# INTRODUCTION TO STATISTICS

**LECTURE 4** 

# LAST TIME

Continuous random variables

• 
$$P(X \le x)$$
,  $P(X > x)$ ,  $P(X = x)$ 

• CDFs

• Uniform distribution

## **TODAY**

Probability density function (PDF)

PDF and CDF

- Properties of continuous random variables
  - Mean
  - Variance
- + 1 more continuous distribution

**CDFs** 

**CDF** 
$$F(x) = P(X \le x)$$

1. F(5)

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$$F(x) = P(X \le x)$$

1. 
$$P(X \le 5)$$

**CDF** 
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1. F(5)

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

**CDF** 
$$F(x) = P(X \le x)$$

#### 1. F(5)

2. 1 - F(3)

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

**CDF** 
$$F(x) = P(X \le x)$$

1. F(5)

- 2. 1 F(3)
- 3. F(2) F(-1)

1. 
$$P(X \le 5)$$

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$$P(X > 3)$$

**CDF** 
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#### 1. F(5)

2. 
$$1 - F(3)$$

3. 
$$F(2) - F(-1)$$

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

3. 
$$P(-1 < X \le 2)$$

**CDF** 
$$F(x) = P(X \le x)$$

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- 2. 1 F(3)
- 3. F(2) F(-1)

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

3. 
$$P(-1 < X \le 2)$$

4. 
$$P(-5 < X \le -3)$$

**CDF** 
$$F(x) = P(X \le x)$$

#### 1. F(5)

2. 
$$1 - F(3)$$

3. 
$$F(2) - F(-1)$$

4. 
$$F(-3) - F(-5)$$

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

3. 
$$P(-1 < X \le 2)$$

4. 
$$P(-5 < X \le -3)$$

**CDF** 
$$F(x) = P(X \le x)$$

1. F(5)

- 2. 1 F(3)
- 3. F(2) F(-1)
- 4. F(-3) F(-5)
- 5. F(5) F(3) + F(-1) F(-5)

#### **Probability mass function**

1.  $P(X \le 5)$ 

2. P(X > 3)

3.  $P(-1 < X \le 2)$ 

4.  $P(-5 < X \le -3)$ 

#### **CDF** $F(x) = P(X \le x)$

#### 1. F(5)

2. 
$$1 - F(3)$$

3. 
$$F(2) - F(-1)$$

4. 
$$F(-3) - F(-5)$$

5. 
$$F(5) - F(3) + F(-1) - F(-5)$$

1. 
$$P(X \le 5)$$

2. 
$$P(X > 3)$$

3. 
$$P(-1 < X \le 2)$$

4. 
$$P(-5 < X \le -3)$$

5. 
$$P(3 < X \le 5 \text{ or } -5 < X \le -1)$$

# PROBABILITY DENSITY FUNCTION

**PDFs** 

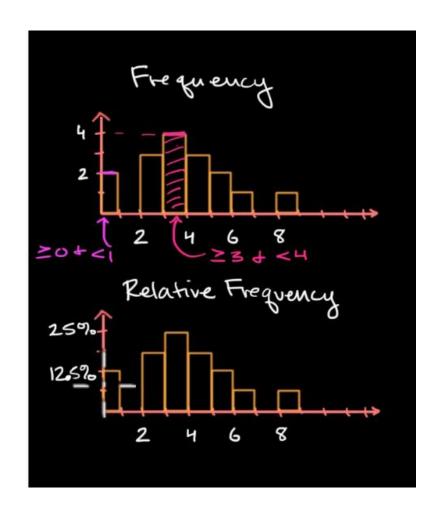
# WATCH THE VIDEO

https://youtu.be/PUvUQMQ7xQk

# NOW, LET'S DISCUSS IT STEP BY STEP

### **HISTOGRAMS**

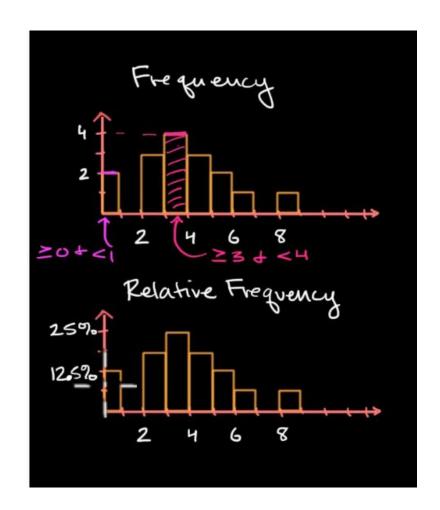
What's the difference between 'frequency' and 'relative frequency' histograms?



## **HISTOGRAMS**

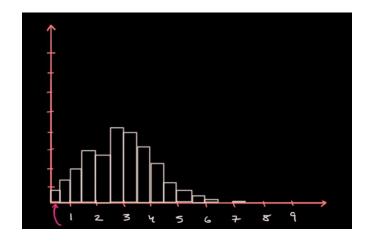
What's the difference between 'frequency' and 'relative frequency' histograms (as stated in the video)?

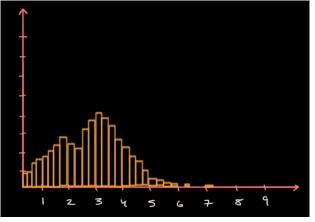
When one is better than another?

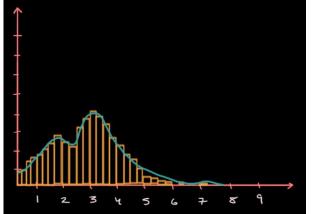


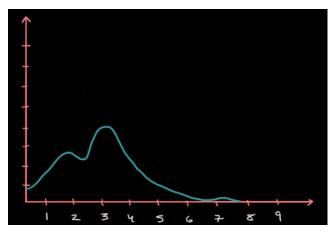
# **DENSITY: ESTIMATION**

How did they obtain the density?



