

# Continuous Variables

Chris Piech  
CS109, Stanford University





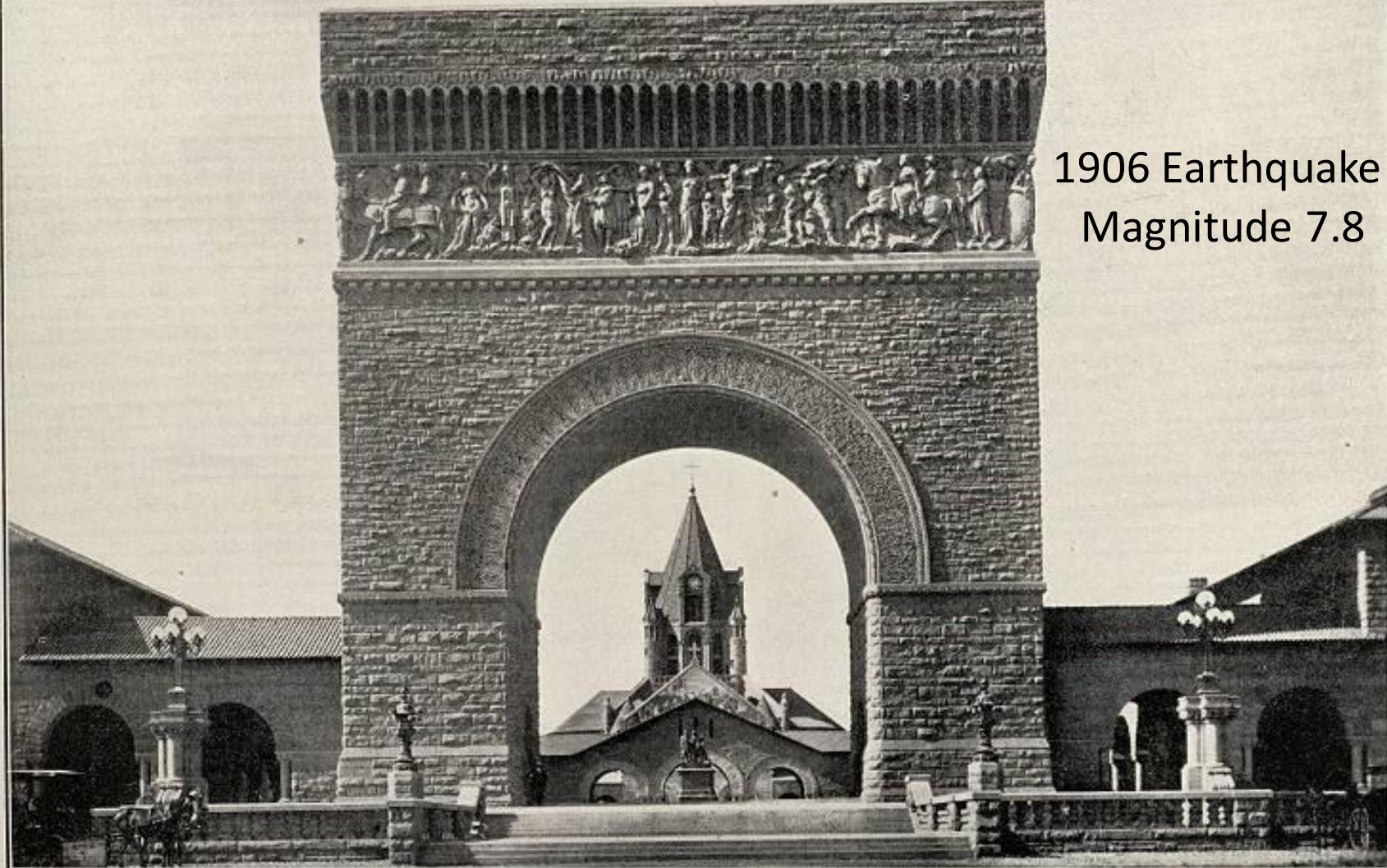


ILL. No. 65. MEMORIAL ARCH, WITH CHURCH IN BACKGROUND, STANFORD UNIVERSITY, SHOWING TYPES OF CARVED WORK WITH THE SANDSTONE.

1906 Earthquake  
Magnitude 7.8



ILL. No. 65. MEMORIAL ARCH, WITH CHURCH IN BACKGROUND, STANFORD UNIVERSITY, SHOWING TYPES OF CARVED WORK WITH THE SANDSTONE.



1906 Earthquake  
Magnitude 7.8

ILL. No. 65. MEMORIAL ARCH, WITH CHURCH IN BACKGROUND, STANFORD UNIVERSITY, SHOWING TYPES OF CARVED WORK WITH THE SANDSTONE.

How long until the next “big one”?

Piech & Cain, CS109, Stanford University

# Review

# Binomial Random Variable

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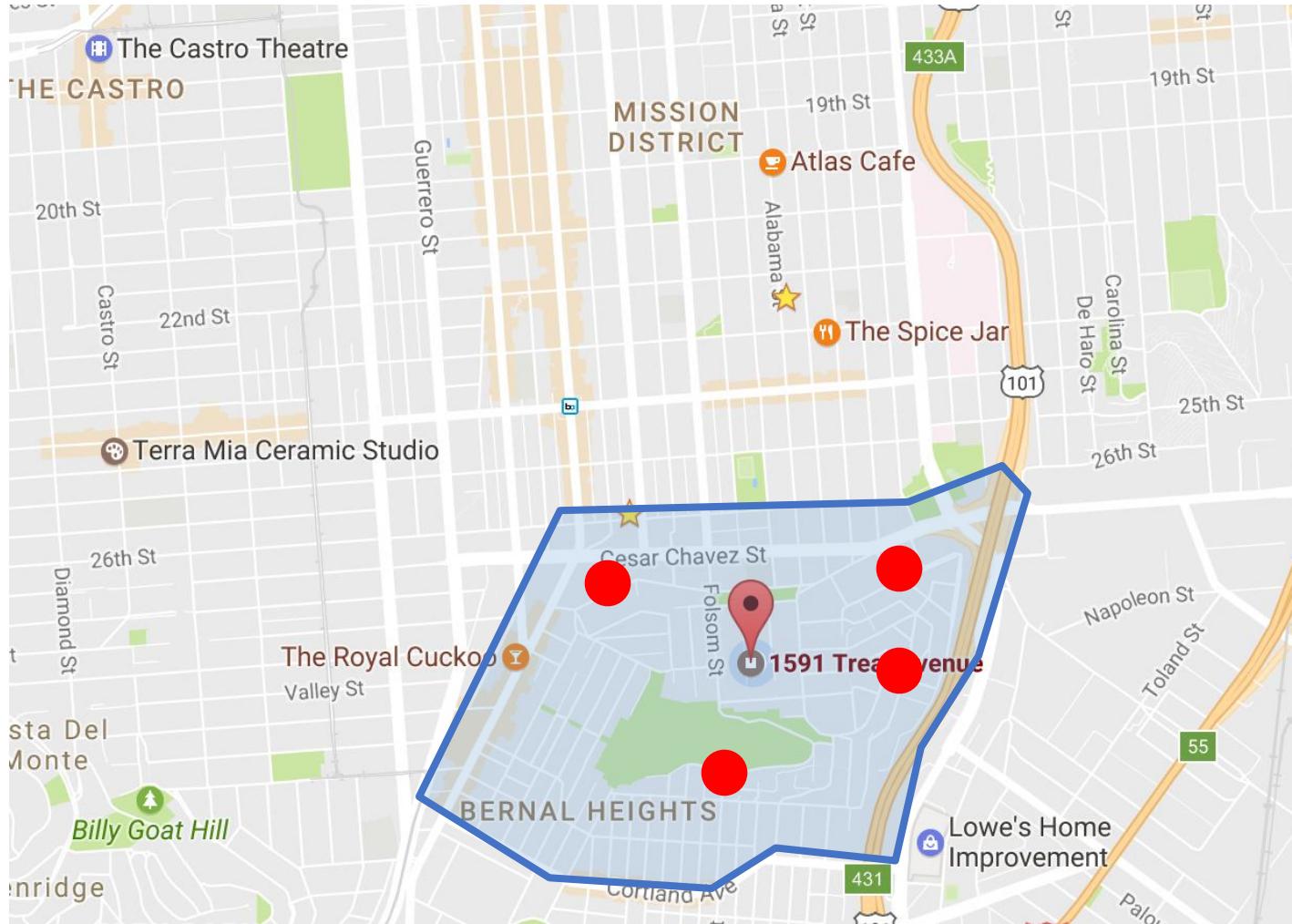
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(H, H, T, T, H, T, T, T, H, T)  
(H, H, T, T, H, T, T, T, T, H)

The number of **successes**, in  $n$  independent **trials**, where each **trial** is a **success** with probability  $p$ :



# Poisson Random Variable

Probability of ***k*** requests from this area in the next 1 min



# Poisson Random Variable

## Poisson Random Variable

**Notation:**  $X \sim \text{Poi}(\lambda)$

**Description:** Number of events in a fixed time frame if (a) the events occur with a constant mean rate and (b) they occur independently of time since last event.

**Parameters:**  $\lambda \in \{0, 1, \dots\}$ , the constant average rate.

**Support:**  $x \in \{0, 1, \dots\}$

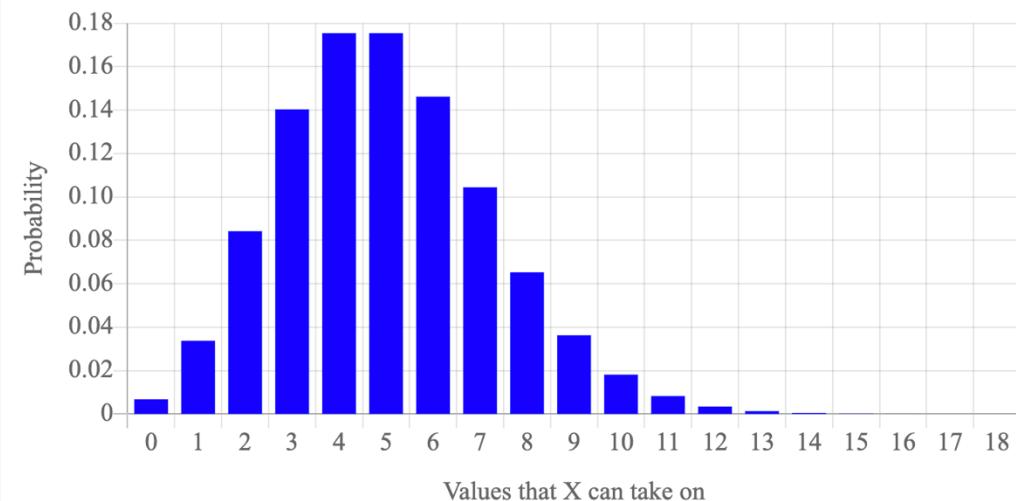
**PMF equation:**  $\Pr(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$

**Expectation:**  $E[X] = \lambda$

**Variance:**  $\text{Var}(X) = \lambda$

**PMF graph:**

Parameter  $\lambda$ :



# Discrete Random Variables

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$$X \sim \text{Bern}(p)$$

Successes in one trial

$$X \sim \text{Geo}(p)$$

Trials until one success

$$X \sim \text{Poi}(\lambda)$$

Events in one time interval

$$Y \sim \text{Bin}(n, p)$$

Successes in  $n$  trials

$$Y \sim \text{NegBin}(r, p)$$

Trials until  $r$  success



# Goal: Be Able to Use a New Random Variable

---

Can you use a novel random variable type if I give you its “fact sheet”?



# Goal: Be Able to Use a New Random Variable

You are learning about servers...



You read about the MD1 queue...

You find a paper that says the length of a server “busy period” is distributed as a Borel with parameter  $\mu = 0.2 \dots$

The screenshot shows the Wikipedia page for the Borel distribution. The page header is "Borel distribution". The content includes a summary, a table of parameters, and a detailed explanation of the distribution's properties and interpretation. A blue oval highlights the table of parameters. The table is as follows:

Parameters	$\mu \in [0, 1]$
Support	$n \in \{1, 2, 3, \dots\}$
pmf	$\frac{e^{-\mu n} (\mu n)^{n-1}}{n!}$
Mean	$\frac{1}{1 - \mu}$
Variance	$\frac{\mu}{(1 - \mu)^3}$

The page also features a sidebar with navigation links and a "Definition" section with a formula for the probability mass function:

$$P_\mu(n) = \Pr(X = n) = \frac{e^{-\mu n} (\mu n)^{n-1}}{n!}$$

for  $n = 1, 2, 3 \dots$



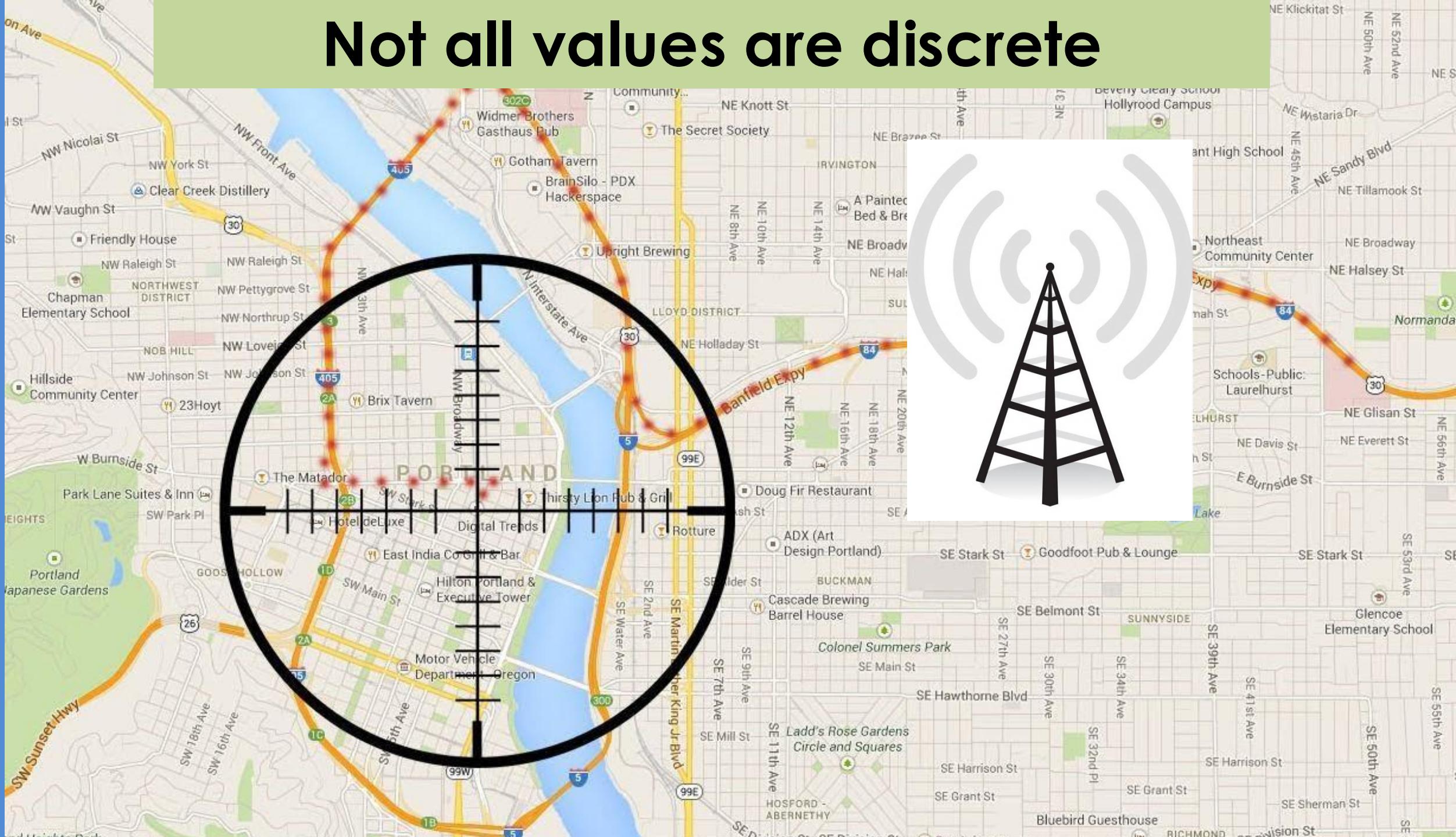
# Learning Goals

1. Integrate a density function (PDF) to get a probability
2. Use a cumulative function (CDF) to get a probability



Big hole in our knowledge

# Not all values are discrete



# Can't Talk About Continuous Values

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Say the average rate of earthquakes is 1 every 100 years.

We **can** talk about the probability distribution of different numbers of earthquakes next year.

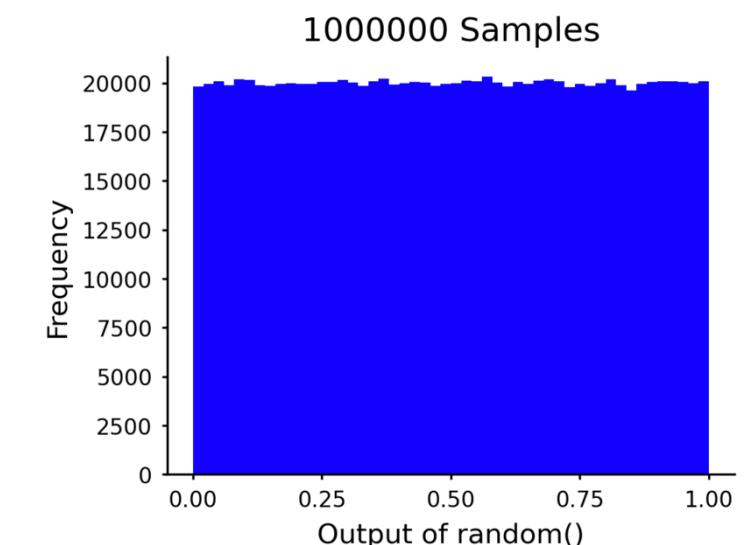
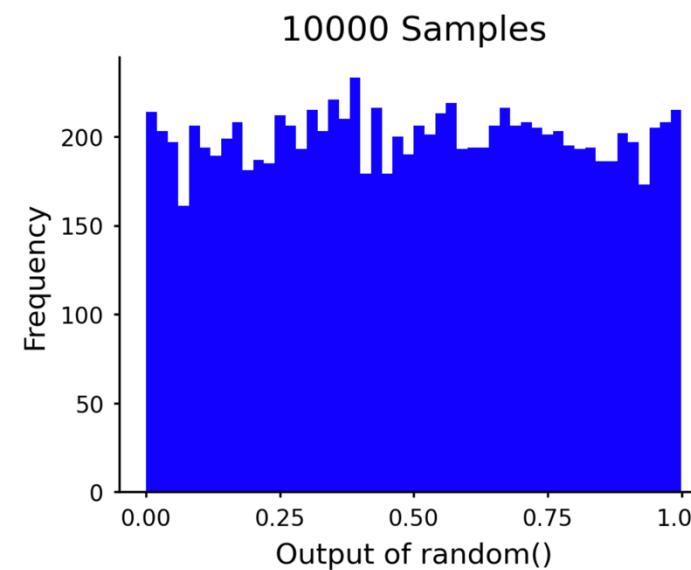
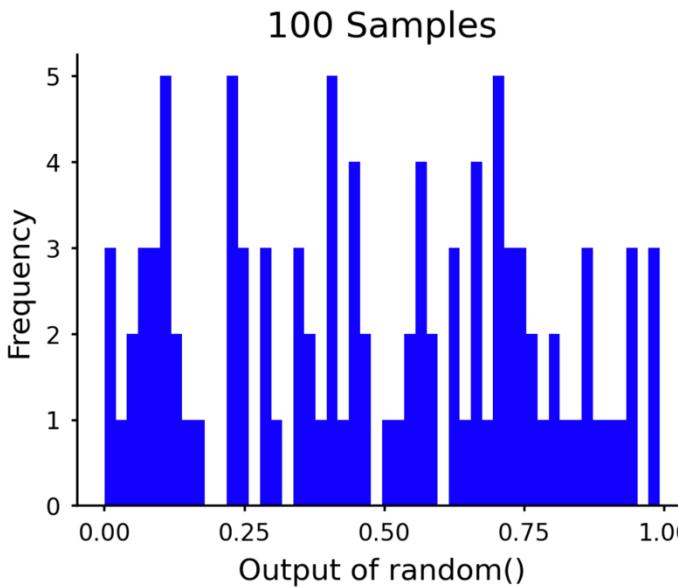
We **can't** talk about the probability distribution of the amount of time until the next earthquake.



random( ) ?

