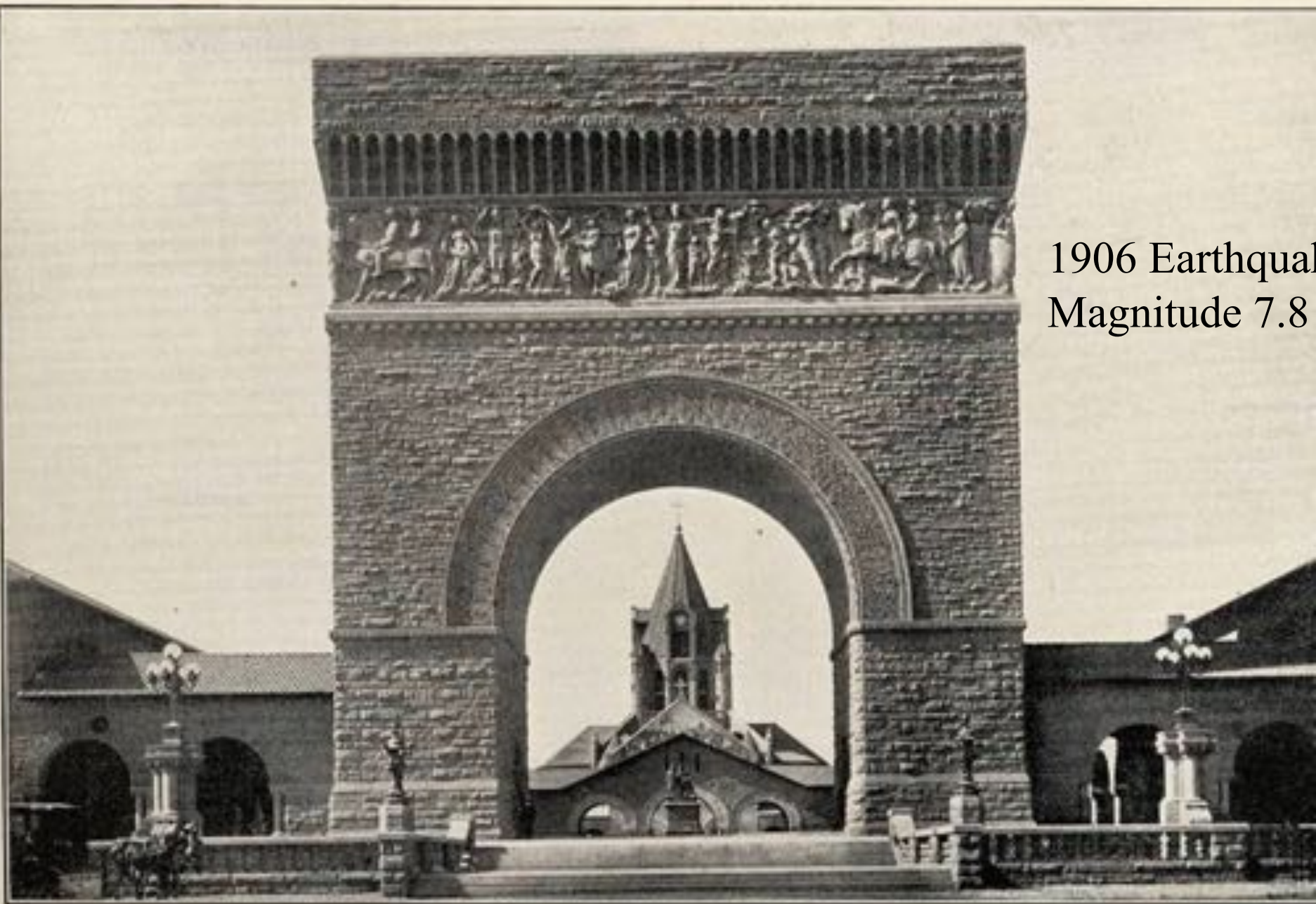




Continuous Variables

Chris Piech
CS109, Stanford University



1906 Earthquake
Magnitude 7.8

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

Learning Goals

1. Comfort using new discrete random variables
2. Integrate a density function (PDF) to get a probability
3. Use a cumulative function (CDF) to get a probability



Discrete Distributions

Don't have to derive all of the following distributions.
We want you to get a sense of how random
variables work.

Grid of Random Variables

	number of successes	time to get successes	
One trial	$X \sim \text{Ber}(p)$ \uparrow $n = 1$	$X \sim \text{Geo}(p)$ \uparrow $r = 1$	One success
Several trials	$X \sim \text{Bin}(n, p)$	$X \sim \text{NegBin}(r, p)$	Several successes
Interval of time	$X \sim \text{Poi}(\lambda)$	$X \sim \text{Exp}(\lambda)$	One success

Geometric Random Variable

- X is **Geometric** Random Variable: $X \sim \text{Geo}(p)$
 - X is number of independent trials until first success
 - p is probability of success on each trial
 - X takes on values 1, 2, 3, ..., with probability:

$$P(X = n) = (1 - p)^{n-1}p$$

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

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



Negative Binomial Random Variable

- X is **Negative Binomial** RV: $X \sim \text{NegBin}(r, p)$
 - X is number of independent trials until r successes
 - p is probability of success on each trial
 - X takes on values $r, r + 1, r + 2 \dots$, with probability:

$$P(X = n) = \binom{n-1}{r-1} p^r (1-p)^{n-r}, \text{ where } n = r, r+1, \dots$$

- $E[X] = r/p$ $\text{Var}(X) = r(1-p)/p^2$
- Note: $\text{Geo}(p) \sim \text{NegBin}(1, p)$



New Handout

Chris Piech
CS 109

Handout
Oct 12, 2018

All Discrete Distributions

Bernoulli

An indicator variable that takes on the value 1 or 0. Often the variable is defined to be 1 if an underlying event has occurred, 0 otherwise.

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

Parameters: p : The probability of the variable being 1

Range(X): $\{0, 1\}$

pmf: $\Pr(X = k) = \begin{cases} p & \text{if } k = 1 \\ (1 - p) & \text{if } k = 0 \end{cases}$

$E[X]$: p

$\text{Var}(X)$: $p(1 - p)$

Note: Sometimes in machine learning algorithms a derivable version of the PMF is used:
 $f(X = k) = p^k(1 - p)^{1-k}$

Binomial

A variable which represents the number of successes in a fixed number of independent trials. The probability of success must be the same for each trial.

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

Parameters: n : the number of trials
 p : the probability of success in each trial

Range(X): $\{0, 1, \dots, n\}$

pmf: $\Pr(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$

$E[X]$: np

$\text{Var}(X)$: $np(1 - p)$

Note: $\text{Bin}(1, p) = \text{Bern}(p)$

Poisson

The number of events occurring in a fixed interval of time or space if these events occur independently with a constant rate.

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

Parameters: λ : the rate of events in one interval

Range(X): $\{0, 1, \dots, \infty\}$

pmf: $\Pr(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}$

$E[X]$: λ

$\text{Var}(X)$: λ

Note: The Poisson is the number of events in an interval of time. The Exponential is a continuous distribution which is the time until the next event. They have the same parameter (λ).

Geometric

The number of independent Bernoulli trials until the first success.

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

Parameters: p : the probability of success of each trial

Range(X): $\{1, 2, \dots, \infty\}$

pmf: $\Pr(X = k) = (1 - p)^{k-1} p$

$E[X]$: $1/p$

$\text{Var}(X)$: $\frac{1-p}{p^2}$

Etc...

Discrete Distributions

Bernoulli:

- indicator of coin flip $X \sim \text{Ber}(p)$

Binomial:

- # successes in n coin flips $X \sim \text{Bin}(n, p)$

Poisson:

- # successes in n coin flips $X \sim \text{Poi}(\lambda)$

Geometric:

- # coin flips until success $X \sim \text{Geo}(p)$

Negative Binomial:

- # trials until r successes $X \sim \text{NegBin}(r, p)$

Zipf:

- The popularity rank of a random word, from a natural language
- $X \sim \text{Zipf}(s)$

Discrete Distributions

Bernoulli:

- indicator of coin flip $X \sim \text{Ber}(p)$

Binomial:

- # successes in n coin flips $X \sim \text{Bin}(n, p)$

Poisson:

- # successes in n coin flips $X \sim \text{Poi}(\lambda)$

Geometric:

- # coin flips until success $X \sim \text{Geo}(p)$

Negative Binomial:

- # trials until r successes $X \sim \text{NegBin}(r, p)$

Zipf:

- The popularity rank of a random word, from a natural language
- $X \sim \text{Zipf}(s)$



Bit Coin Mining

SHA-256 Hash(Data
Fixed , Salt
Choice)



Number that looks like random bits



You “mine a bitcoin” if, for given data D , you find a number N such that Hash(D , N) produces a string that starts with g zeroes.

Midterm Question: Bit Coin Mining

You “mine a bitcoin” if, for given data D , you find a number N such that $\text{Hash}(D, N)$ produces a string that starts with g zeroes.

(a) What is the probability that the first number you try will produce a bit string which starts with g zeroes (in other words you mine a bitcoin)?

(b) How many different numbers do you expect to have to try before you mine five bitcoins?

Dating at Stanford

Each person you date has a 0.2 probability of being someone you spend your life with. What is the average number of people one will date? What is the standard deviation?



Equity in the Courts

Berghuis v. Smith

If a group is underrepresented in a jury pool, how do you tell?

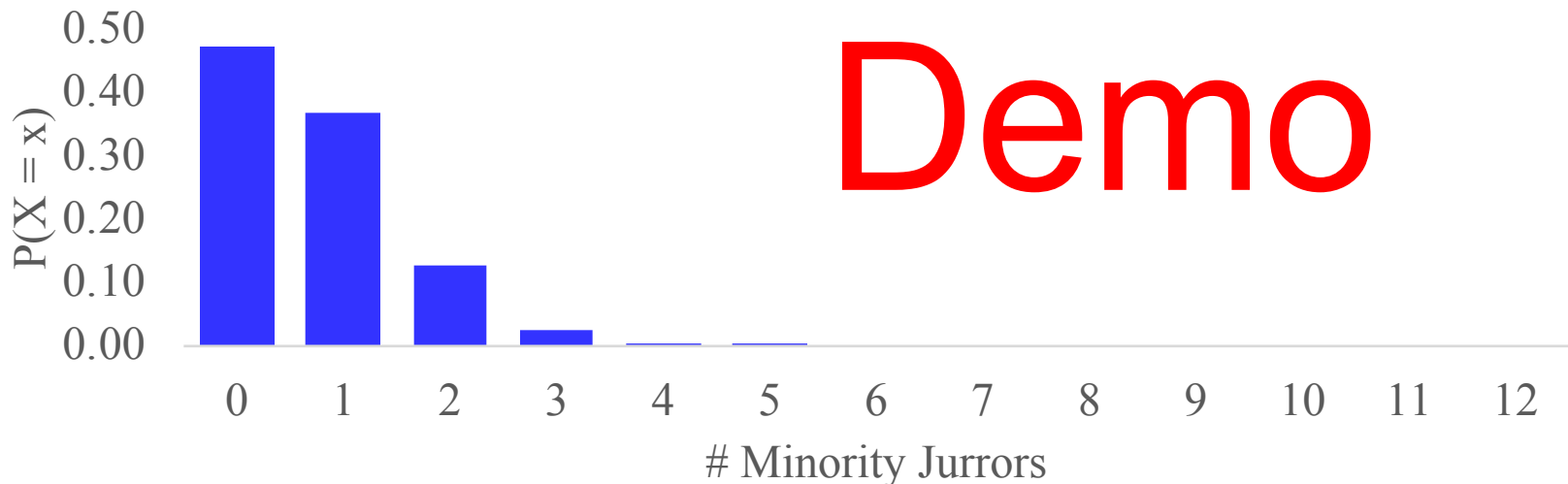
- Article by Erin Miller –January 22, 2010
- Thanks to (former CS109er) Josh Falk for this article

Justice Breyer [Stanford Alum] opened the questioning by invoking the binomial theorem. He hypothesized a scenario involving **“an urn with a thousand balls, and sixty are blue, and nine hundred forty are purple, and then you select them at random... twelve at a time.”** According to Justice Breyer and the binomial theorem, if the purple balls were under represented jurors then **“you would expect... something like a third to a half of juries would have at least one minority person”** on them.

Justin Breyer Meets CS109

- Approximation using Binomial distribution
 - Assume $P(\text{blue ball})$ constant for every draw = $60/1000$
 - $X = \#$ blue balls drawn. $X \sim \text{Bin}(12, 60/1000 = 0.06)$
 - $P(X \geq 1) = 1 - P(X = 0) \approx 1 - 0.4759 = 0.5240$

In Breyer's description, should actually expect just over half of juries to have at least one non-white person on them



Learning Goal: Use new RVs

You are learning about servers...