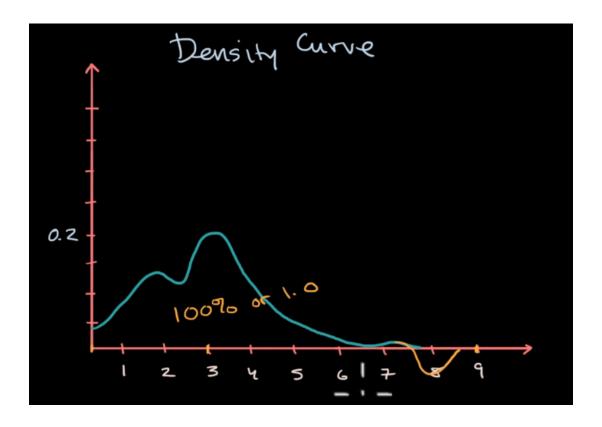






### **DENSITY: BASIC PROPERTIES**

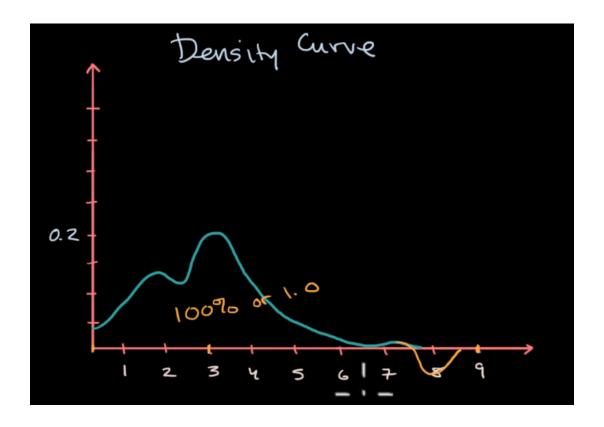
• Which important properties of a density function were mentioned in the video?



# **DENSITY: BASIC PROPERTIES**

• Which important properties of a density function were mentioned in the video?

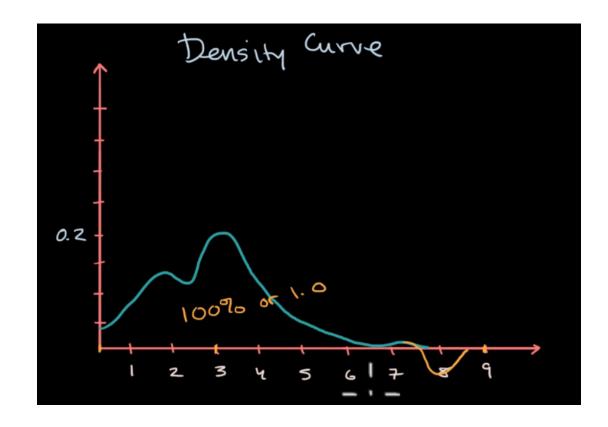
1. area under the curve is 1;



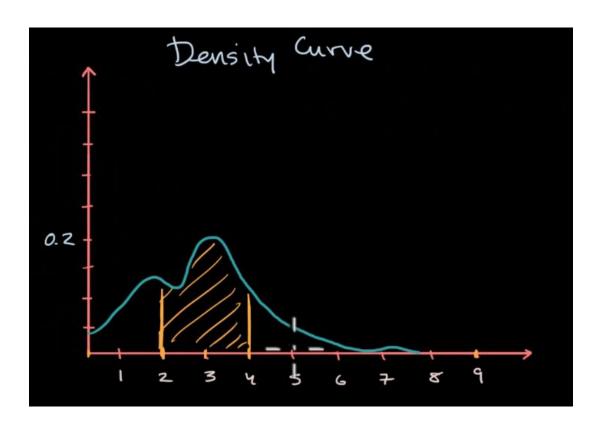
# **DENSITY: BASIC PROPERTIES**

• Which important properties of a density function were mentioned in the video?

- 1. area under the curve is 1;
- 2.  $p(x) \ge 0$ .

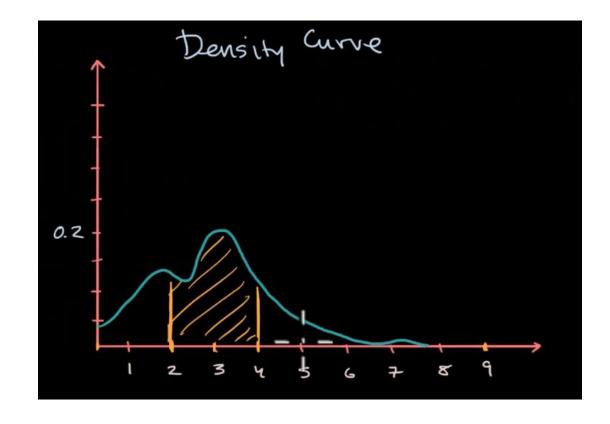


• Which probability does the area of the region corresponds to?



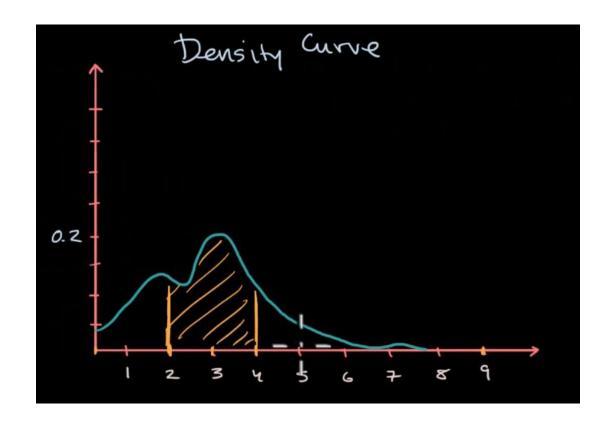
• Which probability does the area of the region corresponds to?