# The `random()` Function

- Outputs values between 0 and 1
- All possible values are equally likely
- This is a continuous random variable!



```
import random

samples_small = []
for i in range(100):
    samples_small.append(random.random())

samples_medium = []
for i in range(10000):
    samples_medium.append(random.random())

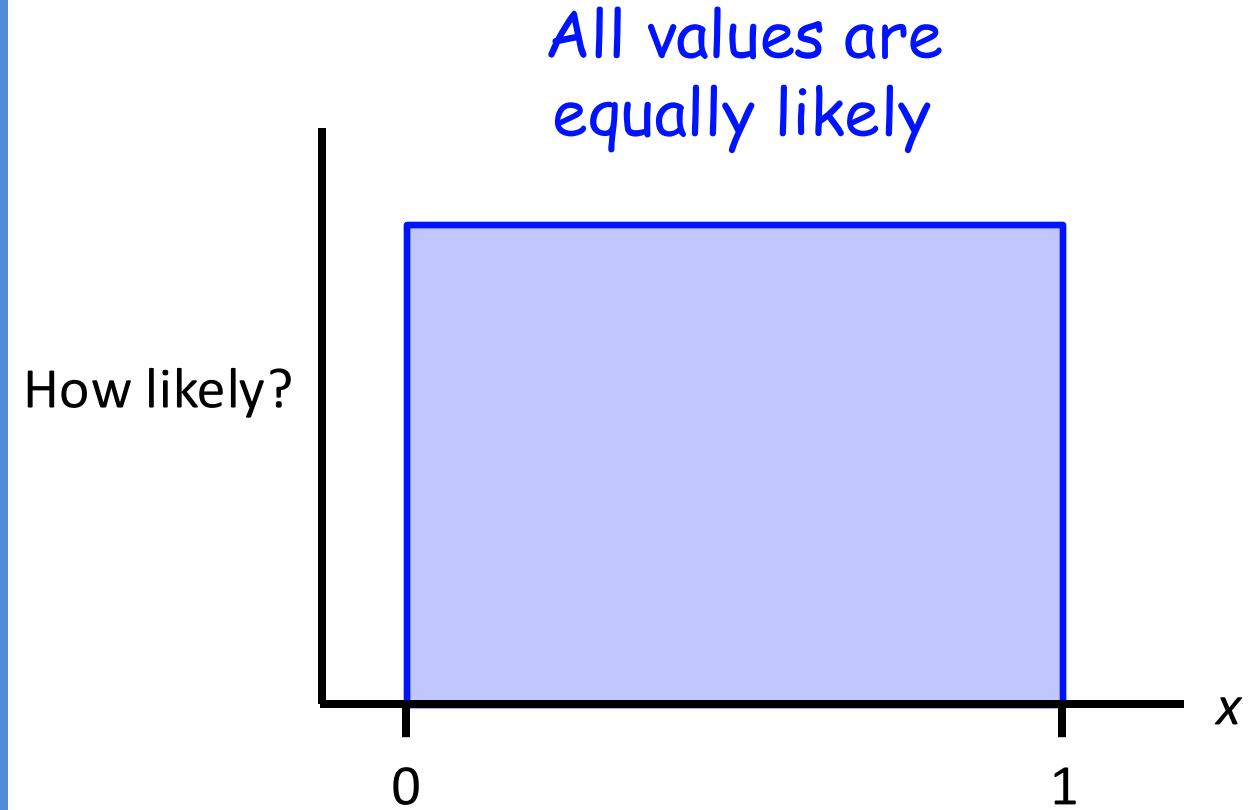
samples_large = []
for i in range(1000000):
    samples_large.append(random.random())
```

# $X \sim \text{Uniform}(0,1)$ : A Continuous Random Variable

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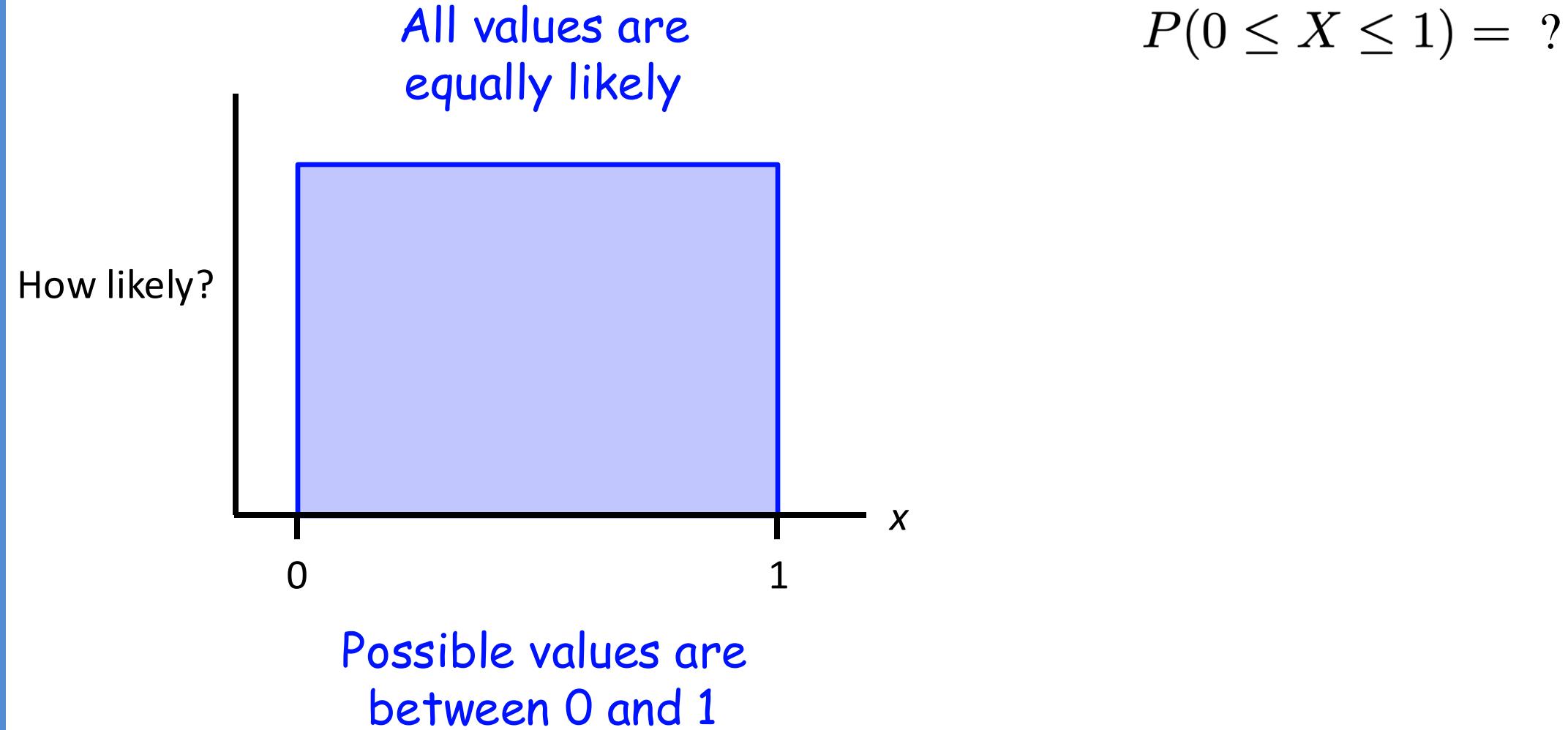
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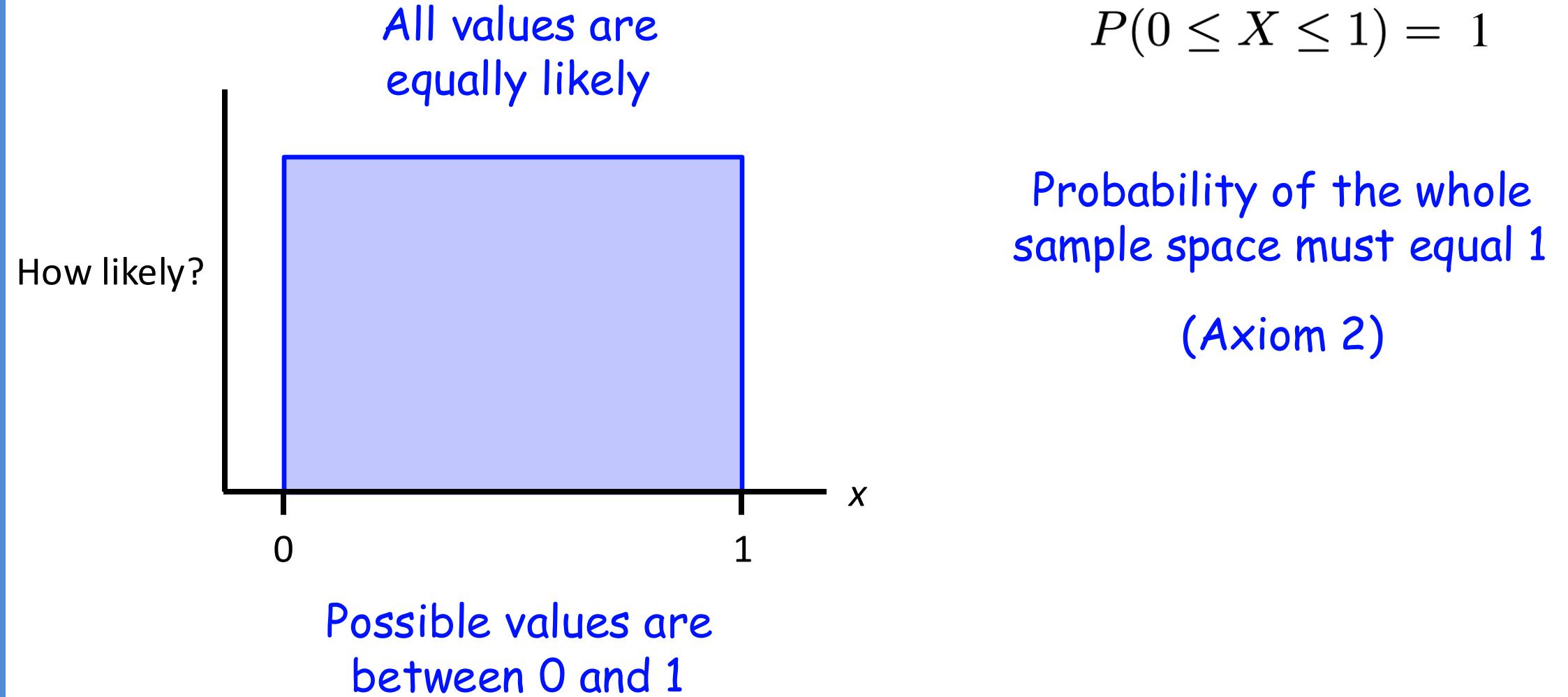
Possible values are  
between 0 and 1



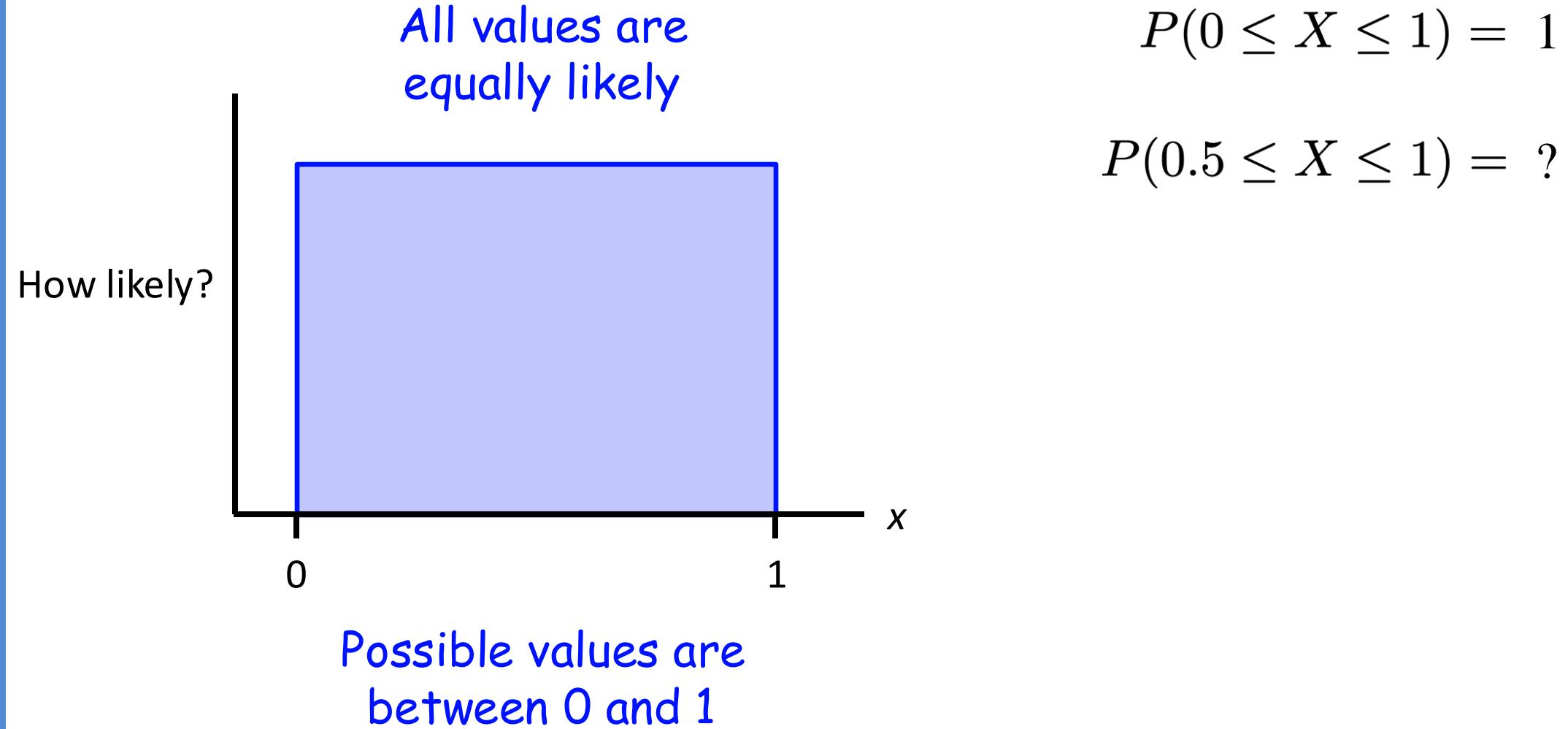
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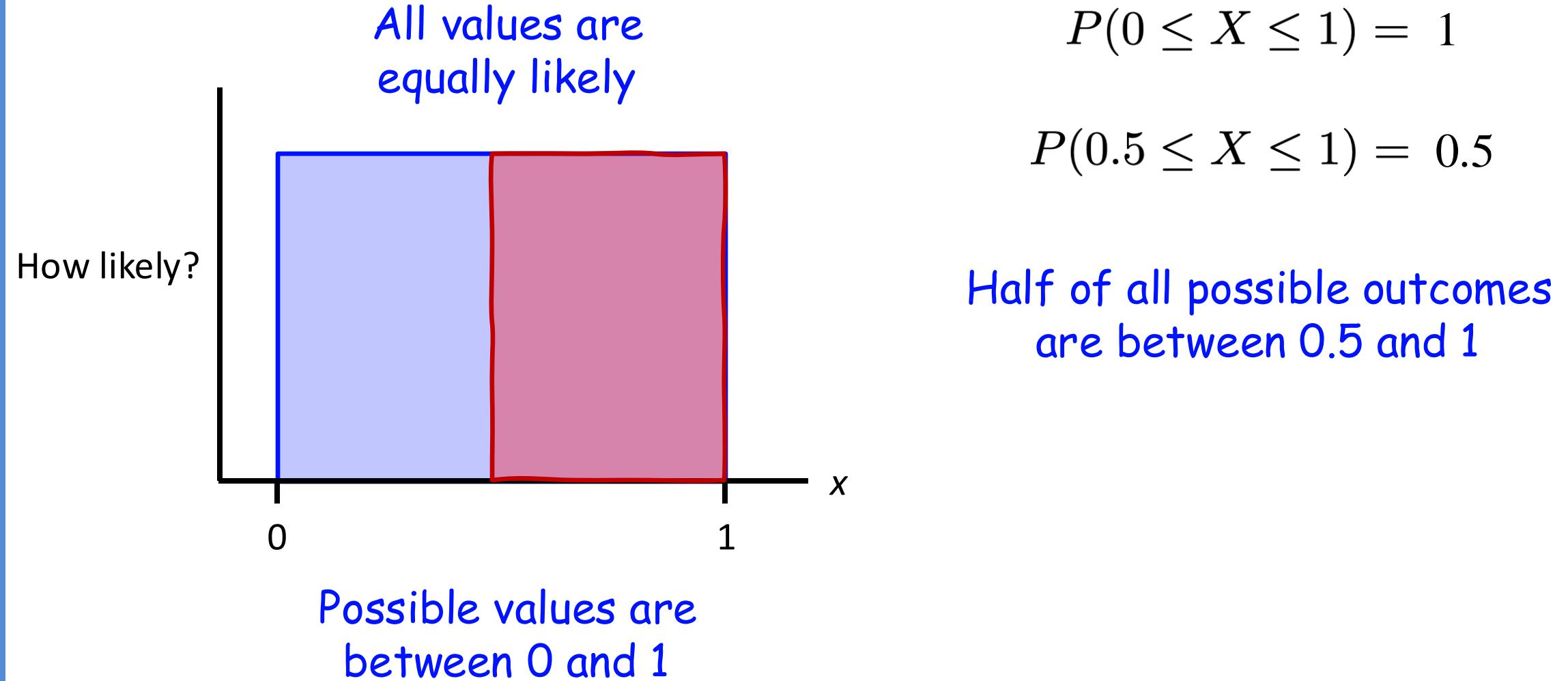
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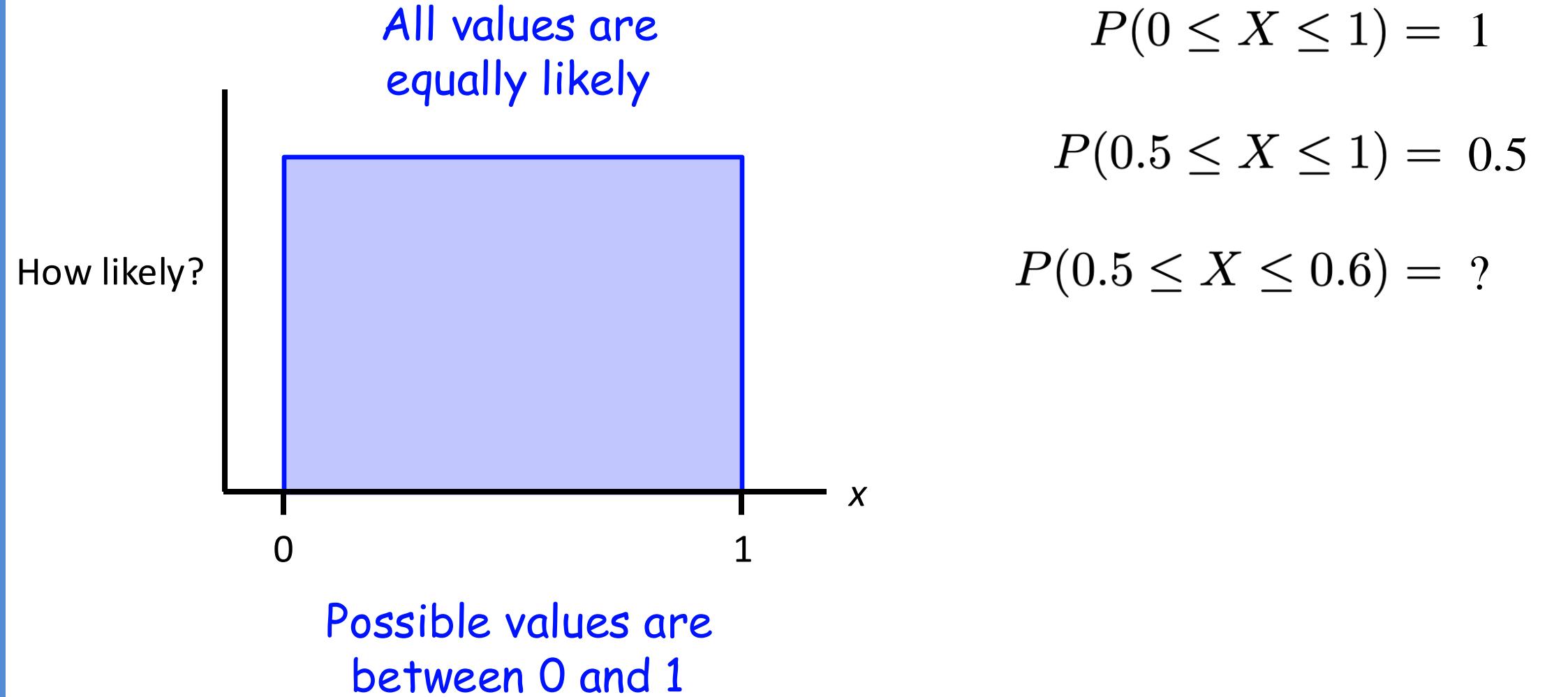
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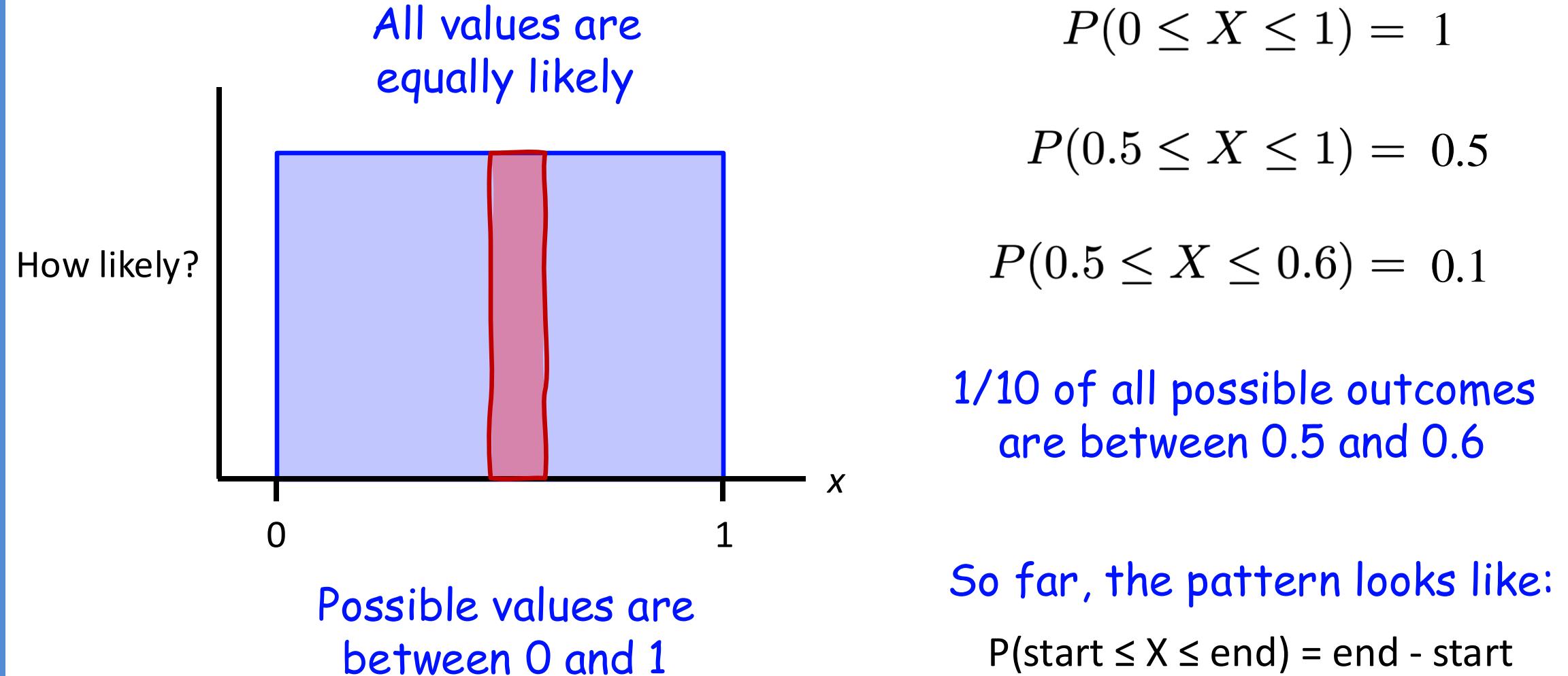
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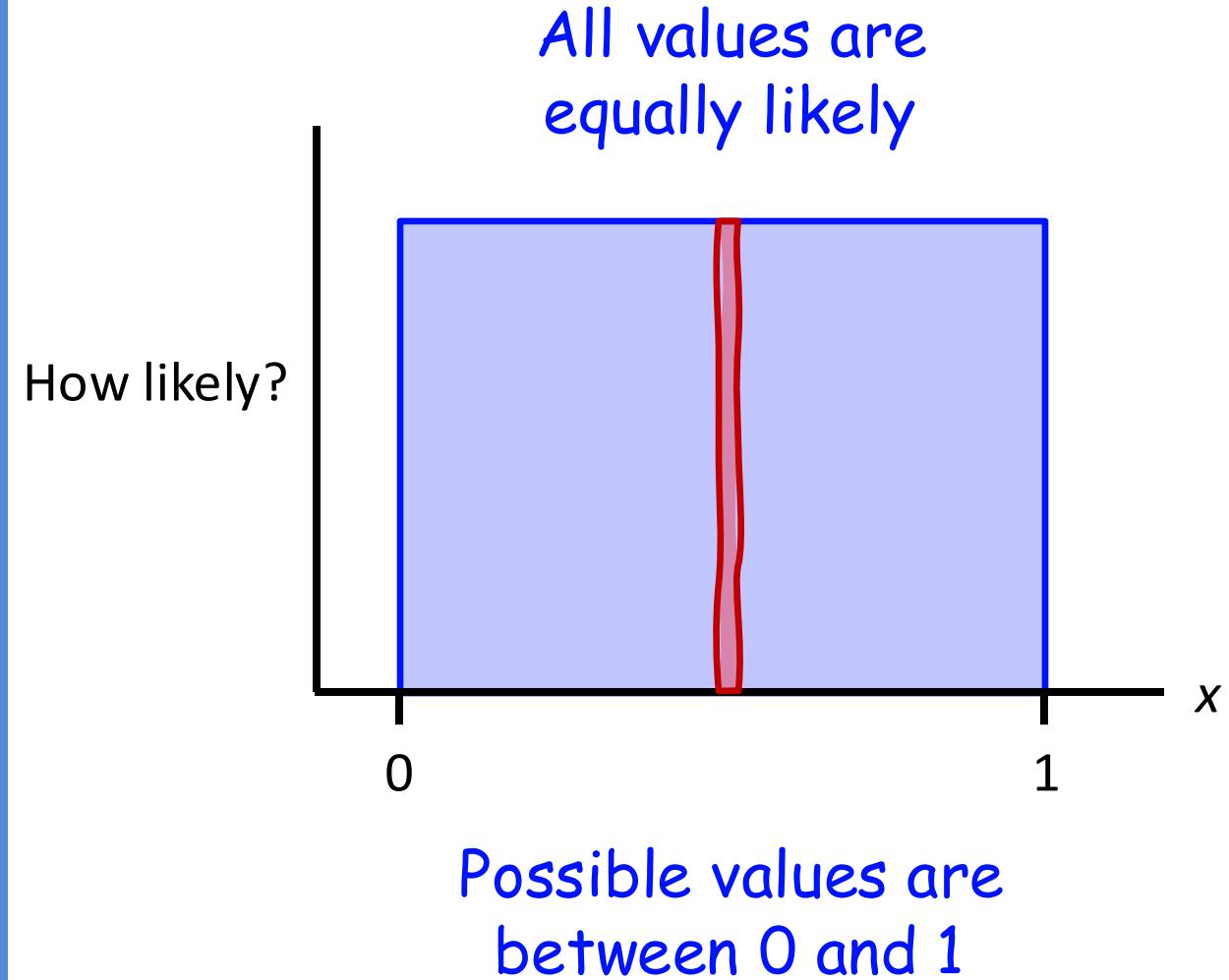
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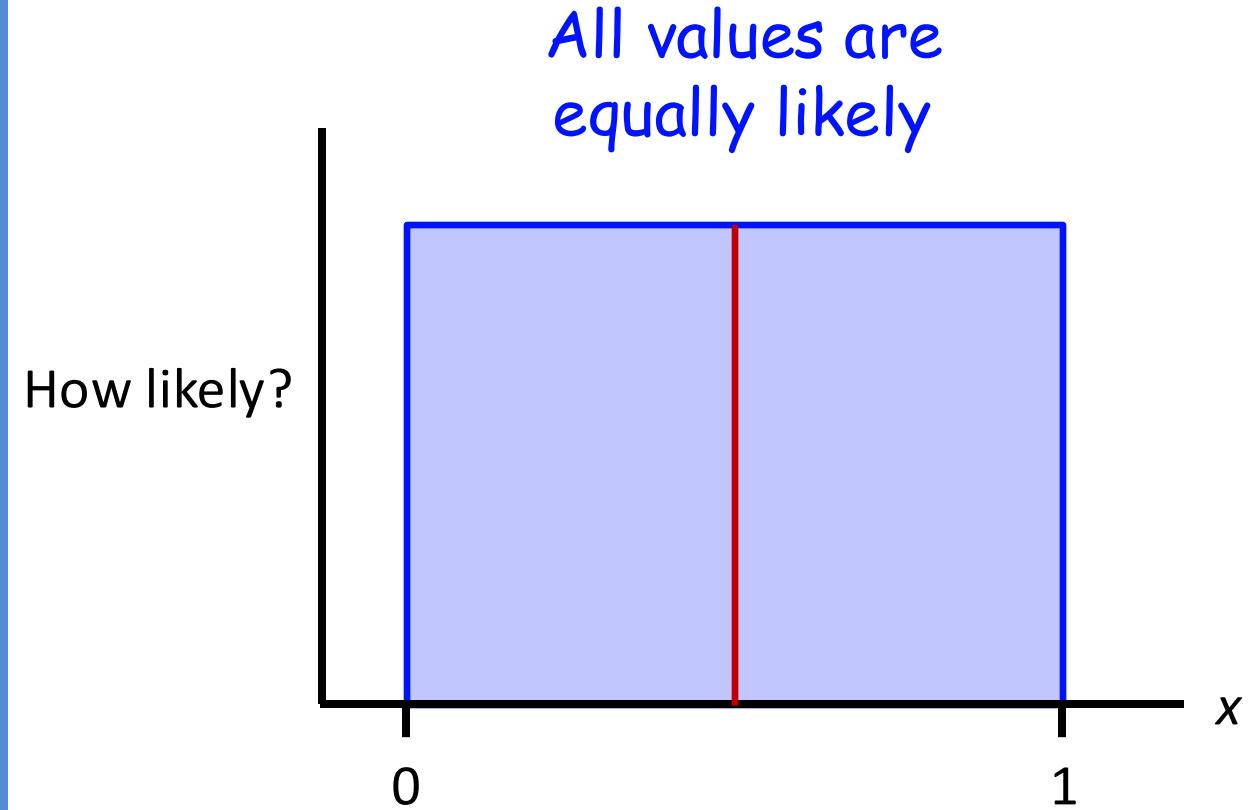


