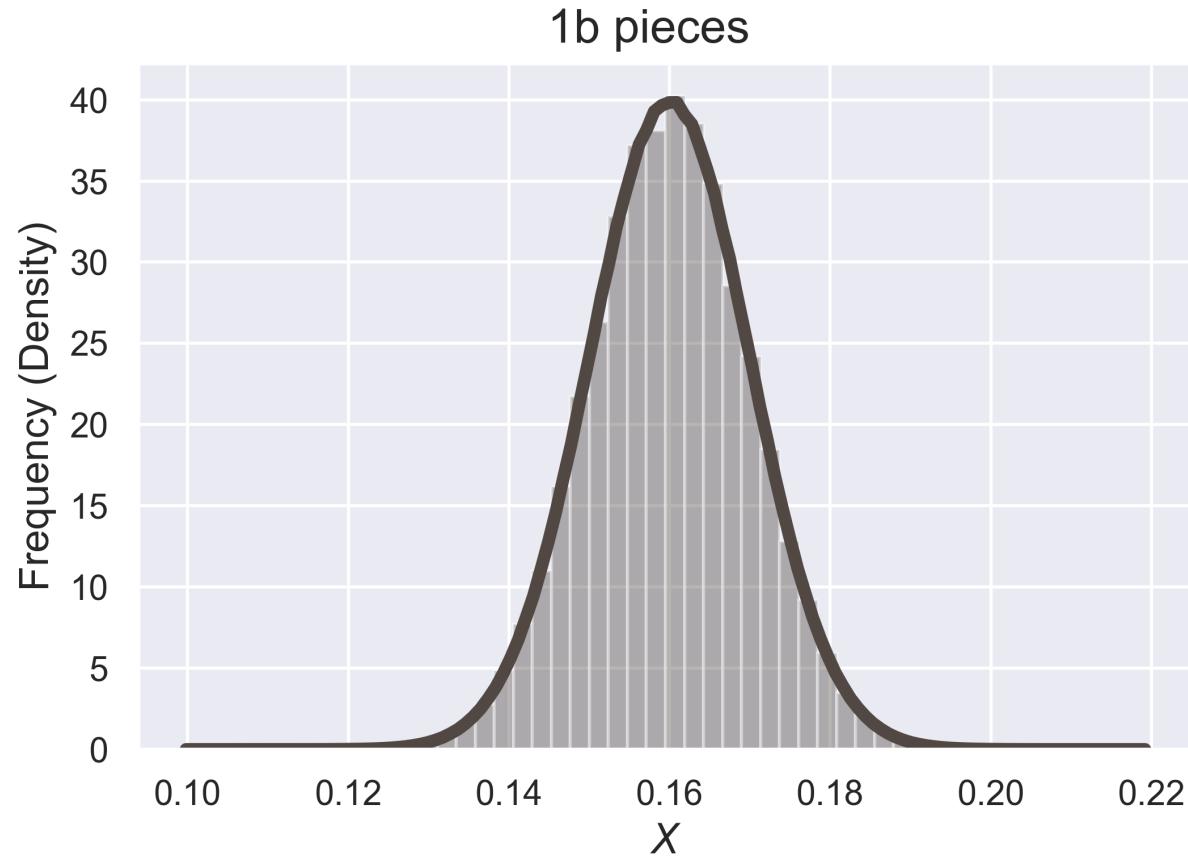
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- A single standard Hershey's Kiss weighs 0.16 ounces.



# Normal Density Distribution (Gaussian Distribution)



- Three-sigma limits is a statistical calculation that refers to data within three standard deviations from a mean.
- In business applications, three-sigma refers to processes that operate efficiently and produce items of the highest quality.
- Three-sigma limits are used to set the upper and lower control limits in statistical quality control charts.
- Control charts are used to establish limits for a manufacturing or business process that is in a state of statistical control.

The normal density function with parameters  $\mu$  and  $\sigma$   
expectation:  $\mu$  standard deviation:  $\sigma$

# Normal Density Distribution

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- The normal density function with parameters  $\mu$  and  $\sigma$ :

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}.$$

- Its cumulative distribution

$$F_X(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}\sigma} e^{-(u-\mu)^2/2\sigma^2} du$$

- The standard normal distribution  $Z$  is the normal density function with  $\mu = 0$  and  $\sigma = 1$ . A general normal distribution  $X$  can be written as