$$P(2 < X \le 4)$$



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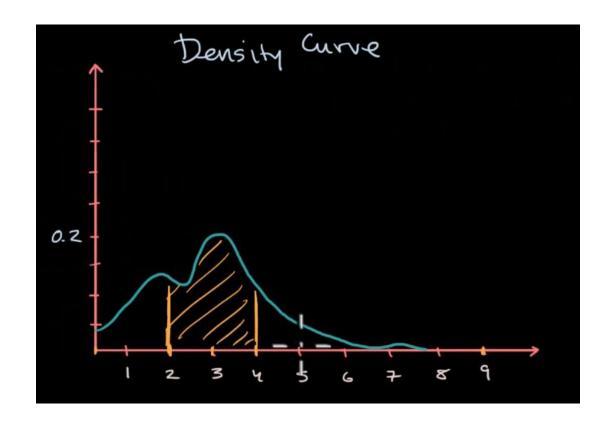
$$P(2 < X \le 4)$$



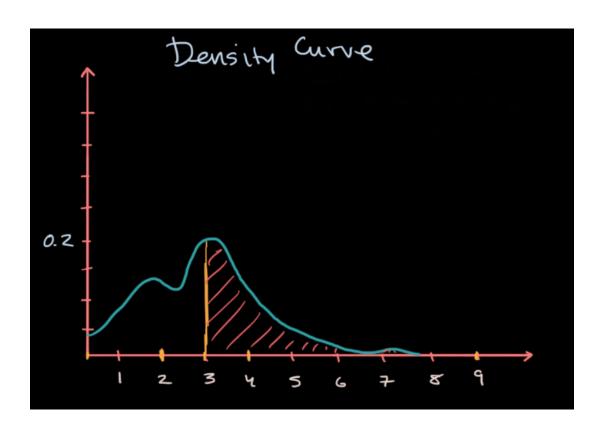
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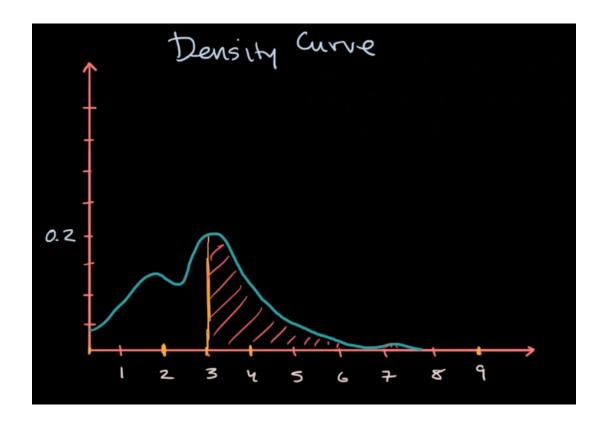
$$F(4) - F(2)$$



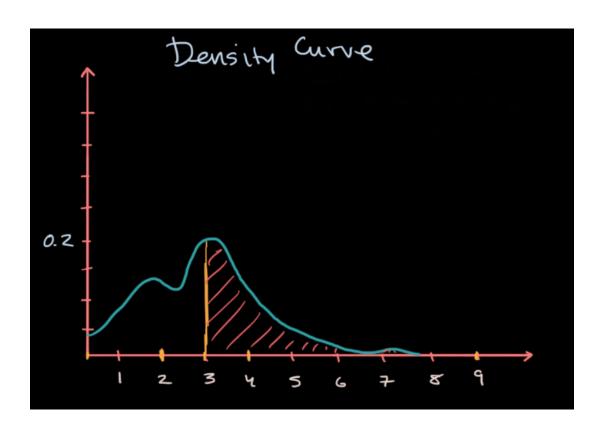
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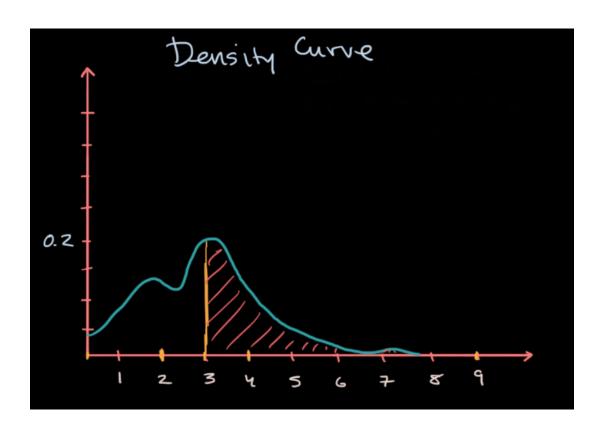


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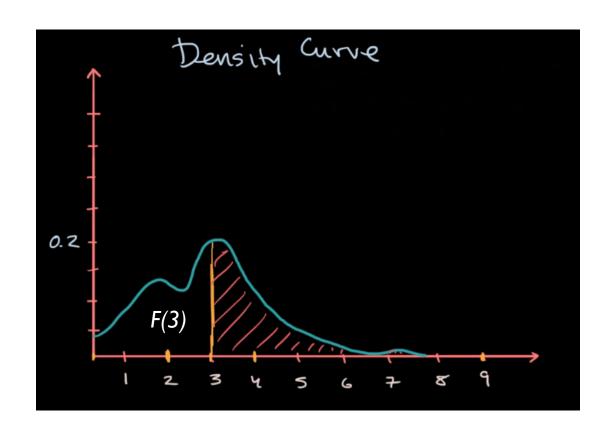
• Which probability does the area of the region corresponds to?