$$P(0 \leq X \leq 1) = 1$$
$$P(0.5 \leq X \leq 1) = 0.5$$
$$P(0.5 \leq X \leq 0.6) = 0.1$$
$$P(0.5 \leq X \leq 0.5001) = 0.0001$$

As we get more precise,  
probabilities keep shrinking...



# $X \sim \text{Uniform}(0,1)$ : A Continuous Random Variable



All values are  
equally likely

Possible values are  
between 0 and 1

$$P(0 \leq X \leq 1) = 1$$

$$P(0.5 \leq X \leq 1) = 0.5$$

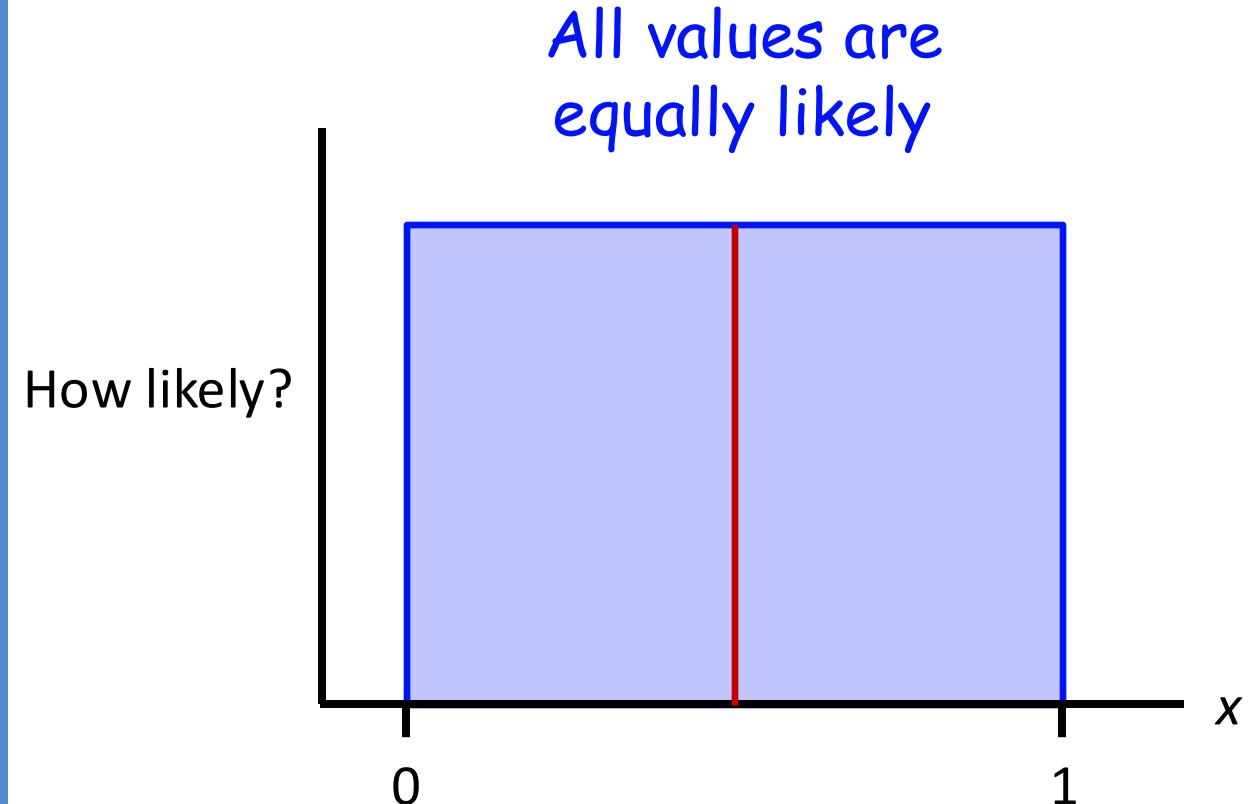
$$P(0.5 \leq X \leq 0.6) = 0.1$$

$$P(0.5 \leq X \leq 0.5001) = 0.0001$$

$$P(X = 0.5) = ?$$



# $X \sim \text{Uniform}(0,1)$ : A Continuous Random Variable



All values are  
equally likely

Possible values are  
between 0 and 1

$$P(0 \leq X \leq 1) = 1$$

$$P(0.5 \leq X \leq 1) = 0.5$$

$$P(0.5 \leq X \leq 0.6) = 0.1$$

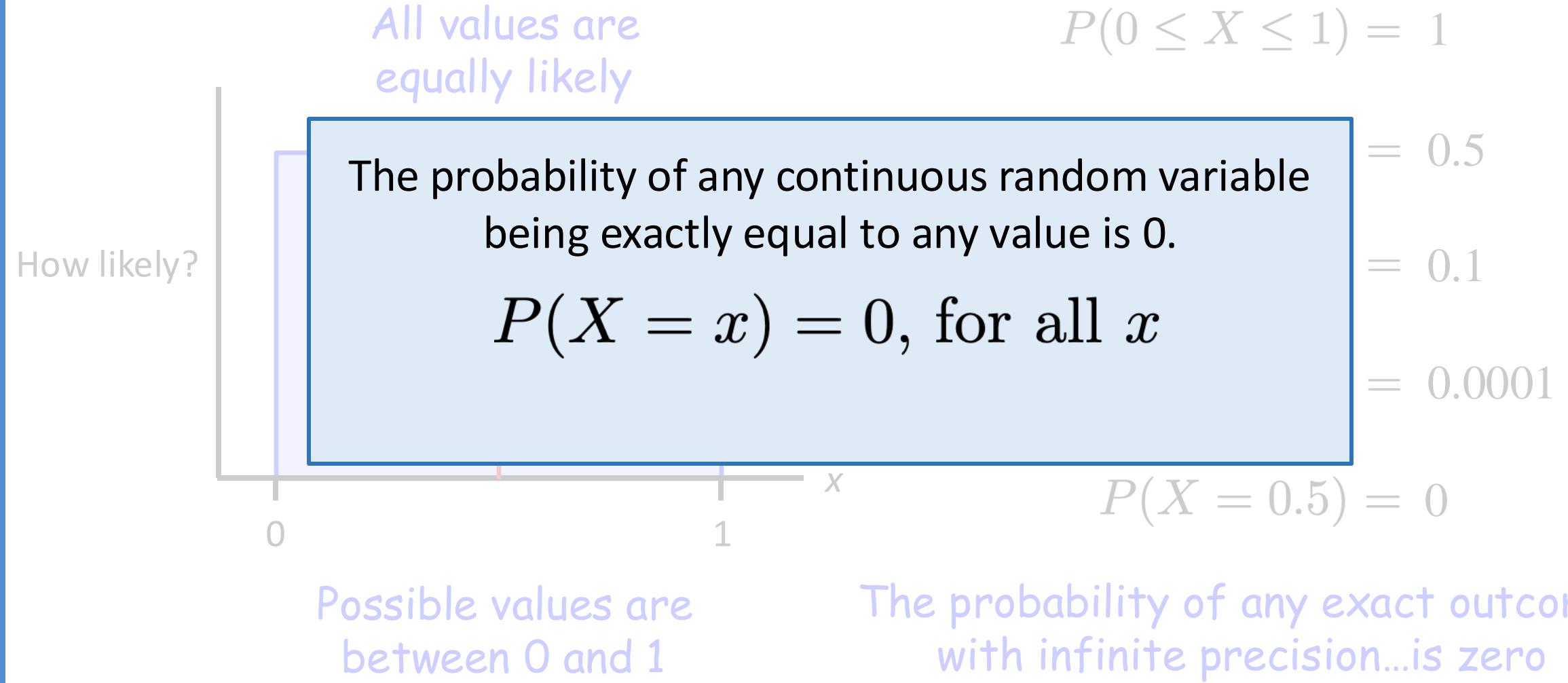
$$P(0.5 \leq X \leq 0.5001) = 0.0001$$

$$P(X = 0.5) = 0$$

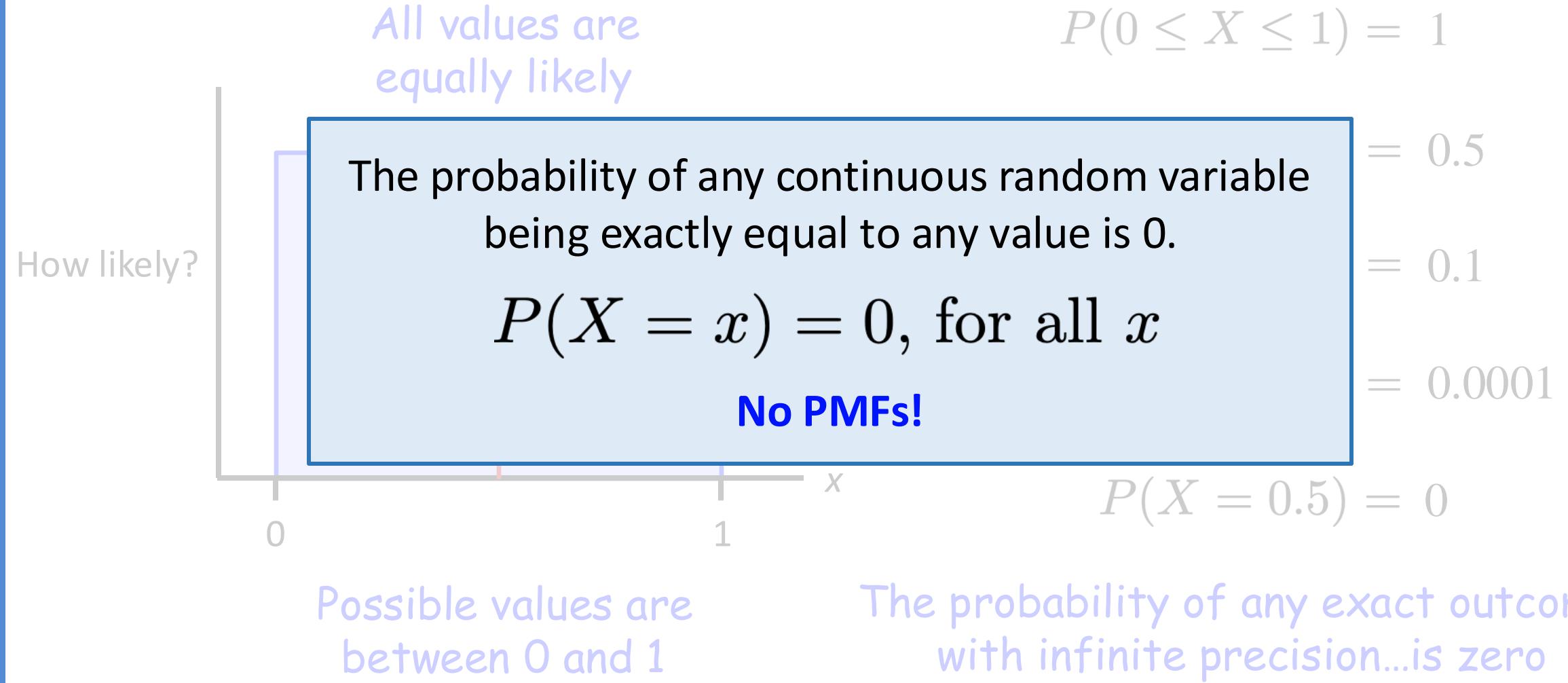
The probability of any exact outcome,  
with infinite precision...is zero



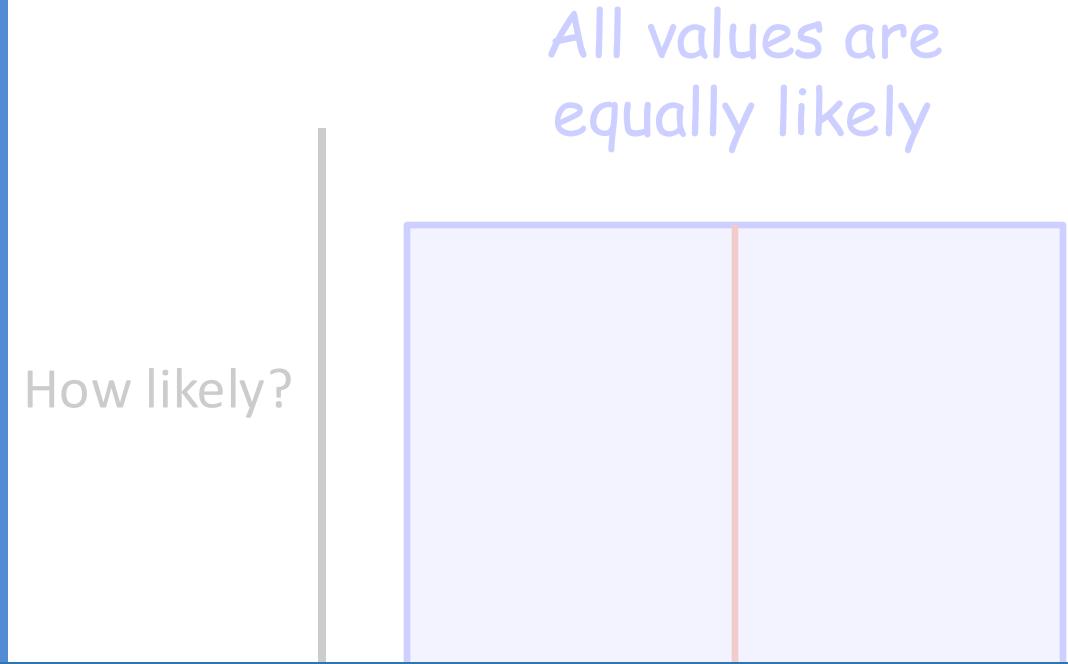
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The only way to talk about probabilities of outcomes for *continuous* random variables is using ranges of possible values.

$$P(0 \leq X \leq 1) = 1$$
$$P(0.5 \leq X \leq 1) = 0.5$$
$$P(0.5 \leq X \leq 0.6) = 0.1$$
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$$P(X = 0.5) = 0$$

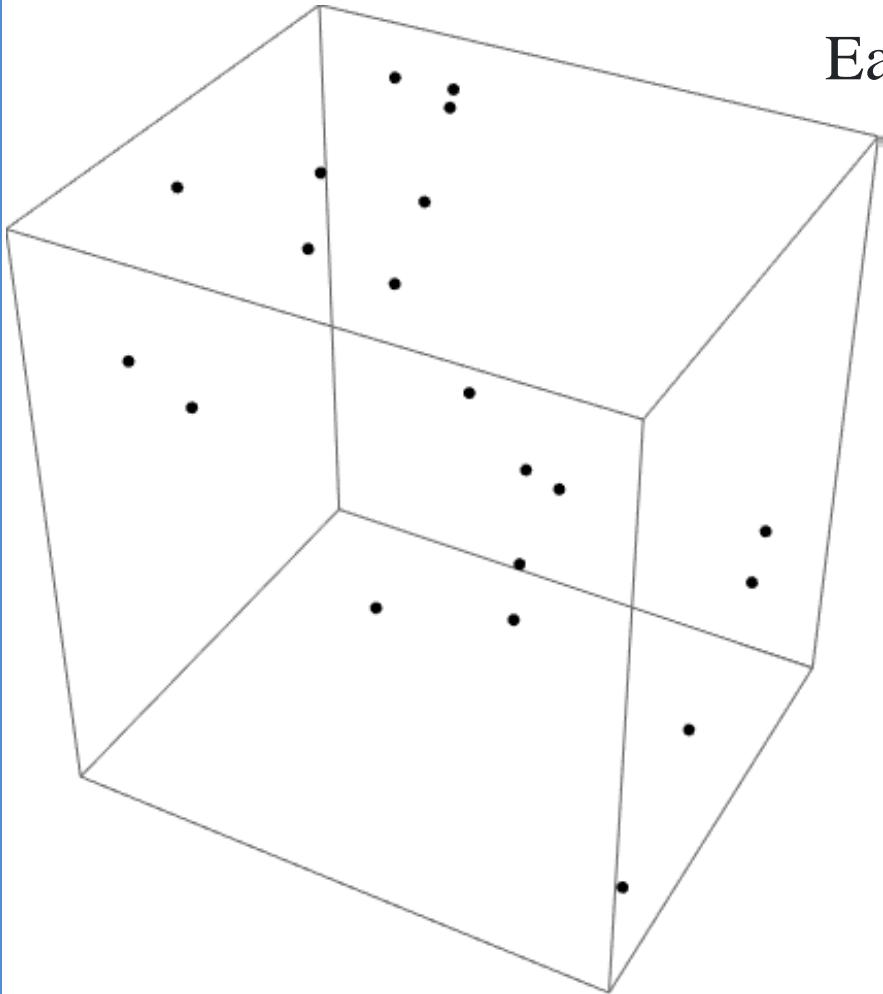
Probability of any exact outcome,  
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# Curse of Dimensionality

