Week 3

### Continuous Random Variables

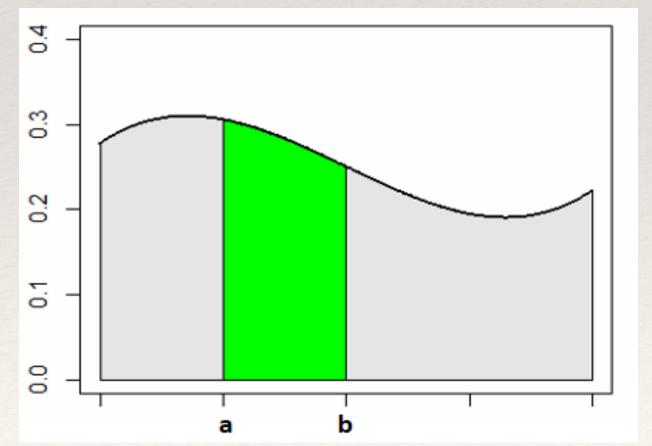
ST 314
Introduction to Statistics for Engineers



## Continuous Probability Distributions

For a continuous random variable X, the **Probability Density Function (PDF)**, denoted by f(X), is a mathematical expression for the shape of the distribution of X.

We use the pdf to calculate probabilities between two values of X and expected values of X.



\* The probability density function is a continuous function of the random variable X such that

$$P(c \le X \le d) = \int f(x) dx \qquad P(x=c) = 0$$

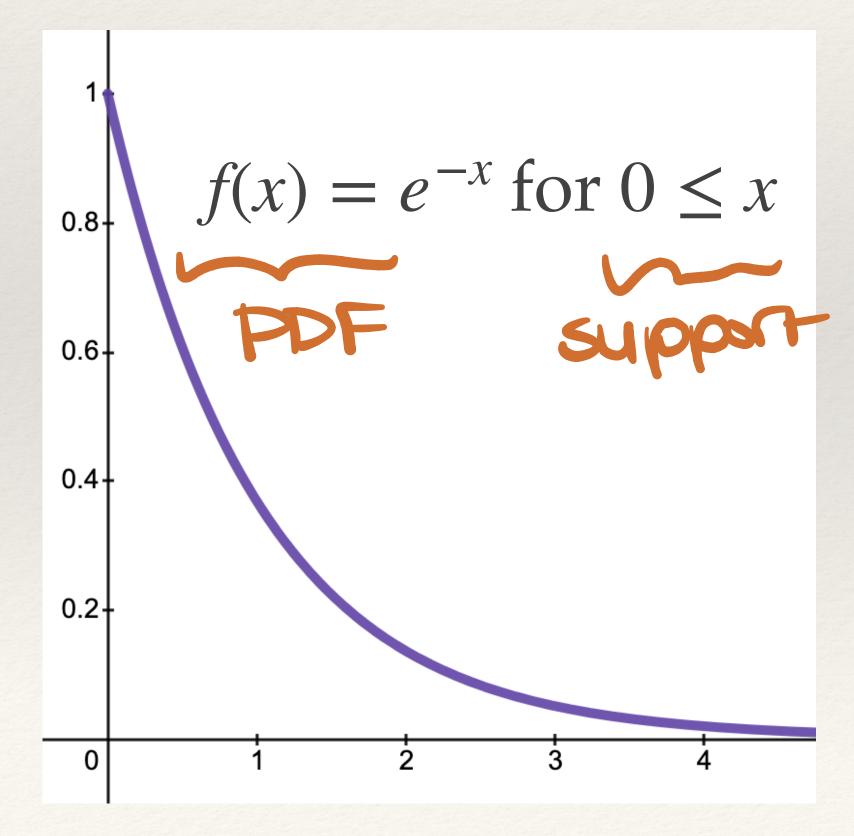
$$P(X=c) = P(c \le X \le c) = \int f(x) dx$$