$$1 - F(3)$$



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$$1 - F(3)$$



# THERE IS A MISTAKE IN A VIDEO THOUGH...

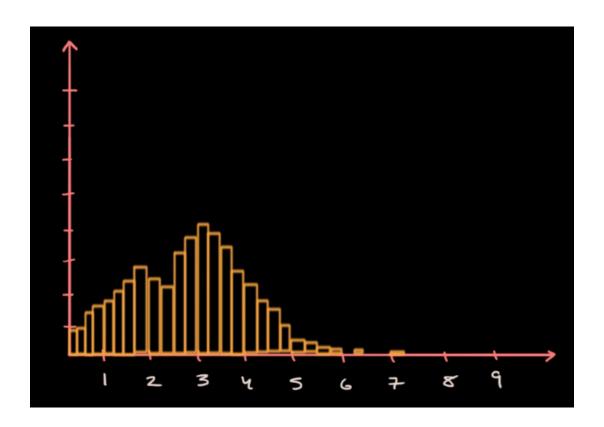
Google Classroom -> [Python] From histograms to density

## **NORMALIZING BIN HEIGHTS**

• Total area must sum up to 1:

$$\sum_{i} area(bar_i) = 1$$

• What is the area of each bar for such a histogram?



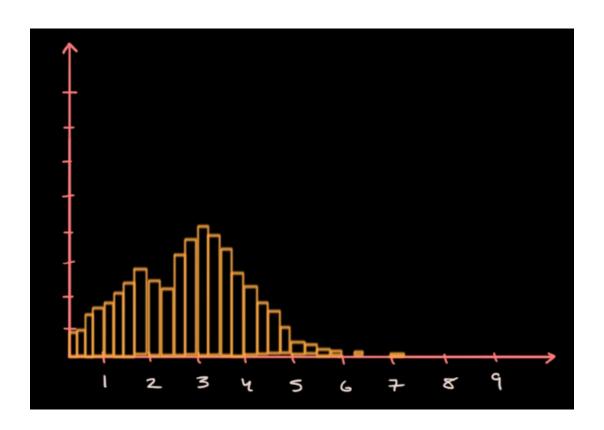
## **NORMALIZING BIN HEIGHTS**

• Total area must sum up to 1:

$$\sum_{i} area(bar_i) = 1$$

• What is the area of each bar for such a histogram?

bin height x bin width



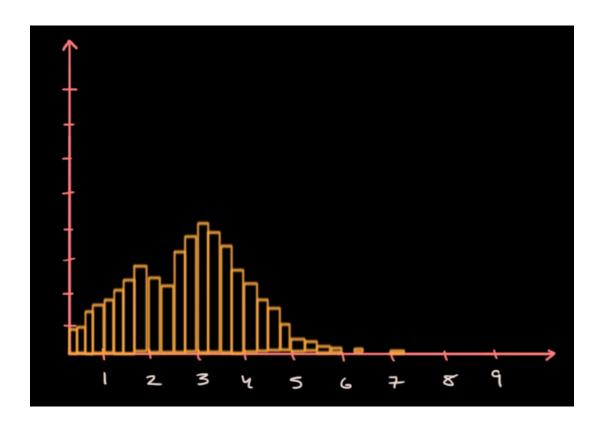
## **NORMALIZING BIN HEIGHTS**

• Total area must sum up to 1:

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• What is the area of each bar for such a histogram?

bin height x bin width relative frequency x bin width



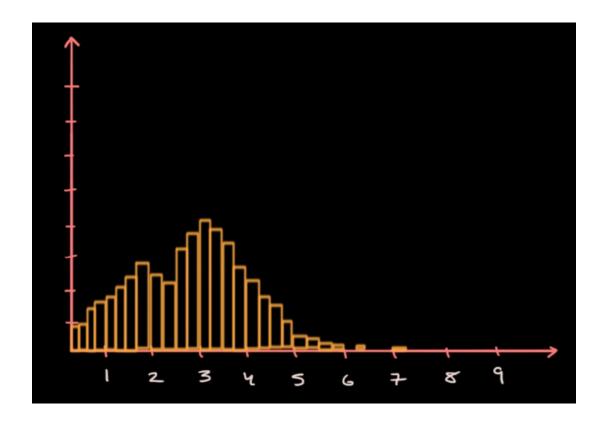
#### **NORMALIZING BIN HEIGHTS**

Total area must sum up to 1:

$$\sum_{i} area(bar_i) = 1$$

• What is the area of each bar for such a histogram?

bin height x bin width relative frequency x bin width  $\rightarrow$  0 when bin width  $\rightarrow$  0



#### **NORMALIZING BIN HEIGHTS**

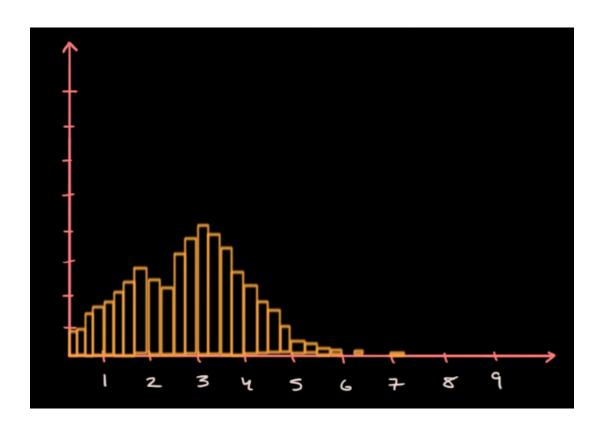
Total area must sum up to 1:

$$\sum_{i} area(bar_i) = 1$$

• What is the area of each bar for such a histogram?

bin height x bin width

(relative frequency/bin width) x bin width



#### **NORMALIZING BIN HEIGHTS**

Total area must sum up to 1:

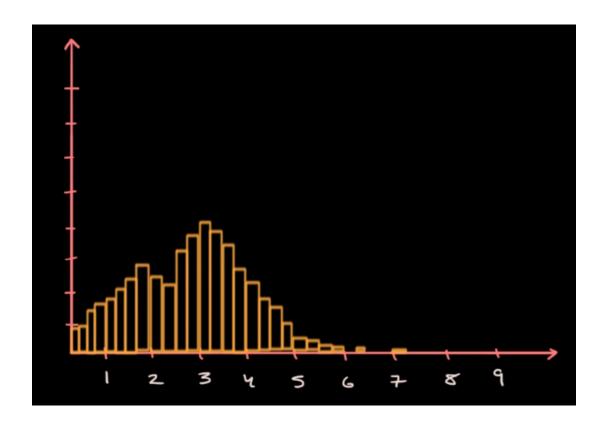
$$\sum_{i} area(bar_i) = 1$$

• What is the area of each bar for such a histogram?

