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# 09: Continuous RVs

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Jerry Cain  
January 29, 2024

[Lecture Discussion on Ed](#)



# Continuous RVs

# People heights

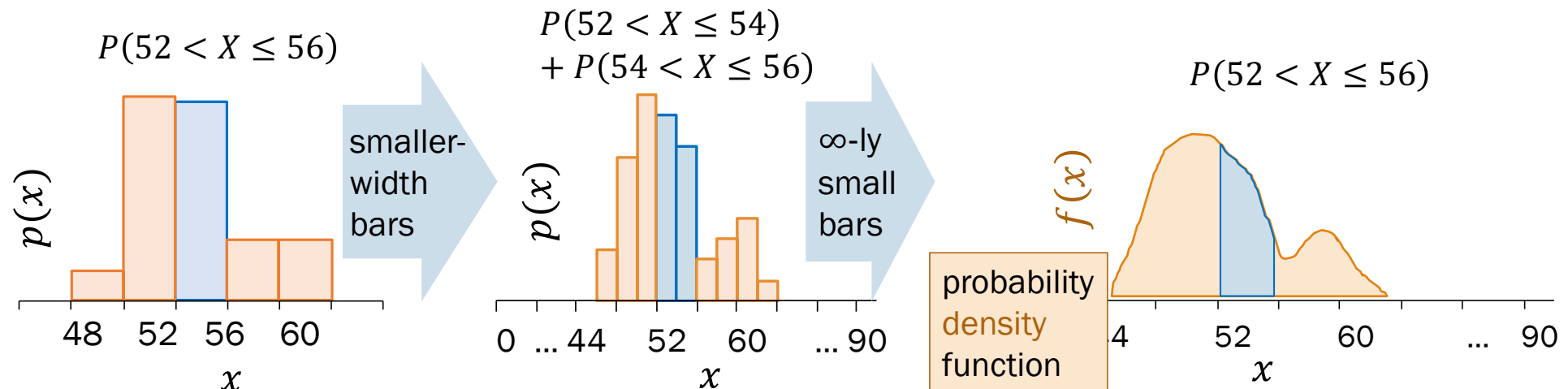
You are volunteering at the local elementary school fundraiser.

- To buy a t-shirt for your friend Vanessa, you need to know her height.

1. What is the probability that your friend is 54.0923857234 inches tall?

Essentially 0

2. What is the probability that Vanessa is between 52–56 inches tall?



# Continuous RV definition

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A random variable  $X$  is **continuous** if there is a **probability density function**  $f(x) \geq 0$  such that for  $-\infty < x < \infty$ :

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

Integrating a PDF must always yield a valid probability, no matter the values of  $a$  and  $b$ . The PDF must also satisfy:

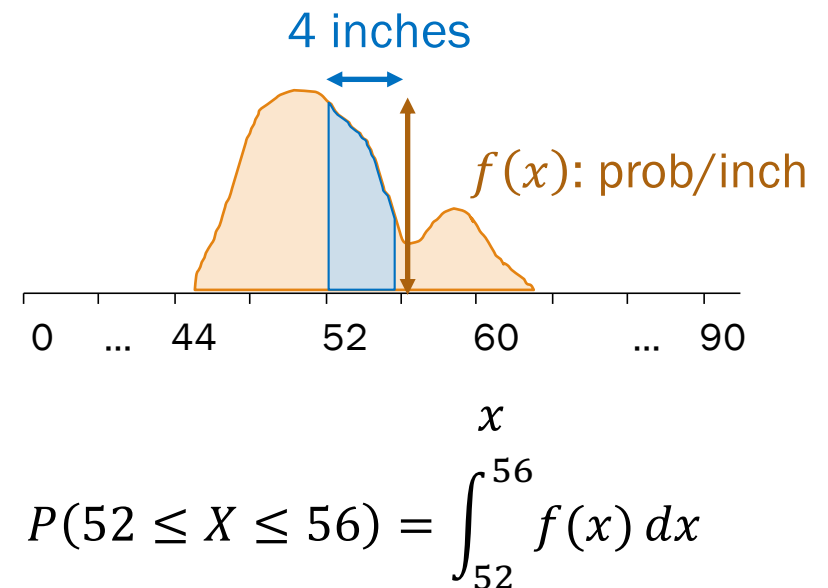
$$\int_{-\infty}^{\infty} f(x) dx = P(-\infty < X < \infty) = 1$$

Note:  $f(x)$  is sometimes written as  $f_X(x)$  to be clear the random variable is  $X$ .

# Main takeaway #1

Integrate  $f(x)$  to get probabilities.

PDF Units: probability per units of  $X$



# PMF vs PDF

**Discrete** random variable  $X$

Probability mass function (PMF):

$$p(x)$$

To get probability:

$$P(X = x) = p(x)$$

**Continuous** random variable  $X$

Probability density function (PDF):

$$f(x)$$

To get probability:

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

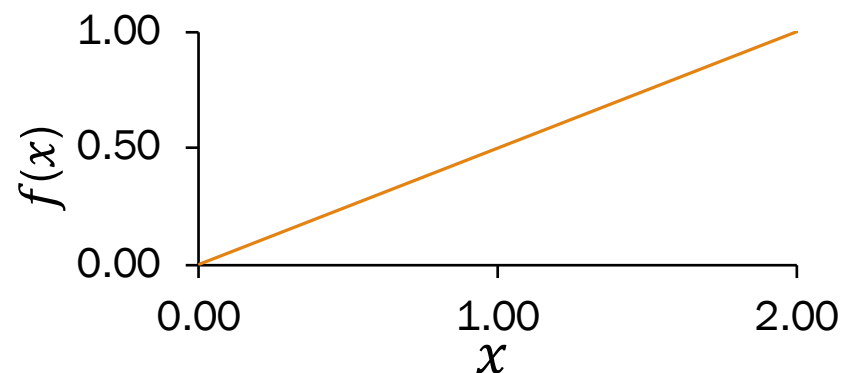
Both are measures of how likely  $X$  is to take on a value or some range of values.

# Computing probability

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

Let  $X$  be a continuous RV with PDF:

$$f(x) = \begin{cases} \frac{x}{2} & \text{if } 0 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$



What is  $P(X \geq 1)$ ?

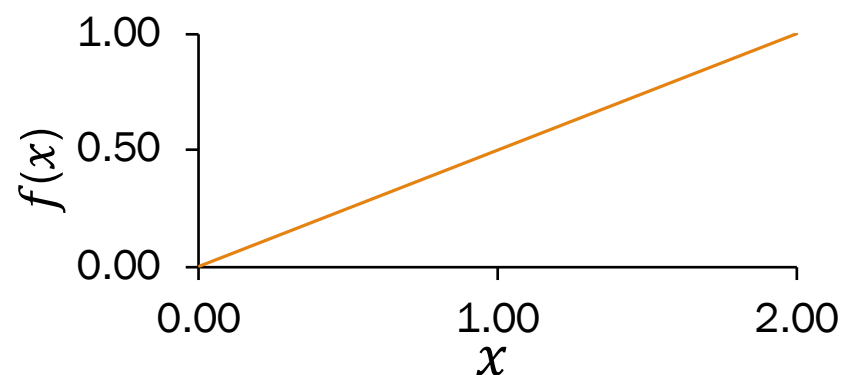


# Computing probability

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

Let  $X$  be a continuous RV with PDF:

$$f(x) = \begin{cases} \frac{x}{2} & \text{if } 0 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$



What is  $P(X \geq 1)$ ?