$$f(x) \ge 0$$

$$f(x) \ge 0$$

$$f(x) dx = 1$$

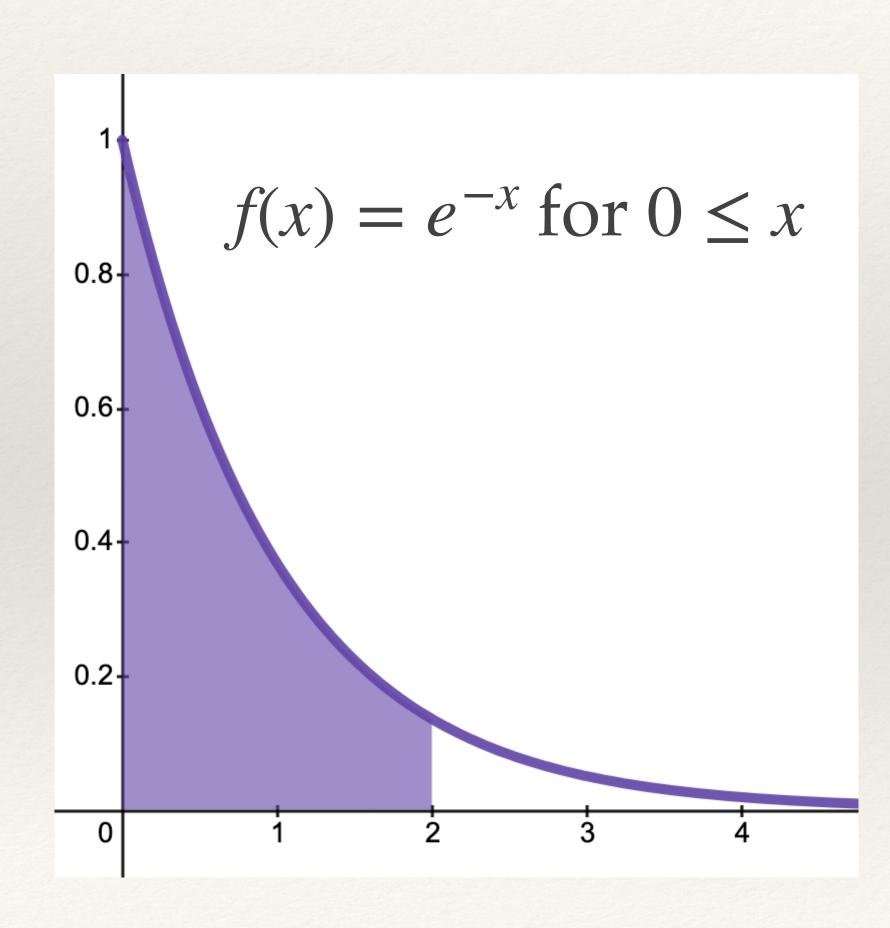
Let *X* be the random variable that represents the number of minutes between customers who enter the drive thru of a fast food restaurant. Let's model the time between customers with the following probability density function.



Does the area under f(x) equal 1?