A random *point* of dimension  $d$  is a list of  $d$  random values:  $[X_1 \dots X_d]$   
 $X_i \sim \text{Uni}(0, 1)$  for all  $i$

Each value  $X_i$  is independent of other values



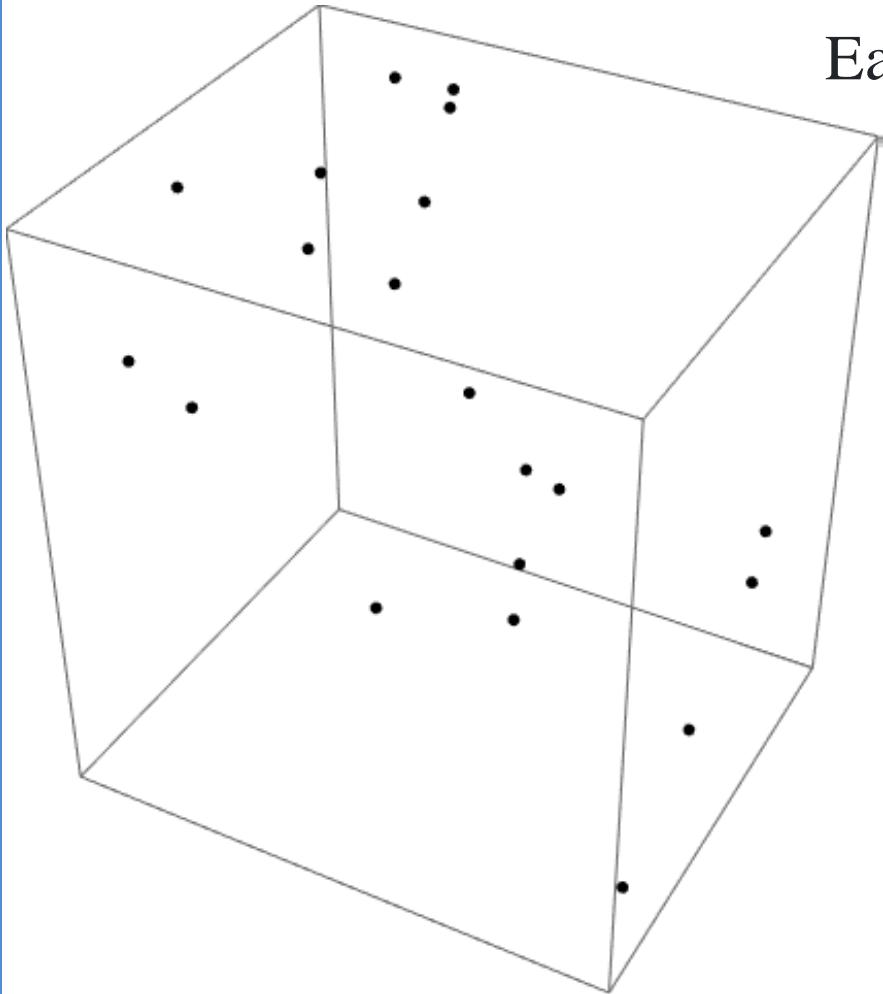
$X_i$  is close to an edge if  $X_i$  is less than 0.01 **or**  $X_i$  is greater than 0.99. What is the probability that  $X_i$  is close to an edge?



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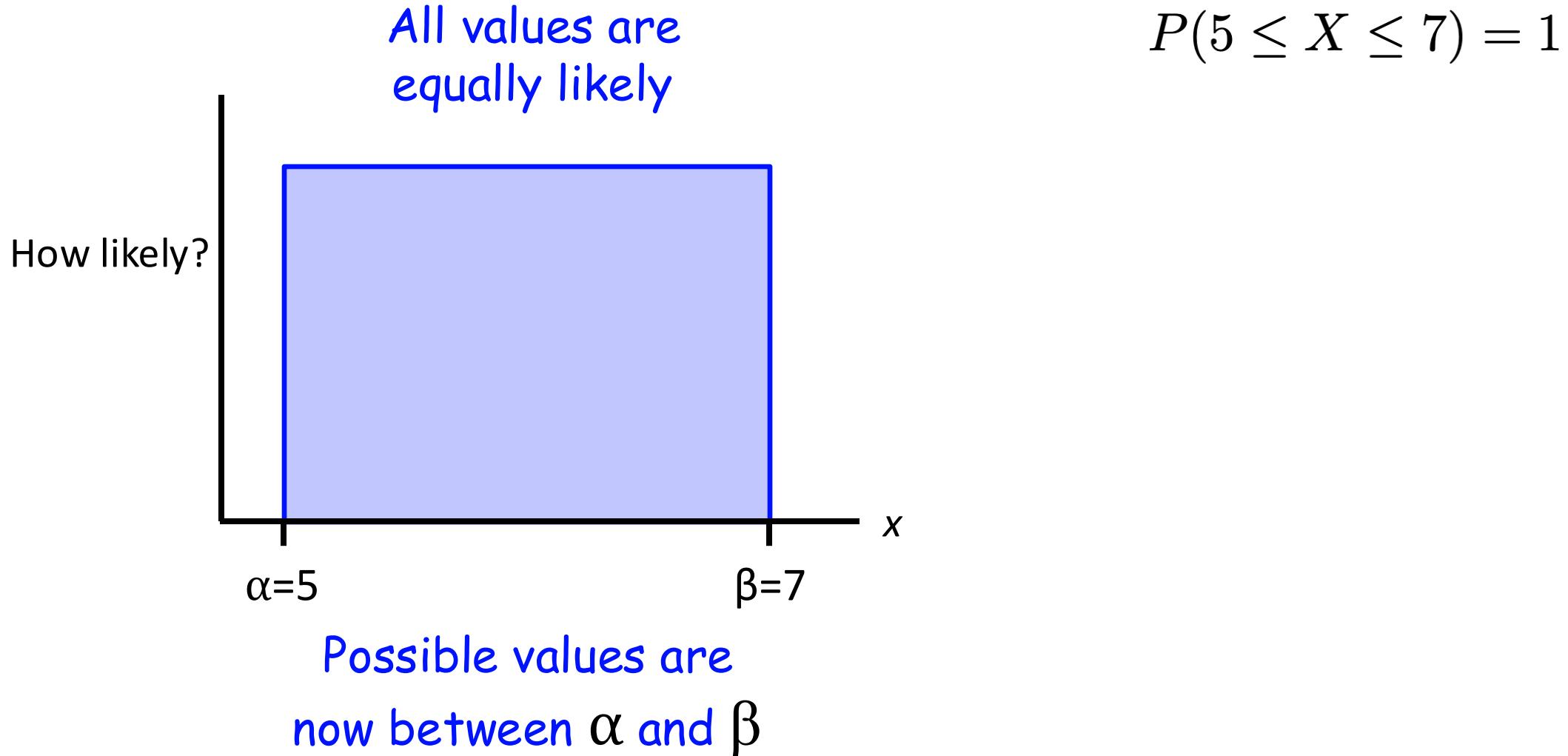
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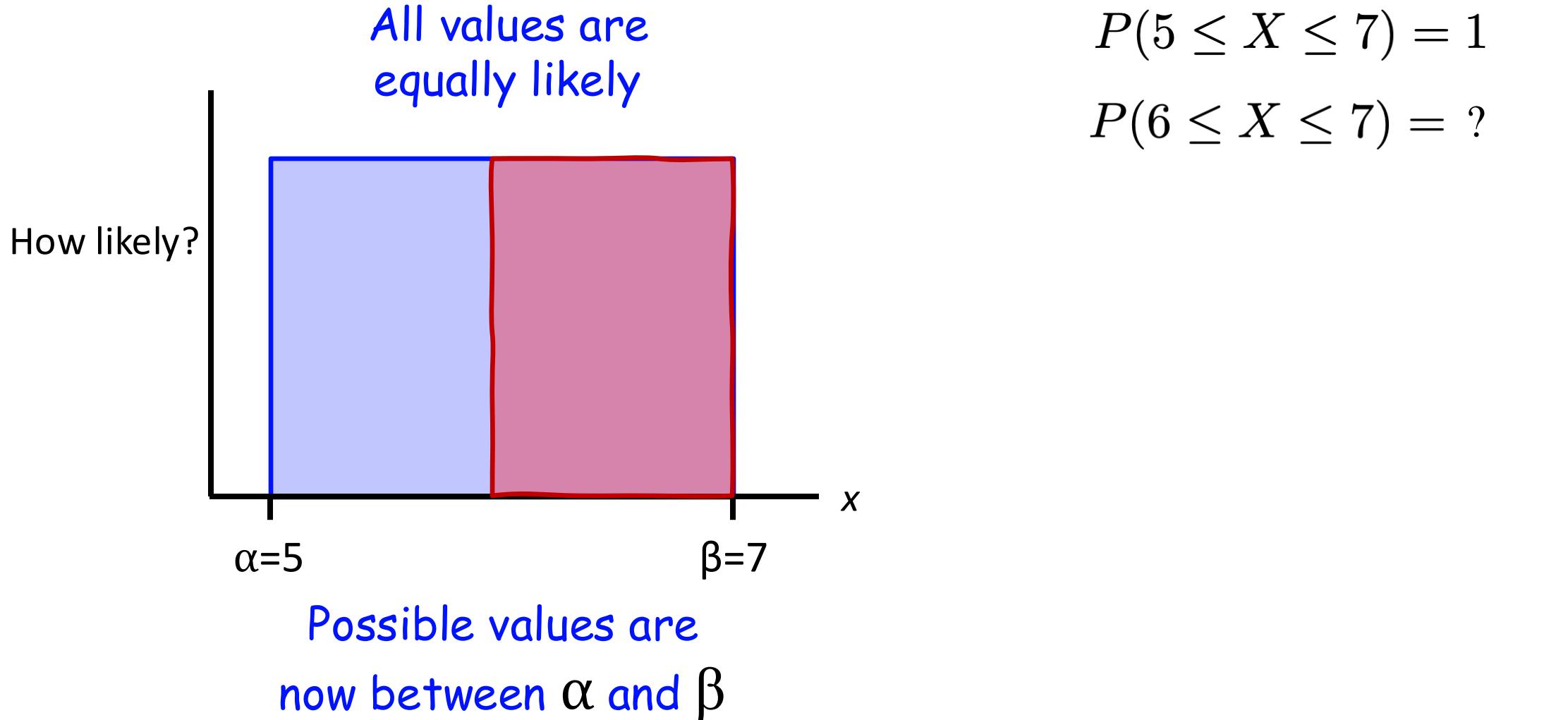
A random *point*  $[X_1 \dots X_{100}]$  of dimension 100 is close to an edge if *any* of its values are close to an edge. What is the probability that a 100 dimensional point is close to an edge?



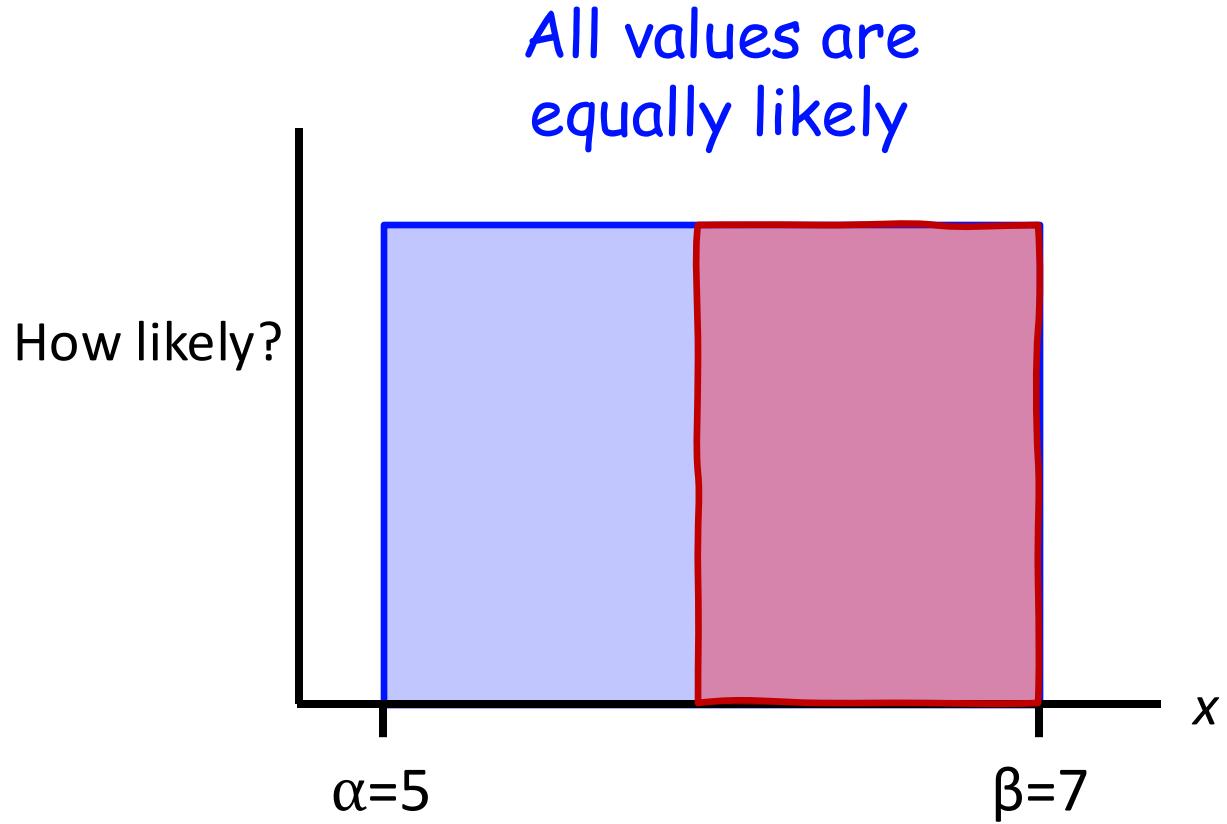
# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



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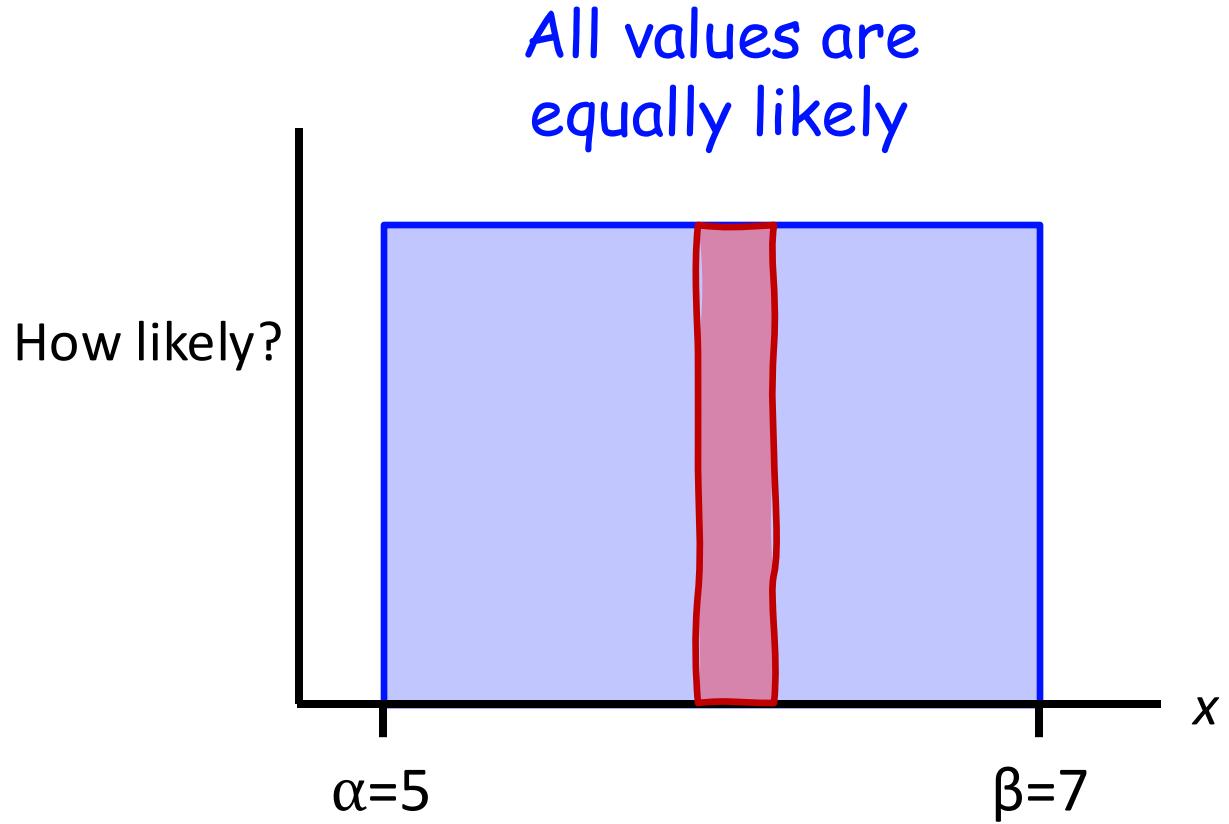
All values are  
equally likely

Possible values are  
now between  $\alpha$  and  $\beta$

$$P(5 \leq X \leq 7) = 1$$
$$P(6 \leq X \leq 7) = 0.5$$



# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



Possible values are  
now between  $\alpha$  and  $\beta$

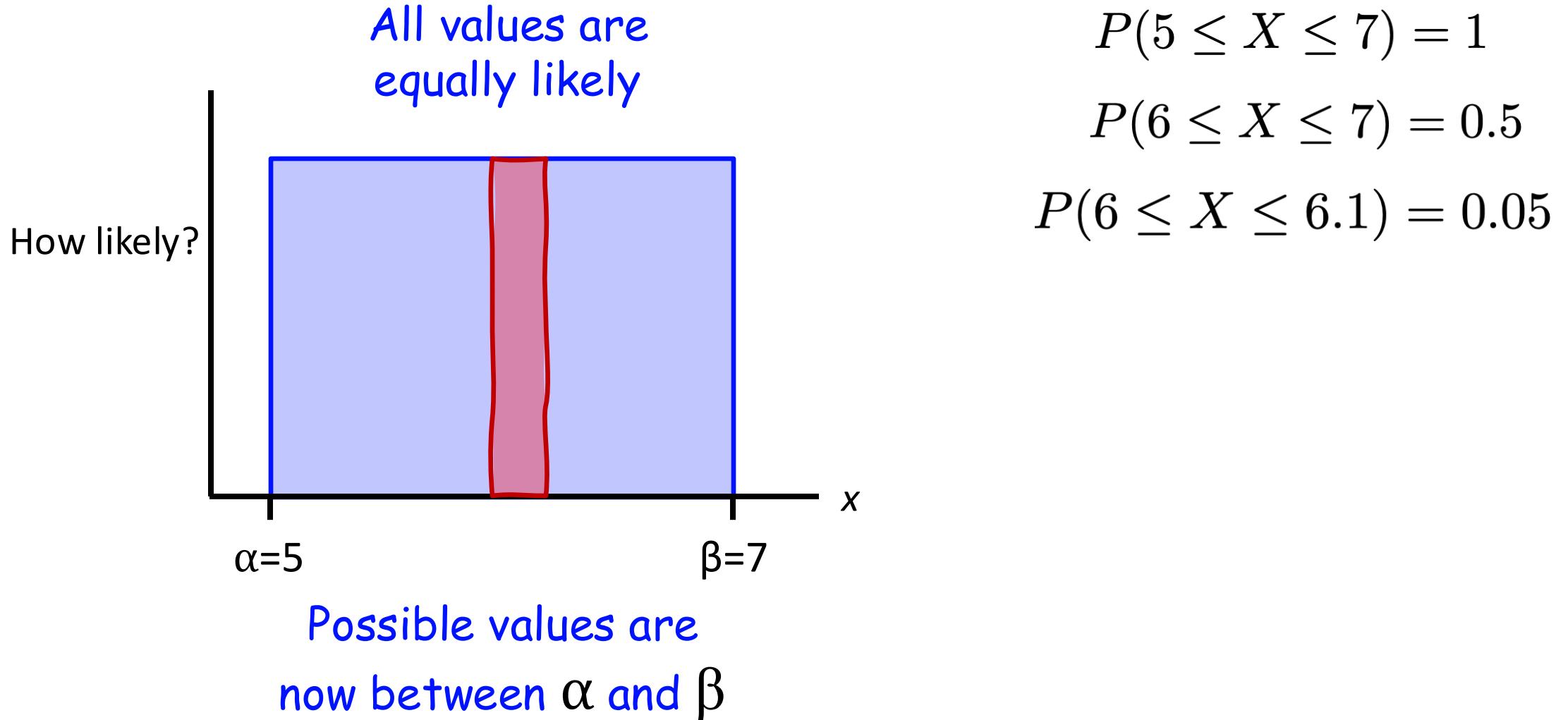
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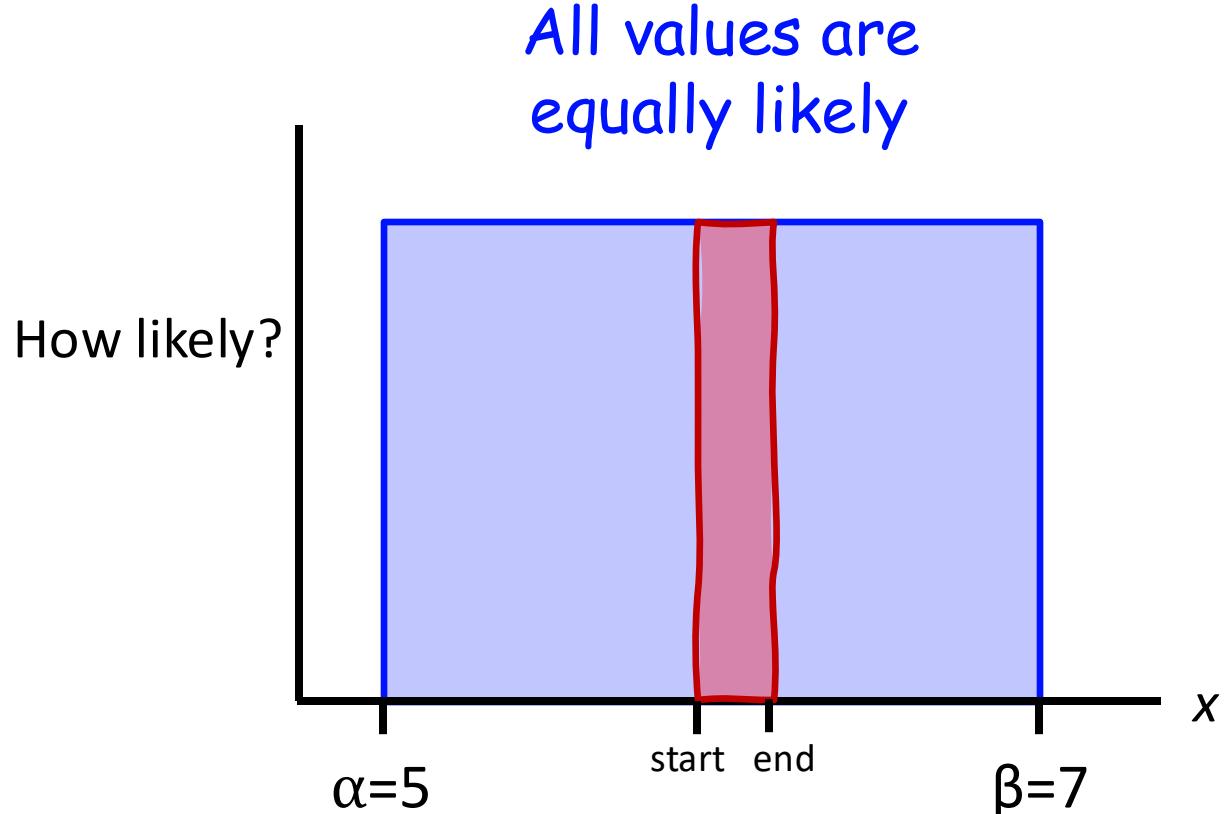
$$P(6 \leq X \leq 6.1) = ?$$



# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



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Possible values are now between  $\alpha$  and  $\beta$

$$P(5 \leq X \leq 7) = 1$$

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$$P(6 \leq X \leq 6.1) = 0.05$$

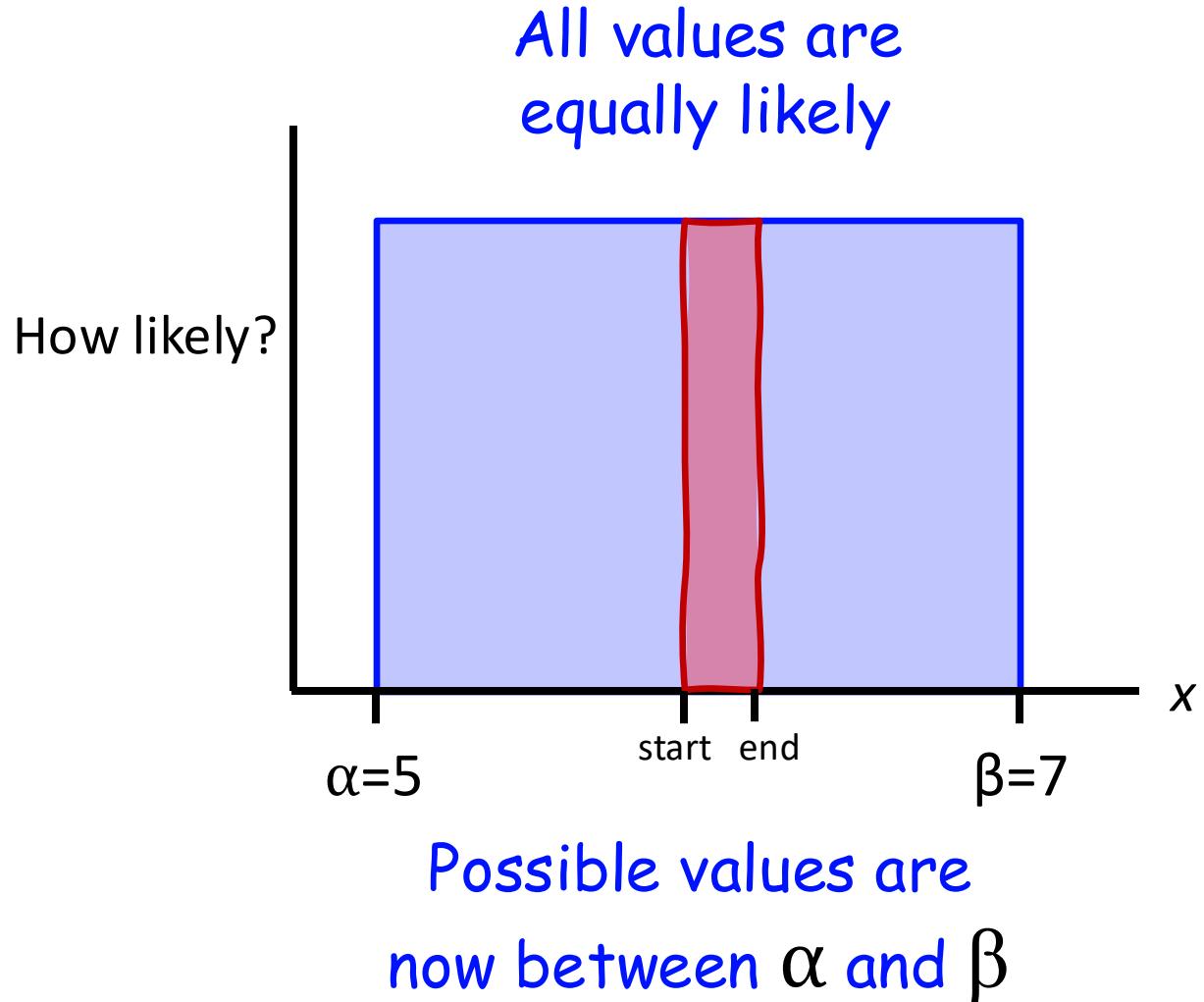
For Uniform(0,1):

$$P(\text{start} \leq X \leq \text{end}) = \text{end} - \text{start}$$

Does that still work?



# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



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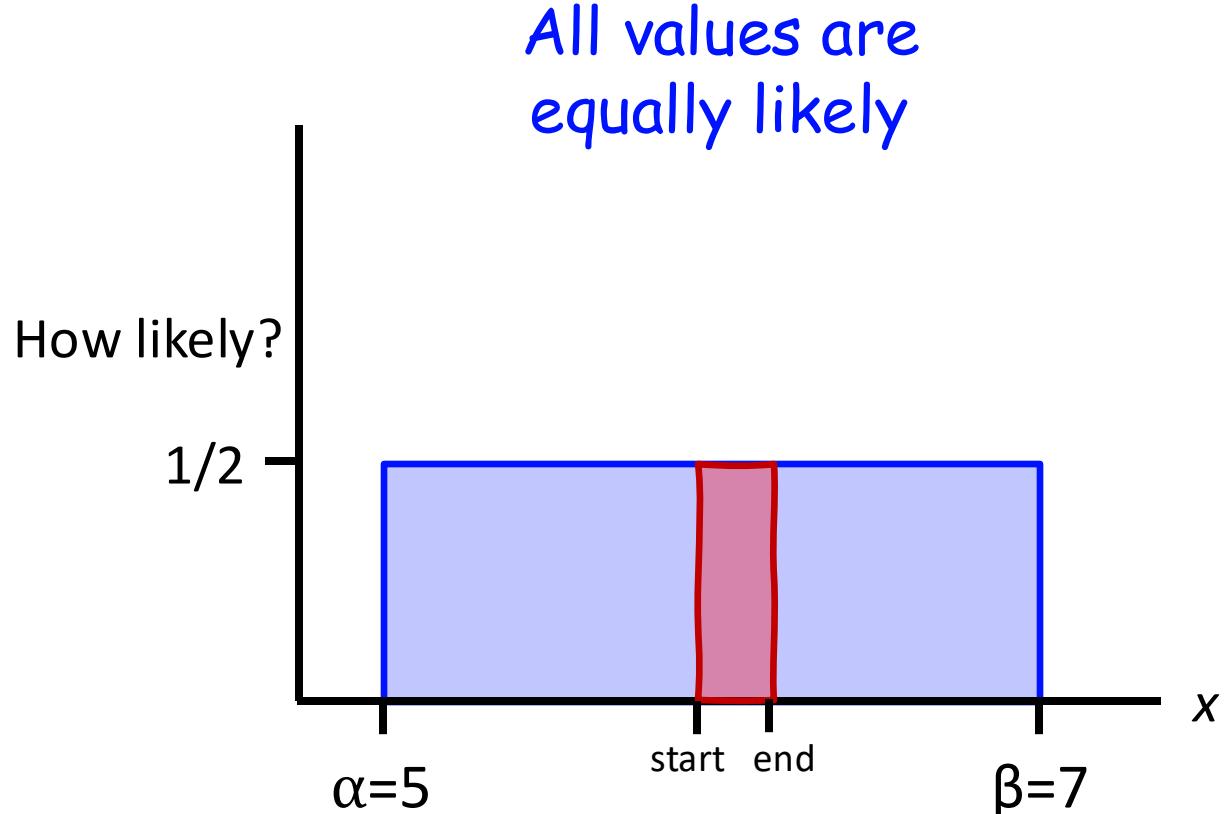
$$P(\text{start} \leq X \leq \text{end}) = \text{end} - \text{start}$$

Does that still work? No!

Need to divide by 2?



# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



Possible values are now between  $\alpha$  and  $\beta$

$$P(5 \leq X \leq 7) = 1$$

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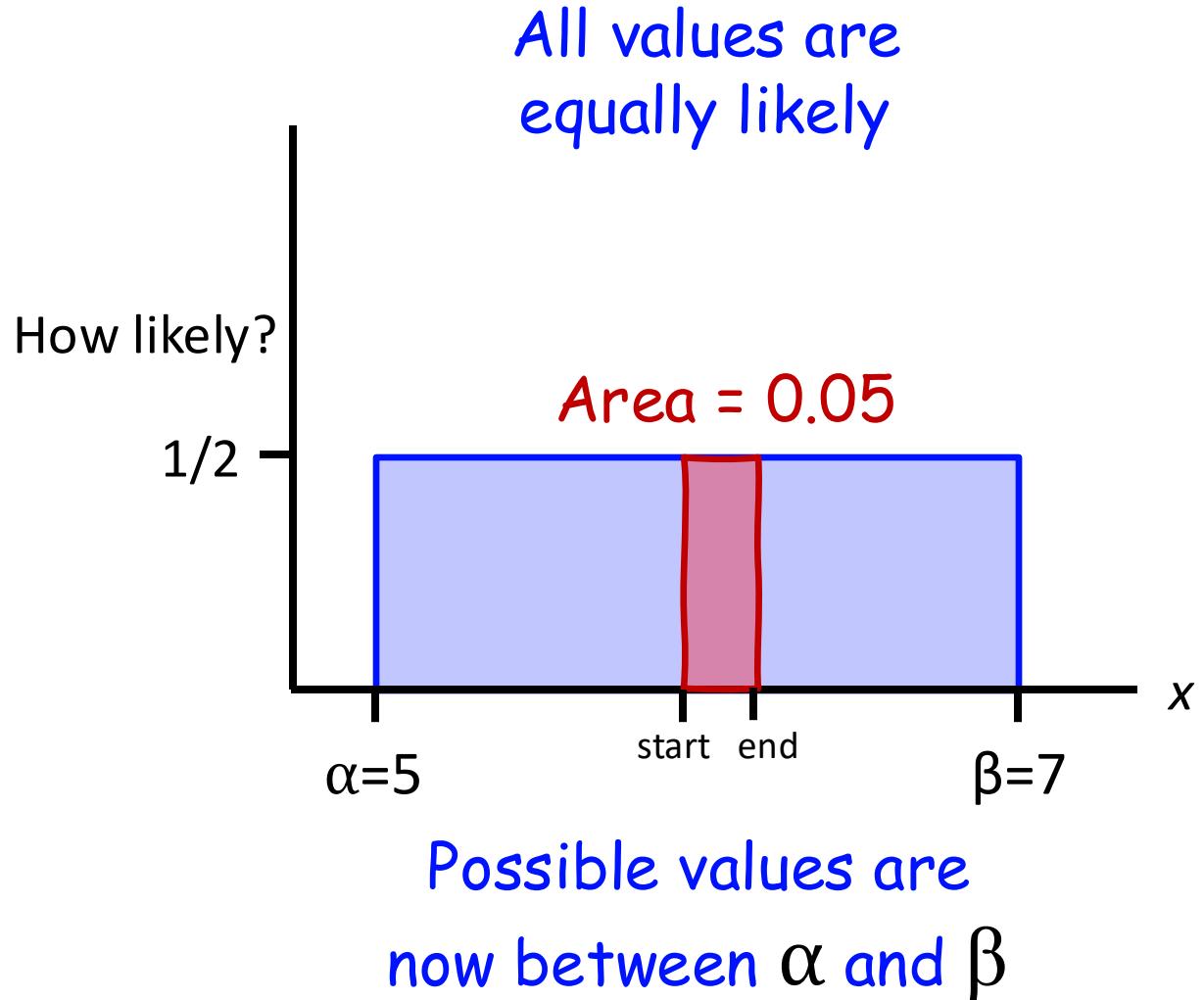
$$P(6 \leq X \leq 6.1) = 0.05$$

**For Uniform( $\alpha, \beta$ ):**

$$P(\text{start} \leq X \leq \text{end}) = \frac{\text{end} - \text{start}}{\beta - \alpha}$$



# $X \sim \text{Uniform}(\alpha, \beta)$ : More General Case



$$P(5 \leq X \leq 7) = 1$$

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For Uniform( $\alpha, \beta$ ):

$$P(\text{start} \leq X \leq \text{end}) = \frac{\text{end} - \text{start}}{\beta - \alpha}$$

If we set  $y = 1/2$  between  $\alpha$  and  $\beta$ , then probabilities are "slices of the whole box"



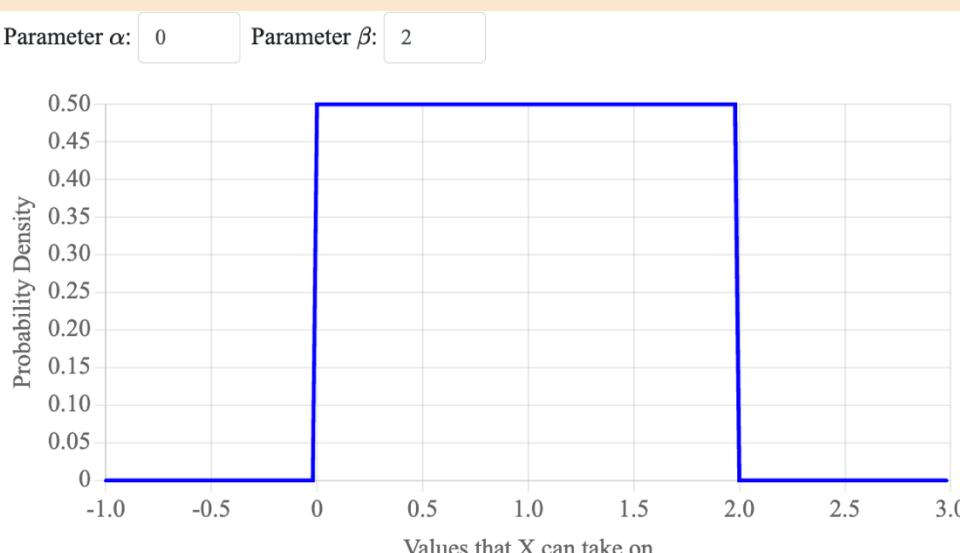
# Uniform Random Variable

A **Uniform** random variable  $X$  takes on a value, with equal likelihood between  $\alpha$  and  $\beta$ .

$$X \sim \text{Uni}(\alpha, \beta)$$

$$P(x_1 < X < x_2) = \frac{x_2 - x_1}{\beta - \alpha}$$

Support:  $[\alpha, \beta]$



Examples:

- Result of python `random()`
- Random points

Can we generalize to other continuous  
random variables?

# Riding the Marguerite



# Riding the Margueritte

---



You're running to the bus stop. You don't know exactly when the bus arrives.

You have a probability distribution for bus arrival times -- some times are more likely than others.

You show up at 2:15pm. What is  $P(\text{wait} < 5 \text{ min})$ ?



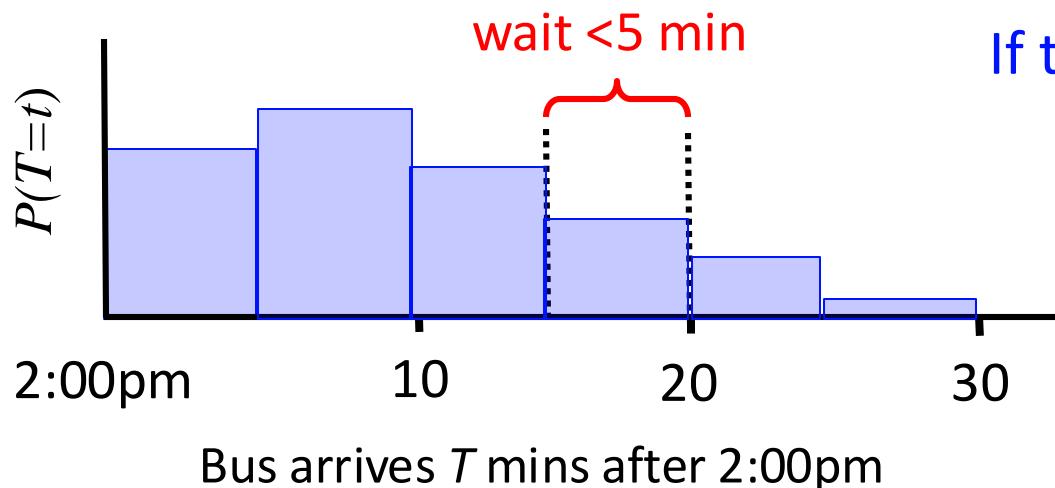
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If time was discrete: a PMF could look like this.



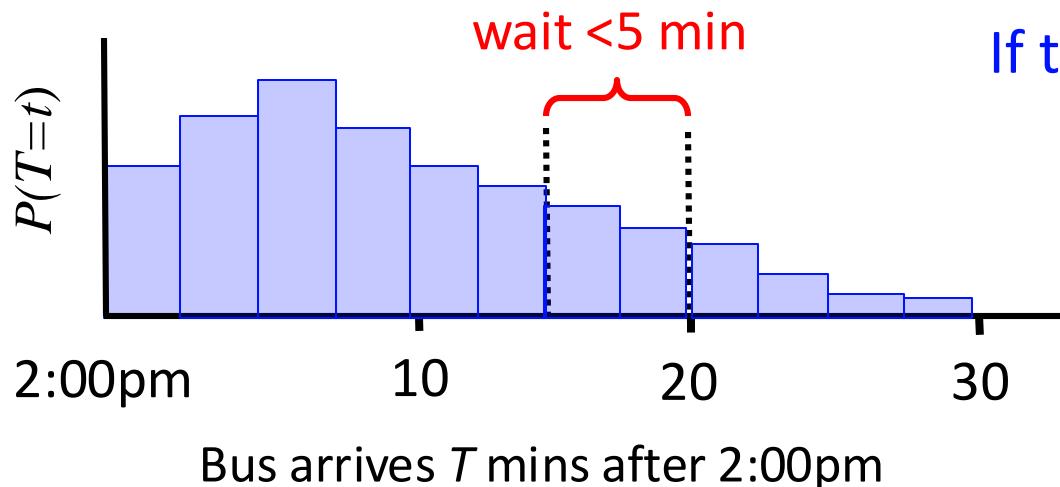
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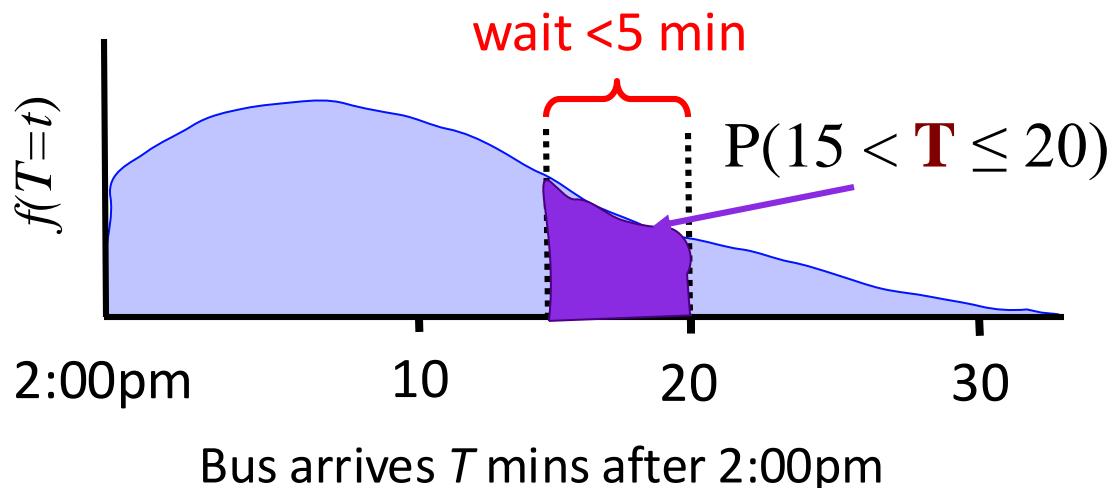
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When interval sizes tend towards 0:

- Time is now a **continuous** variable
- The **probability mass function (PMF)** becomes a **derivative** called a **probability density function (PDF)**
- Probability are now calculated as **area under the curve**



# Time For Integrals!!!!



# Probability Density Function

---



The **probability density function** (PDF) of a continuous random variable represents the relative likelihood of various values.