You read about the MD1 queue...

You find a paper that says the server "busy period" is distributed as a Borel with parameter $\mu = 0.2$...

W Borel distribution - Wikipedia

https://en.wikipedia...

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Borel distribution

From Wikipedia, the free encyclopedia

The **Borel distribution** is a discrete probability distribution, arising in contexts including branching processes and queueing theory. It is named after the French mathematician Émile Borel.

If the number of offspring that an organism has is Poisson-distributed, and if the average number of offspring of each organism is no bigger than 1, then the descendants of each individual will ultimately become extinct. The number of descendants that an individual ultimately has in that situation is a random variable distributed according to a Borel distribution.

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	$\frac{\mu}{(1 - \mu)^3}$

Contents [hide]

- Definition
- Derivation and branching process interpretation
- Queueing theory interpretation
- Properties
- Borel-Tanner distribution
- References
- External links

Definition [edit]

A discrete random variable X is said to have a Borel distribution^{[1][2]} with parameter $\mu \in [0, 1]$ if the probability mass function of X is given by

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

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

Derivation and branching process interpretation [edit]

Big hole in our knowledge

Not all values are discrete



random() ?

Riding the Marguerite



Riding the Marguerite



You are running to the bus stop.
You don't know exactly when the bus arrives. You have a distribution of probabilities.

You show up at 2:20pm.

What is $P(\text{wait} < 5 \text{ minutes})$?

What is the probability that the bus arrives at:
2:17pm and 12.12333911102389234 seconds?

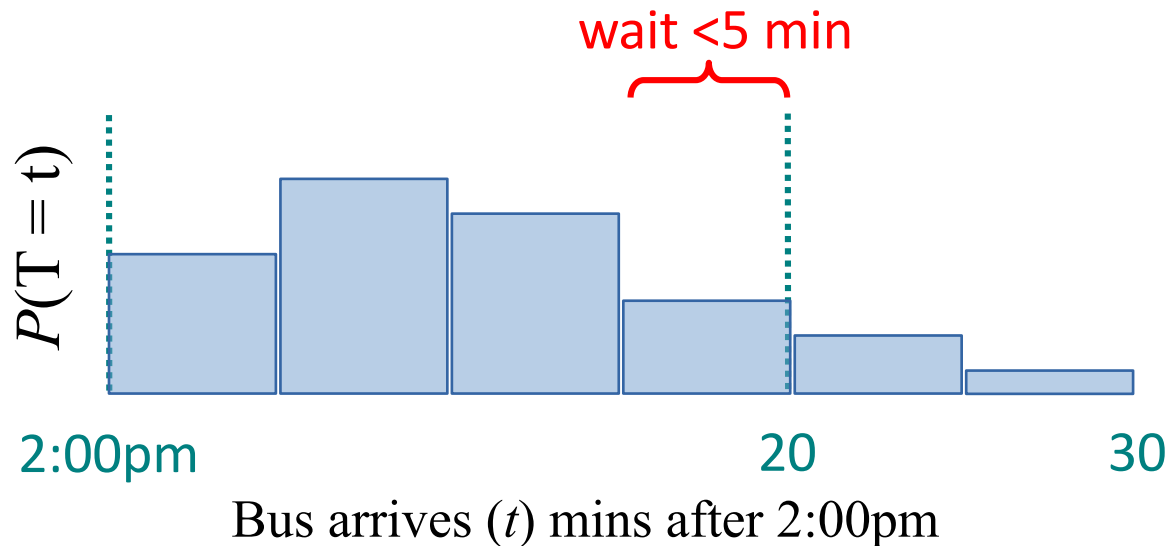
Riding the Marguerite



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You show up at 2:15pm.

What is $P(\text{wait} < 5 \text{ minutes})$?



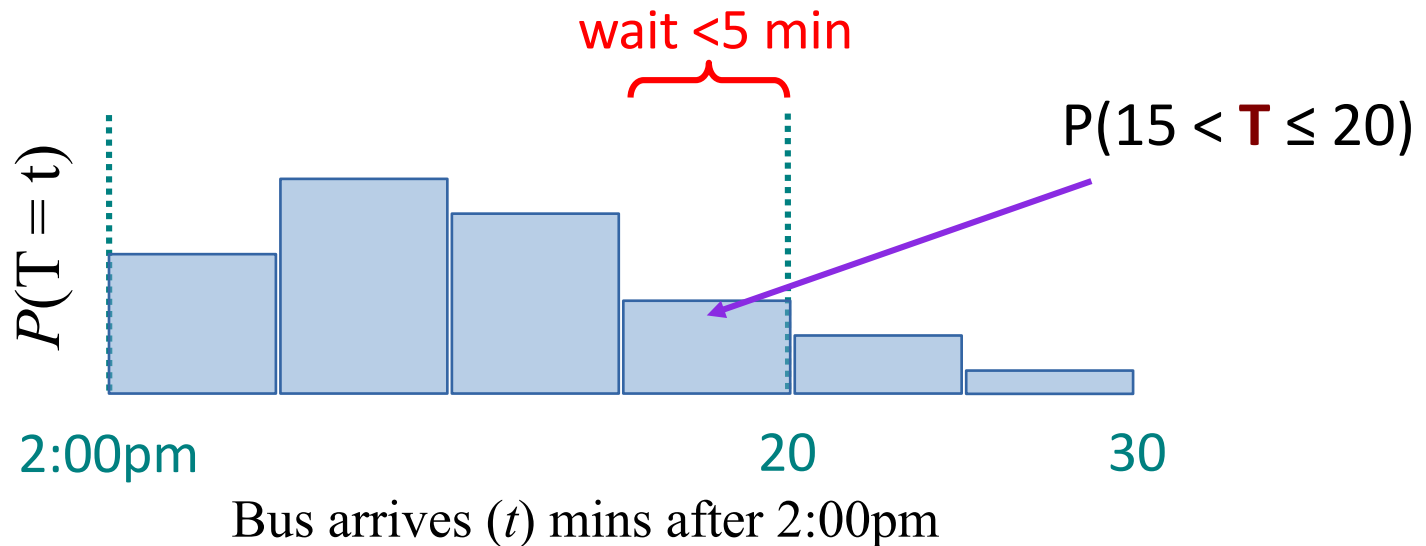
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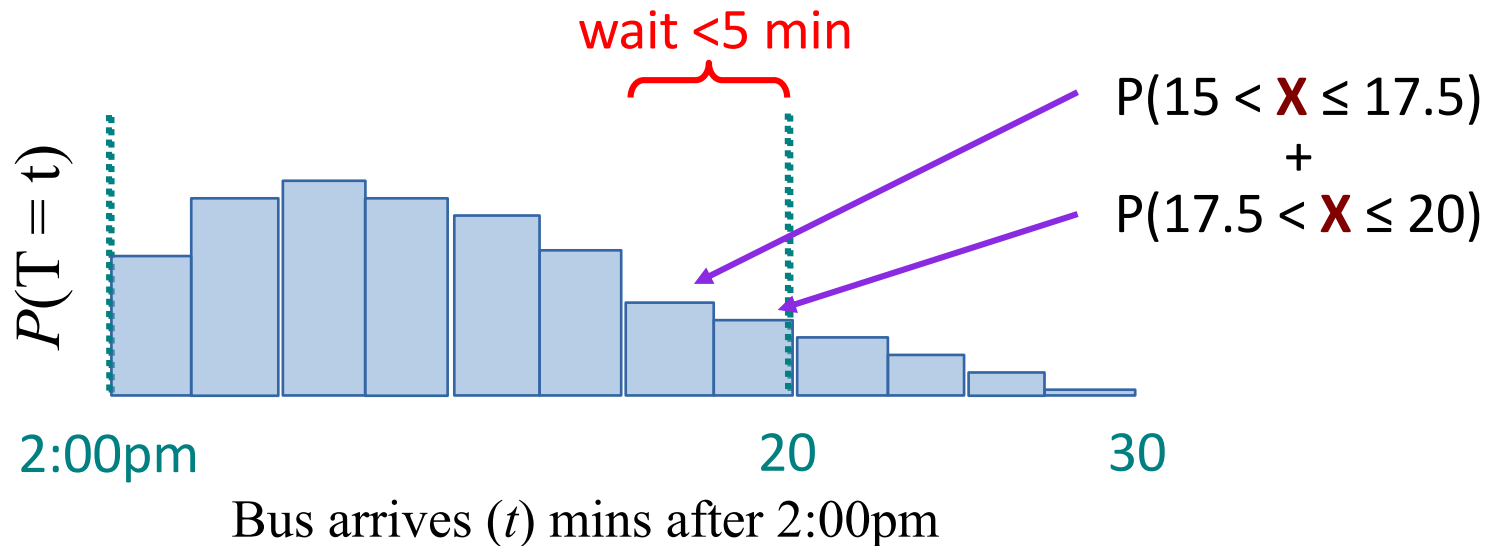
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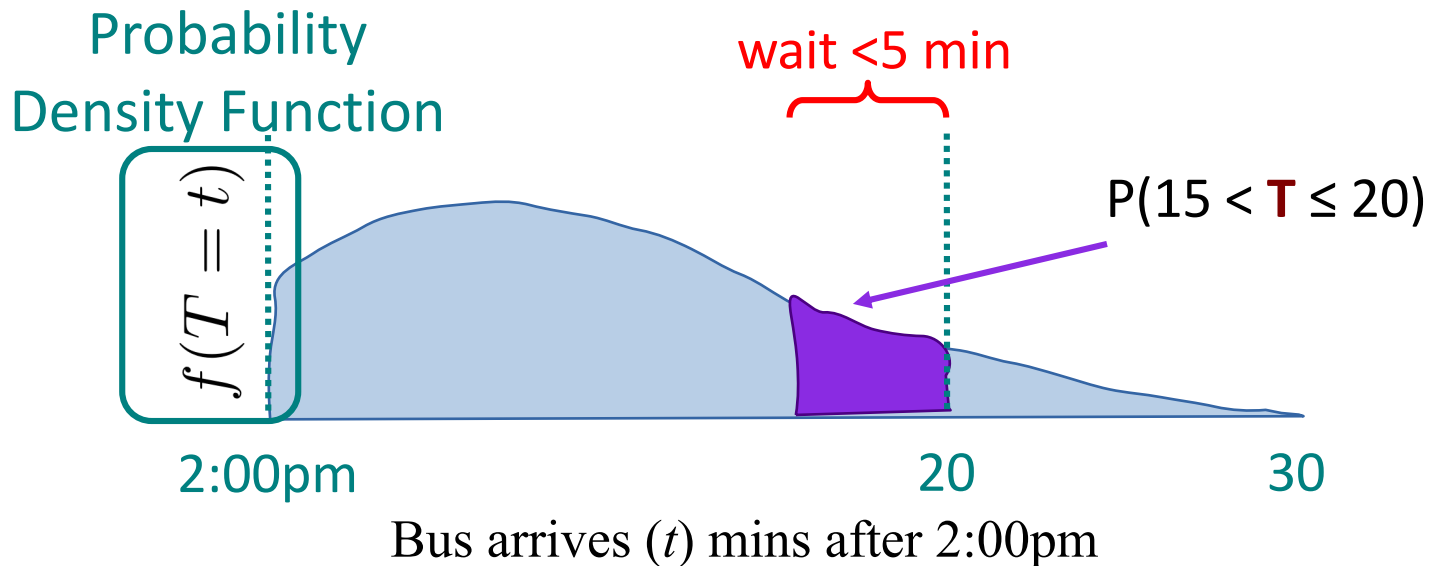
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You show up at 2:15pm.

What is $P(\text{wait} < 5 \text{ minutes})$?



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!


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

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!

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


This is another way to write the PDF

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!