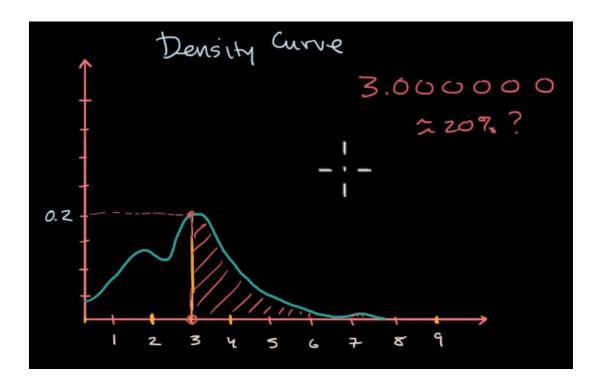
bin height x bin width

(relative frequency/bin width) x bin width

always sums up to 1

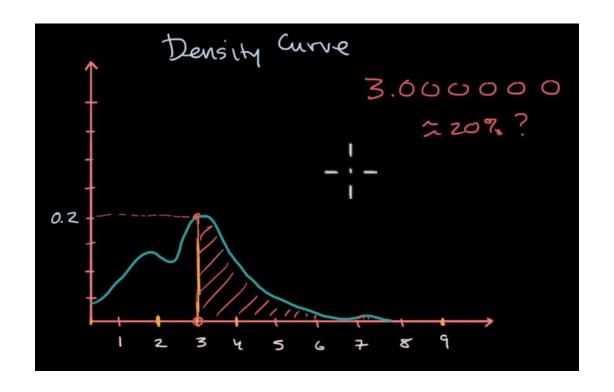


What is P(X = 3)?



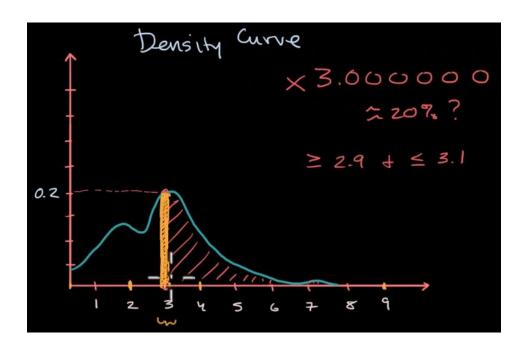
What is P(X = 3)?

$$P(X = 3) = 0$$



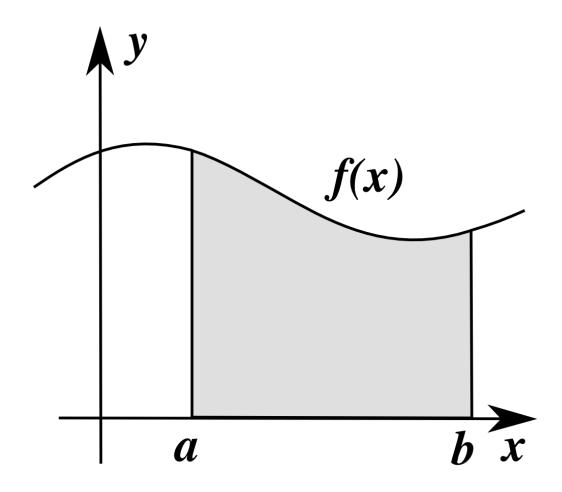
## RELATION BETWEEN CDF AND DENSITY

How did they estimate the area under the curve?



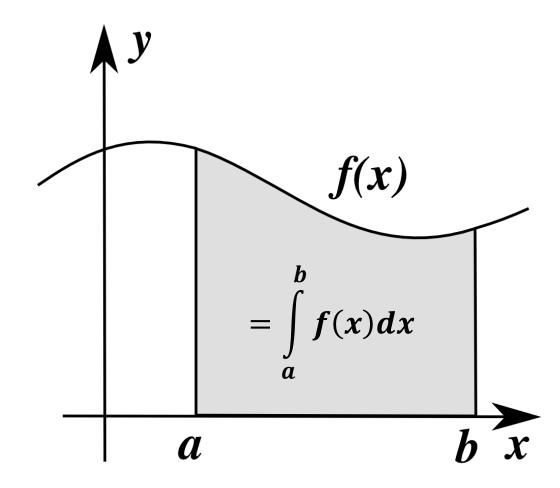
How did they estimate the area under the curve?

Do you remember how to compute it exactly?



How did they estimate the area under the curve?