Strategy 1: Integrate

$$\begin{aligned} P(1 \leq X < \infty) &= \int_1^{\infty} f(x) dx = \int_1^2 \frac{1}{2} x dx \\ &= \frac{1}{2} \left( \frac{1}{2} x^2 \right) \Big|_1^2 = \frac{1}{2} \left[ 2 - \frac{1}{2} \right] = \frac{3}{4} \end{aligned}$$

Strategy 2: Know triangles

$$1 - \frac{1}{2} \left( \frac{1}{2} \right) = \frac{3}{4}$$

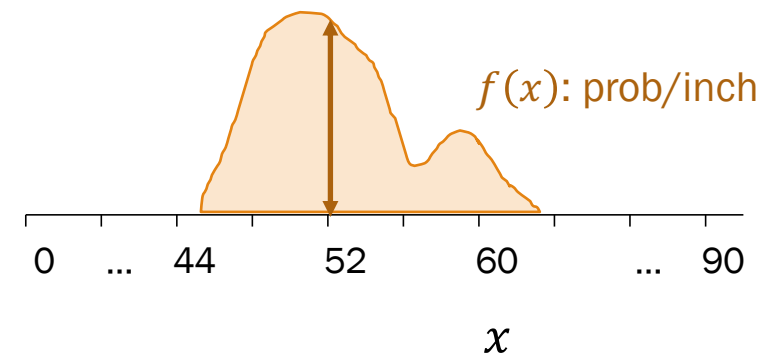
Wait! Is this even legal?

$$P(0 \leq X < 1) = \int_0^1 f(x) dx ??$$



## Main takeaway #2

For a continuous random variable  $X$  with PDF  $f(x)$ ,  
$$P(X = c) = \int_c^c f(x)dx = 0.$$



Contrast with PMF in discrete case:  $P(X = c) = p(c)$

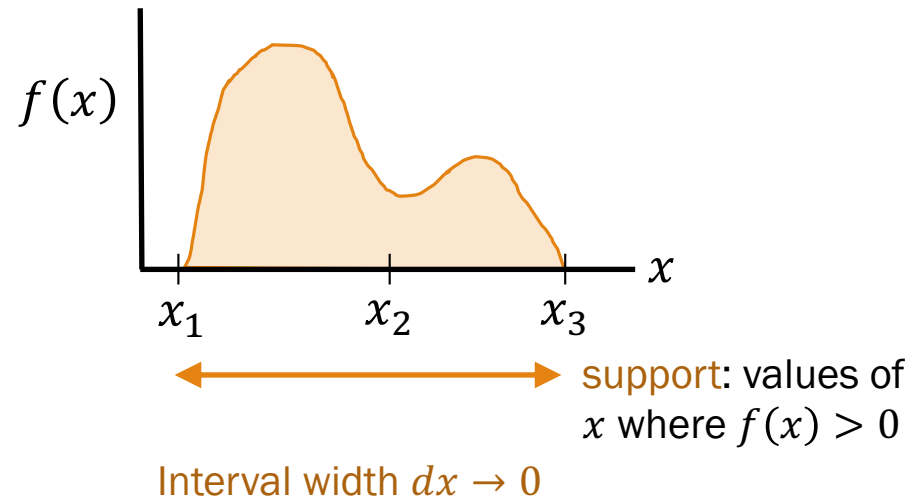
# PDF Properties

For a **continuous** RV  $X$  with PDF  $f$ ,

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

True/False:

- ★ 1.  $P(X = c) = 0$
- ★ 2.  $P(a \leq X \leq b) = P(a < X < b) = P(a \leq X < b) = P(a < X \leq b)$
- ✗ 3.  $f(x)$  is a probability
- ★ 4. In the graphed PDF above,  
 $P(x_1 \leq X \leq x_2) > P(x_2 \leq X \leq x_3)$



It's a probability density!

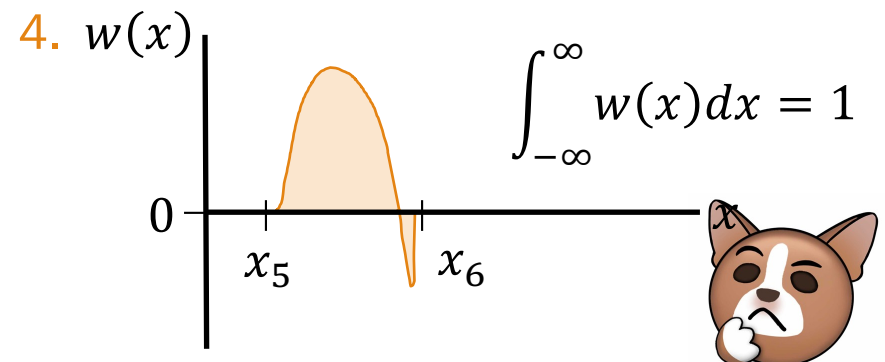
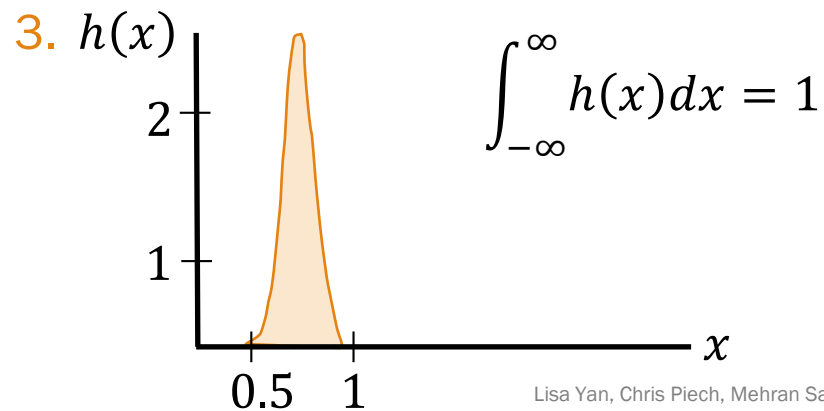
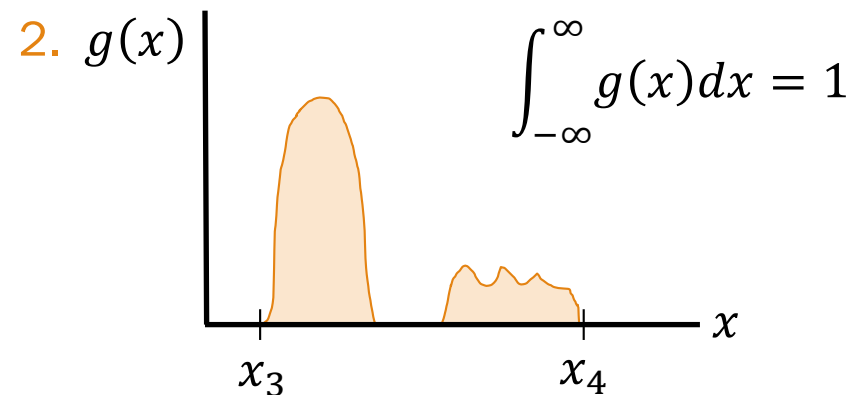
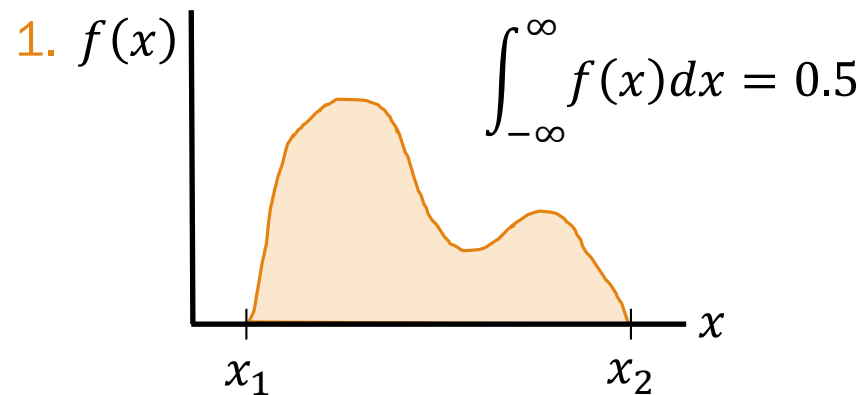
Compare area under the curve