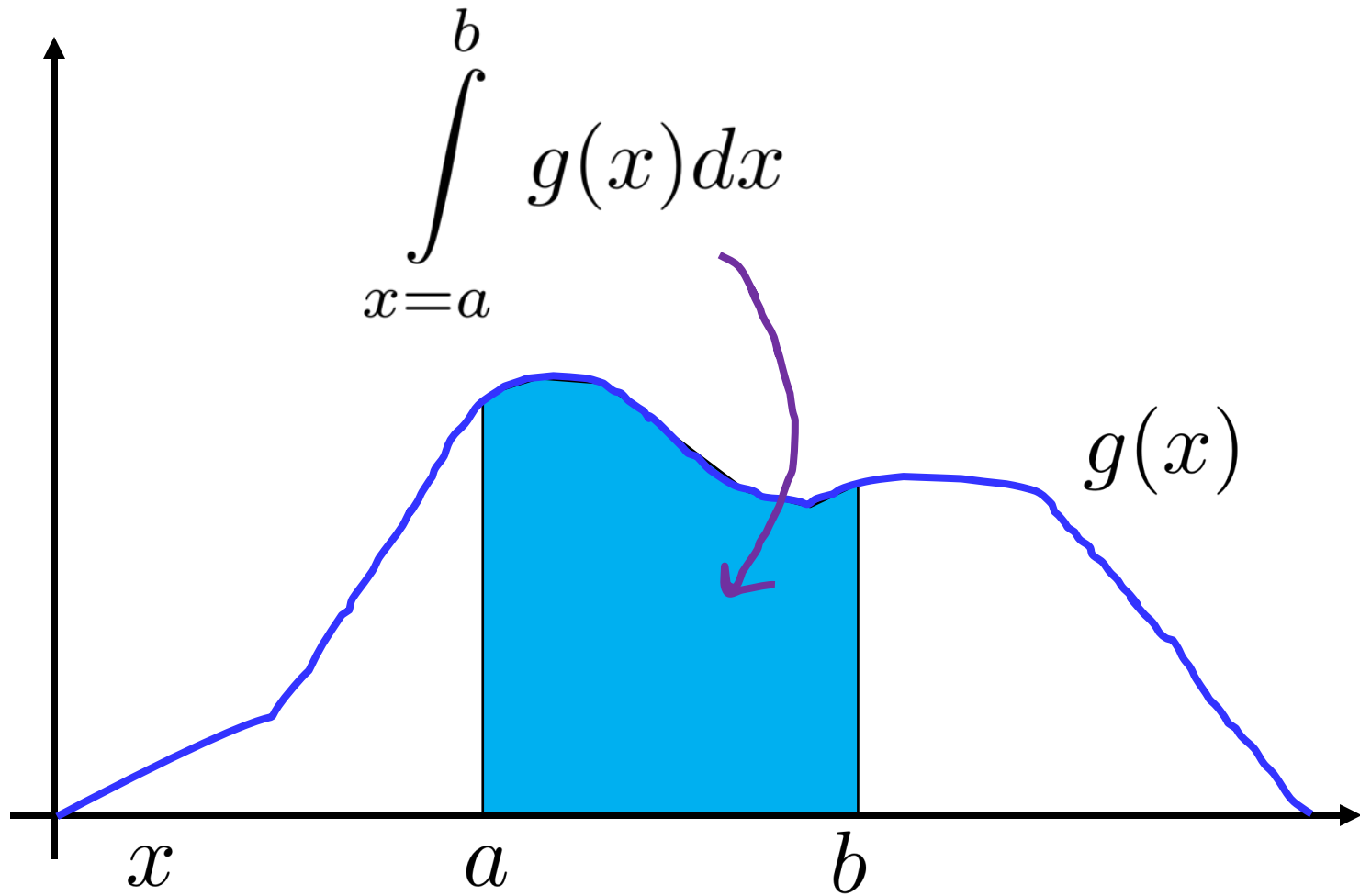
$$P(a < X < b) = \int_{x=a}^b f(X = x) dx$$

Integrals



*loving, not scary

Integrals



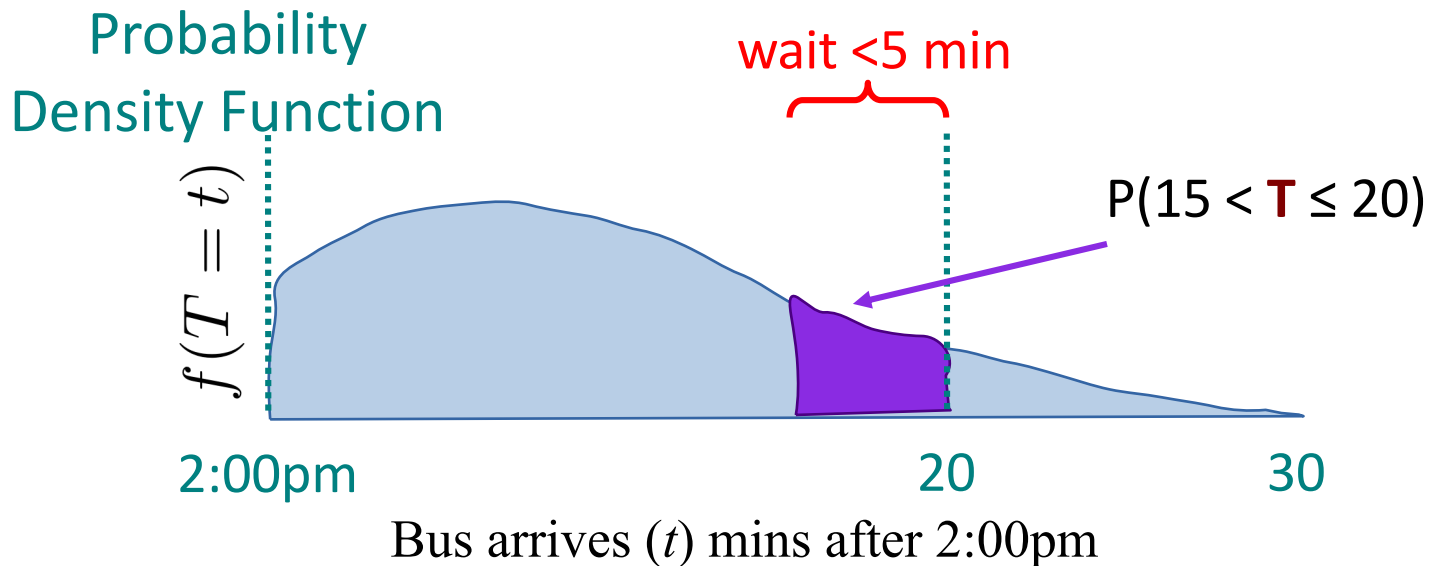
Riding the Marguerite



You are running to the bus stop.
You don't know exactly when the bus arrives. You have a distribution of probabilities.

You show up at 2:15pm.

What is $P(\text{wait} < 5 \text{ minutes})$?



Properties of PDFs

The integral of a PDF gives a probability. Thus:

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

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

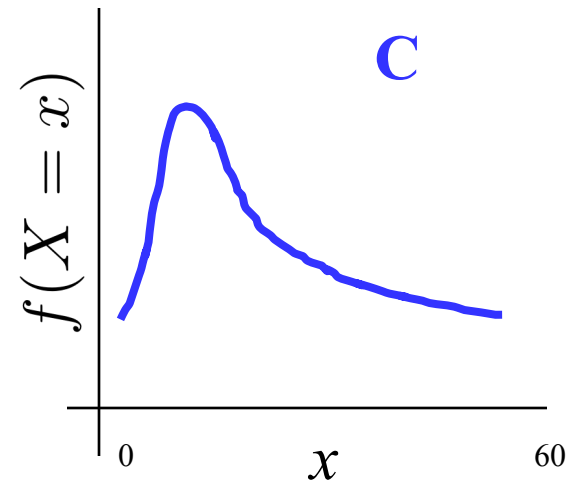
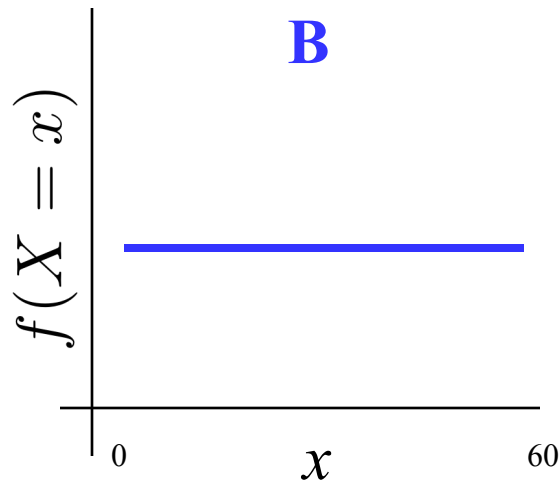
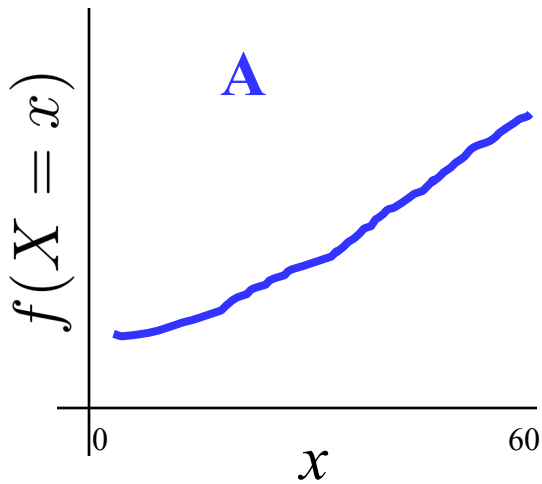
What do you get if you
integrate over a
probability *density* function?

A probability!

Probability Density Function

Probability density functions articulate *relative* belief.

Let X be a random variable which is the # of minutes after 2pm that the bus arrives at the stop:



Which of these represent that you think the arrival is more likely to be close to 3:00pm