Do you remember how to compute it exactly?



$$\int f(x)dx = F(x) + C \iff \frac{d}{dx}F(x) = f(x)$$

$$\int f(x)dx = F(x) + C \iff$$

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

$$\int x dx = \frac{x^2}{2} + C$$

$$\int f(x)dx = F(x) + C \iff$$

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

$$\int x dx = \frac{x^2}{2} + C$$

$$\int \frac{1}{x} dx = \log x + C$$

$$\int f(x)dx = F(x) + C \iff$$

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

$$\int_{a}^{b} f(x)dx = F(b) - F(a)$$

$$\int x dx = \frac{x^2}{2} + C$$

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$$\int x dx = \frac{x^2}{2} + C$$

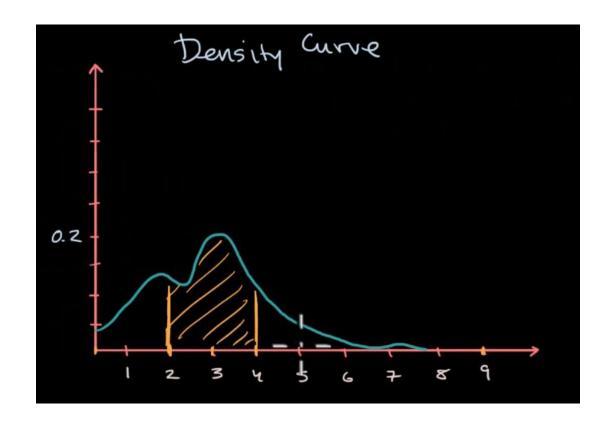
$$\int \frac{1}{x} dx = \log x + C$$

$$\int_0^1 x dx = \frac{1^2}{2} - \frac{0^2}{2} = \frac{1}{2}$$

• Which probability does the area of the region corresponds to?

$$P(2 < X \le 4)$$

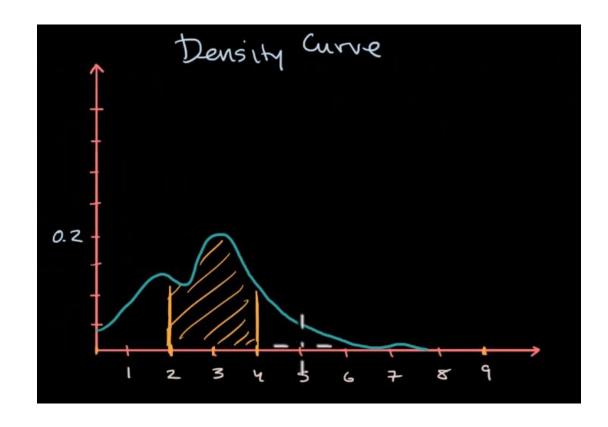
$$F(4) - F(2)$$



• Which probability does the area of the region corresponds to?

$$\int_{2}^{4} p(x) dx = P(2 < X \le 4)$$

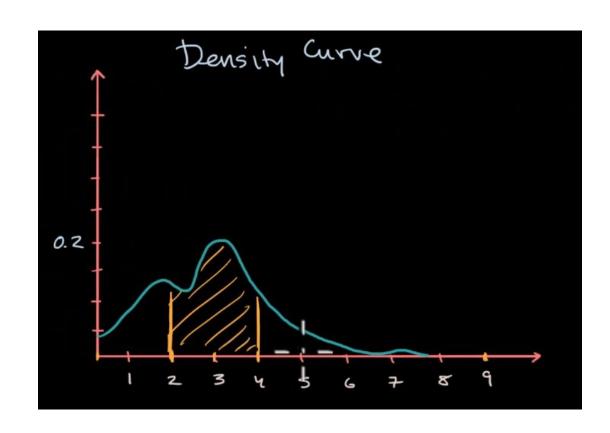
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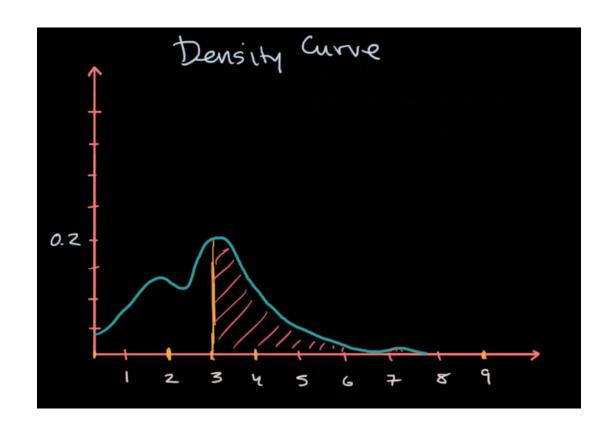
$$\int_{2}^{4} p(x) dx = P(2 < X \le 4)$$

$$\int_{2}^{4} p(x) dx = F(4) - F(2)$$



• Which probability does the area of the region corresponds to?

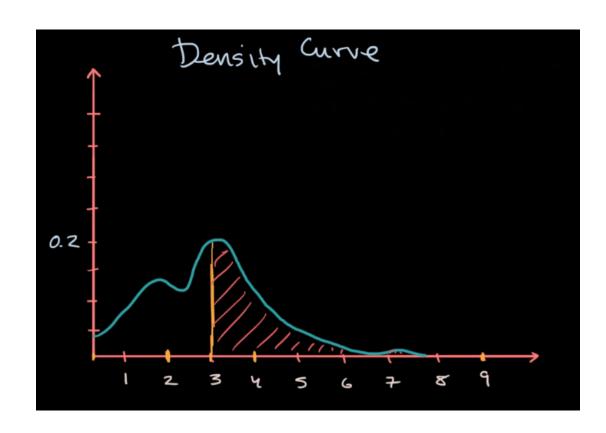
$$1 - F(3)$$



• Which probability does the area of the region corresponds to?

$$\int_3^{+\infty} p(x) dx = P(X > 3)$$

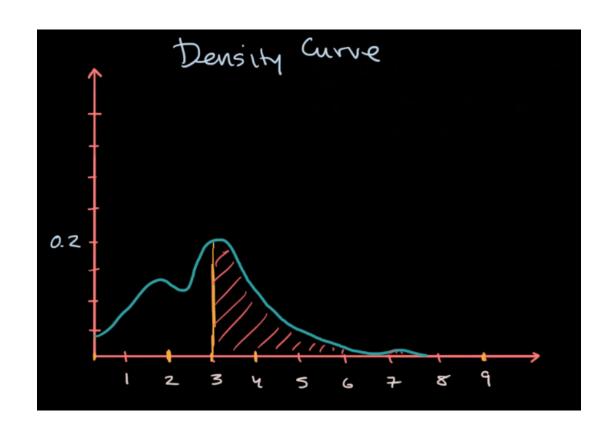
$$1 - F(3)$$



• Which probability does the area of the region corresponds to?