# Determining valid PDFs

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

Which of the following functions are valid PDFs?





# Uniform RV

# Uniform Random Variable

def A **Uniform** random variable  $X$  is defined as follows:

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

PDF

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

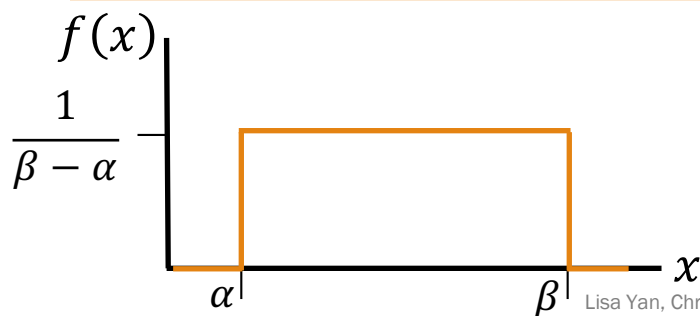
Support:  $[\alpha, \beta]$   
(sometimes defined  
over  $(\alpha, \beta)$ )

Expectation

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

Variance

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

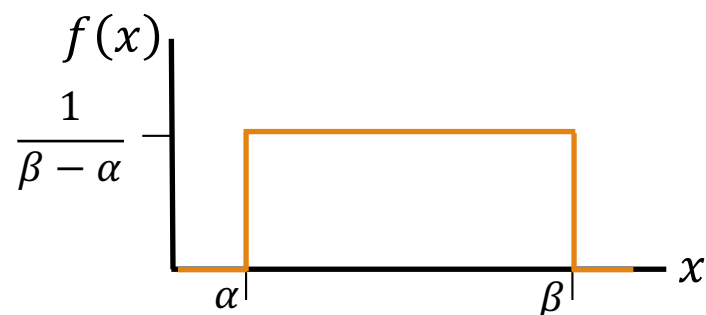


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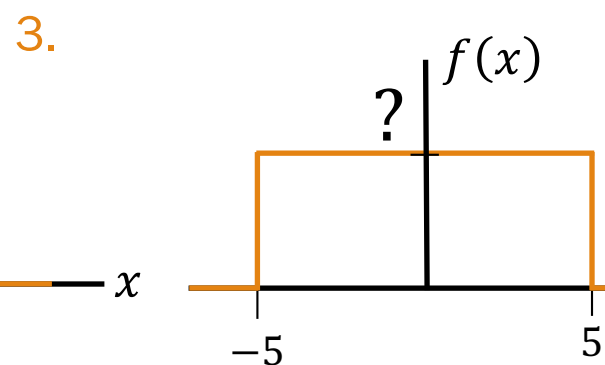
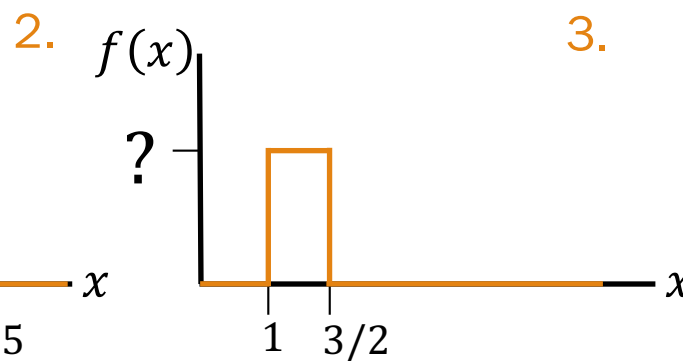
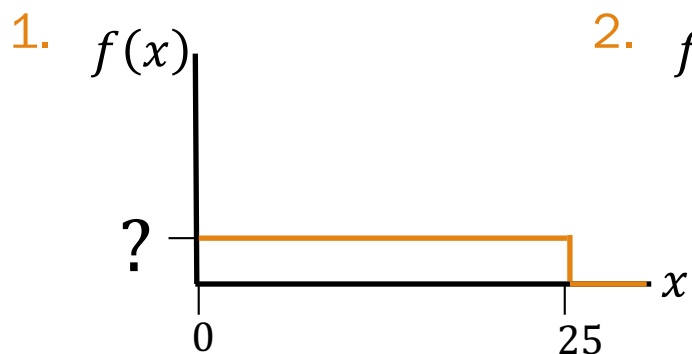
# Quick check

If  $X \sim \text{Uni}(\alpha, \beta)$ , the PDF of  $X$  is:

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



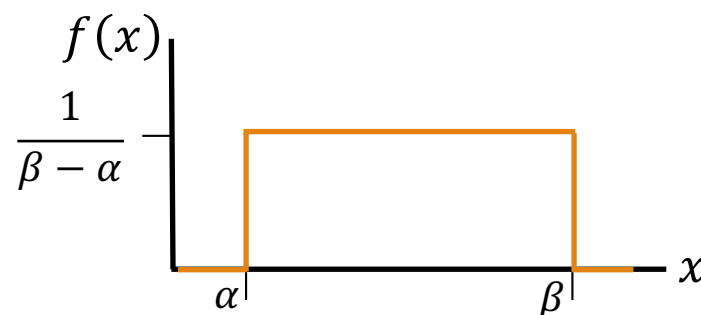
What is  $\frac{1}{\beta - \alpha}$  if the following graphs are PDFs of Uniform RVs  $X$ ?