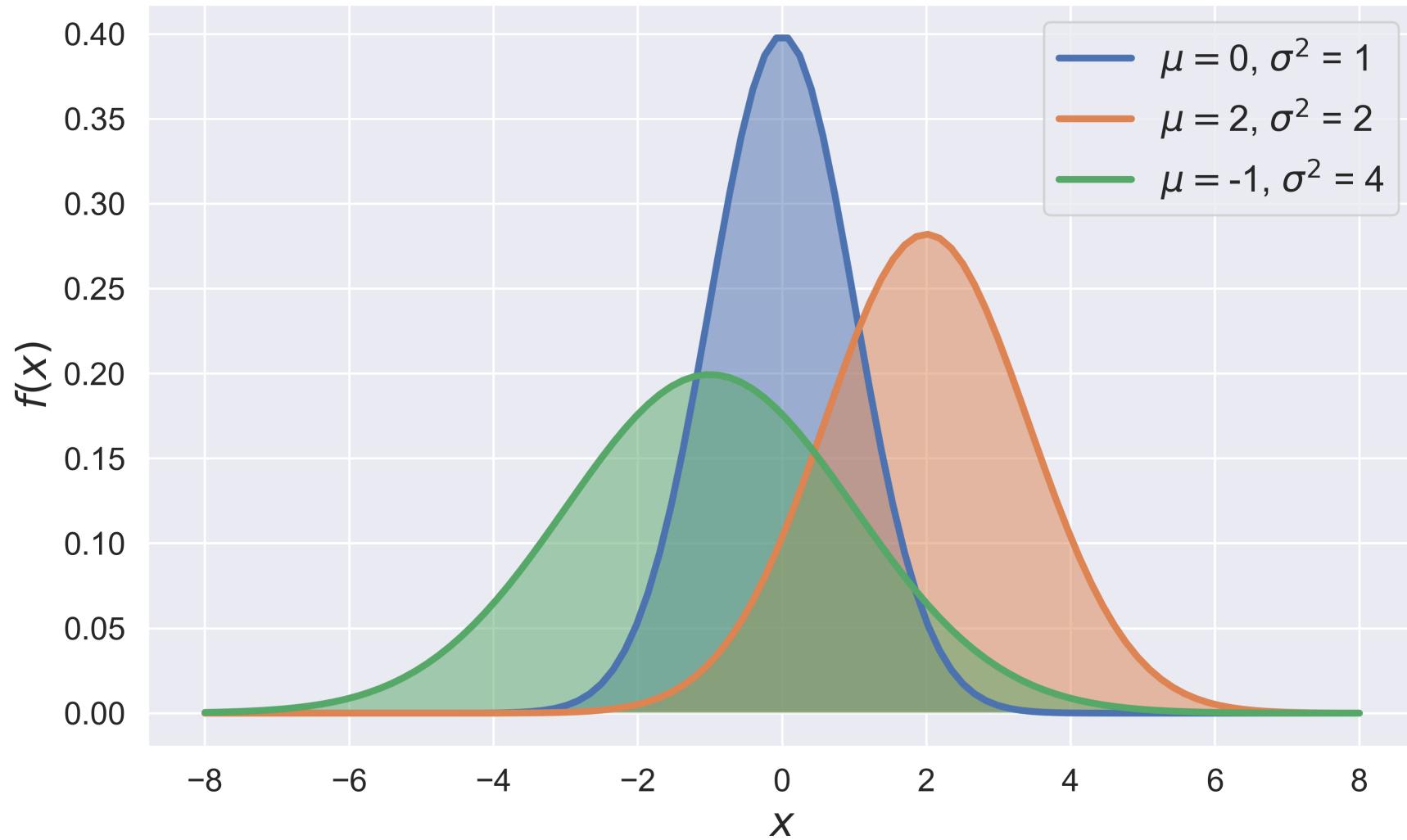
$$X = \sigma Z + \mu$$

# Bell Curve

It is often called a "Bell Curve" because it looks like a bell.

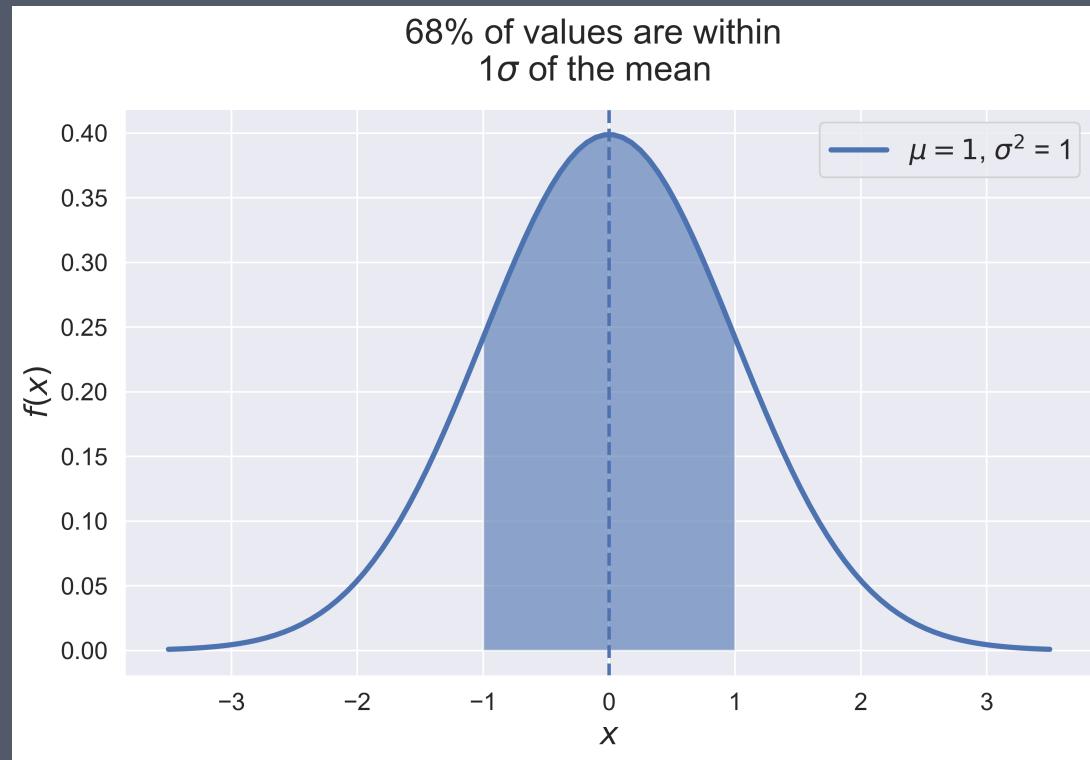
Standard normal distribution $Z$
$\mu = 0$ and $\sigma = 1$
General normal distribution $X$
$X = \sigma Z + \mu$