$$\int_3^{+\infty} p(x) dx = P(X > 3)$$

$$\int_{3}^{+\infty} p(x) dx = 1 - F(3) =$$

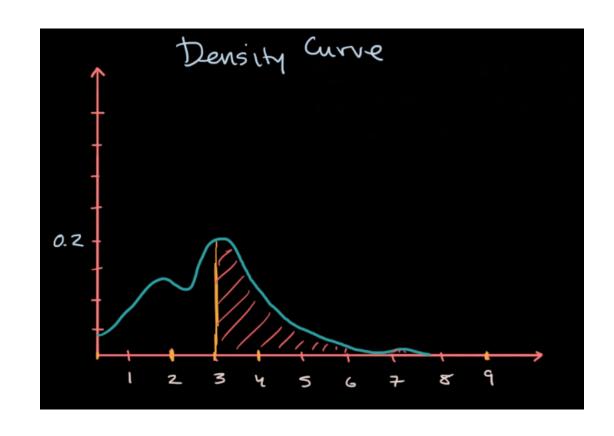


• Which probability does the area of the region corresponds to?

$$\int_3^{+\infty} p(x) dx = P(X > 3)$$

$$\int_{3}^{+\infty} p(x)dx = 1 - F(3) =$$

$$= 1 - \int_{-\infty}^{3} p(x)dx$$



#### **CDF AND PDF**

• 
$$\int_{2}^{4} p(x) dx = F(4) - F(2)$$

• 
$$\int_{-\infty}^{3} p(x) dx = F(3)$$

#### **CDF AND PDF**

• 
$$\int_{2}^{4} p(x) dx = F(4) - F(2)$$

$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
$$\frac{d}{dx}F(x) = p(x)$$

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

**CDF:** 
$$F(x) = P(X \le x)$$

# $F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases}$

#### • PDF:

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

**CDF:** 
$$F(x) = P(X \le x)$$

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases} \qquad p(x) = \begin{cases} 0, & x < a \\ x - a \\ b - a \end{cases}$$

$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
• PDF:

$$p(x) = \begin{cases} \\ \end{cases}$$

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

**CDF:** 
$$F(x) = P(X \le x)$$

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases} \qquad p(x) = \begin{cases} x \le a \text{ or } x \ge b \\ a < b < b \end{cases}$$

$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
• PDF:

$$p(x) = \begin{cases} x \le a \text{ or } x \ge b \\ a < b < b \end{cases}$$

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$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
• PDF:

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ & a < b < b \end{cases}$$

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$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases} \qquad p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b - a}, & a < b < b \end{cases}$$

$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
• PDF:

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

### SHORT QUIZ

CDFs and PDFs

1. 
$$P(X \le 3)$$

2. 
$$\int_3^{+\infty} p(x) dx$$

3. 
$$F(3)$$

4. 
$$\int_{-\infty}^{3} p(x) dx$$

1. 
$$P(X \le 3)$$

$$2. \int_3^{+\infty} p(x) dx$$

3. 
$$F(3)$$

4. 
$$\int_{-\infty}^{3} p(x) dx$$

1. 
$$P(X > 5)$$

2. 
$$\int_{5}^{+\infty} p(x) dx$$

3. 
$$\int_{-\infty}^{5} p(x) dx$$

4. 
$$1 - F(5)$$

1. 
$$P(X > 5)$$

$$2. \int_5^{+\infty} p(x) dx$$

3. 
$$\int_{-\infty}^{5} p(x) dx$$

4. 
$$1 - F(5)$$

1. 
$$\int_3^5 p(x) dx$$

2. 
$$P(3 < X \le 5)$$

3. 
$$F(3) - F(5)$$

4. 
$$\int_{-\infty}^{5} p(x)dx - \int_{-\infty}^{3} p(x)dx$$

1. 
$$\int_3^5 p(x) dx$$

2. 
$$P(3 < X \le 5)$$

3. 
$$F(3) - F(5)$$

4. 
$$\int_{-\infty}^{5} p(x)dx - \int_{-\infty}^{3} p(x)dx$$

## PROPERTIES OF CONTINUOUS RANDOM VARIABLES

•  $X \sim U(a, b)$ 

• 
$$X \sim U(a, b)$$

• CDF:  

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases}$$

• 
$$X \sim U(a, b)$$

• CDF:  

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases}$$

• PDF:  

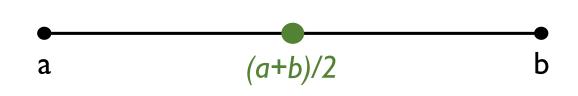
$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

• 
$$X \sim U(a, b)$$

• 
$$E(X) = \frac{(a+b)}{2}$$

• CDF:  

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ \frac{1}{a}, & x > b \end{cases}$$



### • PDF:

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

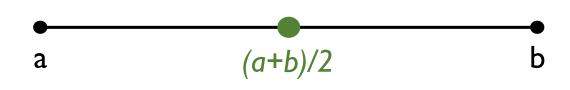
• 
$$X \sim U(a, b)$$

$$E(X) = \frac{(a+b)}{2}$$

$$Var(X) = ?$$

• CDF:  

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \le x \le b \\ 1, & x > b \end{cases}$$
•  $Var(X) = ?$ 



• PDF:

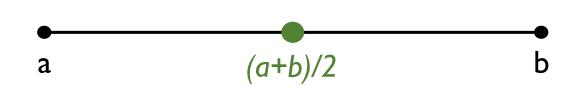
$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

• 
$$X \sim U(a, b)$$

• 
$$E(X) = \frac{(a+b)}{2}$$
  
•  $Var(X) = ?$ 

• CDF:  

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases}$$



• PDF:

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

But how to compute expected value and variance of a continuous variable?