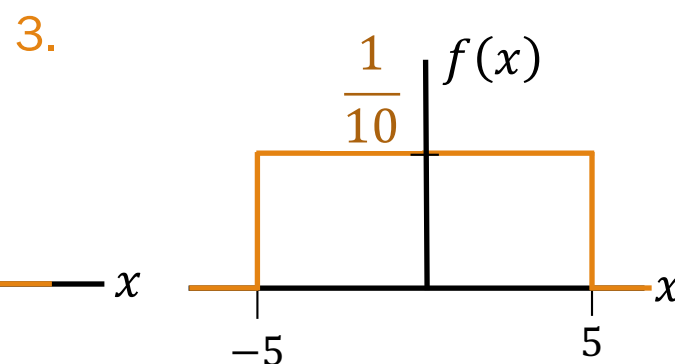
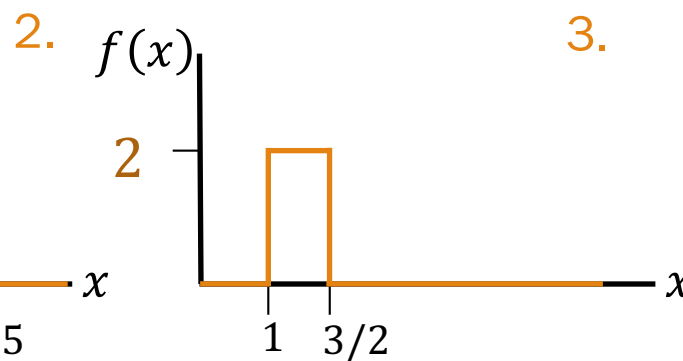
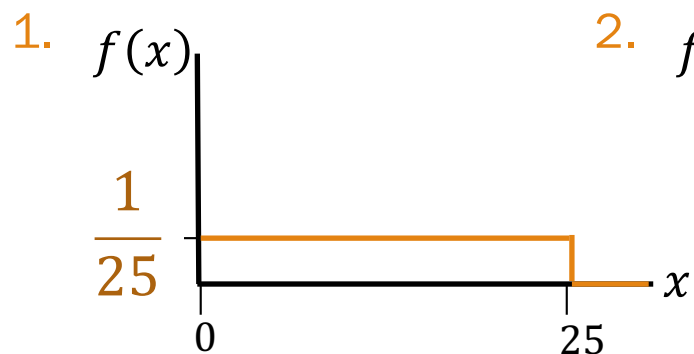
# Quick check

If  $X \sim \text{Uni}(\alpha, \beta)$ , the PDF of  $X$  is:

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



What is  $\frac{1}{\beta - \alpha}$  if the following graphs are PDFs of Uniform RVs  $X$ ?



# Expectation and Variance

Discrete RV  $X$

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

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

Continuous RV  $X$

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

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

Both continuous and discrete RVs

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

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

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

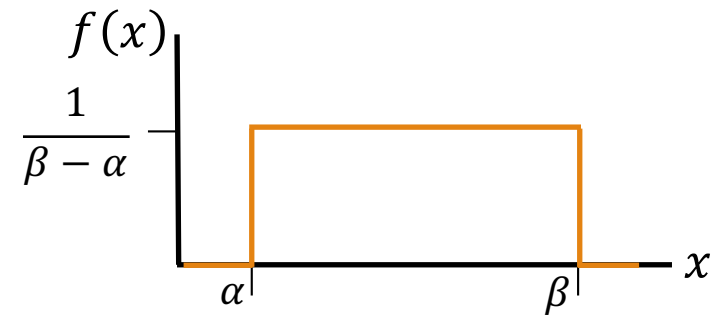
} Linearity of  
Expectation  
} Properties of  
variance

$$\text{TL;DR: } \sum_{x=a}^b \Rightarrow \int_a^b$$



# Uniform RV expectation

$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} x \cdot f(x) dx \\ &= \int_{\alpha}^{\beta} x \cdot \frac{1}{\beta - \alpha} dx \\ &= \frac{1}{\beta - \alpha} \cdot \frac{1}{2} x^2 \Big|_{\alpha}^{\beta} \\ &= \frac{1}{\beta - \alpha} \cdot \frac{1}{2} (\beta^2 - \alpha^2) \\ &= \frac{1}{2} \cdot \frac{(\beta + \alpha)(\beta - \alpha)}{\beta - \alpha} = \frac{\alpha + \beta}{2} \end{aligned}$$



Interpretation:  
Average the start & end

# Uniform Random Variable

def An Uniform random variable  $X$  is defined as follows:

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

PDF

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

Support:  $[\alpha, \beta]$   
(sometimes defined  
over  $(\alpha, \beta)$ )