$$\int_{0}^{\infty} e^{-x} dx = -e^{-x} \Big|_{0}^{\infty}$$

$$= 0 - (-1) = 1$$



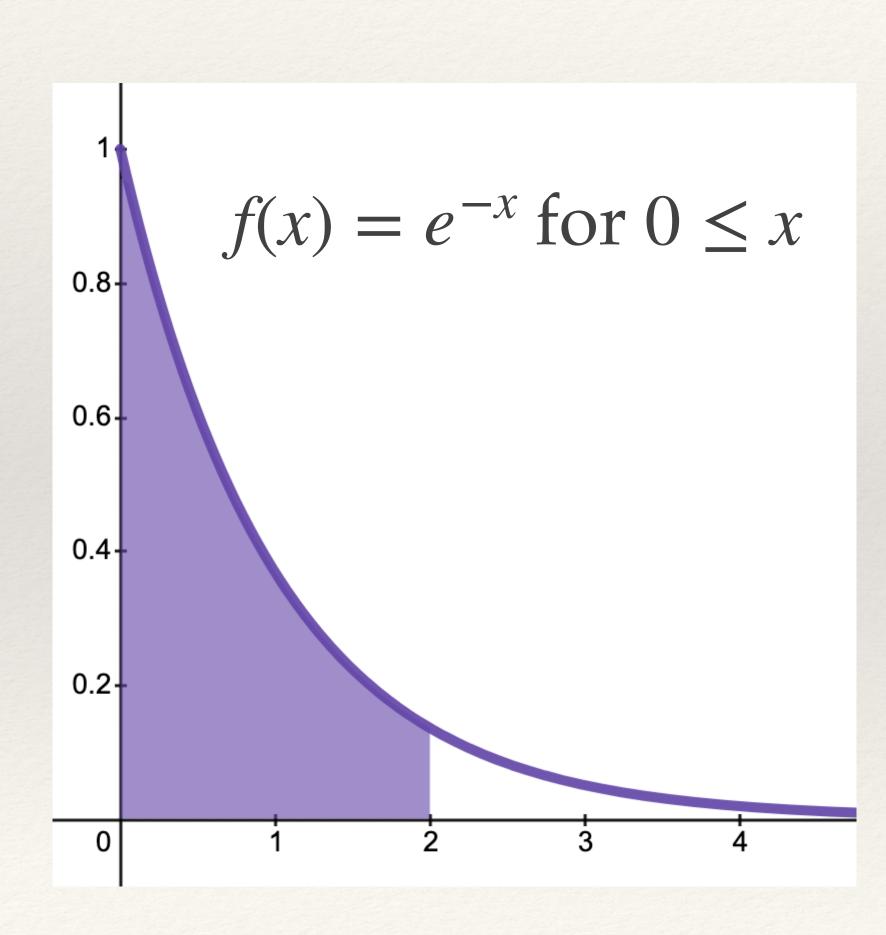
A customer has just entered the drive thru. Use the PDF to determine the probability that the next customer will arrive within 2 minutes.

$$P(0 \le X \le 2) = \int_{0}^{2} e^{-x} dx$$

$$= -e^{-x} |_{0}^{2}$$

$$= -e^{-x} + 1$$

$$= 0.8047$$



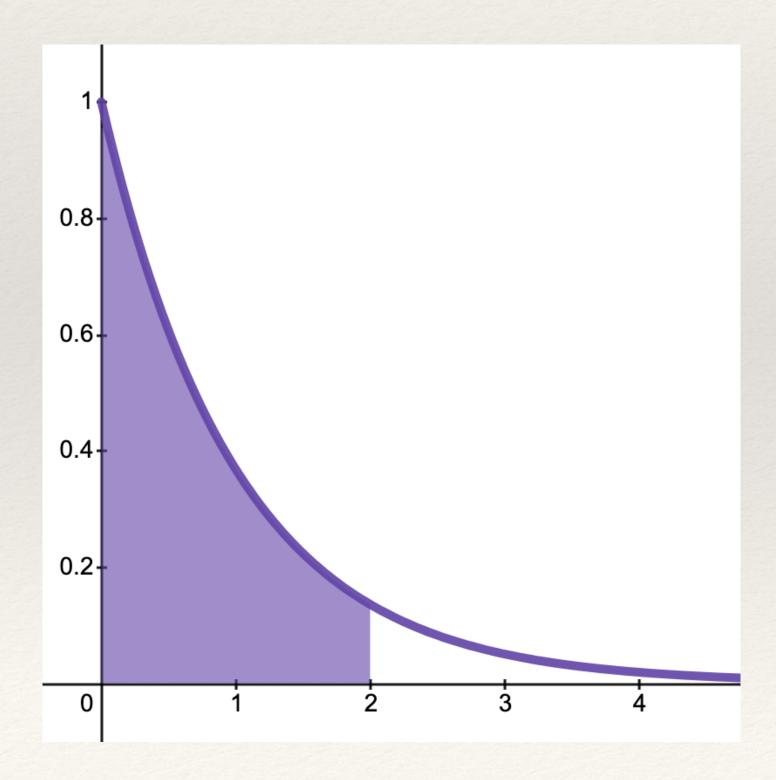
A customer has just entered the drive thru. Use the PDF to determine the probability that the next customer will arrive within 2 minutes.

$$P(X > 2) = P(X \ge 2)$$

$$= \int_{2}^{\infty} e^{-x} dx$$

#### Cumulative Distribution Function

For a continuous random variable X, the **Cumulative Density Function (CDF)** is used to calculate the probability the random variable X will be less than or equal to some real number x.