Density function



# Standard Deviation

The standard deviation is a measure of how spread out numbers are.



## Standard normal distribution $Z$

$$\mu = 0 \text{ and } \sigma = 1$$

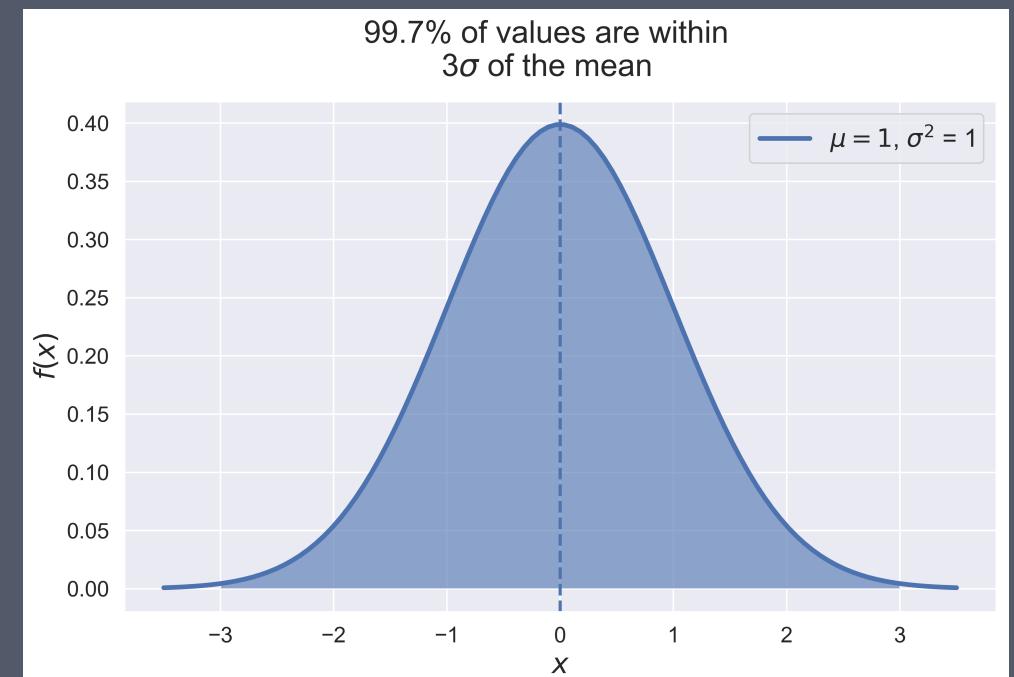
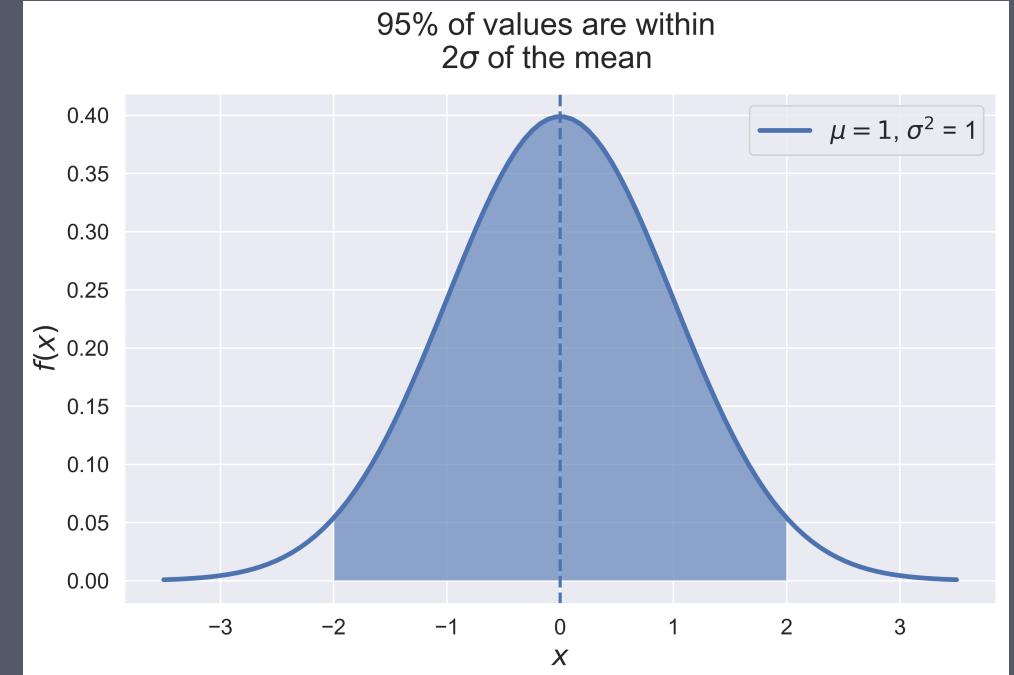
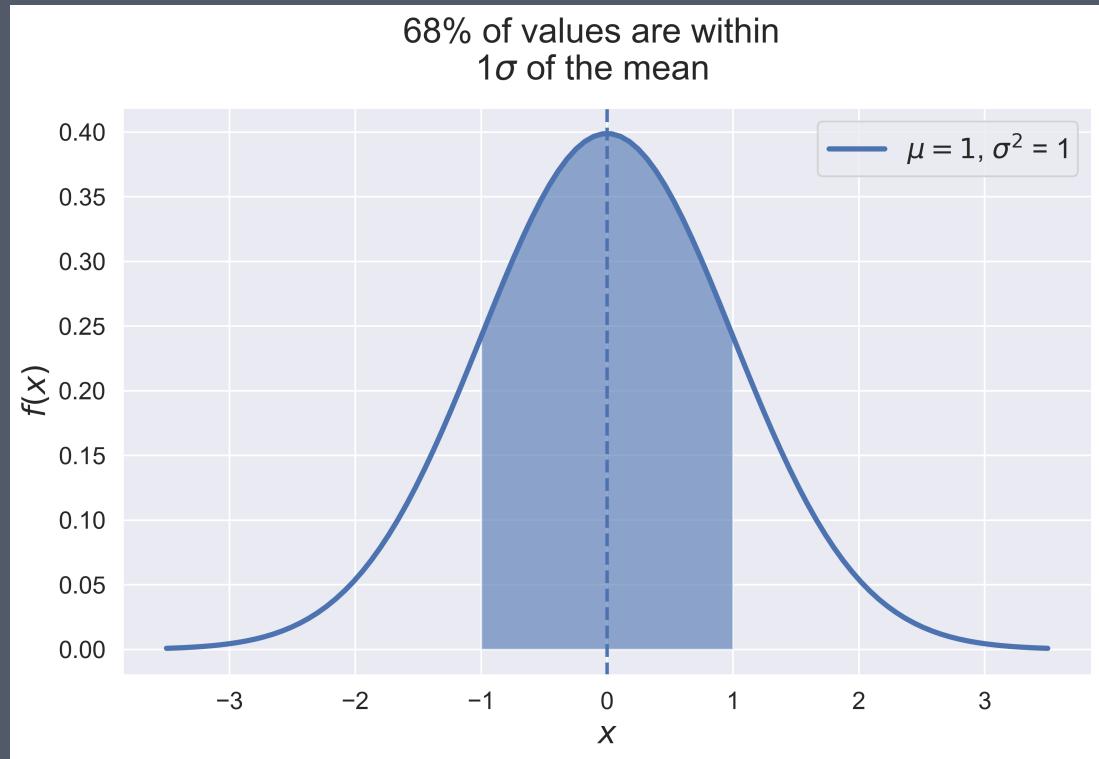
$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}.$$