Units of probability *divided by units of X*.  
**Integrate it** to get probabilities!

$$P(a < X < b) = \int_{x=a}^b f(X = x) dx$$



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$$P(a < X < b) = \int_{x=a}^b f_X(x) dx$$



This is another way to write the PDF



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# PDFs like $f(X = x)$ vs. PMFs like $P(X = x)$

---

$P(X = x)$

“The probability that a **discrete** random variable X takes on the value x.”

$f(X = x)$

“The **derivative** of the probability that a **continuous** random variable X takes at the value x.”

*They are both measures of how **likely** X is to take on the value x.  
Sometimes called the **distribution** function.*



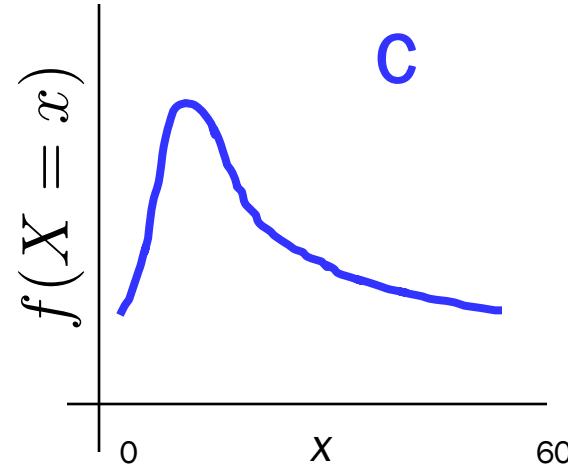
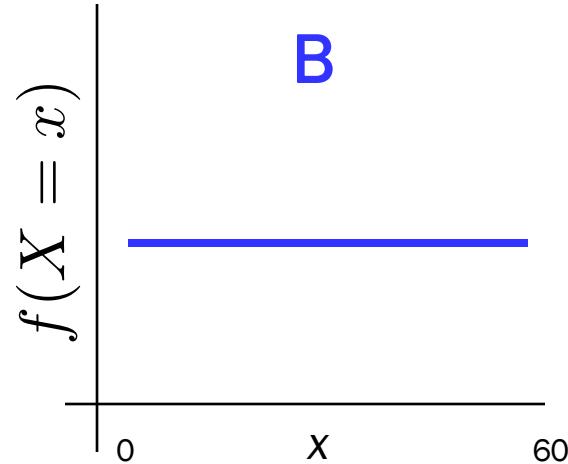
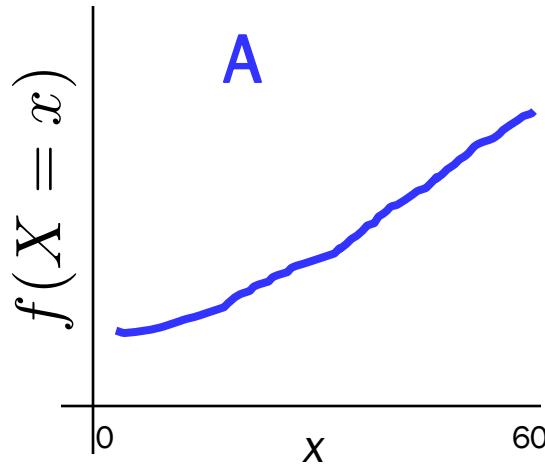
What do you get if you  
integrate over a  
**probability *density* function?**

A probability!

# The Relative Values of PDFs Are Meaningful

Probability density functions are derivatives that articulate *relative* belief.

Let  $X$  be the # of minutes after 2pm that the bus arrives at a stop.



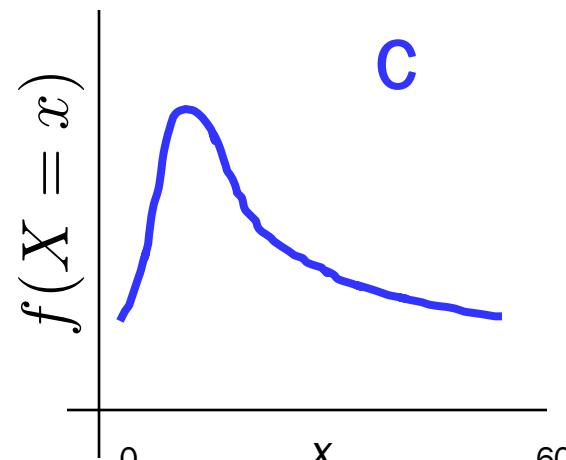
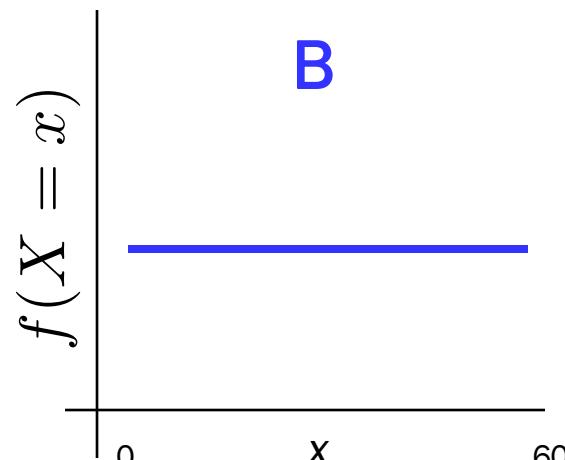
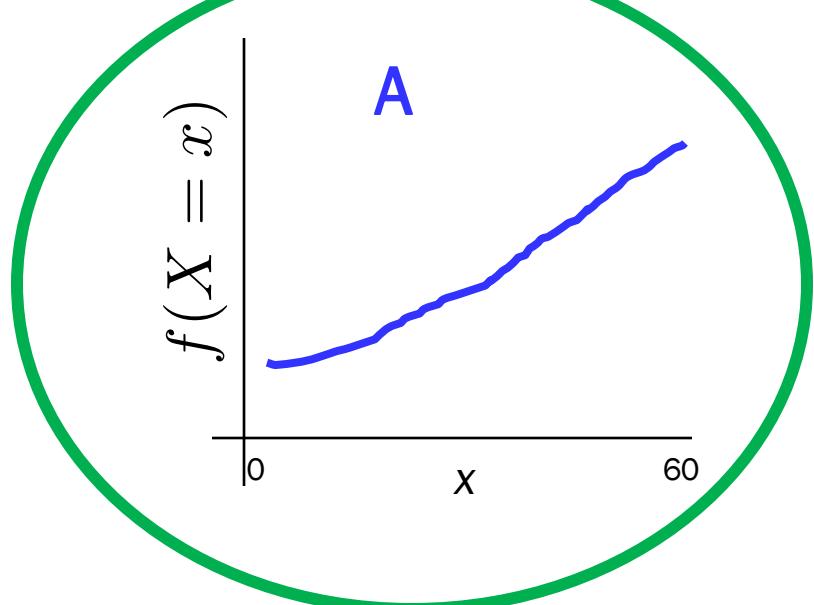
Which of these represent that the bus's arrival  
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---

The relative value of probability densities is meaningful

# Truths of Probability For Continuous Random Variables

---

**Truth 1:**

$$P(a < X < b) = \int_{x=a}^b f(X = x) \, dx$$

Area under the curve!



# Truths of Probability For Continuous Random Variables

**Truth 1:**

$$P(a < X < b) = \int_{x=a}^b f(X = x) dx$$

Area under the curve!

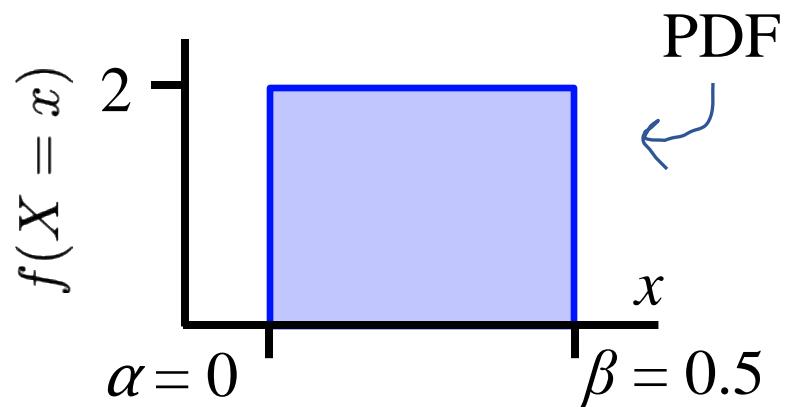
**Truth 2:**

$$0 \leq \int_{x=a}^b f(X = x) dx \leq 1$$

Since the integral is a probability (Axiom 1)

Can a PDF ever have a value  $> 1$ ?

Yes!



# Truths of Probability For Continuous Random Variables

---

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**Truth 3:**

$$\int_{x=-\infty}^{\infty} f(X = x) \, dx = 1$$

That's all possible values (Axiom 2)



# Truths of Probability For Continuous Random Variables

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That's all possible values (Axiom 2)

**Truth 4:**

$$P(X = x) = 0$$

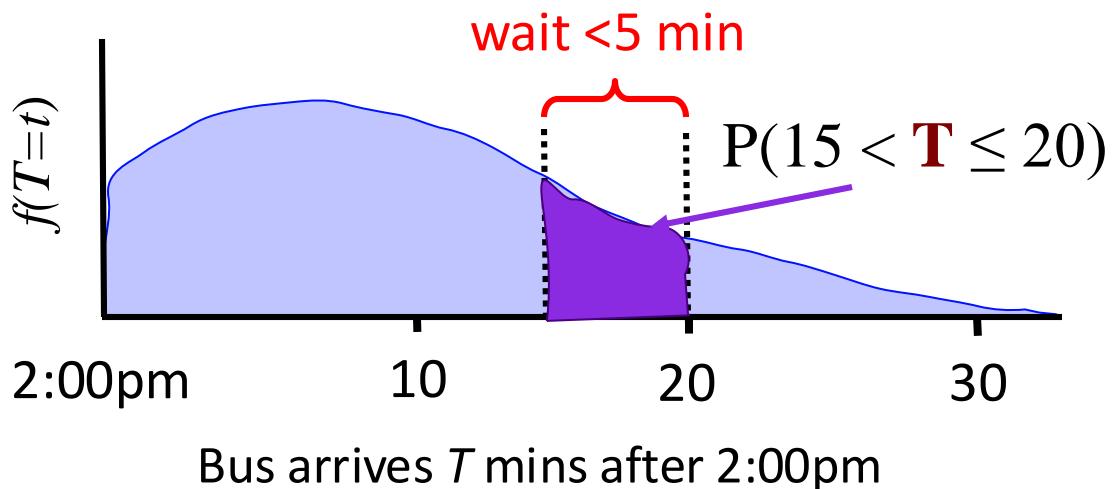
What a time to be alive...



# PDFs Need an Integral



What is the probability that the bus arrives at: 12.12332343234 mins after 2pm?



What do you get if you  
integrate over a  
**probability *density* function?**

A probability!

# Pedagogic Pause

# Quantum Example



Consider a random  $5000 \times 5000$  matrix, where each element in the matrix is  $\text{Uniform}(0,1)$ . What is the probability that a selected eigenvalue ( $\lambda$ ) of the matrix is greater than 0?\*

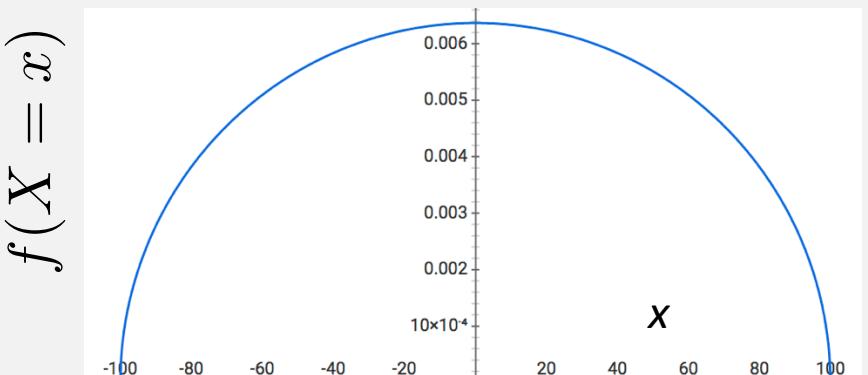
\* With help from Wigner, Chris is rephrased this problem

# Rephrased as a Standard Continuous Problem

Let  $X$  be a continuous random variable<sup>1</sup>

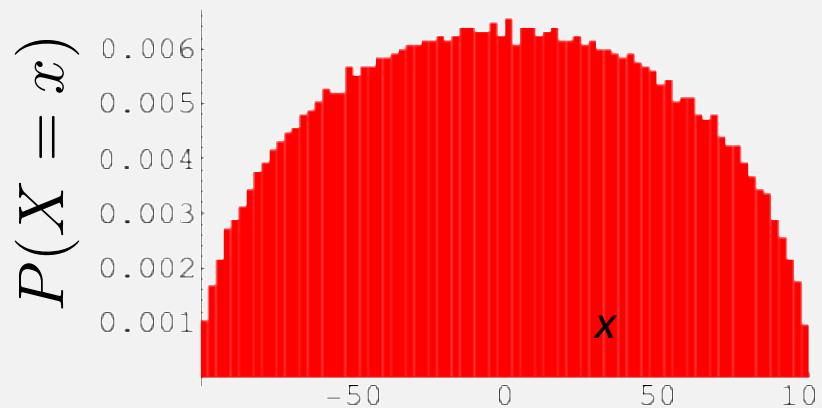
Theory

$$f(X = x) = \frac{1}{15708} \sqrt{100^2 - x^2}$$



Practice

From simulations



$$P(X > 0) = ?$$

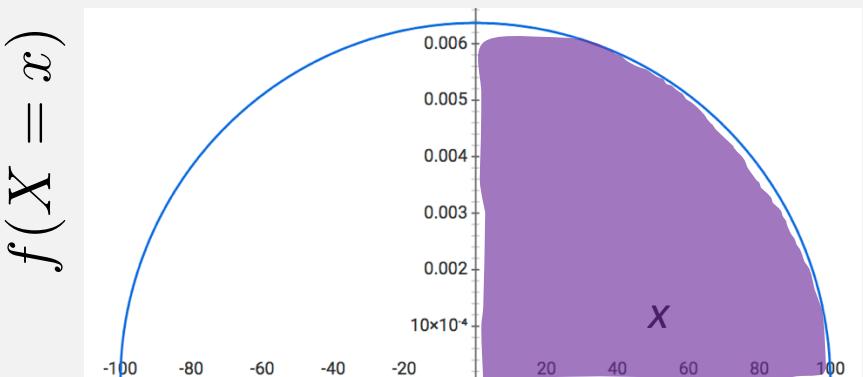


# Simple Example from Quantum Physics

Let  $X$  be a continuous random variable<sup>1</sup>

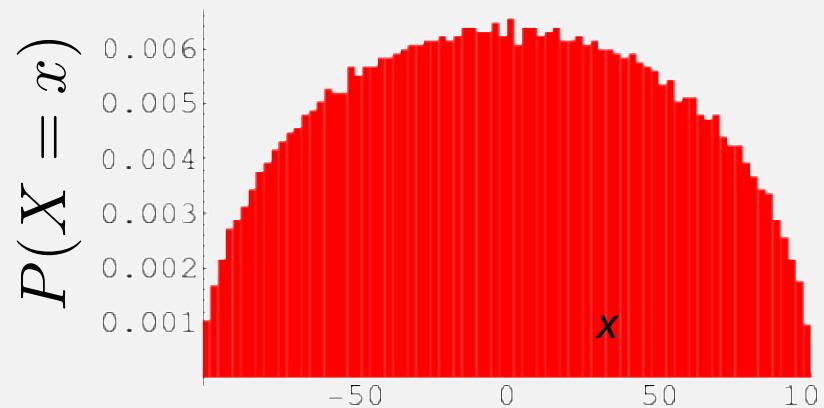
Theory

$$f(X = x) = \frac{1}{15708} \sqrt{100^2 - x^2}$$



Practice

From simulations



Approach #1: Integrate over the PDF

$$P(X > 0) = \int_0^{100} f(X = x) dx$$

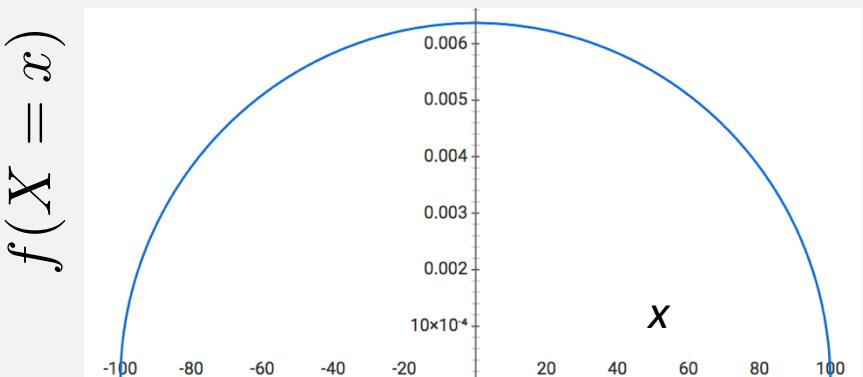


# Simple Example from Quantum Physics

Let  $X$  be a continuous random variable<sup>1</sup>

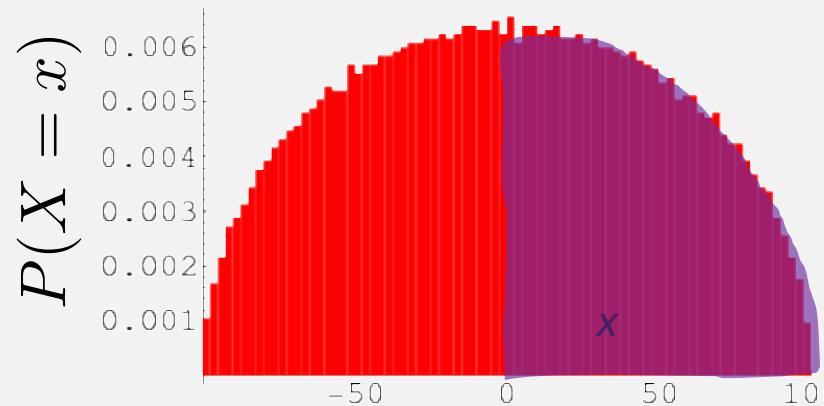
Theory

$$f(X = x) = \frac{1}{15708} \sqrt{100^2 - x^2}$$



Practice

From simulations



Approach #2: Discrete Approximation

$$P(X > 0) \approx \sum_{i=0}^{100} P(X = i)$$

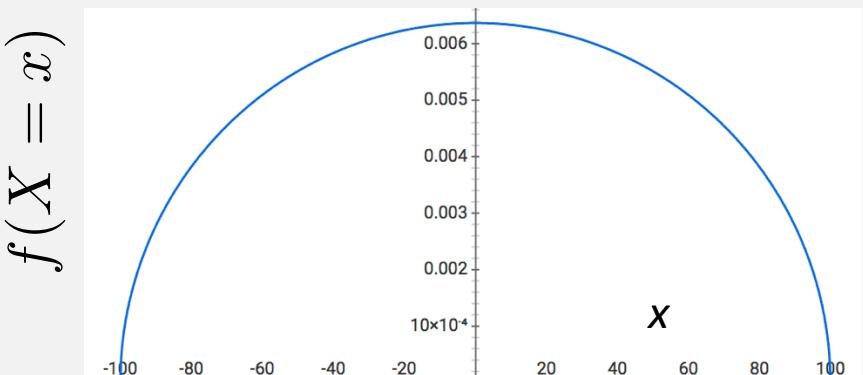


# Simple Example from Quantum Physics

Let  $X$  be a continuous random variable<sup>1</sup>

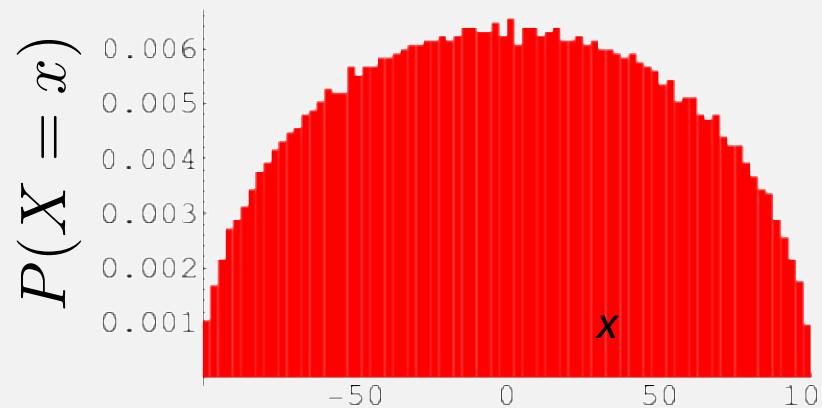
Theory

$$f(X = x) = \frac{1}{15708} \sqrt{100^2 - x^2}$$



Practice

From simulations



Approach #3: Know Semi-Circles

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



You are ready for the classic  
continuous random variables

# Exponential Random Variable

Consider an experiment that lasts a duration of time until success occurs.

def An **Exponential** random variable  $X$  is the amount of time until success.

$$X \sim \text{Exp}(\lambda)$$

Support:  $[0, \infty)$

PDF

Expectation

Variance

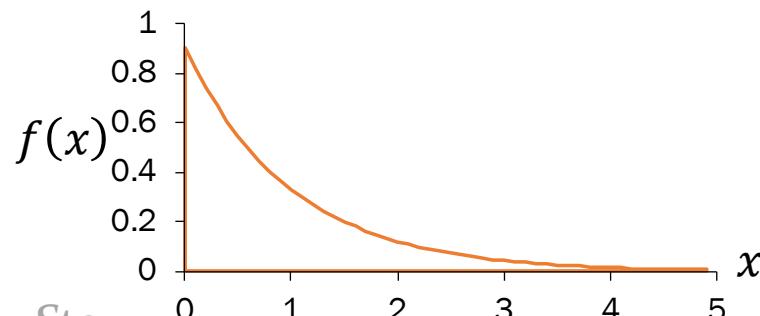
$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

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

$$\text{Var}(X) = \frac{1}{\lambda^2}$$

Examples:

- Time until next earthquake
- Time for request to reach web server
- Time until end of cell phone contract





---

The process for an Exponential and a Poisson are the **same**. So is the parameter  $\lambda$ . The question is different

1906 Earthquake  
Magnitude 7.8



ILL. No. 65. MEMORIAL ARCH, WITH CHURCH IN BACKGROUND, STANFORD UNIVERSITY, SHOWING TYPES OF CARVED WORK WITH THE SANDSTONE

# How Many Earthquakes

---

Based on historical data, major earthquakes (magnitude 8.0+) happen at a **rate of 0.002** per year\*. What is the probability of **zero major earthquakes magnitude next year?**

---

$X$  = Number of major earthquakes next year

$$X \sim \text{Poi}(\lambda = 0.002)$$

$$P(X = 0) = \frac{\lambda^0 e^{-\lambda}}{0!} = \frac{0.002^0 e^{-0.002}}{0!} \approx 0.998$$



# How Long Until the Next Earthquake

---

Based on historical data, major earthquakes (magnitude 8.0+) happen at a **rate of 0.002** per year\*. What is the probability of **a major earthquake in the next 30 years?**

---

$Y$  = Years until the next earthquake of magnitude 8.0+

$$Y \sim \text{Exp}(\lambda = 0.002)$$

$$\begin{aligned}f_Y(y) &= \lambda e^{-\lambda y} \\&= 0.002e^{-0.002y}\end{aligned}$$

$$P(Y < 30) = \int_0^{30} 0.002e^{-0.002y} dy$$

--

\*In California, according to the USGS, 2015



# Integral Review

---

$$\int e^{cx} dx = \frac{1}{c} e^{cx}$$



# How Long Until the Next Earthquake

---

Based on historical data, major earthquakes (magnitude 8.0+) happen at a **rate of 0.002** per year\*. What is the probability of a major earthquake in the next 30 years?