$$F(x) = P(X \le x) = \int f(t)dt$$

$$-\infty$$
change to lower bound of the support

#### Cumulative Distribution Function

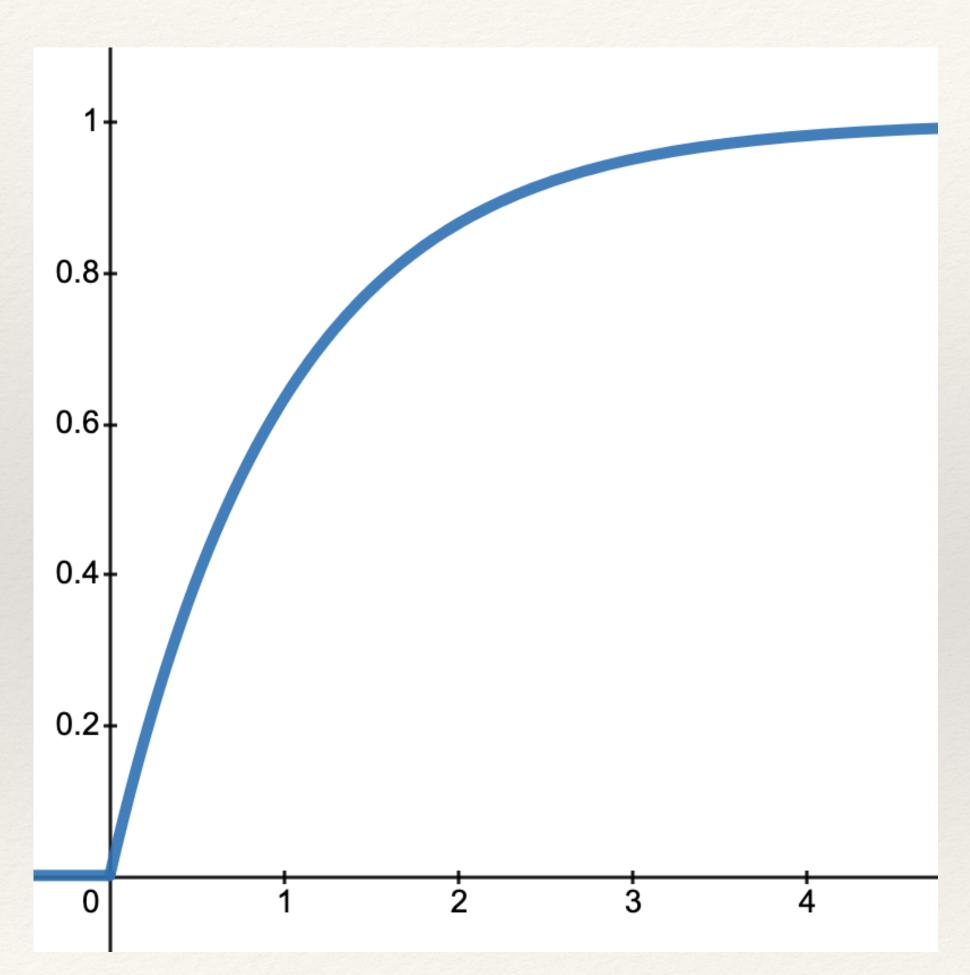
Properties
$$F(x) = P(X \le x) = \int_{-\infty}^{x} f(t) dt \quad (def).$$

$$D \le F(x) \le 1$$

$$P(c \le X \le d) = F(d) - F(c) = \int_{c}^{d} f(x) dx$$

$$F'(x) = F(x)$$
1st derivative of CDF equals the PDF

### Cumulative Distribution Function (CDF)



The random variable *X* has the following probability density function. Determine its CDF.