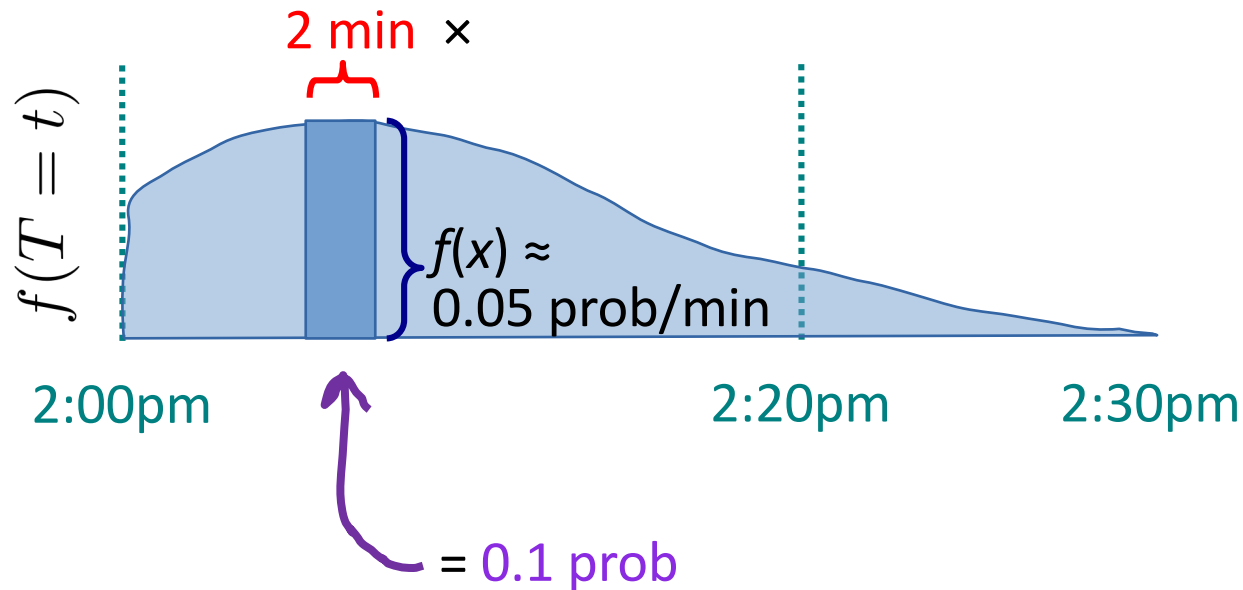


The ratio of probability densities is meaningful

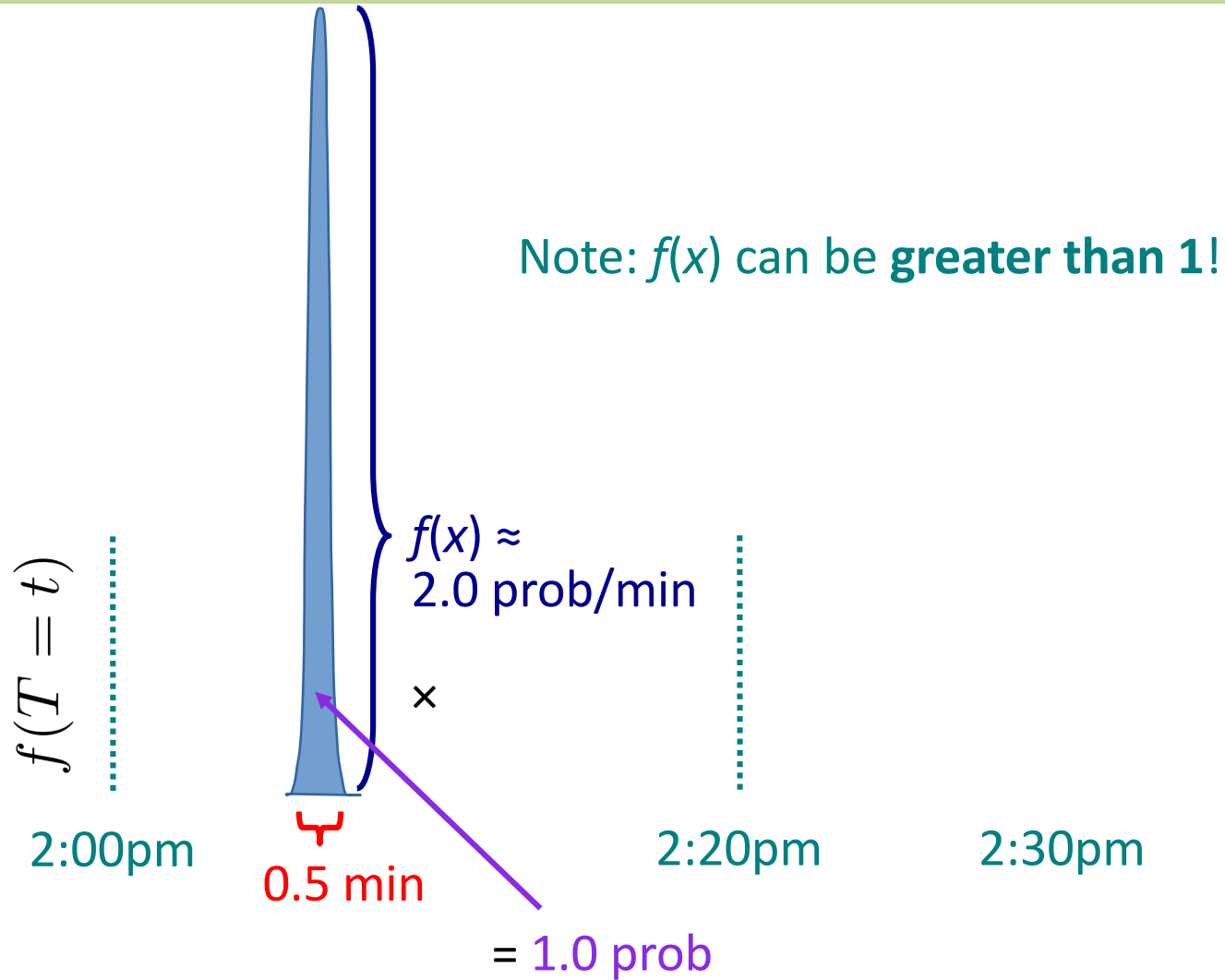


$f(X = x)$ is **Not** a Probability

Rather, it has “units” of:
probability divided by units of X .



$f(X = x)$ is **Not** a Probability



$f(X = x)$ vs $P(X = x)$

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

$$P(X = x)$$

Aka the PMF

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

$$f(X = x)$$

Aka the PDF

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

Simple Example



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

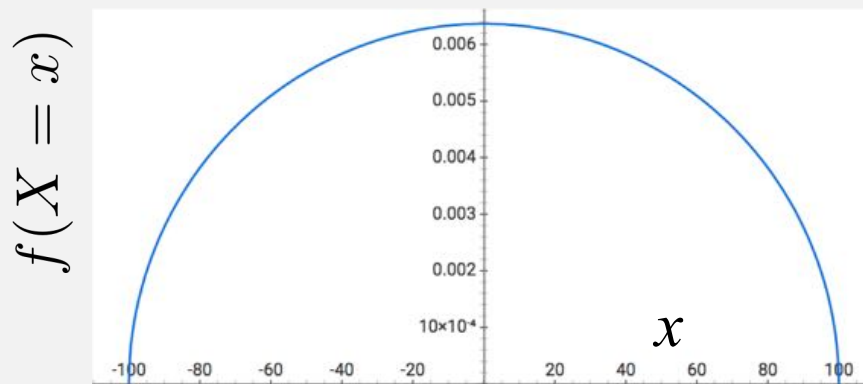
* With help from Wigner, Chris is going to rephrase this problem

Simple Example from Quantum Physics

Let X be a continuous random variable¹

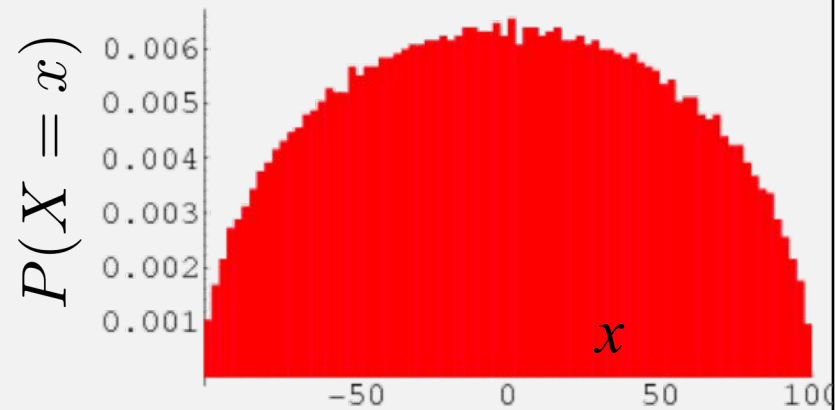
Theory

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



Practice

From simulations



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

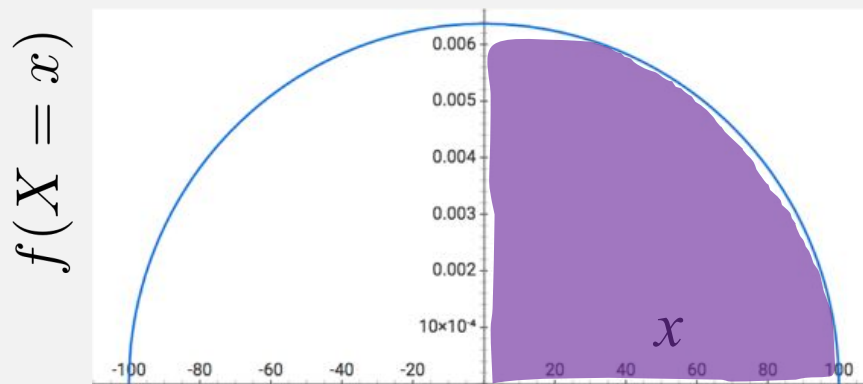
¹ X represents the eigenvalue of a 5000x5000 matrix of uniform random variables

Simple Example from Quantum Physics

Let X be a continuous random variable:

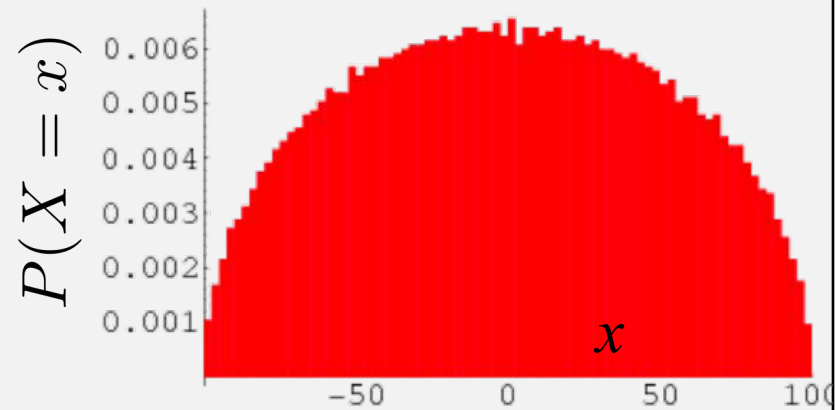
Theory

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



Practice

From simulations



Approach #1: Integrate over the PDF

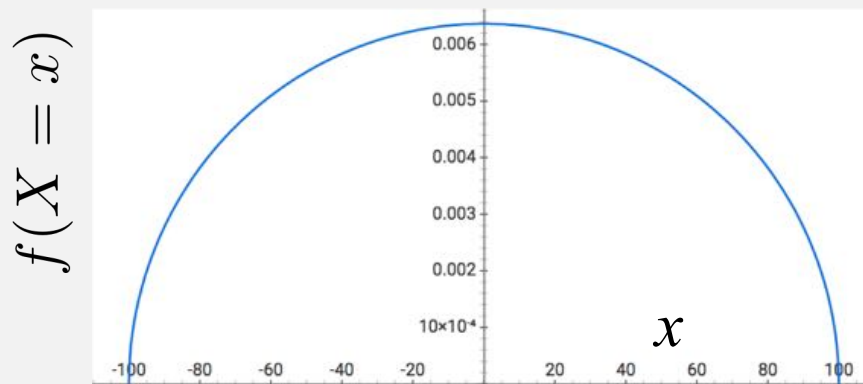
$$P(X > 0) = \int_0^{100} f(X = x) \, dx$$

Simple Example from Quantum Physics

Let X be a continuous random variable:

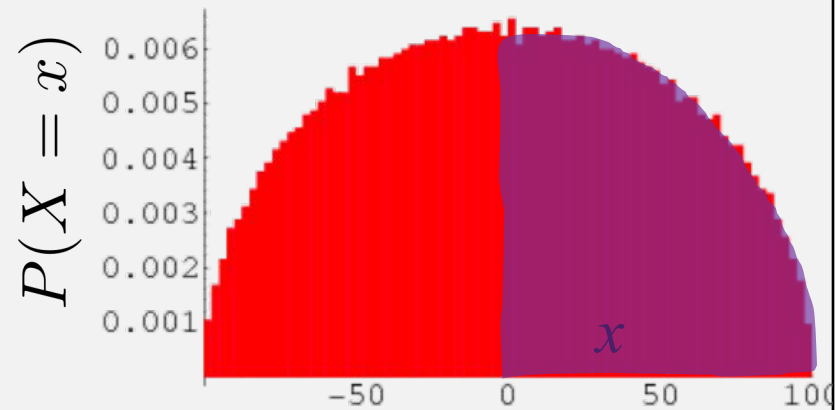
Theory

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



Practice

From simulations



Approach #2: Discrete Approximation

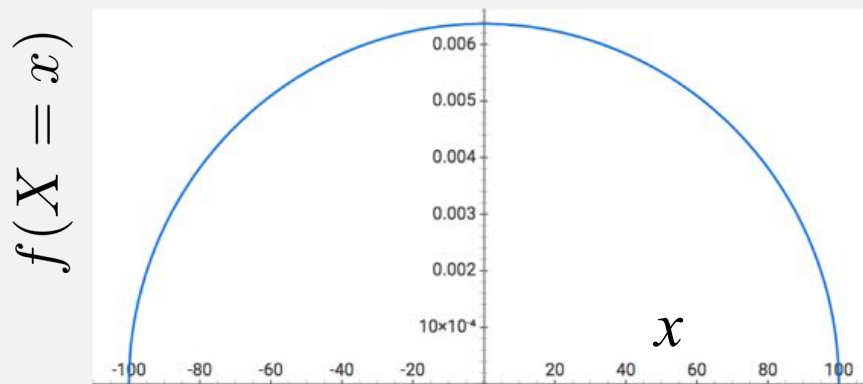
$$P(X > 0) \approx \sum_{i=0}^{100} P(X = i)$$

Simple Example from Quantum Physics

Let X be a continuous random variable:

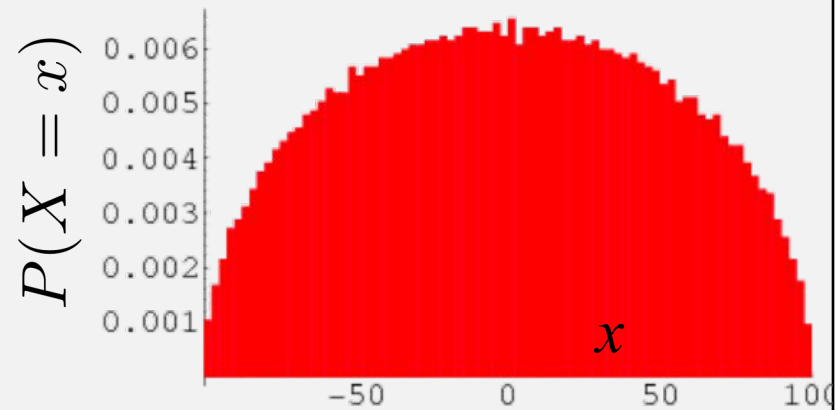
Theory

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



Practice

From simulations



Approach #3: Know Semi-Circles

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

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

A probability!