within  $1\sigma$  of the mean

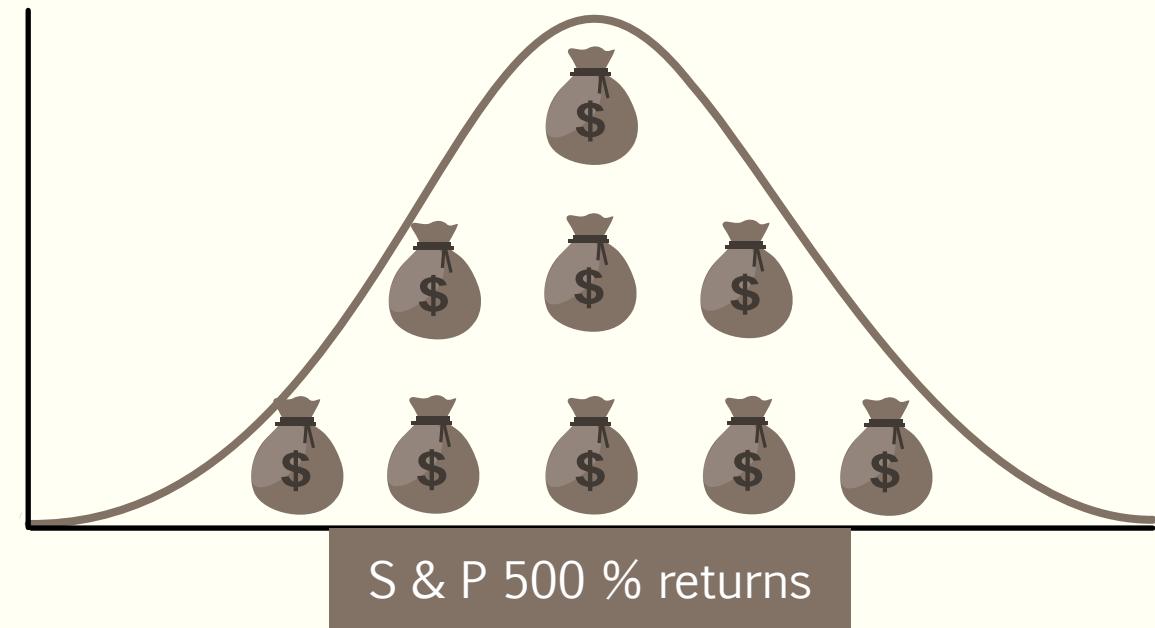
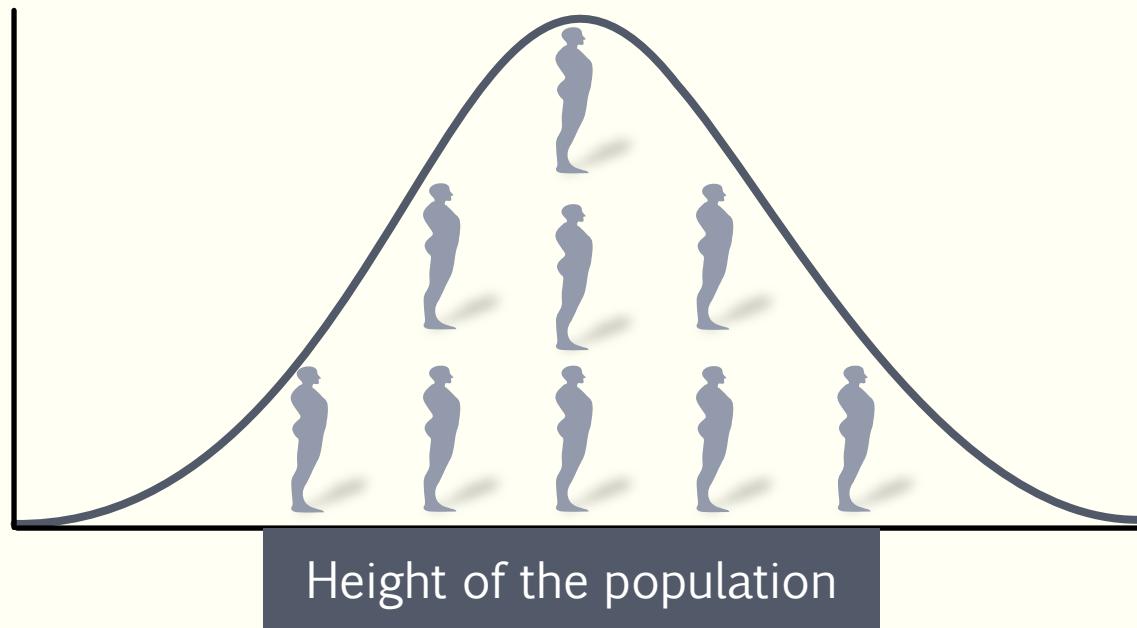
$$\begin{aligned} P(|Z - \mu| \leq \sigma) &= P(|Z| \leq 1) \\ &= \int_{\mu-\sigma}^{\mu+\sigma} \frac{1}{\sqrt{2\pi}\sigma} e^{-(u-\mu)^2/2\sigma^2} du \\ &= \int_{-1}^1 \frac{1}{\sqrt{2\pi}} e^{-u^2/2} du \end{aligned}$$