Expectation

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

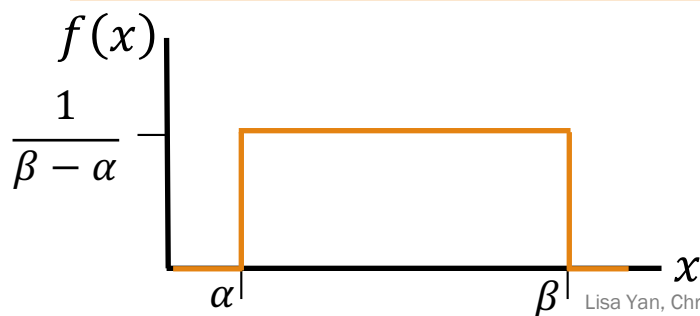
Just now

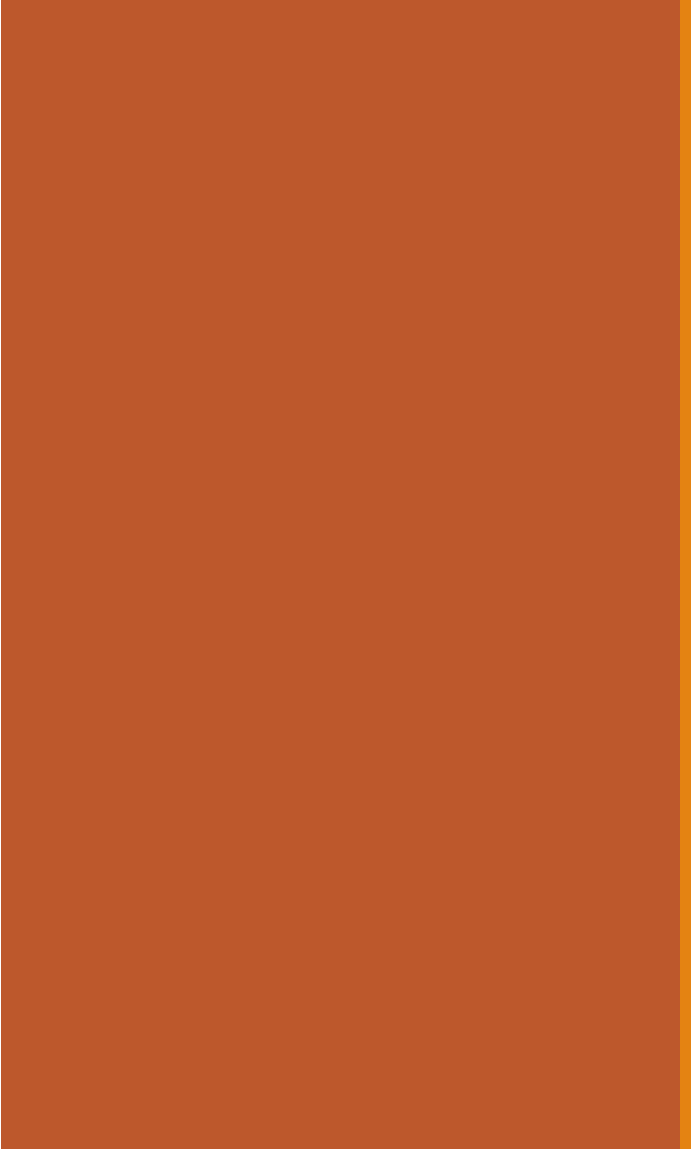
Variance

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



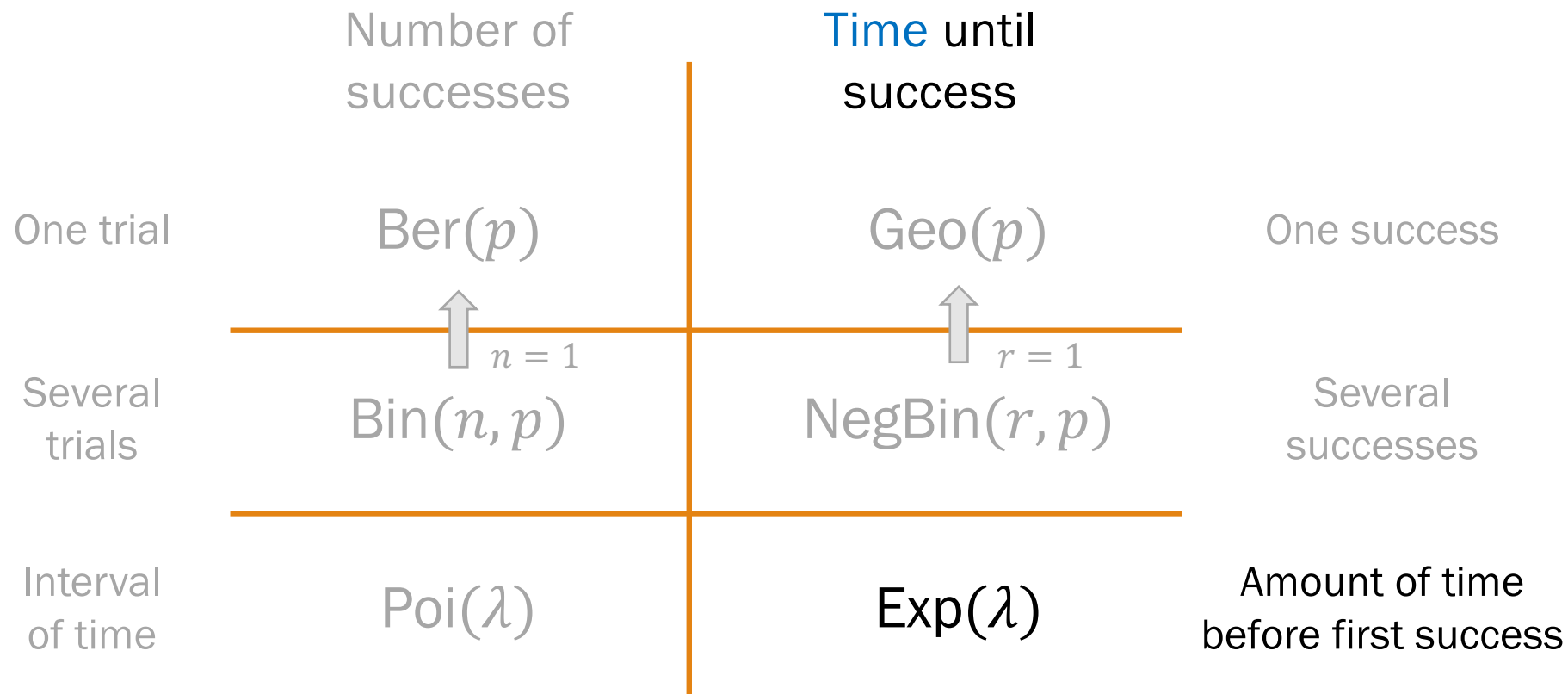
On your own!





# Exponential RV

# Grid of 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.

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

PDF

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

Expectation

$$E[X] = \frac{1}{\lambda} \quad (\text{in extra slides})$$

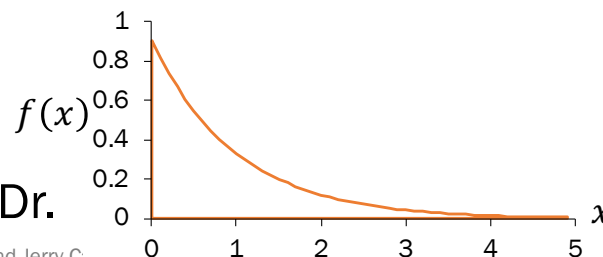
Support:  $[0, \infty)$

Variance

$$\text{Var}(X) = \frac{1}{\lambda^2} \quad (\text{on your own})$$

Examples:

- Time until next earthquake
- Time for request to reach web server
- Time until water main break on Campus Dr.



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# Interpreting $\text{Exp}(\lambda)$

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

$$X \sim \text{Exp}(\lambda) \quad \text{Expectation} \quad E[X] = \frac{1}{\lambda}$$

Based on the expectation  $E[X]$ , what are the units of  $\lambda$ ?



# Interpreting $\text{Exp}(\lambda)$

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

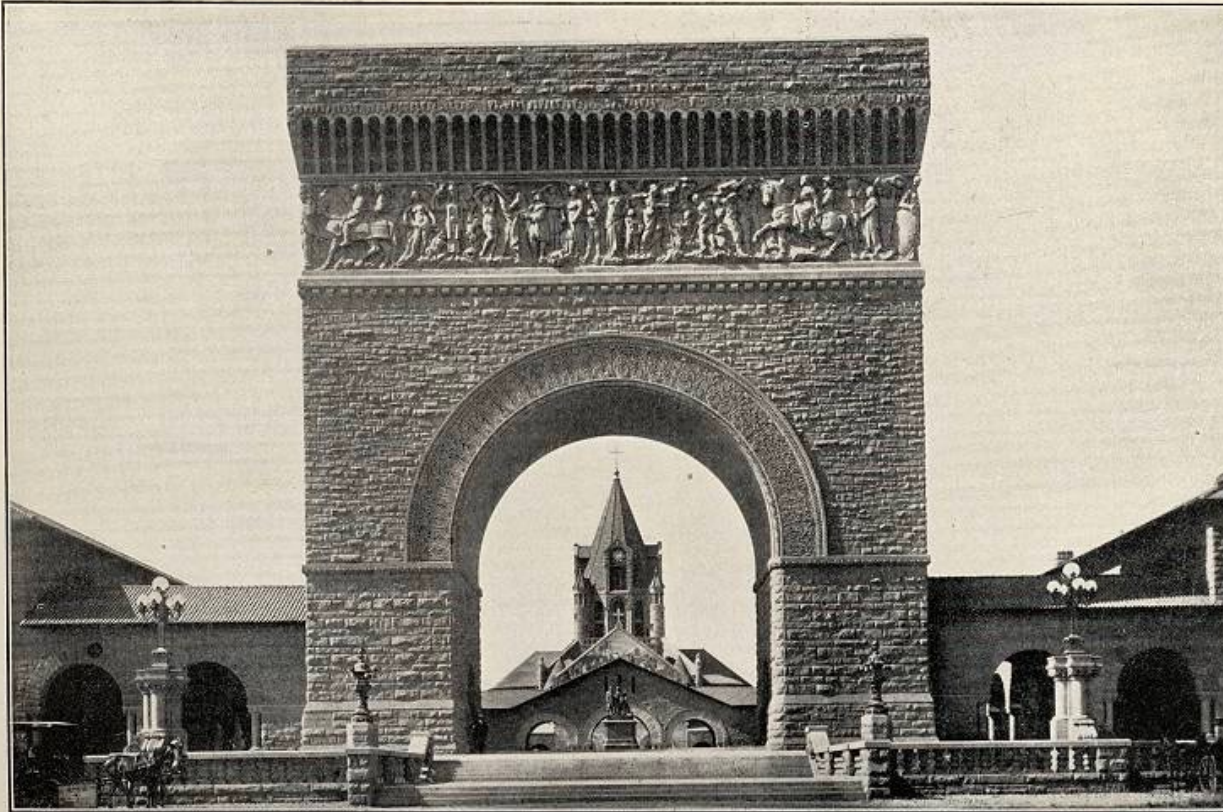
$$X \sim \text{Exp}(\lambda) \quad \text{Expectation} \quad E[X] = \frac{1}{\lambda}$$