Uniform Random Variable

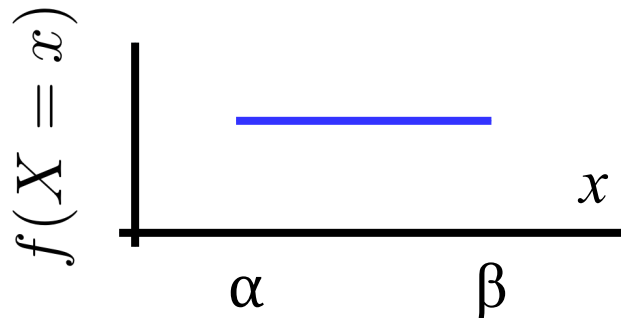
A **uniform** random variable is **equally likely** to be any value in an interval.

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



Probability Density

$$f(X = x) = \begin{cases} \frac{1}{\beta - \alpha} & \alpha \leq x \leq \beta \\ 0 & \text{otherwise} \end{cases}$$



Properties

$$E[X] = \frac{\beta - \alpha}{2}$$

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

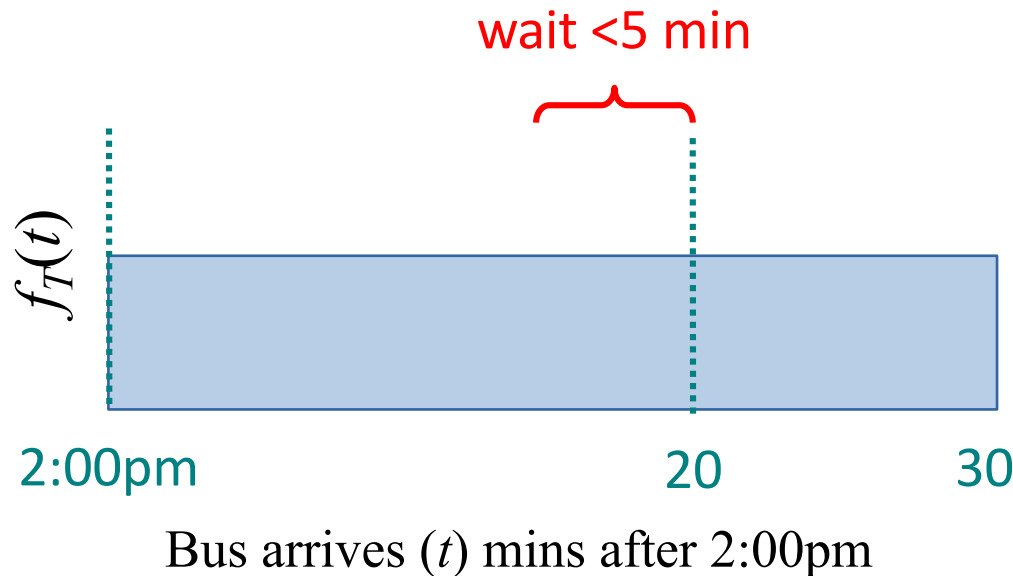
Uniform Bus



You are running to the bus stop.
You don't know exactly when the bus arrives. **You believe all times between 2 and 2:30 are equally likely.**

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

$$T \sim \text{Uni}(\alpha = 0, \beta = 30)$$



$$\begin{aligned} P(\text{Wait} < 5) &= \int_{15}^{20} \frac{1}{\beta - \alpha} dx \\ &= \frac{x}{\beta - \alpha} \Big|_{15}^{20} \\ &= \frac{x}{30 - 0} \Big|_{15}^{20} = \frac{5}{30} \end{aligned}$$

Expectation and Variance

For discrete RV X :

$$E[X] = \sum_x x \cdot p(X = x)$$

$$E[g(X)] = \sum_x g(x) \cdot p(X = x)$$

$$E[X^n] = \sum_x x^n \cdot p(X = x)$$

For continuous RV X :

$$E[X] = \int_{-\infty}^{\infty} x \cdot f_X(x)$$

$$E[g(X)] = \int_{-\infty}^{\infty} g(x) \cdot f_X(x)$$

$$E[X^n] = \int_{-\infty}^{\infty} x^n \cdot f_X(x)$$

For both discrete and continuous RVs:

$$E[aX + b] = aE[X] + b$$

$$\text{Var}(X) = E[(x - \mu)^2] = E[X^2] - (E[X])^2$$

$$\text{Var}(aX + b) = a^2 \text{Var}(X)$$

Expectation of Uniform

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

$$E[X] = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

$$= \int_{\alpha}^{\beta} x \cdot \frac{1}{\beta - \alpha} dx$$

$$= \frac{1}{\beta - \alpha} \left[\frac{1}{2} x^2 \right]_{\alpha}^{\beta}$$

$$= \frac{1}{\beta - \alpha} \left[\frac{\beta^2}{2} - \frac{\alpha^2}{2} \right]$$

$$= \frac{1}{2} \frac{1}{\beta - \alpha} (\beta + \alpha)(\beta - \alpha)$$

just average
the start and
end!

$$= \frac{1}{2}(\alpha + \beta)$$

Exponential Random Variable

- X is an **Exponential RV**: $X \sim \text{Exp}(\lambda)$ Rate: $\lambda > 0$

- Probability Density Function (PDF):

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

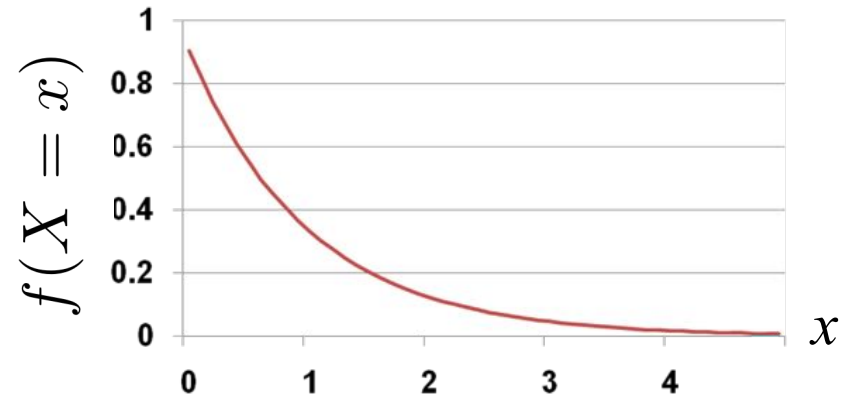
- $E[X] = \frac{1}{\lambda}$

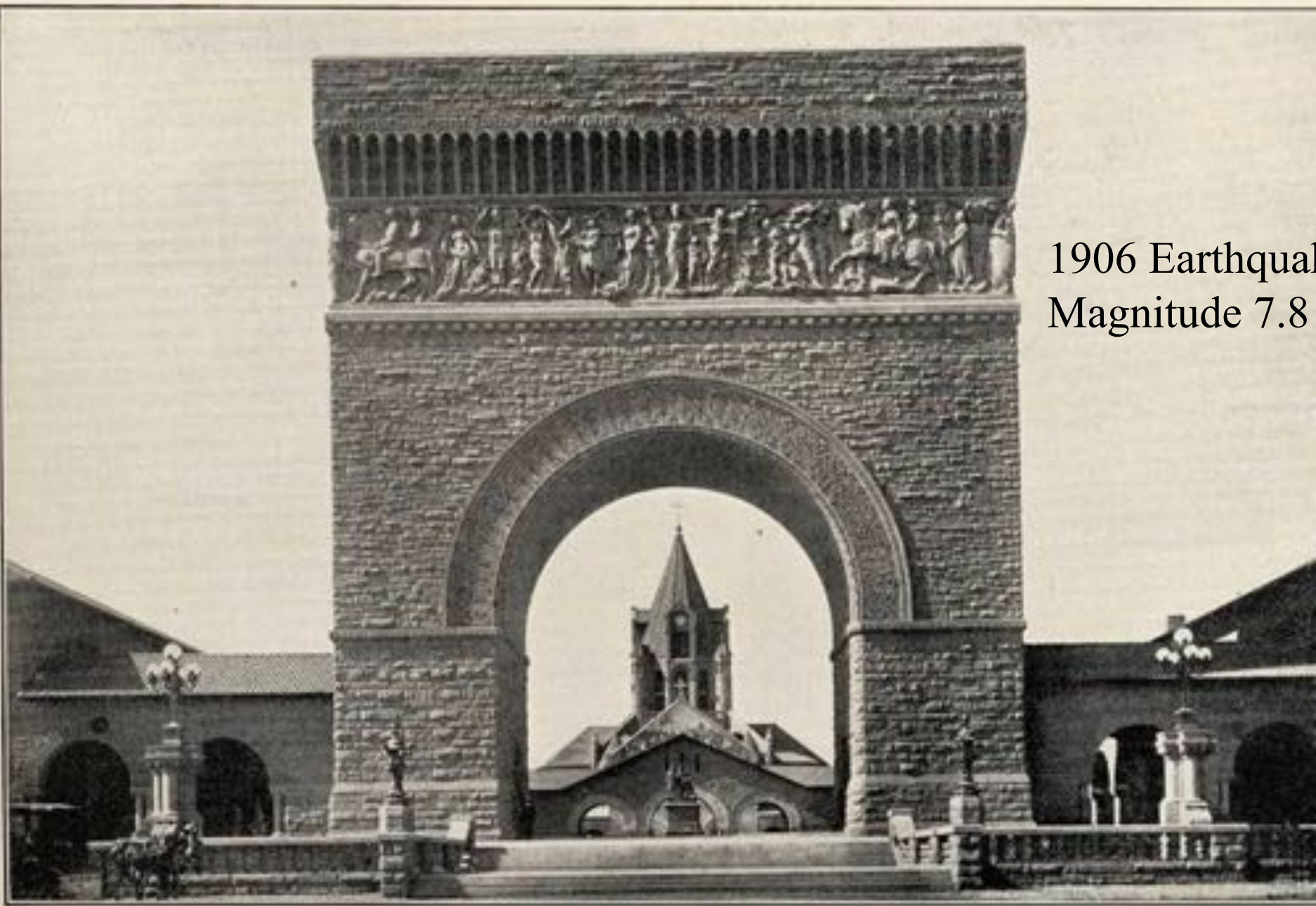
- $\text{Var}(X) = \frac{1}{\lambda^2}$

- Support: $0 \leq x \leq \infty$

- Represents time until some event

- Earthquake, request to web server, end cell phone contract, etc.





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 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+

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

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

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

Integral Review

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

How Long Until 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+

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

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

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

How Long Until 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 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

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

$$x = 2$$


$$0.03125$$

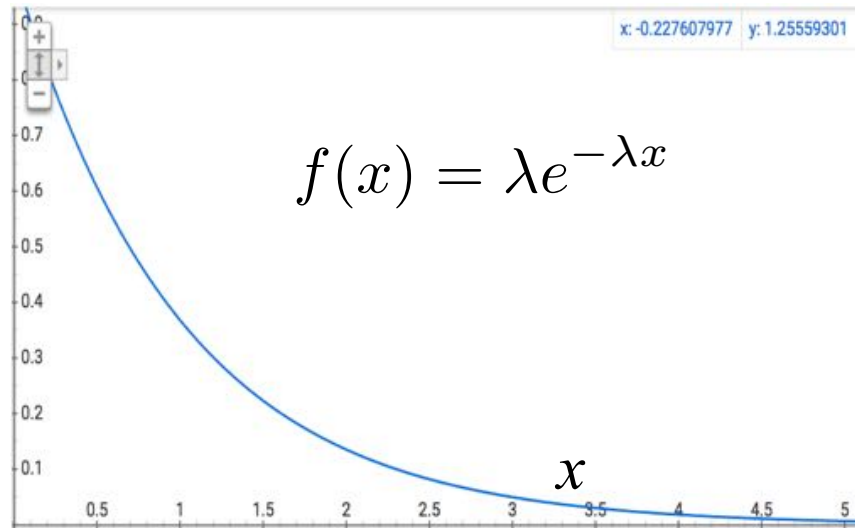

CDF of an Exponential

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

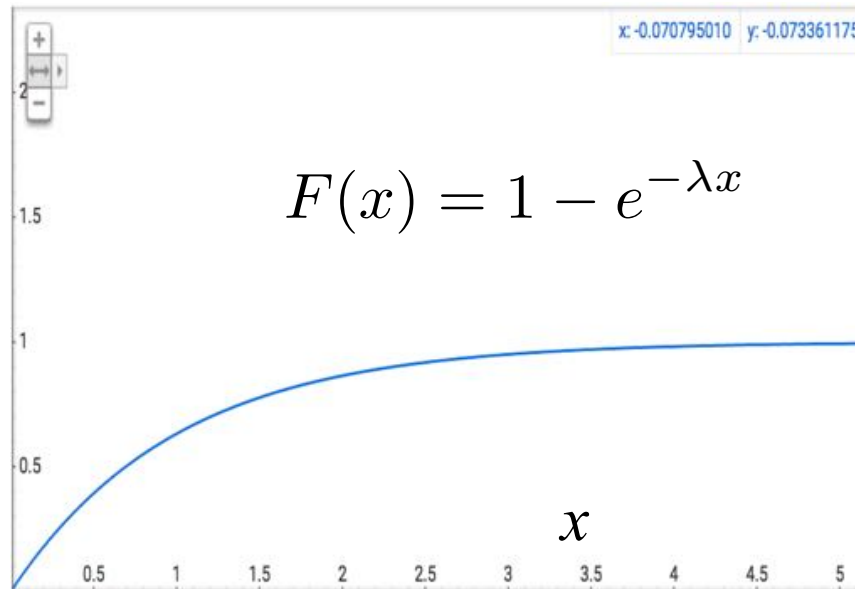
$$\begin{aligned} P(X < x) &= \int_{y=-\infty}^x f(y) \, dy \\ &= \int_{y=0}^x \lambda e^{-\lambda y} \, dy \\ &= \frac{\lambda}{\lambda} \left[-e^{-\lambda y} \right]_0^x \\ &= [-e^{-\lambda x}] - [-e^{\lambda 0}] \\ &= 1 - e^{-\lambda x} \end{aligned}$$

CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



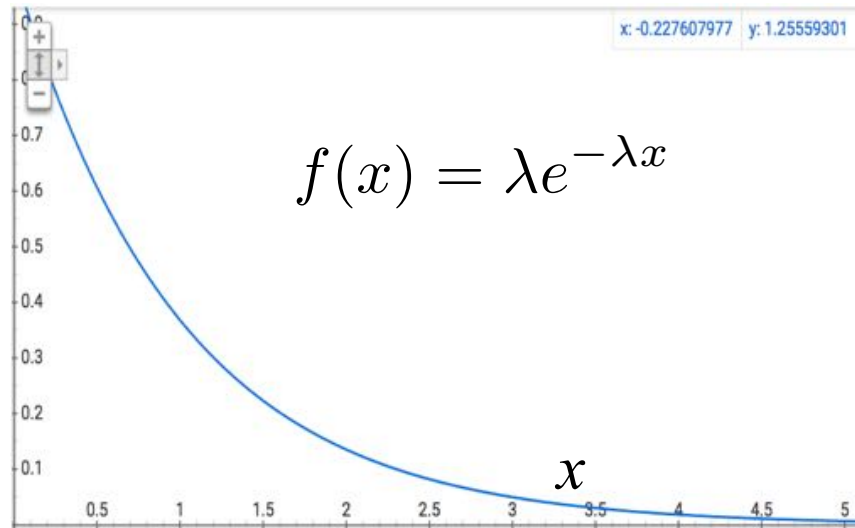
*Cumulative
density
function*



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

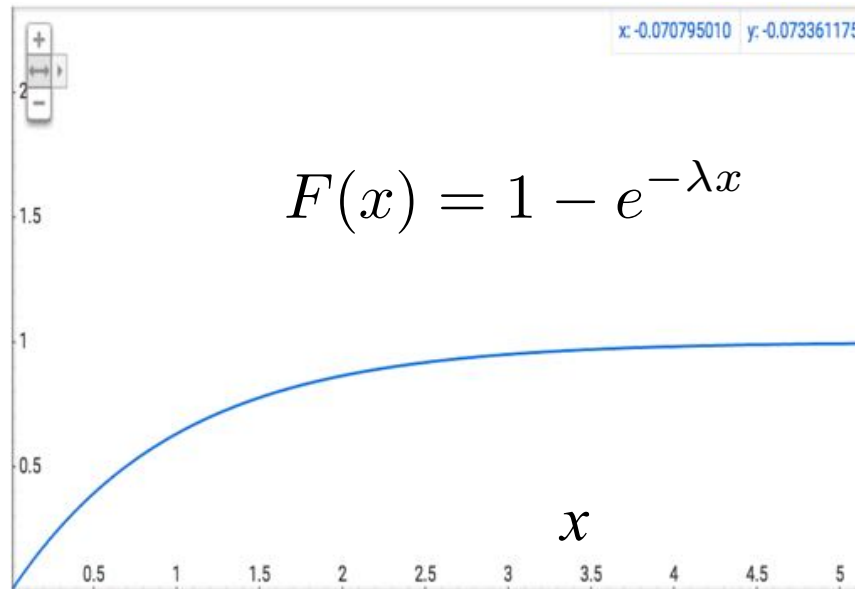
CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



$$P(X < 2)$$

*Cumulative
density
function*

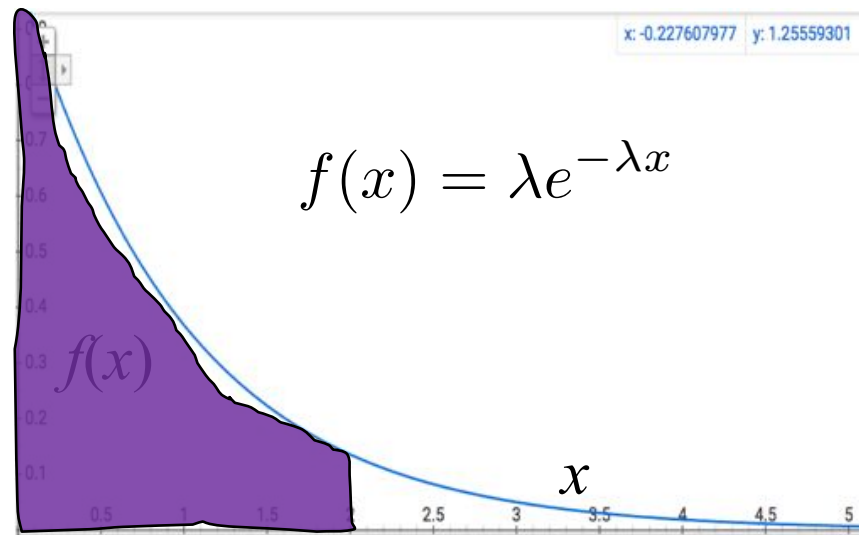


$$F_X(x) = P(X < x)$$

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

CDF: $X \sim \text{Exp}(\lambda = 1)$

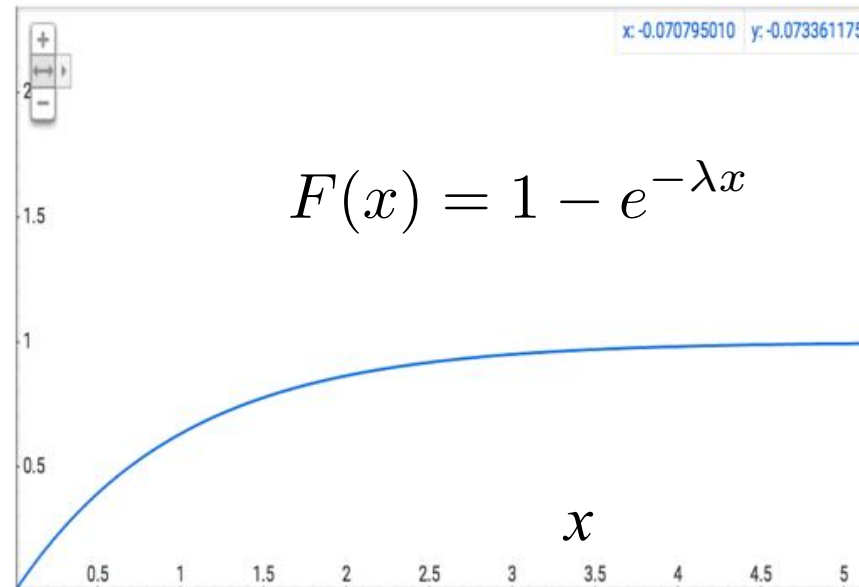
*Probability
density
function*



$$P(X < 2)$$

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

*Cumulative
density
function*

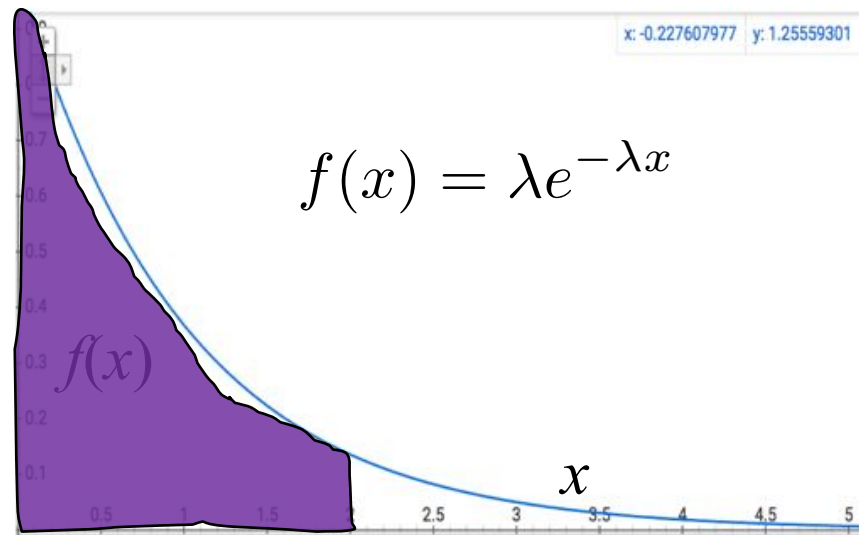


$$F_X(x) = P(X < x)$$

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

CDF: $X \sim \text{Exp}(\lambda = 1)$

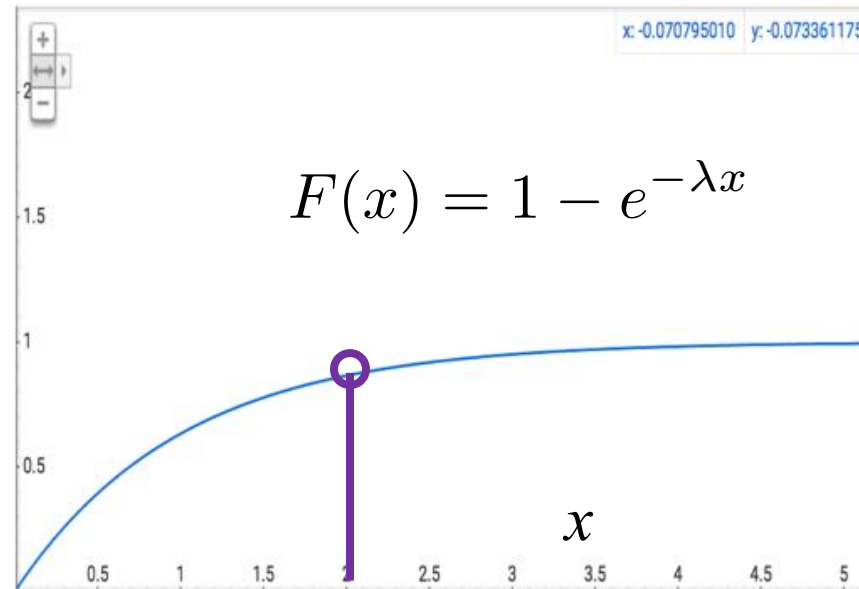
*Probability
density
function*



$$P(X < 2)$$

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

*Cumulative
density
function*



or

$$= F(2)$$

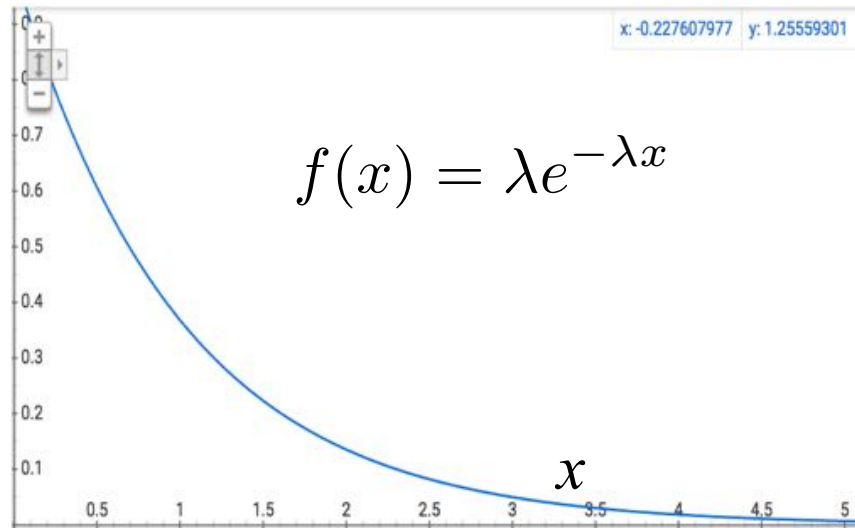
$$= 1 - e^{-2}$$
$$\approx 0.84$$

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

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

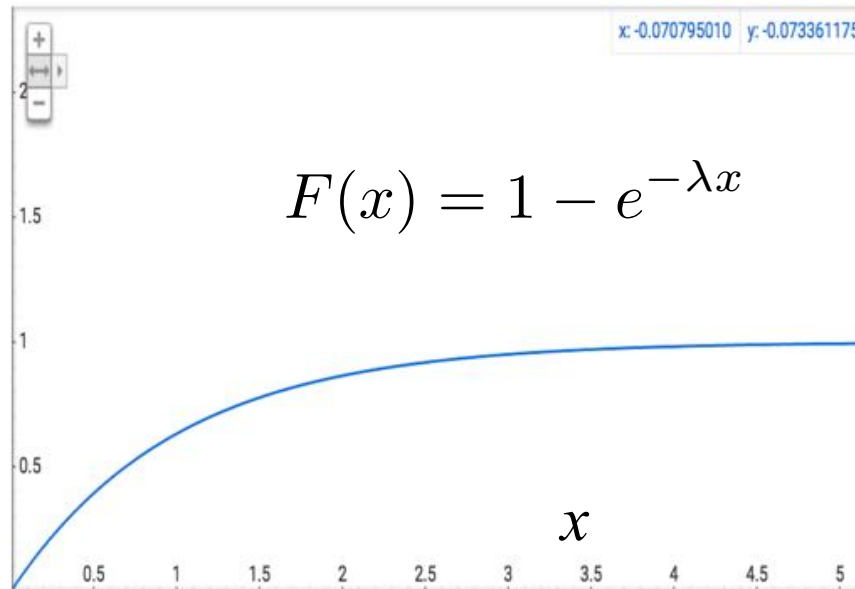
CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