# Standard Deviation

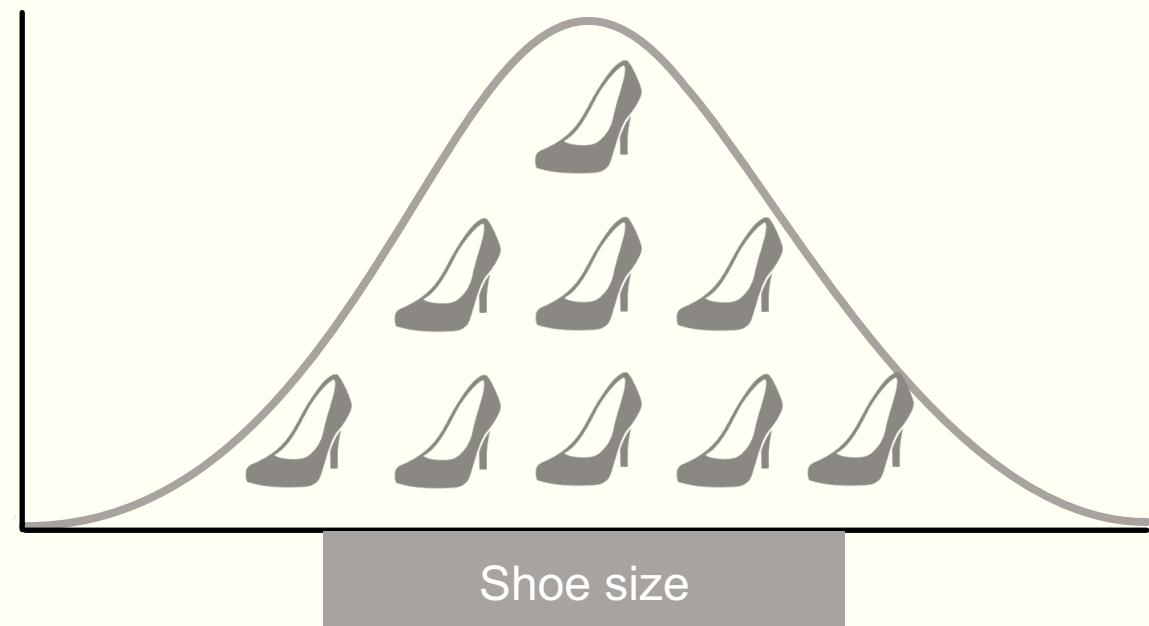
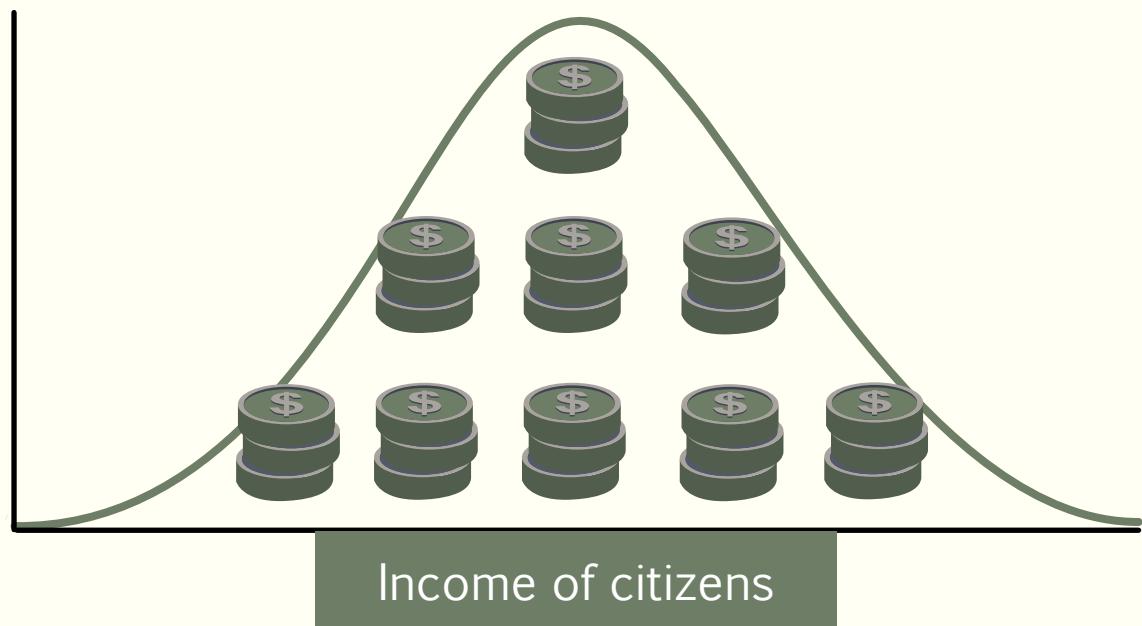
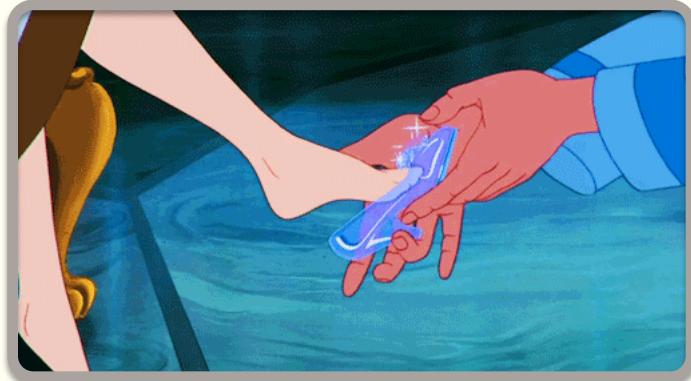
The standard deviation is a measure of how spread out numbers are.