---

$Y$  = Years until the next earthquake of magnitude 8.0+

$$Y \sim \text{Exp}(\lambda = 0.002) \quad f_Y(y) = \lambda e^{-\lambda y} \\ = 0.002^{-0.002y}$$

$$P(Y < 30) = \int_0^{30} 0.002e^{-0.002y} dy \\ = 0.002 \left[ -500e^{-0.002y} \right]_0^{30} \\ = \frac{500}{500} (-e^{-0.06} + e^0) \approx 0.058$$



# How Long Until the Next Earthquake

---

Based on historical data, major earthquakes (magnitude 8.0+) happen at a **rate of 0.002** per year\*. What is the **expected number of years until the next earthquake?**

---

$Y$  = Years until the next earthquake of magnitude 8.0+

$$Y \sim \text{Exp}(\lambda = 0.002)$$

$$E[Y] = \frac{1}{\lambda} = \frac{1}{0.002} = 500$$



# How Long Until the Next Earthquake

---

Based on historical data, major earthquakes (magnitude 8.0+) happen at a **rate of 0.002** per year\*. What is the **standard deviation of years until the next earthquake?**

---

$Y$  = Years until the next earthquake of magnitude 8.0+

$$Y \sim \text{Exp}(\lambda = 0.002)$$

$$\text{Var}(Y) = \frac{1}{\lambda^2} = \frac{1}{0.002^2} = 250,000 \text{ years}^2$$

$$\text{Std}(Y) = \sqrt{\text{Var}(X)} = 500 \text{ years}$$



Is there a way to avoid integrals?

# Cumulative Density Function

---

A cumulative density function (CDF) is a “closed form” equation for the probability that a random variable is less than a given value

$$F(x) = P(X < x)$$



If you learn how to use a cumulative density function, you can avoid integrals!

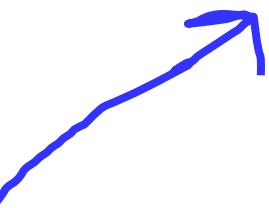
$$F_X(x)$$

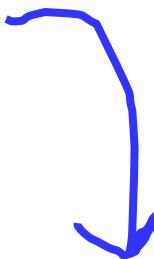
This is also shorthand notation for the PMF

# Cumulative Density Function

---

$$F(x) = P(X < x)$$

$$x = 2$$



$$0.03125$$



# CDF of an Exponential

---

$$F_X(x) = 1 - e^{-\lambda x}$$

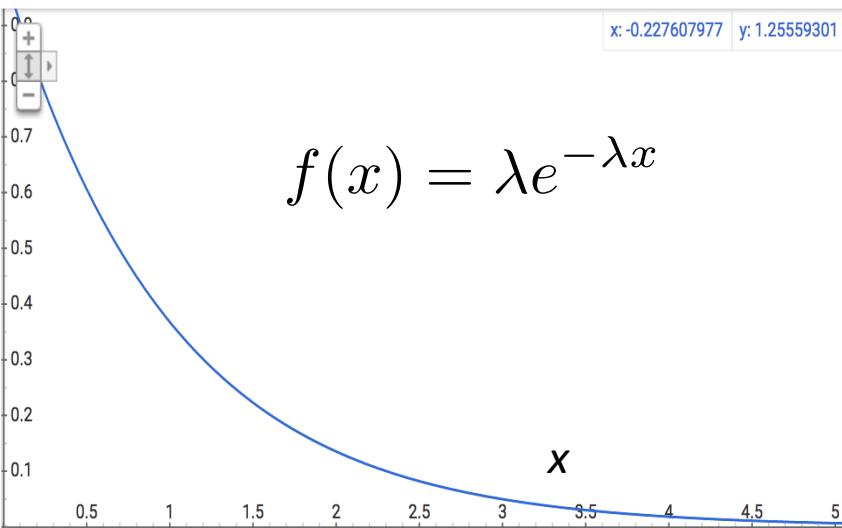
---

$$\begin{aligned} P(X < x) &= \int_{y=-\infty}^x f(y) \, dy \\ &= \int_{y=0}^x \lambda e^{-\lambda y} \, dy \\ &= \frac{\lambda}{\lambda} \left[ -e^{-\lambda y} \right]_0^x \\ &= [-e^{-\lambda x}] - [-e^{\lambda 0}] \\ &= 1 - e^{-\lambda x} \end{aligned}$$



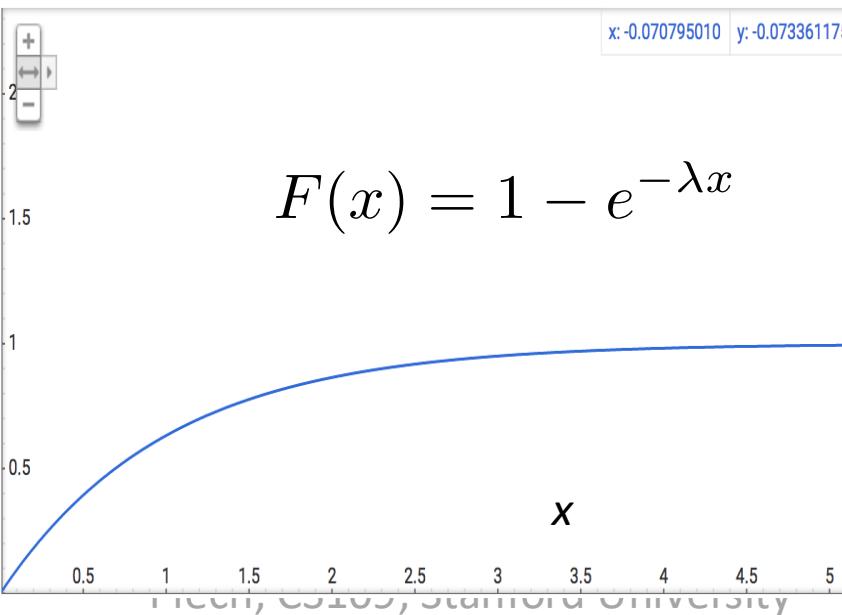
# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

*Probability  
density  
function*



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

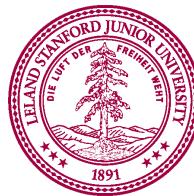
*Cumulative  
density function*



$$F(x) = 1 - e^{-\lambda x}$$

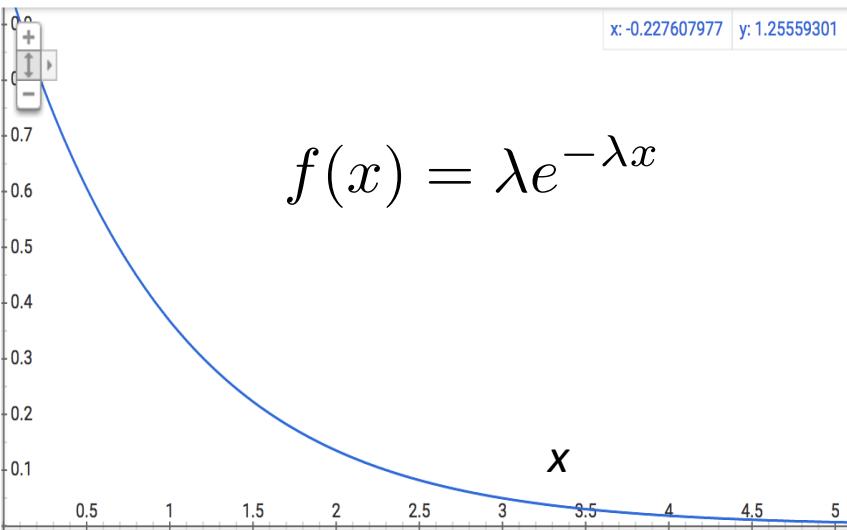
$$F_X(x) = P(X < x)$$

$$= \int_{y=-\infty}^x f(y) dy$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function

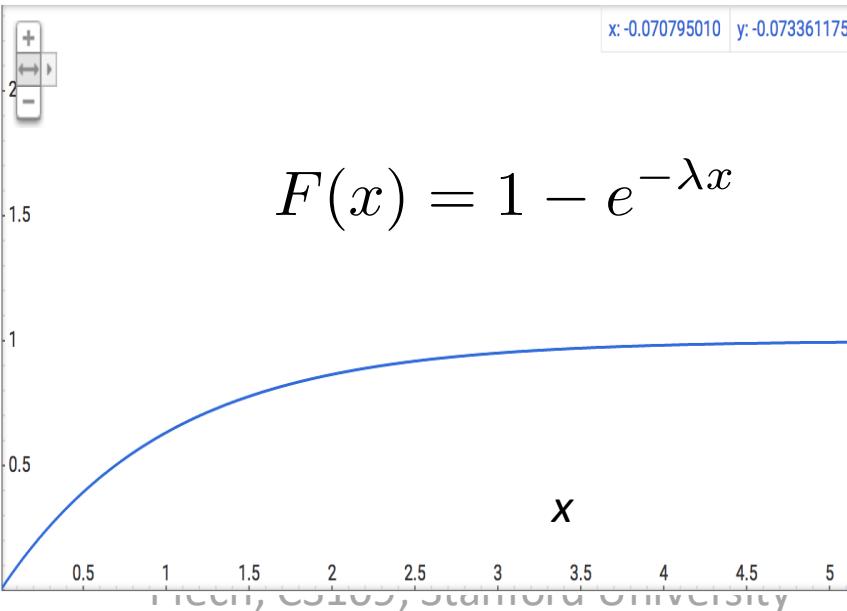


$$P(X < 2)$$

---

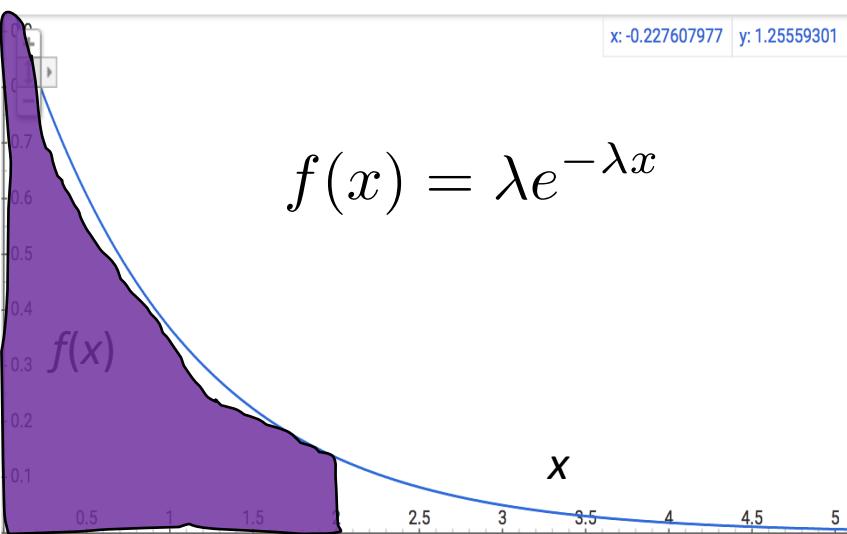
Cumulative density function

$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) dy \end{aligned}$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function

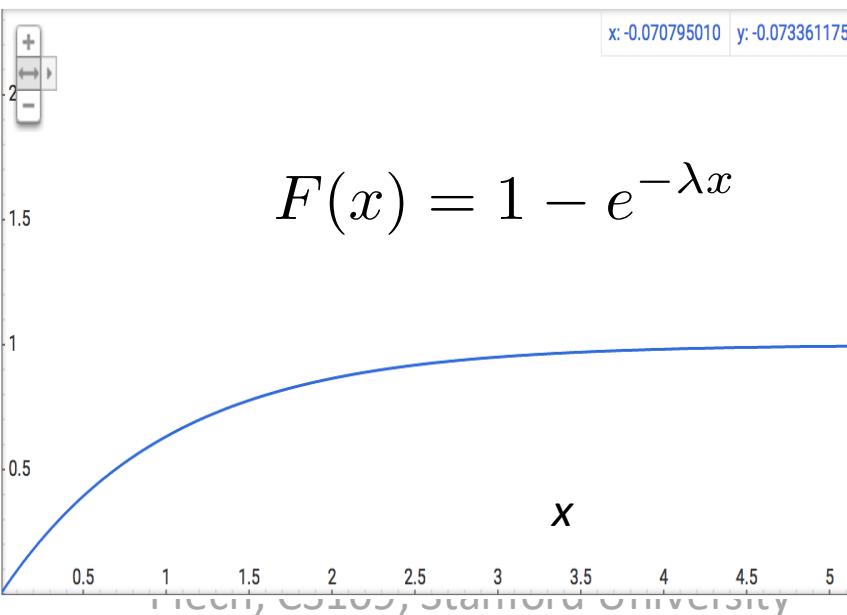


$$f(x) = \lambda e^{-\lambda x}$$

$$\mathbb{P}(X < 2)$$

$$= \int_{x=-\infty}^2 f(x) \, dx$$

Cumulative density function



$$F(x) = 1 - e^{-\lambda x}$$

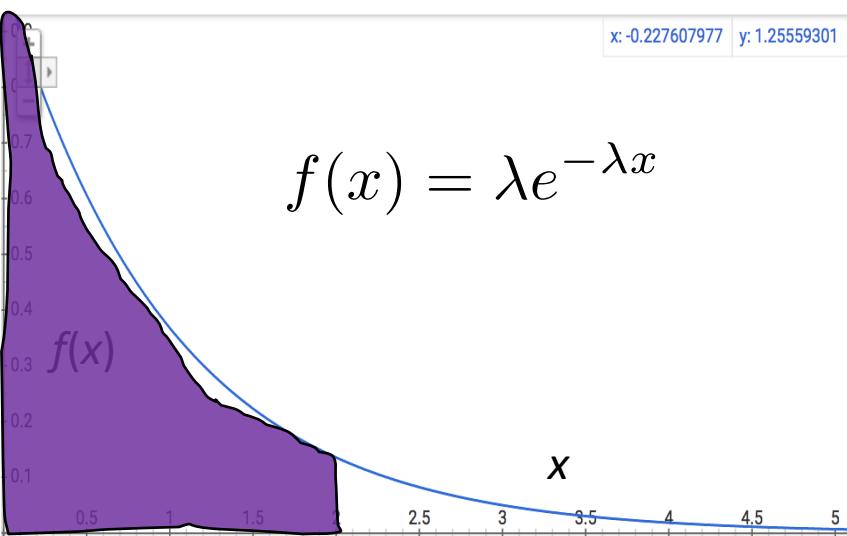
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Probability density function

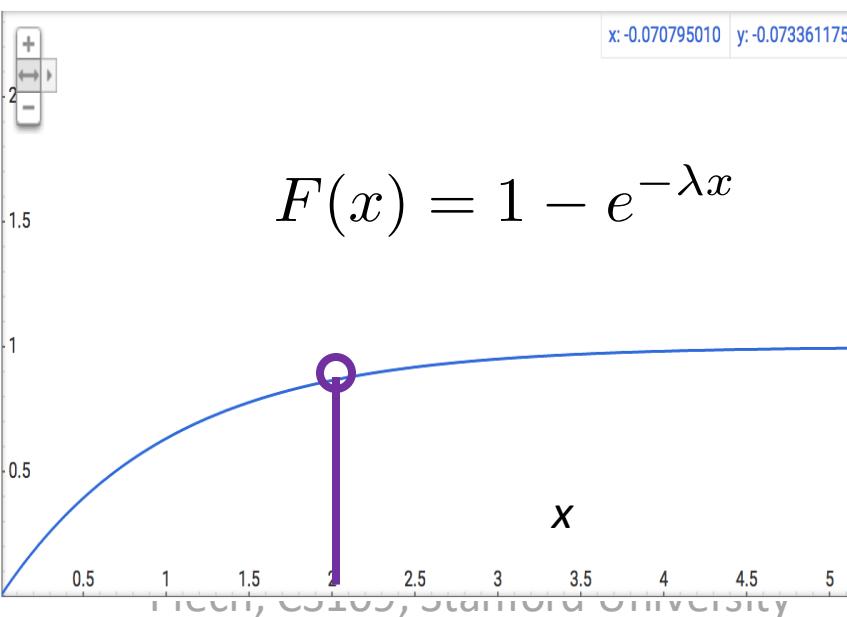


$$f(x) = \lambda e^{-\lambda x}$$

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Cumulative density function



$$F(x) = 1 - e^{-\lambda x}$$

$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) \, dy \end{aligned}$$

or

$$= F(2)$$

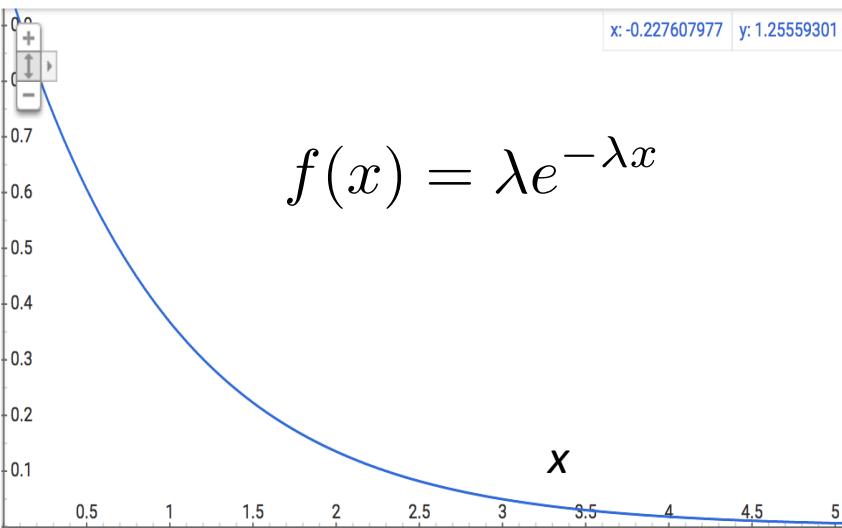
$$= 1 - e^{-2}$$

$$\approx 0.84$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

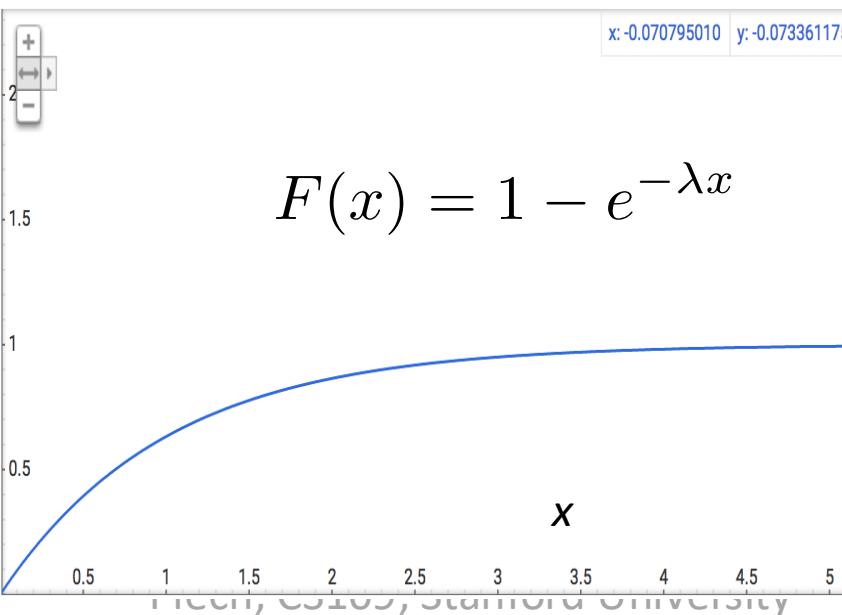
Probability  
density  
function



$$P(X > 1)$$

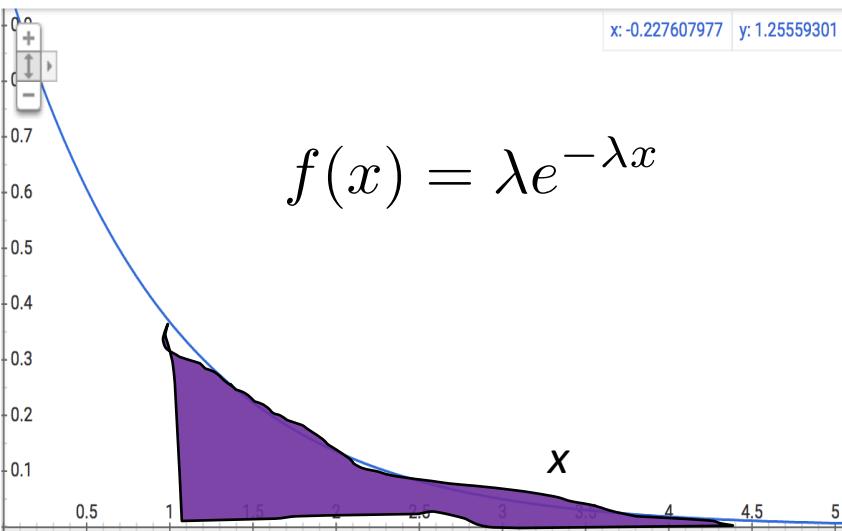
Cumulative  
density function

$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) dy \end{aligned}$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function