$$P(X > 1)$$

*Cumulative
density
function*

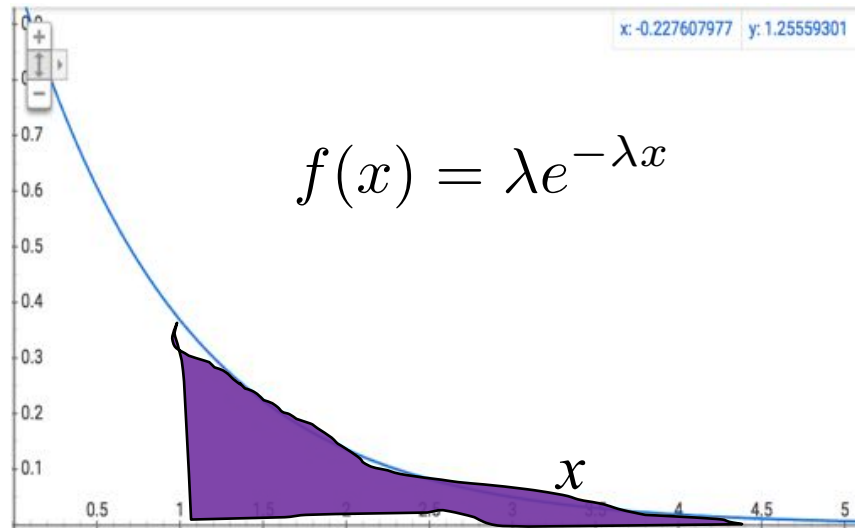


$$F_X(x) = P(X < x)$$

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

CDF: $X \sim \text{Exp}(\lambda = 1)$

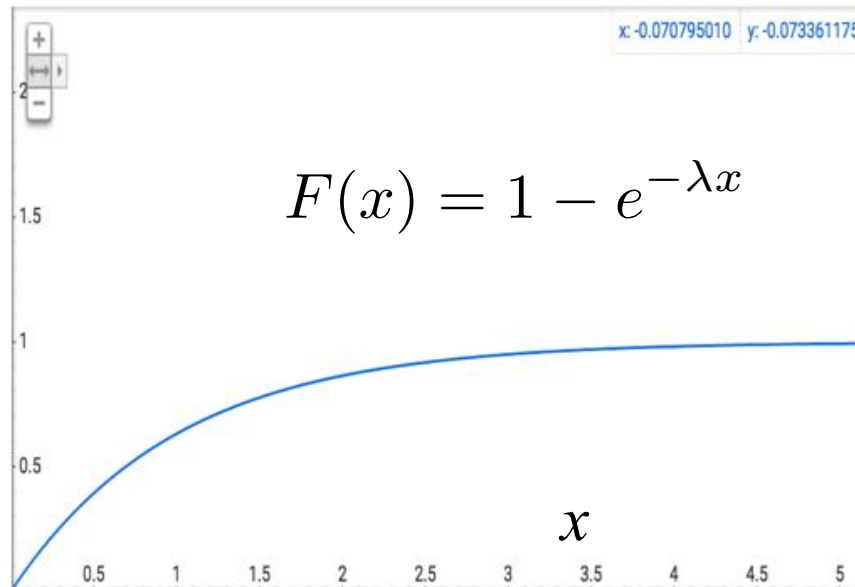
*Probability
density
function*



$$P(X > 1)$$

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

*Cumulative
density
function*

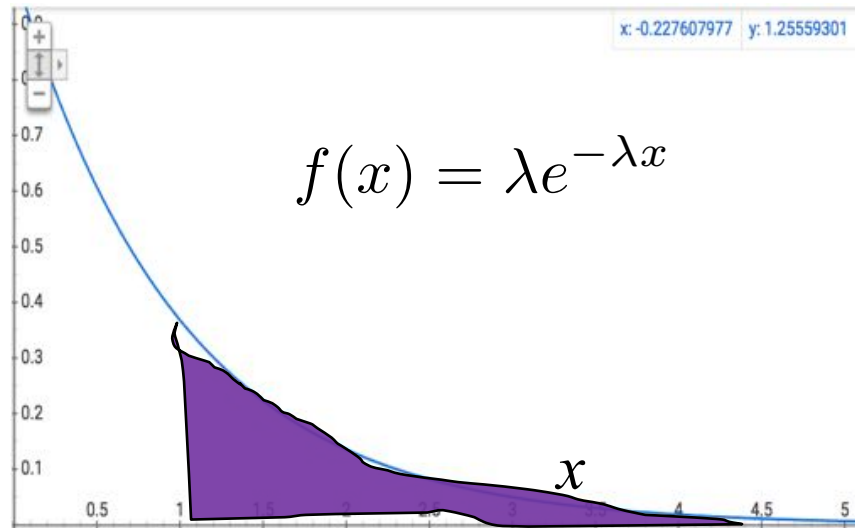


$$F_X(x) = P(X < x)$$

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

CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



$$P(X > 1)$$

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

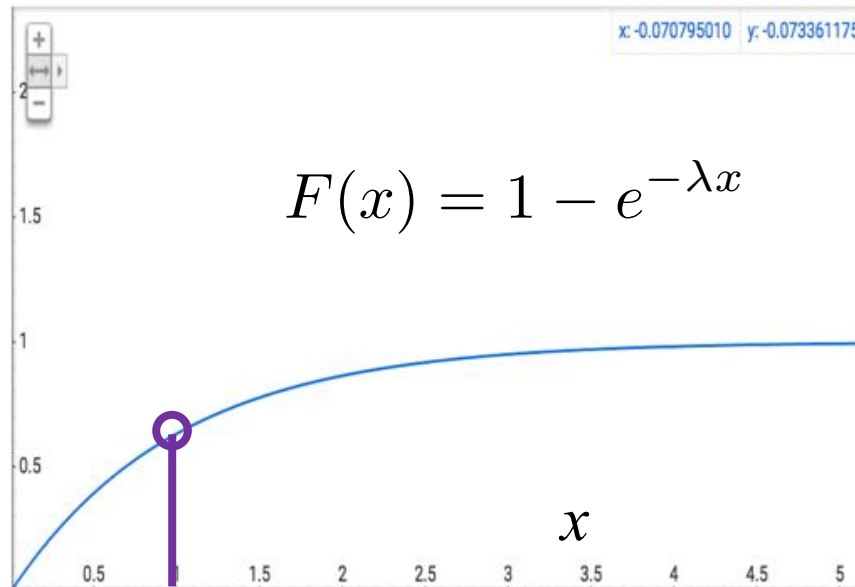
or

$$= 1 - F(1)$$

$$= e^{-1}$$

$$\approx 0.37$$

*Cumulative
density
function*

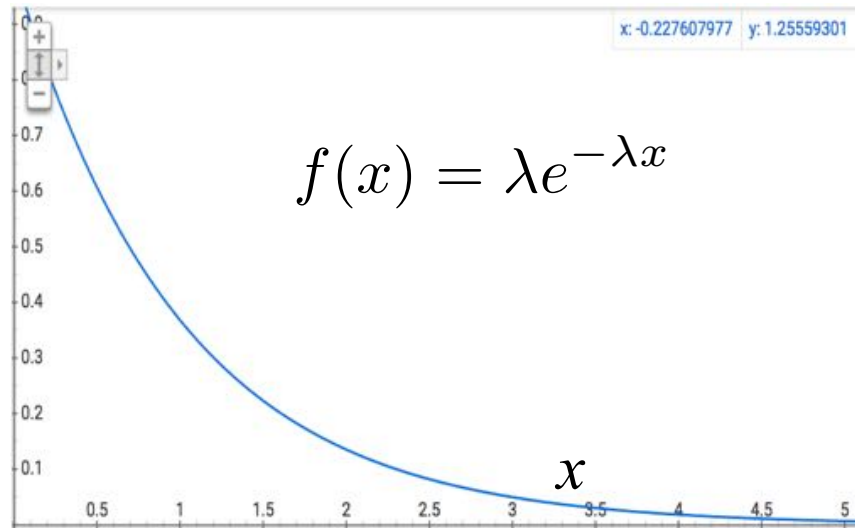


$$F_X(x) = P(X < x)$$

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

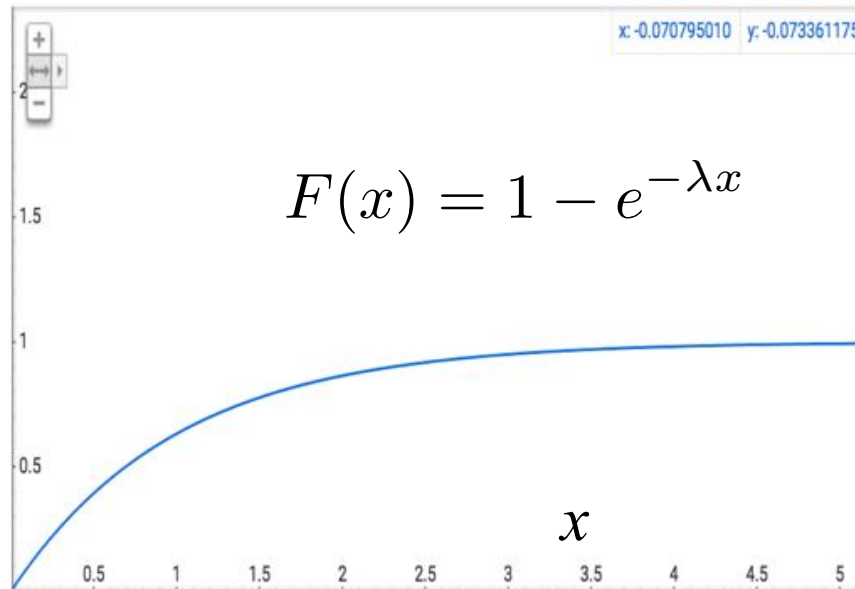
CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