# Examples of Normal Distribution

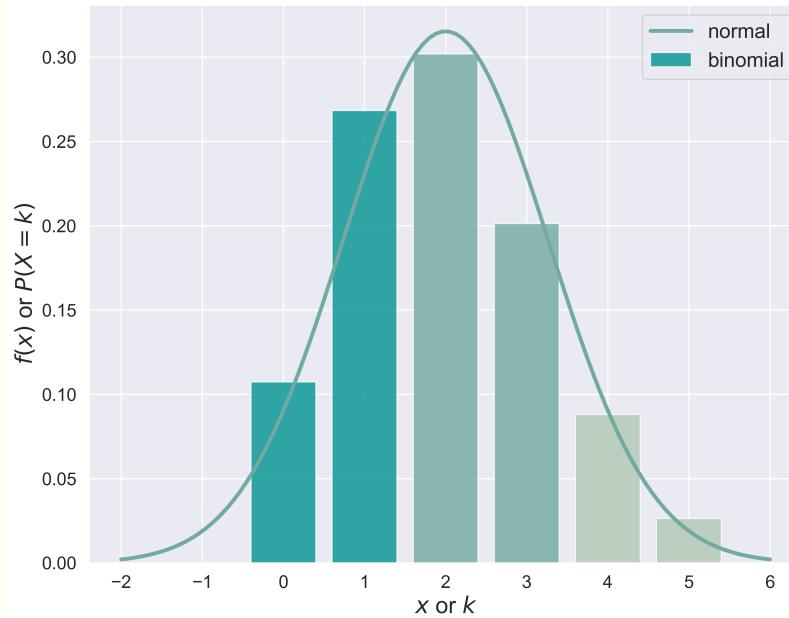


# Examples of Normal Distribution

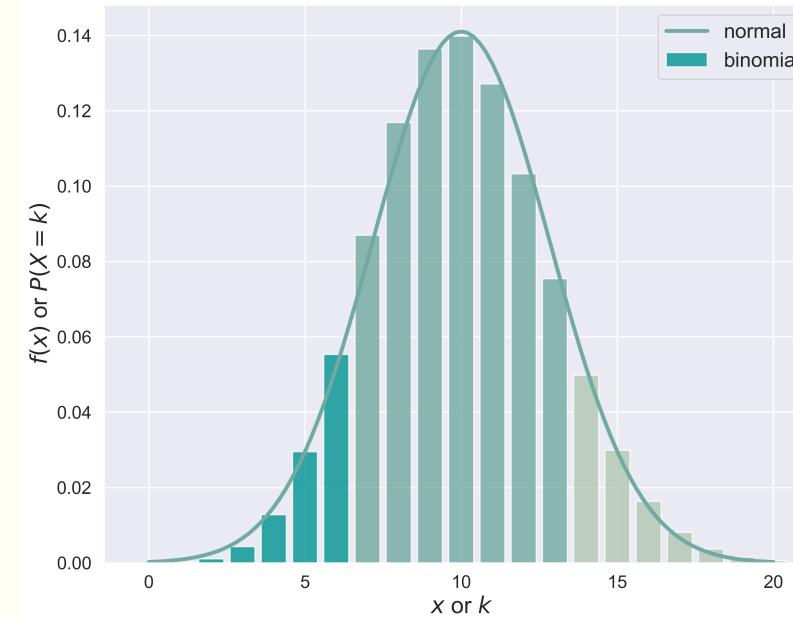


# Binomial Distribution and Normal Distribution

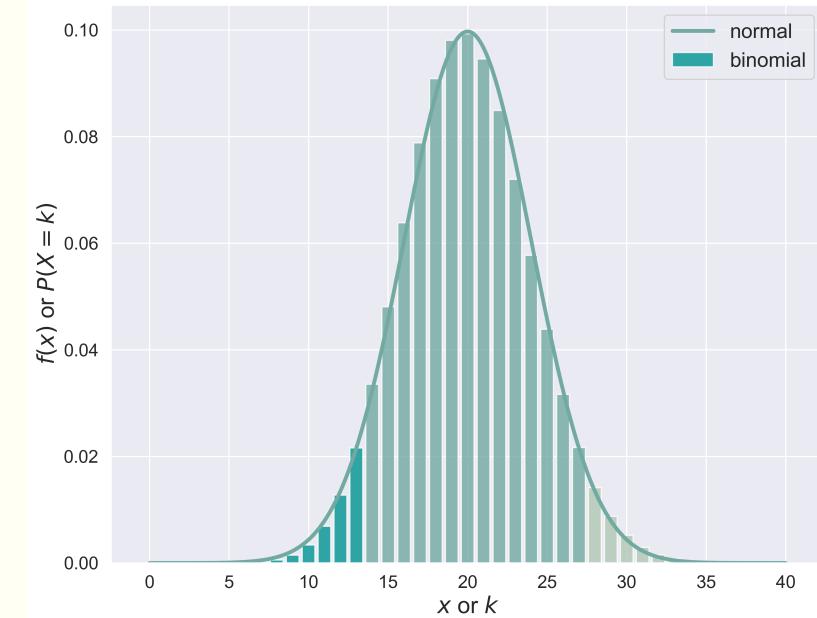
$$\mu = 2, \sigma^2 = 1.6 \\ n = 10, p = 0.2$$