Based on the expectation  $E[X]$ , what are the units of  $\lambda$ ?

e.g., average # of successes per second

For both Poisson and Exponential RVs,  
 $\lambda = \# \text{ successes/time.}$

# Earthquakes



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

1906 Earthquake  
Magnitude 7.8



# Earthquakes

$$X \sim \text{Exp}(\lambda) \quad \begin{aligned} E[X] &= 1/\lambda \\ f(x) &= \lambda e^{-\lambda x} \quad \text{if } x \geq 0 \end{aligned}$$

Major earthquakes (magnitude 8.0+) occur once every 500 years.\*

1. What is the probability of a major earthquake in the next 30 years?

We know on average:

$$\begin{aligned} 500 & \frac{\text{years}}{\text{earthquake}} \\ 0.002 & \frac{\text{earthquakes}}{\text{year}} \\ 1 & \frac{\text{earthquakes}}{500 \text{ years}} \end{aligned}$$

\*In California, according to historical data from USGS, 2015

# Earthquakes

$$X \sim \text{Exp}(\lambda) \quad \begin{aligned} E[X] &= 1/\lambda \\ f(x) &= \lambda e^{-\lambda x} \quad \text{if } x \geq 0 \end{aligned}$$

Major earthquakes (magnitude 8.0+) occur once every 500 years.\*

1. What is the probability of a major earthquake in the next 30 years?

Define events/  
RVs & state goal

Solve

$X$ : when next  
earthquake happens

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

$\lambda$ :  $\text{year}^{-1} = 1/500$

Want:  $P(X < 30)$

Recall

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

\*In California, according to historical data from USGS, 2015

# Earthquakes

$$X \sim \text{Exp}(\lambda) \quad \begin{aligned} E[X] &= 1/\lambda \\ f(x) &= \lambda e^{-\lambda x} \quad \text{if } x \geq 0 \end{aligned}$$

Major earthquakes (magnitude 8.0+) occur once every 500 years.\*

1. What is the probability of a major earthquake in the next 30 years?
2. What is the **standard deviation** of years until the next earthquake?

Define events/  
RVs & state goal

Solve

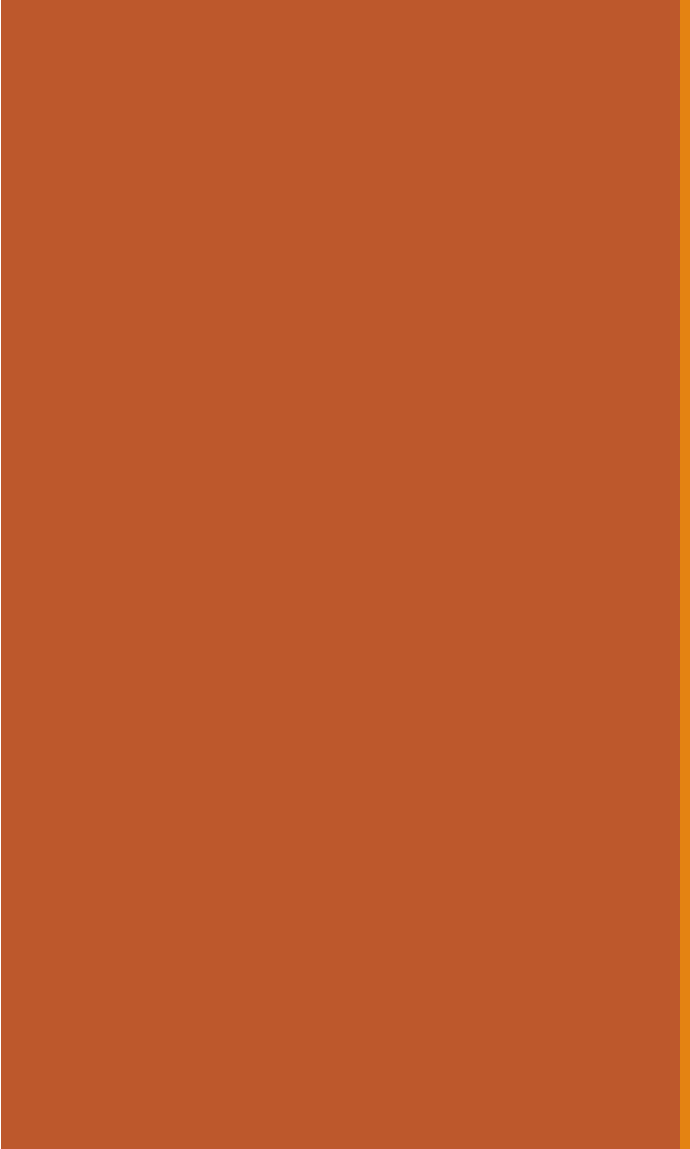
$X$ : when next  
earthquake happens

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

$\lambda$ : year<sup>-1</sup>

Want:  $P(X < 30)$

\*In California, according to historical data from USGS, 2015  
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# Cumulative Distribution Functions

# Cumulative Distribution Function (CDF)

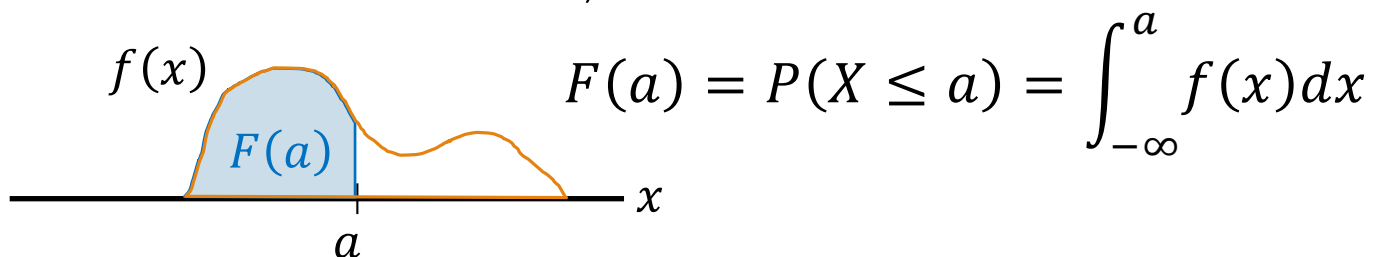
For a random variable  $X$ , the cumulative distribution function (CDF) is defined as

$$F(a) = F_X(a) = P(X \leq a), \text{ where } -\infty < a < \infty$$

For a discrete RV  $X$ , the CDF is:

$$F(a) = P(X \leq a) = \sum_{\text{all } x \leq a} p(x)$$

For a continuous RV  $X$ , the CDF is:



CDF is a probability,  
though PDF is not.