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

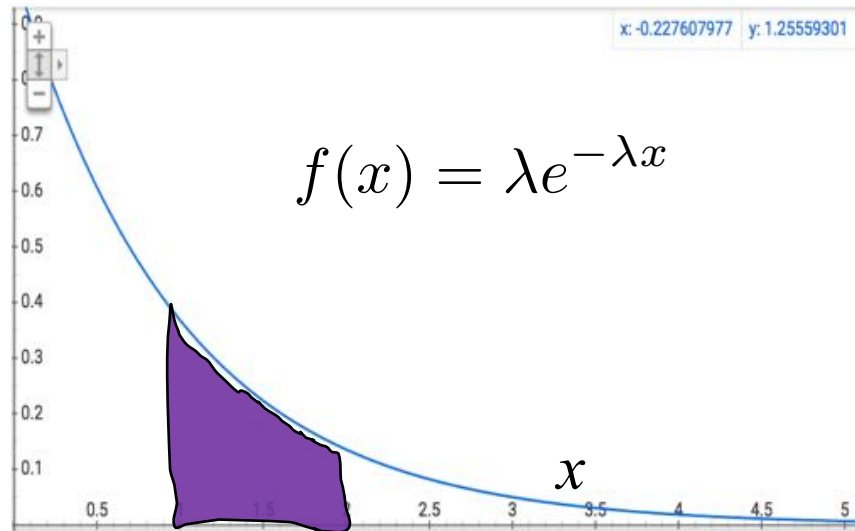
*Cumulative
density
function*



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

CDF: $X \sim \text{Exp}(\lambda = 1)$

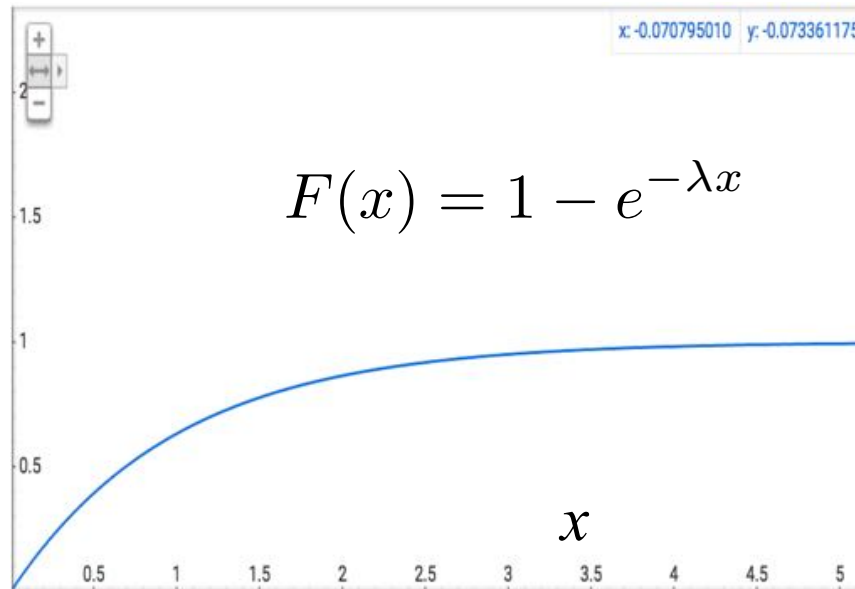
*Probability
density
function*



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

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

*Cumulative
density
function*

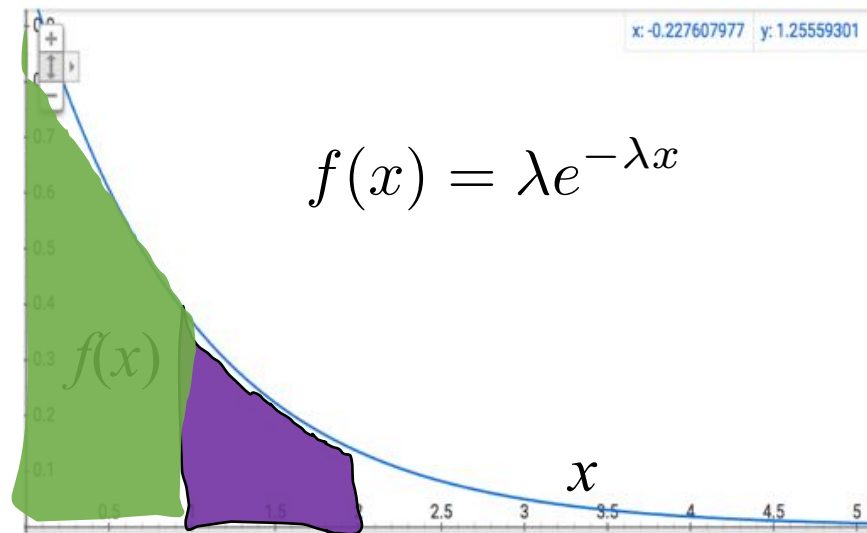


$$F_X(x) = P(X < x)$$

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

CDF: $X \sim \text{Exp}(\lambda = 1)$

*Probability
density
function*



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

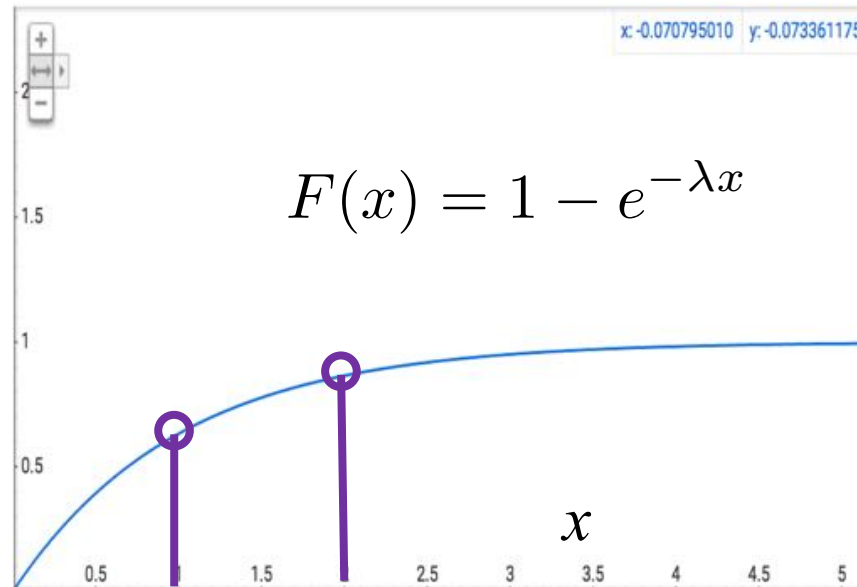
$$= \int_{x=1}^2 f(x) dx$$

or

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

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

*Cumulative
density
function*



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

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

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+

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

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

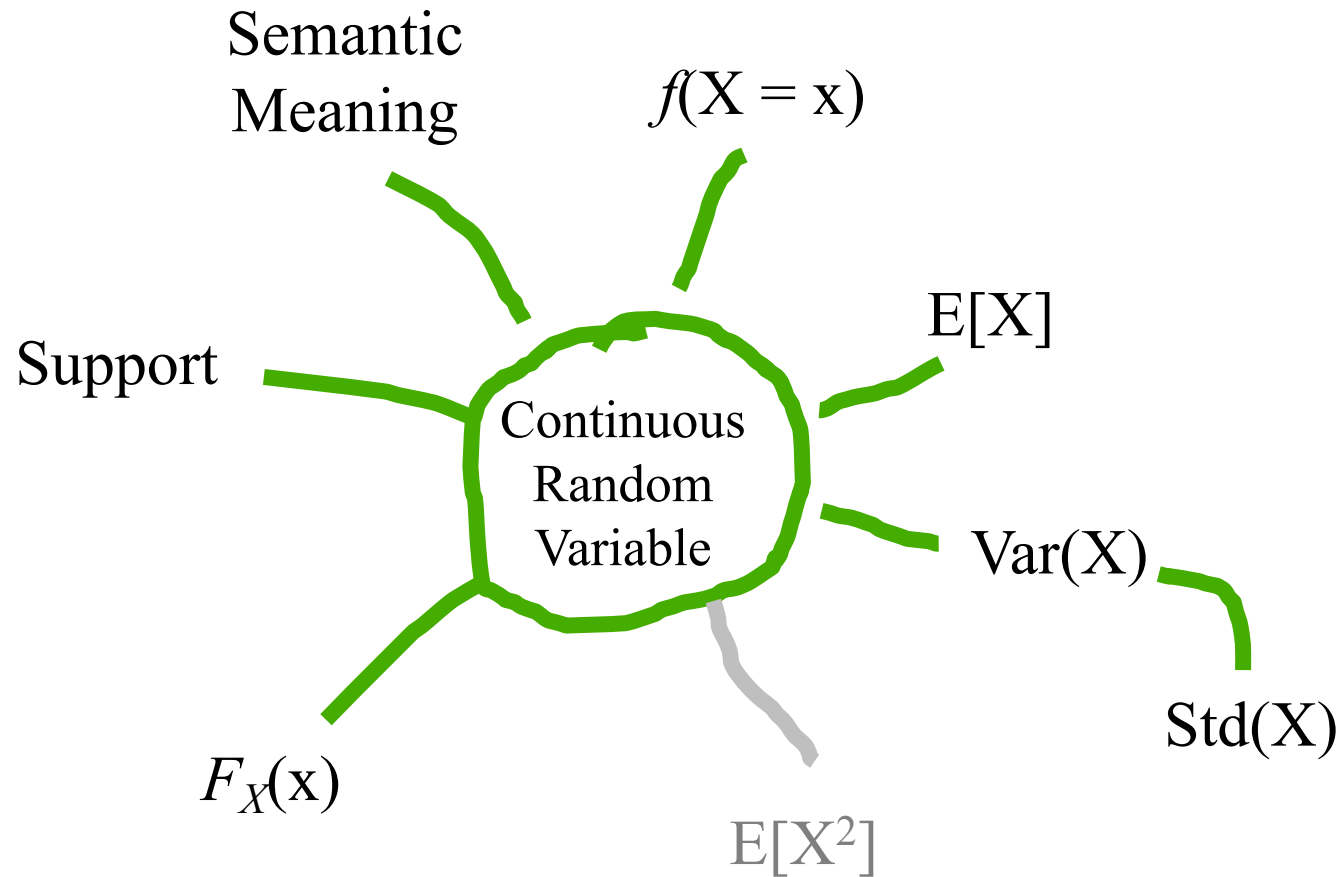
$$P(Y < 4) = F(4)$$

$$= 1 - e^{-0.002 \cdot 4}$$

$$\approx 0.008$$

Feeling lucky?

Properties for Continuous Random Variables



Extra Problems

Visits to a Website

- Say visitor to your web site leaves after X minutes
 - On average, visitors leave site after 5 minutes
 - Assume length of stay is Exponentially distributed
 - $X \sim \text{Exp}(\lambda = 1/5)$, since $E[X] = 1/\lambda = 5$
 - What is $P(X > 10)$?

$$P(X > 10) = 1 - F(10) = 1 - (1 - e^{-\lambda 10}) = e^{-2} \approx 0.1353$$

- What is $P(10 < X < 20)$?

$$P(10 < X < 20) = F(20) - F(10) = (1 - e^{-4}) - (1 - e^{-2}) \approx 0.1170$$

Replacing Your Laptop

- X = # hours of use until your laptop dies
 - On average, laptops die after 5000 hours of use
 - $X \sim \text{Exp}(\lambda = 1/5000)$, since $E[X] = 1/\lambda = 5000$
 - You use your laptop 5 hours/day.
 - What is $P(\text{your laptop lasts 4 years})$?
 - That is: $P(X > (5)(365)(4) = 7300)$

$$P(X > 7300) = 1 - F(7300) = 1 - (1 - e^{-7300/5000}) = e^{-1.46} \approx 0.2322$$

- Better plan ahead... especially if you are cotermining:

$$P(X > 9125) = 1 - F(9125) = e^{-1.825} \approx 0.1612 \quad (5 \text{ year plan})$$

$$P(X > 10950) = 1 - F(10950) = e^{-2.19} \approx 0.1119 \quad (6 \text{ year plan})$$

Exponential is Memoryless

- X = time until some event occurs
 - $X \sim \text{Exp}(\lambda)$
 - What is $P(X > s + t \mid X > s)$?

$$P(X > s + t \mid X > s) = \frac{P(X > s + t \text{ and } X > s)}{P(X > s)} = \frac{P(X > s + t)}{P(X > s)}$$

$$\frac{P(X > s + t)}{P(X > s)} = \frac{1 - F(s + t)}{1 - F(s)} = \frac{e^{-\lambda(s+t)}}{e^{-\lambda s}} = e^{-\lambda t} = 1 - F(t) = P(X > t)$$

$$\text{So, } P(X > s + t \mid X > s) = P(X > t)$$

- After initial period of time s , $P(X > t \mid \bullet)$ for waiting another t units of time until event is same as at start
- “Memoryless” = no impact from preceding period s

Disk Crashes

- X = days of use before your disk crashes

$$f(x) = \begin{cases} \lambda e^{-x/100} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

- First, determine λ to have actual PDF

- Good integral to know: $\int e^u du = e^u$

$$1 = \int \lambda e^{-x/100} dx = -100\lambda \int \frac{-1}{100} e^{-x/100} dx = -100\lambda e^{-x/100} \Big|_0^\infty = 100\lambda \Rightarrow \lambda = \frac{1}{100}$$

- What is $P(50 < X < 150)$?

$$F(150) - F(50) = \int_{50}^{150} \frac{1}{100} e^{-x/100} dx = -e^{-x/100} \Big|_{50}^{150} = -e^{-3/2} + e^{-1/2} \approx 0.383$$

- What is $P(X < 10)$?

$$F(10) = \int_0^{10} \frac{1}{100} e^{-x/100} dx = -e^{-x/100} \Big|_0^{10} = -e^{-1/10} + 1 \approx 0.095$$

Zipf Random Variable

- X is Zipf RV: $X \sim \text{Zipf}(s)$
 - X is the rank index of a chosen word