$$\mu = 10, \sigma^2 = 8.0 \\ n = 50, p = 0.2$$



$$\mu = 20, \sigma^2 = 16.0 \\ n = 100, p = 0.2$$



## Binomial Distribution

$$b(n, p, k) = \binom{n}{k} p^k q^{n-k}$$

## Normal Distribution

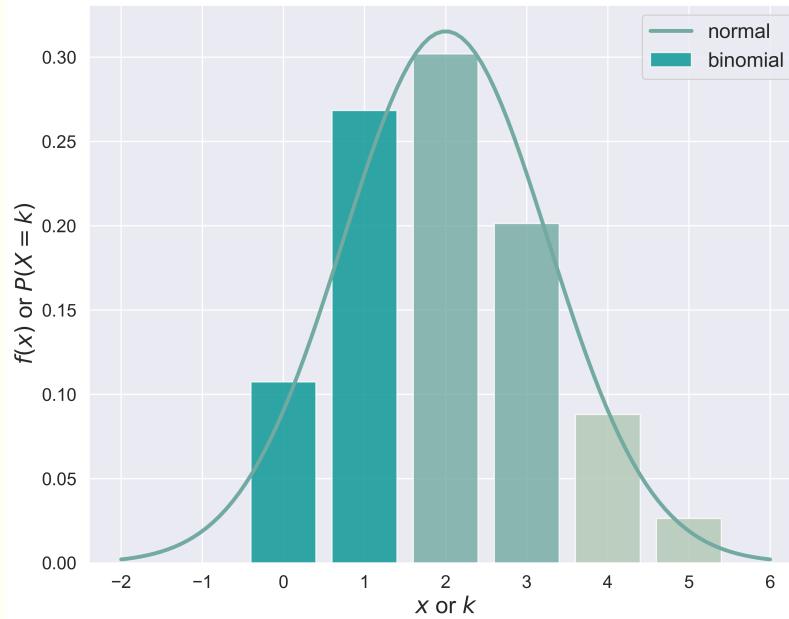
$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}$$

$n \rightarrow \infty$

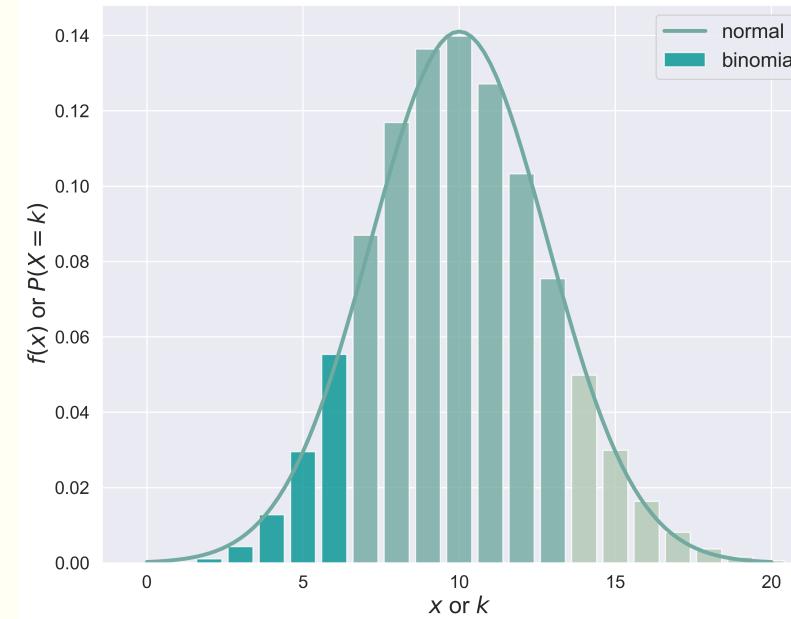
- $np = \dots$
- $np(1 - p) = \dots$

# Binomial Distribution and Normal Distribution

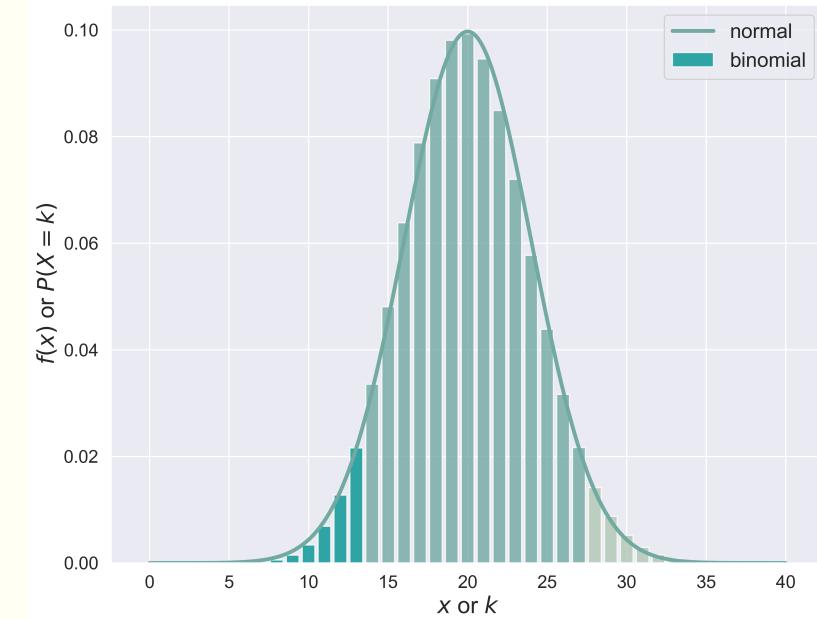
$$\mu = 2, \sigma^2 = 1.6  
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$$\mu = 20, \sigma^2 = 16.0  
n = 100, p = 0.2$$



## Binomial Distribution

$$b(n, p, k) = \binom{n}{k} p^k q^{n-k}$$

## Normal Distribution

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}$$

$n \rightarrow \infty$

- $np = \mu$
- $np(1 - p) = \sigma^2$