If you learn to use  
CDFs, you can avoid  
integrating the PDF.

# Using the CDF for continuous RVs

For a continuous random variable  $X$  with PDF  $f(x)$ , the CDF of  $X$  is

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

Matching (choices are used 0/1/2 times)

- |                         |                  |
|-------------------------|------------------|
| 1. $P(X < a)$           | A. $F(a)$        |
| 2. $P(X > a)$           | B. $1 - F(a)$    |
| 3. $P(X \geq a)$        | C. $F(b) - F(a)$ |
| 4. $P(a \leq X \leq b)$ | D. $F(a) - F(b)$ |



# Using the CDF for continuous RVs

For a continuous random variable  $X$  with PDF  $f(x)$ , the CDF of  $X$  is

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

Matching (choices are used 0/1/2 times)

- |                         |       |                  |              |
|-------------------------|-------|------------------|--------------|
| 1. $P(X < a)$           | ————— | A. $F(a)$        |              |
| 2. $P(X > a)$           | ————— | B. $1 - F(a)$    |              |
| 3. $P(X \geq a)$        | ————— | C. $F(b) - F(a)$ | (next slide) |
| 4. $P(a \leq X \leq b)$ | ————— | D. $F(a) - F(b)$ |              |

# Using the CDF for continuous RVs

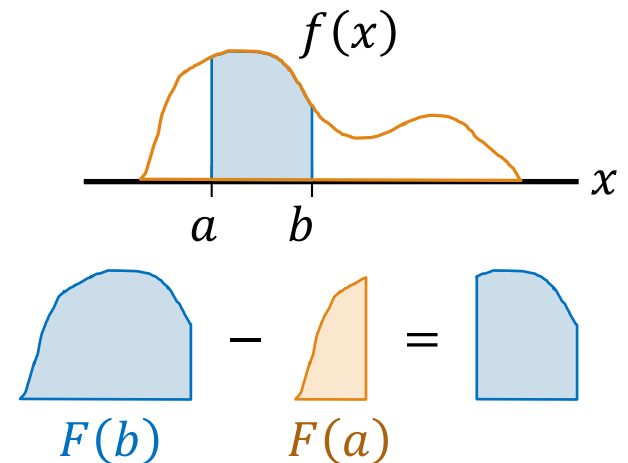
For a continuous random variable  $X$  with PDF  $f(x)$ , the CDF of  $X$  is

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

4.  $P(a \leq X \leq b) = F(b) - F(a)$

Proof:

$$\begin{aligned} F(b) - F(a) &= \int_{-\infty}^b f(x) dx - \int_{-\infty}^a f(x) dx \\ &= \left( \int_{-\infty}^a f(x) dx + \int_a^b f(x) dx \right) - \int_{-\infty}^a f(x) dx \\ &= \int_a^b f(x) dx \end{aligned}$$

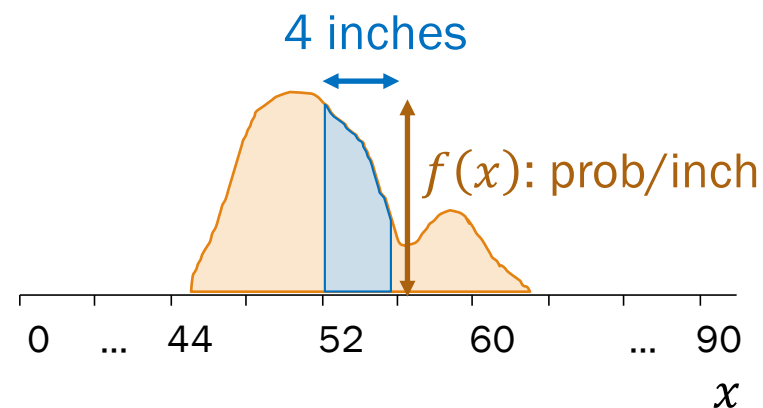




## Addendum to main takeaway #1

Integrate  $f(x)$  to get probabilities.\*

\*If you have  $F(a)$ , you already have probabilities, since  $F(a) = \int_{-\infty}^a f(x)dx$



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

# CDF of an Exponential RV

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

$$X \sim \text{Exp}(\lambda) \quad F(x) = 1 - e^{-\lambda x} \quad \text{if } x \geq 0$$

Proof:

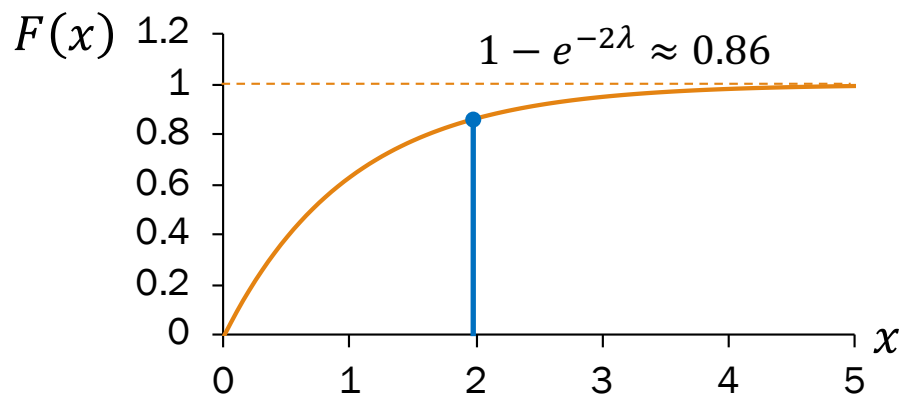
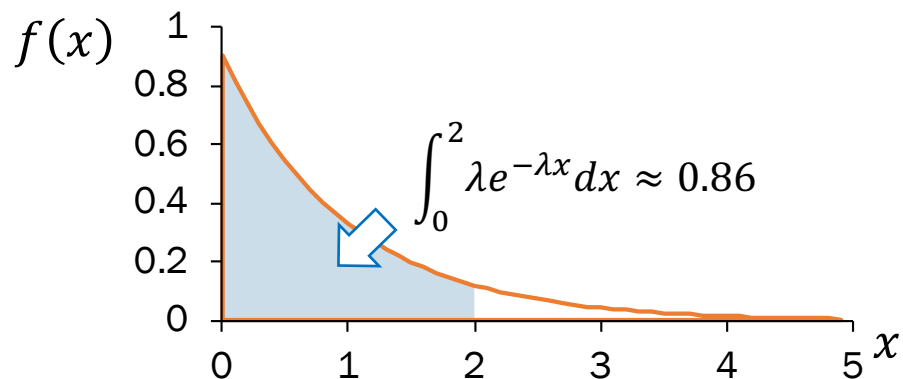
$$\begin{aligned} F(x) &= P(X \leq x) = \int_{y=-\infty}^x f(y) dy = \int_{y=0}^x \lambda e^{-\lambda y} dy \\ &= \lambda \frac{1}{-\lambda} e^{-\lambda y} \Big|_0^x \\ &= -1(e^{-\lambda x} - e^{-\lambda 0}) \\ &= 1 - e^{-\lambda x} \end{aligned}$$