#### DISCRETE RANDOM VARIABLE

# CONTINUOUS RANDOM VARIABLE

• Sum up all the values a random variable can take, multiplying them by their probabilities:

#### **DISCRETE RANDOM VARIABLE**

#### **CONTINUOUS RANDOM VARIABLE**

• Sum up all the values a random variable can take, multiplying them by their probabilities:

$$E(X) = \sum_{X_i} X_i \cdot P(X = X_i)$$

#### DISCRETE RANDOM VARIABLE

• Sum up all the values a random • Same principle: variable can take, multiplying them by their probabilities:

$$E(X) = \sum_{X_i} X_i \cdot P(X = X_i)$$

#### **CONTINUOUS RANDOM VARIABLE**

#### DISCRETE RANDOM VARIABLE

#### • Sum up all the values a random • Same principle: variable can take, multiplying them by their probabilities:

$$E(X) = \sum_{X_i} X_i \cdot P(X = X_i)$$

#### **CONTINUOUS RANDOM VARIABLE**

$$E(X) = \int_{-\infty}^{+\infty} x \cdot p(x) dx$$

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

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

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$$=\frac{1}{b-a}\cdot\frac{b^2-a^2}{2}=\frac{a+b}{2}$$

• 
$$E(X) = \frac{10+0}{2} = 5$$

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• 
$$E(X - 1) = E(X) - 1 = 4$$

• 
$$E(2 \cdot X) = 2 \cdot E(X) = 10$$

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• Imagine that  $X \sim U(0, 10)$ 

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•  $E(X^2) =$ 

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$$E(X^2) = \int_0^{10} x^2 \cdot \frac{1}{10} dx = \frac{1}{10} \cdot \frac{10^3}{3} =$$

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$$E(X^2) = \int_0^{10} x^2 \cdot \frac{1}{10} dx = \frac{1}{10} \cdot \frac{10^3}{3} = \frac{100}{3}$$

#### DISCRETE RANDOM VARIABLE

**CONTINUOUS RANDOM VARIABLE** 

• Expected squared distance between a value and the mean:

$$Var(X) = E\left(\left(X - E(X)\right)^{2}\right)$$

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#### **CONTINUOUS RANDOM VARIABLE**

Same principle.

• 
$$X \sim U(a, b)$$
,  $Var(X) = ?$ 

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- $X \sim U(a, b)$
- $E(X) = \frac{a+b}{2}$
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- $Var(X) = E(X^2) \left(\frac{a+b}{2}\right)^2 =$

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• 
$$Var(X) = E(X^2) - \left(\frac{a+b}{2}\right)^2 = \frac{b^3 - a^3}{3} - \frac{(a+b)^2}{4} = \frac{b^3 - a^3}{4}$$

#### VARIANCE OF A UNIFORM DISTRIBUTION

- $X \sim U(a, b)$
- $E(X) = \frac{a+b}{2}$
- $Var(X) = E\left(\left(X E(X)\right)^2\right) = E(X^2) \left(E(X)\right)^2$

• 
$$Var(X) = E(X^2) - \left(\frac{a+b}{2}\right)^2 = \frac{b^3 - a^3}{3} - \frac{(a+b)^2}{4} = \frac{(a-b)^2}{12}$$

• 
$$X \sim U(a, b)$$

• CDF: 
$$x < a$$

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x - a}{b - a}, & a \le x \le b \\ 1, & x > b \end{cases}$$
• MEAN:
$$E(X) = \frac{(a + b)}{2}$$

$$E(X) = \frac{(a+b)}{2}$$

#### • PDF:

$$p(x) = \begin{cases} 0, & x \le a \text{ or } x \ge b \\ \frac{1}{b-a}, & a < b < b \end{cases}$$

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

$$F(x) =$$

$$E(X) =$$

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$$\begin{aligned}
 x &< 0 \\
 0 &\le x \le 1 \\
 x &> 1
 \end{aligned}$$

$$E(X) = \frac{1-0}{2} = 0.5$$

$$p(x) =$$

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Poisson distribution (discrete):
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Poisson distribution (discrete):
#of events that occur in a given period of time.
# of phone calls received every hour

• Exponential distribution (continuous): time between two Poisson events.

Time between the two phone calls

•  $X \sim Exp(\lambda)$ ,  $\lambda > 0$  – rate parameter.

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· CDF:

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$$X \sim Exp(\lambda)$$
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$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
• CDF:

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

• PDF:

• 
$$X \sim Exp(\lambda)$$
,

• 
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· CDF:

$$F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$$
$$\frac{d}{dx}F(x) = p(x)$$

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

$$x \leq 0$$
 $x < 0$ 

• PDF:

$$p(x) = \begin{cases} x \ge 0 \\ x < 0 \end{cases}$$

•  $X \sim Exp(\lambda)$ ,  $\lambda > 0$  - rate parameter.  $F(x) = P(X \le x) = \int_{-\infty}^{x} p(t)dt$ 

· CDF:

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

• PDF:

$$p(x) = \begin{cases} \lambda \cdot e^{-\lambda x}, & x \ge 0 \\ 0, & x < 0 \end{cases}$$

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- $X \sim Exp(\lambda)$ ,  $\lambda > 0$  rate parameter.
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· VARIANCE:

$$E(X^2) = \int_{-\infty}^{+\infty} x^2 \cdot p(x) dx =$$

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$$E(X^2) = \int_{-\infty}^{+\infty} x^2 \cdot p(x) dx = \int_{0}^{+\infty} x^2 \cdot \lambda e^{-\lambda x} dx = \frac{2}{\lambda^2}$$

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- MEAN:

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$$Var(X) = E(X^{2}) - (E(X))^{2} =$$

- $X \sim Exp(\lambda)$ ,  $\lambda > 0$  rate parameter.
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