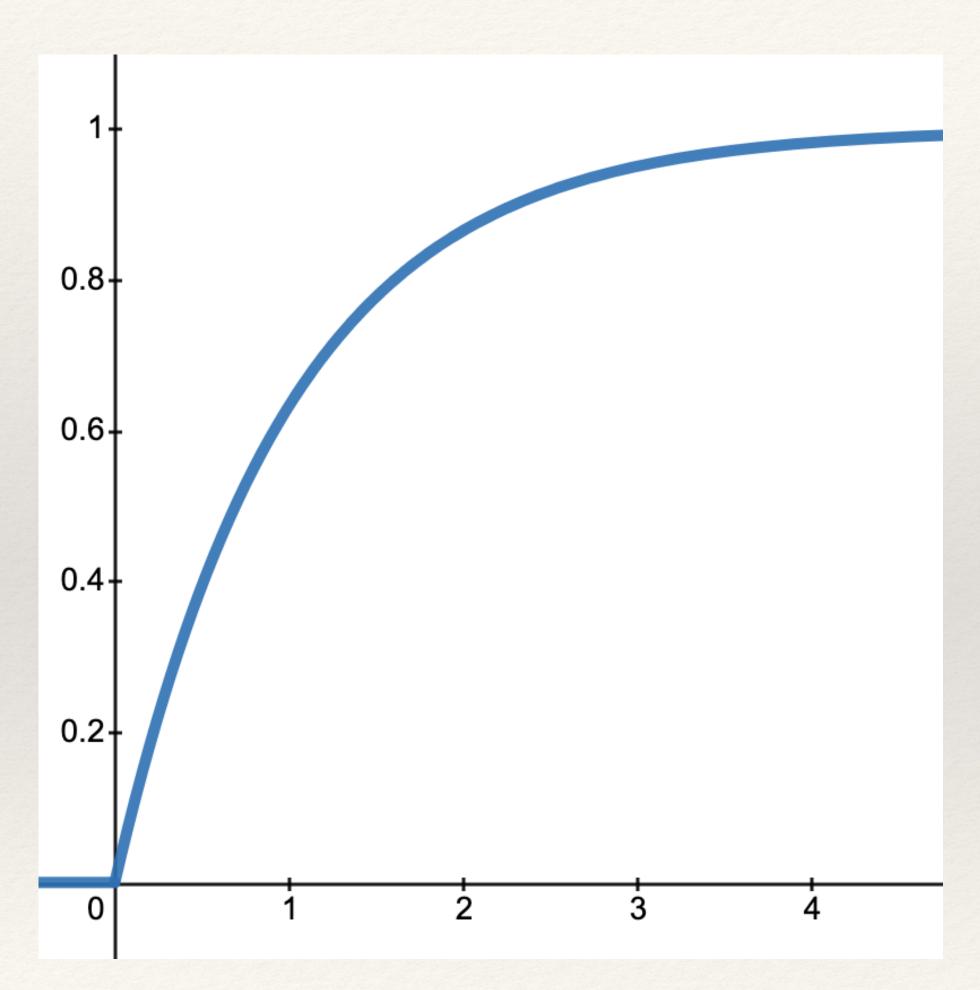
$$f(x) = e^{-x} \text{ for } 0 \le x$$

$$F(x) = \int_{0}^{2\pi} e^{-t} dt$$

$$= -e^{-t} \Big|_{0}^{2\pi} = -e^{-x} - (-1)^{2\pi}$$

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

#### Cumulative Distribution Function (CDF)



A customer has just entered the drive thru. Use the CDF to determine the probability that the next customer will arrive within 2 minutes.

F(x) = 1 - e

$$P(X \le 2) = F(2) = 1 - e$$

$$= 0.8047$$

### Expectation

Let *X* be a continuous random variable with PDF f(x) and possible outcomes from  $-\infty$  to  $\infty$ .

The **expected value** is the value we would expect X to take on. This is the average or mean value, denoted by E(X) or  $\mu_X$ .

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

$$Discrek : E(X) = Z \times P(X=x)$$

#### Variance

The **variance**,  $\sigma_x^2$ , is a measurement of how much spread there is in the distribution of X.

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

The **standard deviation**,  $\sigma_x$ , is the size of a typical deviation away from  $\mu_x$  and has the same units as X and  $\mu_x$ .

$$SD(X) = \sigma_X = \sqrt{var(X)}$$

# Expectation & Variance Example

Suppose *X* is a continuous random variable defined on the interval [0,1] with the following probability density function:  $f(x) = (3x^2)$ 

Find the expectation, 
$$E(X)$$
, and the variance,  $Var(X)$ , of  $X$ .

$$E(X) = \int_{0}^{\infty} x \cdot 3x^{2} dx = \int_{0}^{\infty} 3x^{3} dx = \frac{3}{4}x^{4} = \frac{3}{4}$$