Recall

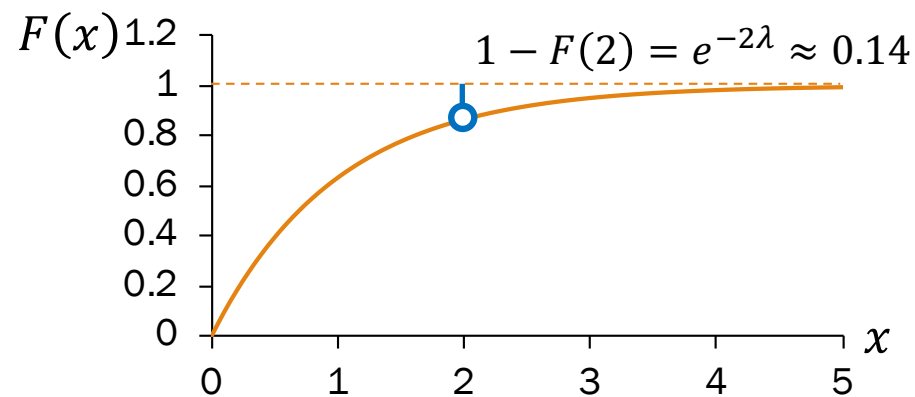
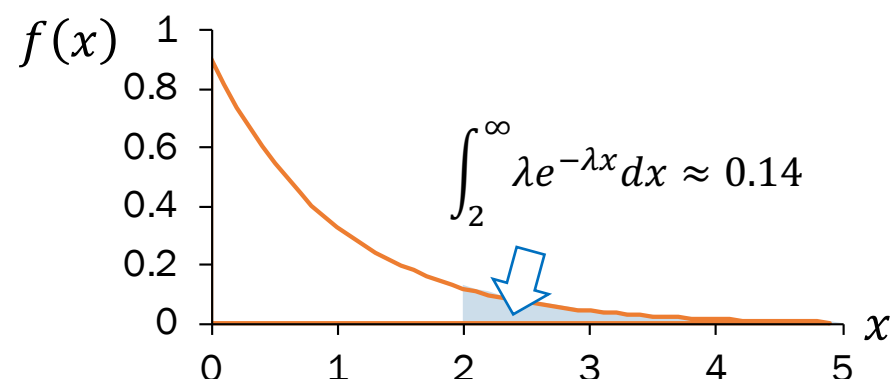
$$\int e^{cx} dx = \frac{1}{c} e^{cx}$$

# PDF/CDF $X \sim \text{Exp}(\lambda = 1)$

$$X \sim \text{Exp}(\lambda) \quad \begin{array}{l} x \geq 0: f(x) = \lambda e^{-\lambda x} \\ F(x) = 1 - e^{-\lambda x} \end{array}$$



$$P(X \leq 2)$$



$$P(X > 2)$$



# Memoryless Property

# Memorylessness: Hurry Up and Wait

A continuous probability distribution is said to be **memoryless** if a random variable  $X$  on that probability distribution satisfies the following for all  $s, t \geq 0$ :

$$P(X \geq s + t \mid X \geq s) = P(X \geq t)$$

- Here,  $s$  represents the time you've already spent waiting.
- The above states that after you've waited  $s$  time units, the probability you'll need to wait an **additional**  $t$  time units is equal to the probability you'd have to wait  $t$  time units without having waited those  $s$  time units in the first place.
- Example: If train arrival is guided by a memoryless random variable, the fact that you've waited 15 minutes doesn't obligate the train to arrive any faster!

# Memorylessness: Hurry Up and Wait

A continuous probability distribution is said to be **memoryless** if a random variable  $X$  on that probability distribution satisfies the following for all  $s, t \geq 0$ :

$$P(X \geq s + t \mid X \geq s) = P(X \geq t)$$

Using the definition of conditional probability, we can show that our Exponential distribution exhibits the memoryless property. Just let  $X \sim \text{Exp}(\lambda)$  and trust the math:

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



# Exercises

# Earthquakes

---

Major earthquakes (magnitude 8.0+) occur independently on average once every 500 years.\*

What is the probability of **zero major earthquakes next year**?



\*In California, according to historical data from USGS, 2015  
Lisa Yan, Chris Piech, Mehran Sahami, and Jerry Cain, CS109, Winter 2024



# Earthquakes

Major earthquakes (magnitude 8.0+) occur independently on average once every 500 years.\*

What is the probability of **zero major earthquakes next year**?

Strategy 1: Exponential RV

Define events/RVs & state goal

$T$ : when first earthquake happens

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

Want:  $P(T > 1) = 1 - F(1)$

Solve

$$P(T > 1) = 1 - (1 - e^{-\lambda \cdot 1}) = e^{-\lambda}$$

\*In California, according to historical data from USGS, 2015

# Earthquakes

$$Y \sim \text{Poi}(\lambda) \quad p(k) = e^{-\lambda} \frac{\lambda^k}{k!}$$

Major earthquakes (magnitude 8.0+) occur independently on average once every 500 years.\*

What is the probability of **zero major earthquakes next year**?

Strategy 1: Exponential RV

Define events/RVs & state goal

$T$ : when first earthquake happens

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

Want:  $P(T > 1) = 1 - F(1)$

Solve

$$P(T > 1) = 1 - (1 - e^{-\lambda \cdot 1}) = e^{-\lambda}$$

Strategy 2: Poisson RV

Define events/RVs & state goal

$N$ : # earthquakes next year

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

$\lambda$ :  $\frac{\text{earthquakes}}{\text{year}}$

Want:  $P(N = 0)$

Solve

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

\*In California, according to historical data from USGS, 2015

# Replacing your laptop

$$X \sim \text{Exp}(\lambda) \quad \begin{aligned} E[X] &= 1/\lambda \\ F(x) &= 1 - e^{-\lambda x} \end{aligned}$$

Let  $X = \#$  hours of use until your laptop dies.

- $X$  is distributed as an Exponential RV, where
- On average, laptops die after 5000 hours of use.
- You use your laptop 5 hours a day.

What is  $P(\text{your laptop lasts 4 years})$ ?



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Define

$X$ : # hours until  
laptop death  
 $X \sim \text{Exp}(\lambda = 1/5000)$

Want:  $P(X > 5 \cdot 365 \cdot 4)$

Solve

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

Better plan ahead if you're co-termining!

- 5-year plan:

$$P(X > 9125) = e^{-1.825} \approx 0.1612$$

- 6-year plan:

$$P(X > 10950) = e^{-2.19} \approx 0.1119$$



Extra

# Expectation of the Exponential

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

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

Expectation

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

Proof:

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx = \int_0^{\infty} x \lambda e^{-\lambda x} dx$$

$$= -x e^{-\lambda x} \Big|_0^{\infty} + \int_0^{\infty} e^{-\lambda x} dx$$

$$= -x e^{-\lambda x} \Big|_0^{\infty} - \frac{1}{\lambda} e^{-\lambda x} \Big|_0^{\infty}$$

$$= [0 - 0] + \left[ 0 - \left( -\frac{1}{\lambda} \right) \right]$$

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

Integration by parts

$$\int x \lambda e^{-\lambda x} dx = \int u \cdot dv$$

$$\begin{array}{ll} u = x & dv = \lambda e^{-\lambda x} dx \\ du = dx & v = -e^{-\lambda x} \end{array}$$

$$\int u \cdot dv = u \cdot v - \int v \cdot du$$

$$-x e^{-\lambda x} - \int -e^{-\lambda x} dx$$