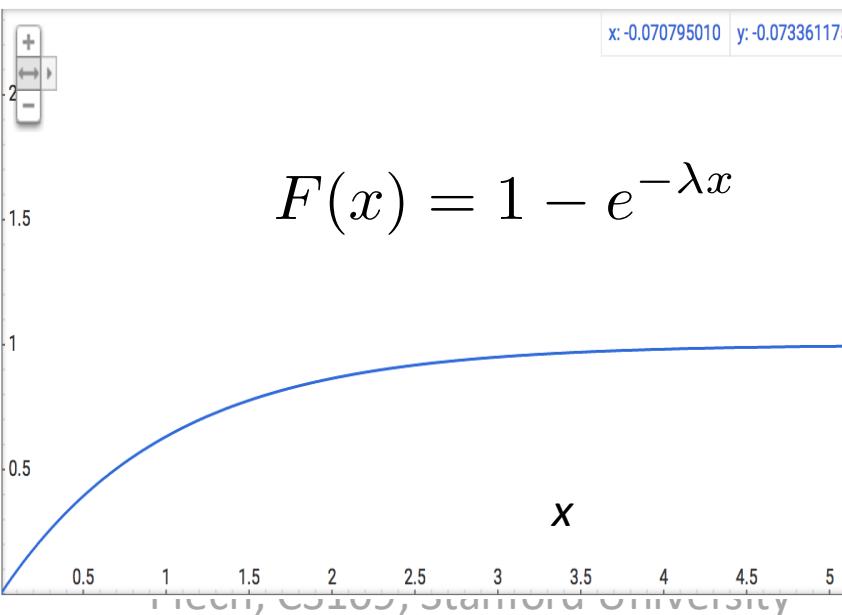


$$f(x) = \lambda e^{-\lambda x}$$

$$\mathbb{P}(X > 1)$$

$$= \int_{x=1}^{\infty} f(x) \, dx$$

Cumulative density function



$$F(x) = 1 - e^{-\lambda x}$$

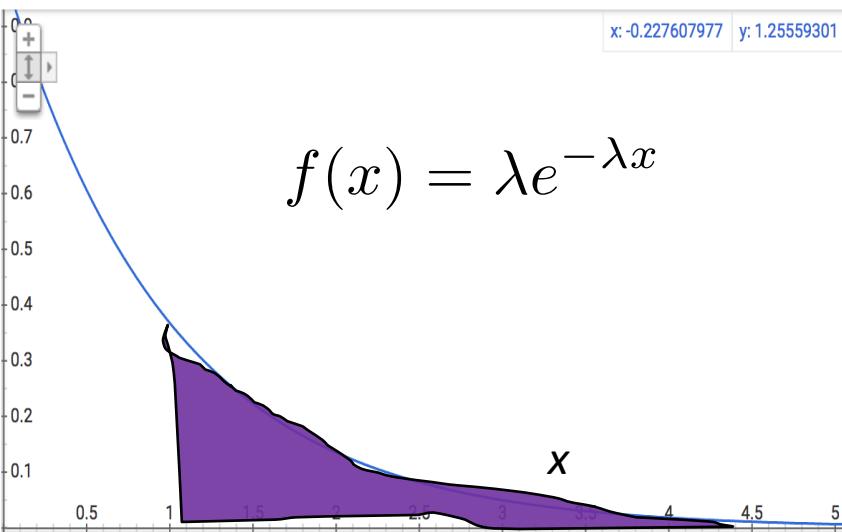
$$F_X(x) = P(X < x)$$

$$= \int_{y=-\infty}^x f(y) \, dy$$



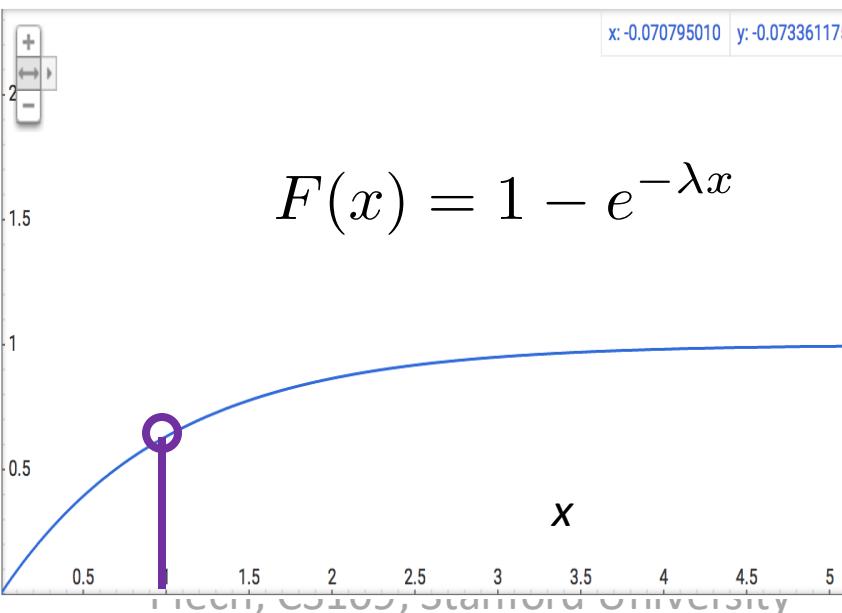
# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function



Cumulative density function

$$F_X(x) = P(X < x) = \int_{y=-\infty}^x f(y) dy$$



$$P(X > 1)$$

$$= \int_{x=1}^{\infty} f(x) dx$$

or

$$= 1 - F(1)$$

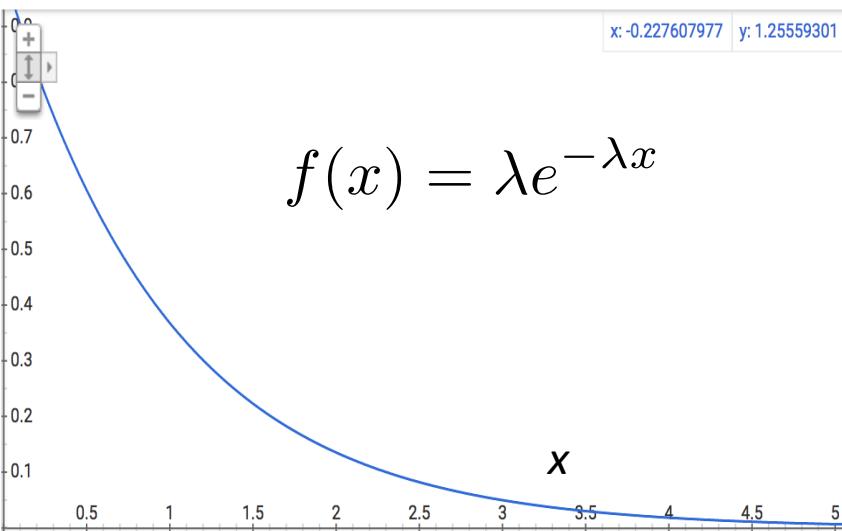
$$= e^{-1}$$

$$\approx 0.37$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

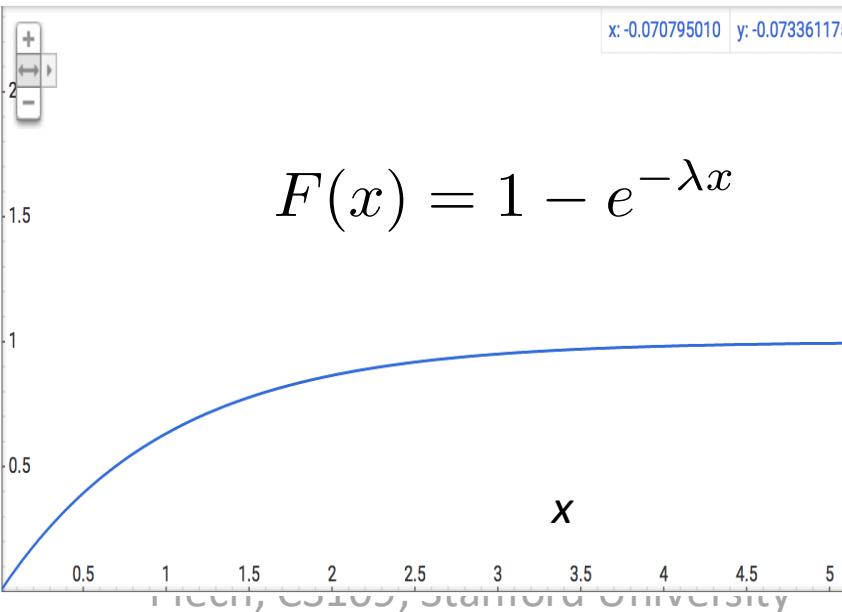
Probability  
density  
function



$$P(1 < X < 2)$$

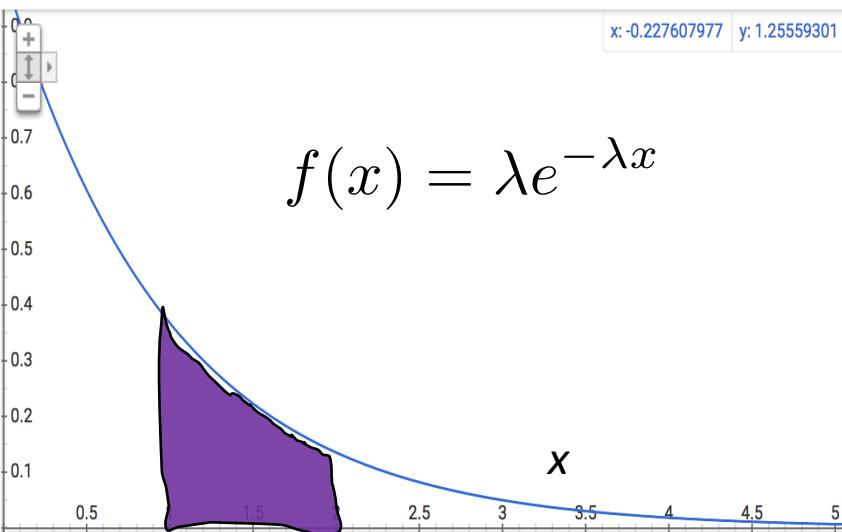
Cumulative  
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$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) dy \end{aligned}$$



# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function

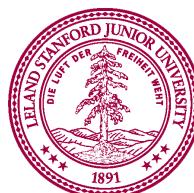
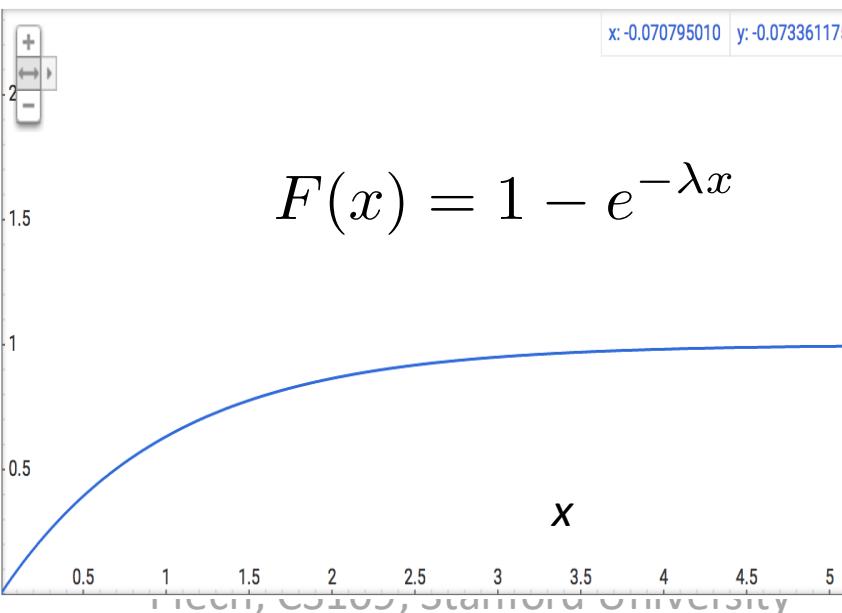


$$P(1 < X < 2)$$

$$= \int_{x=1}^2 f(x) \, dx$$

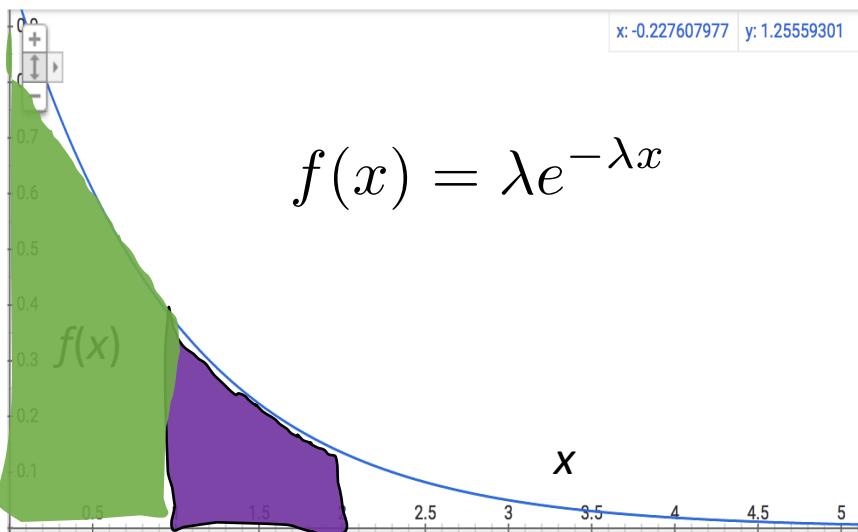
Cumulative density function

$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) \, dy \end{aligned}$$



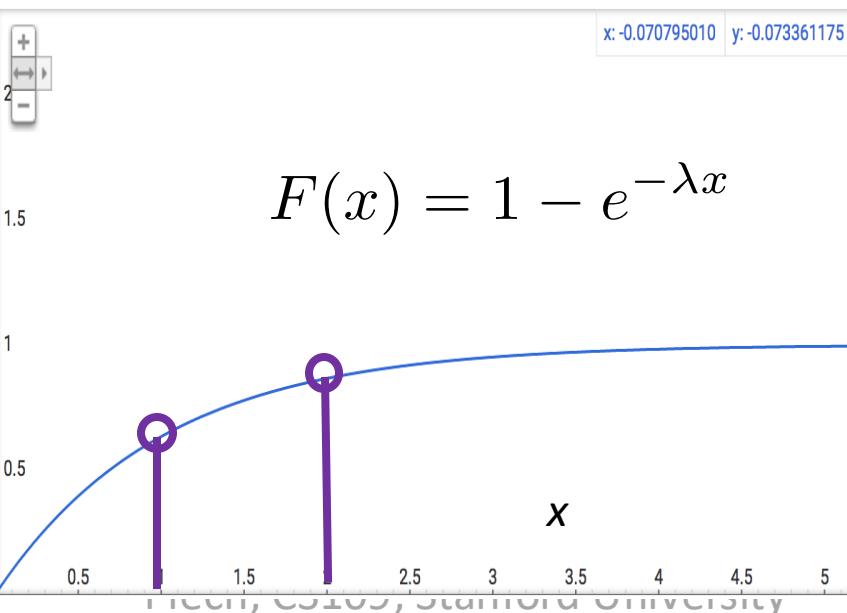
# Using CDF Example. X is $\text{Exp}(\lambda = 1)$

Probability density function



Cumulative density function

$$\begin{aligned} F_X(x) &= P(X < x) \\ &= \int_{y=-\infty}^x f(y) dy \end{aligned}$$



$$P(1 < X < 2)$$

$$= \int_{x=1}^2 f(x) dx$$

or

$$= F(2) - F(1)$$

$$\begin{aligned} &= (1 - e^{-2}) \\ &\quad - (1 - e^{-1}) \\ &\approx 0.23 \end{aligned}$$



# Probability of Earthquake in Next 4 Years?

Based on historical data, earthquakes of magnitude 8.0+ happen at a **rate of 0.002** per year\*. What is the probability of **an major earthquake in the next 4 years?**

$Y$  = Years until the next earthquake of magnitude 8.0+

$$Y \sim \text{Exp}(\lambda = 0.002)$$

$$F(y) = 1 - e^{-0.002y}$$

$$\begin{aligned} P(Y < 4) &= F(4) \\ &= 1 - e^{-0.002 \cdot 4} \\ &\approx 0.008 \end{aligned}$$

Feeling lucky?



# Two Classic Random Variables

## Uniform Random Variable

**Notation:**  $X \sim \text{Uni}(\alpha, \beta)$

**Description:** A continuous random variable that takes on values, with equal likelihood, between  $\alpha$  and  $\beta$

**Parameters:**  $\alpha \in \mathbb{R}$ , the minimum value of the variable.

$\beta \in \mathbb{R}, \beta > \alpha$ , the maximum value of the variable.

**Support:**  $x \in [\alpha, \beta]$

**PDF equation:**  $f(x) = \begin{cases} \frac{1}{\beta-\alpha} & \text{for } x \in [\alpha, \beta] \\ 0 & \text{else} \end{cases}$

**CDF equation:**  $F(x) = \begin{cases} \frac{x-\alpha}{\beta-\alpha} & \text{for } x \in [\alpha, \beta] \\ 0 & \text{for } x < \alpha \\ 1 & \text{for } x > \beta \end{cases}$

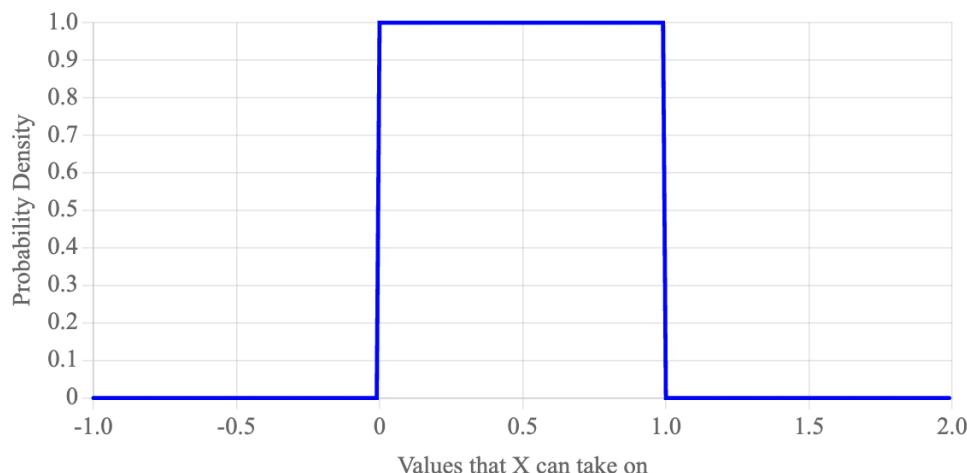
**Expectation:**  $E[X] = \frac{1}{2}(\alpha + \beta)$

**Variance:**  $\text{Var}(X) = \frac{1}{12}(\beta - \alpha)^2$

**PDF graph:**

Parameter  $\alpha$ :

Parameter  $\beta$ :



## Exponential Random Variable

**Notation:**  $X \sim \text{Exp}(\lambda)$

**Description:** Time until next events if (a) the events occur with a constant mean rate and (b) they occur independently of time since last event.

**Parameters:**  $\lambda \in \{0, 1, \dots\}$ , the constant average rate.

**Support:**  $x \in \mathbb{R}^+$

**PDF equation:**  $f(x) = \lambda e^{-\lambda x}$

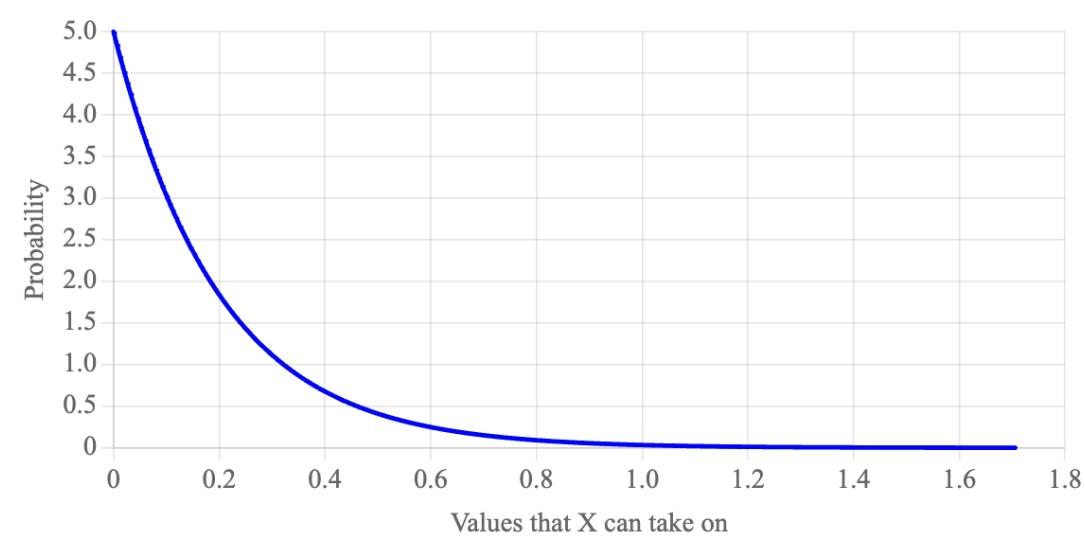
**CDF equation:**  $F(x) = 1 - e^{-\lambda x}$

**Expectation:**  $E[X] = 1/\lambda$

**Variance:**  $\text{Var}(X) = 1/\lambda^2$

**PDF graph:**

Parameter  $\lambda$ :



Pset 1  
Grades Sat at noon!

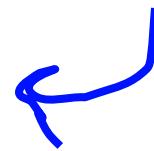
# What is a check in CS109?

---

Functionality and style grades for the assignments use the following scale:

The modal grade in CS109

- ✓ + Satisfies all requirements of the assignment / full credit. An “A”
- ✓ Meets most requirements, but with some problems. An “A-” / “B+”
- ✓ - Has more serious problems, such as not explaining work.
- Progress was made.



If your overall score is a ✓ + you rocked the PSet.



# Small nit: Avoid answers that are just equations

Numeric Answer:

Enter your answer

Check Answer

Explanation:

Block LaTeX

Image

B

↔

I

U

$$|E| = \binom{f}{1} \binom{u-f}{s-1} = 50 \cdot \binom{11950}{249}$$

A small group of students made this mistake on pset1.



That is all folks!