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

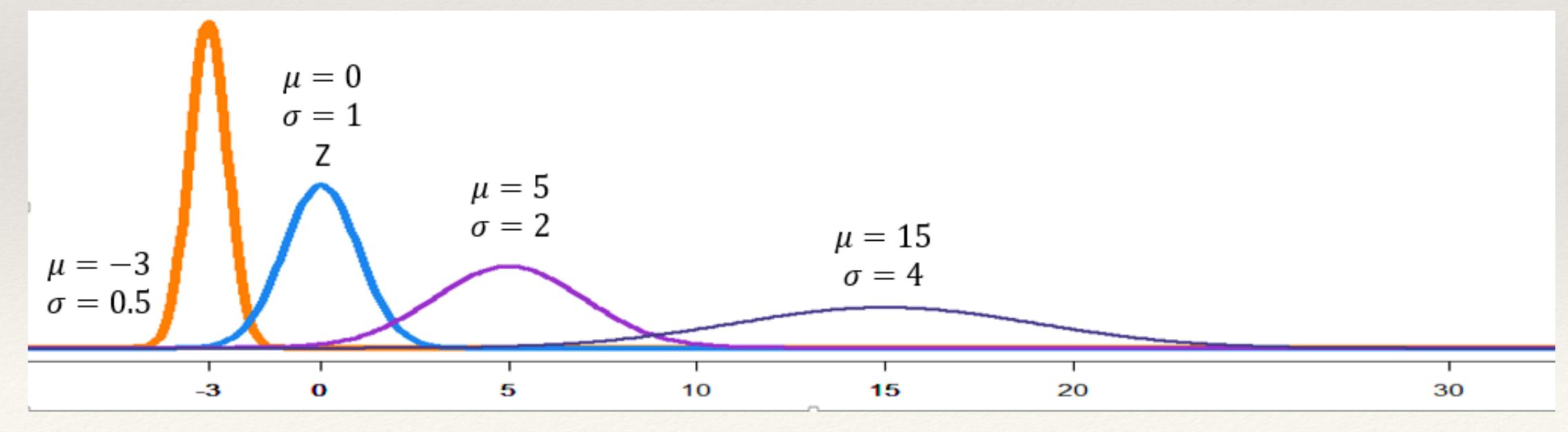
$$E(X^{2}) = \int x^{2} \cdot 3x^{2} dx = \int 3x^{4} dx = \frac{3}{5}x^{5} \Big|_{0}^{1} = \frac{3}{5}$$

$$Var(X) = \frac{3}{5} - \left(\frac{3}{4}\right)^2 = 0.0375$$

$$SD(X) = -\sqrt{Var(X)} = -\sqrt{0.0075}$$

#### The Normal Distribution

- \* Defined by two parameters: U, T X~N(u, 5)
- \* Symmetric, single-peaked (uni-modal), bell-shaped
- \* Models many naturally occurring random variables



#### Normal PDF

A random variable *X* that follows a Normal distribution has a probability density function:

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

$$* E(X) = 1$$

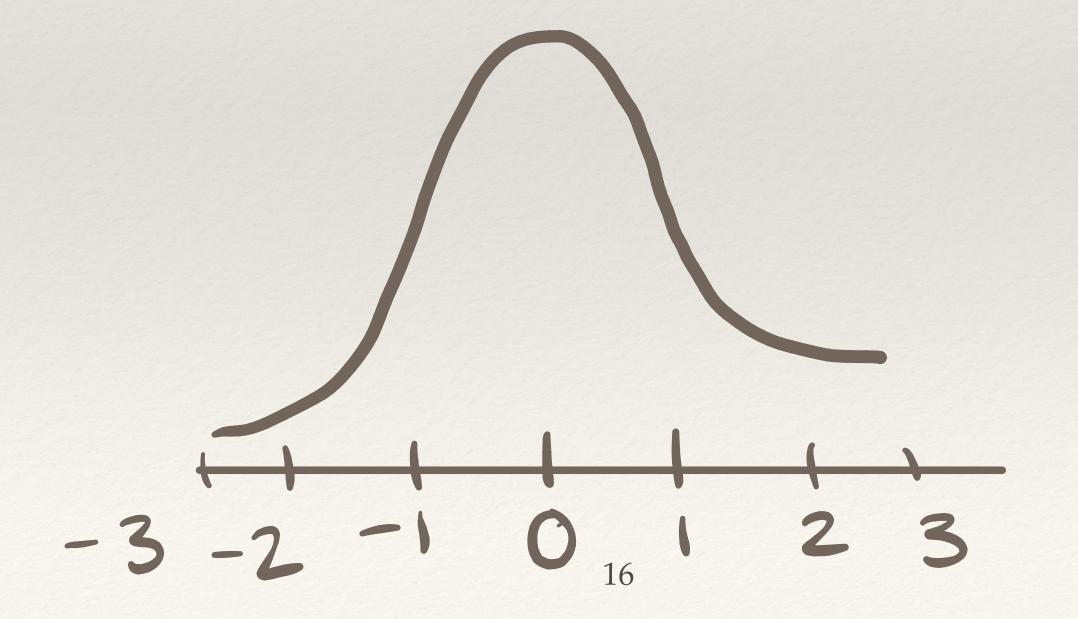
$$* Var(X) = 1 SD(X) = 1$$

### Standardizing Normal Distributions

For a Normal random variable *X*, a *z*-score represents the number of standard deviations any observation *x* is from the mean:

#### The Standard Normal Distribution

- \* Any normally distributed random variable can be "standardized" or transformed into a **standard normal distribution**.
- \* The **standard normal** random variable, denoted by *Z*, has  $\mu = \underline{\hspace{0.5cm}}$  and  $\sigma = \underline{\hspace{0.5cm}}$  .



### Normal Random Variable Example

For a particular bridge, recorded vehicle speeds are normally distributed with a mean of 58 mph and a standard deviation of 10 mph.

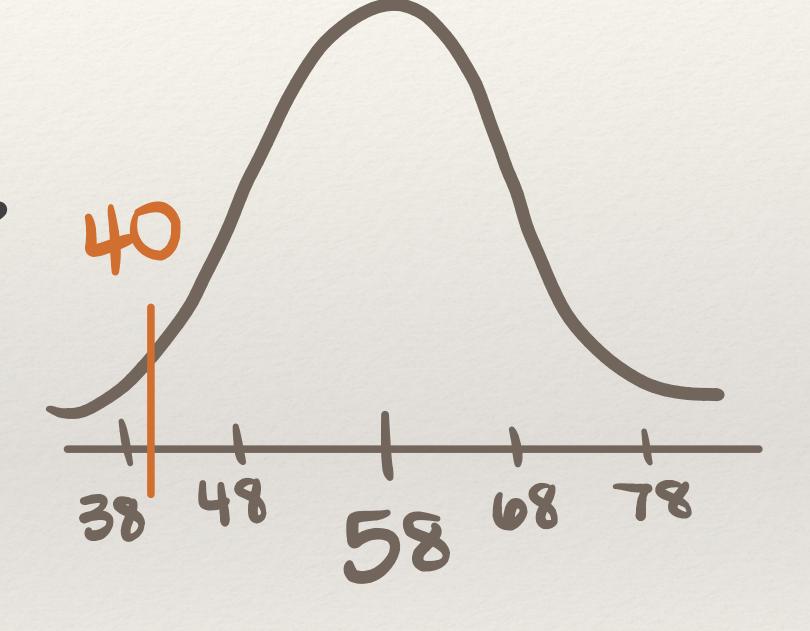
Suppose a randomly chosen vehicle is going 40 miles per hour.

How many standard deviations away from the mean is 40 mph?

$$\mu = 58$$

$$\sigma = 10$$

$$Z = x - \mu = 40 - 58 = -1.8$$



### Normal Random Variable Example

For a particular bridge, recorded vehicle speeds are normally distributed with a mean of 58 mph and a standard deviation of 10 mph.

Suppose a randomly chosen vehicle is going 40 miles per hour.

What is the probability of a randomly selecting a vehicle going less than 40 mph?
$$P(X < 40) = \int_{1}^{40} \frac{1}{2(10)^2} dx$$

What is the probability of a randomly selecting a vehicle going between 40 and 65 mph?