

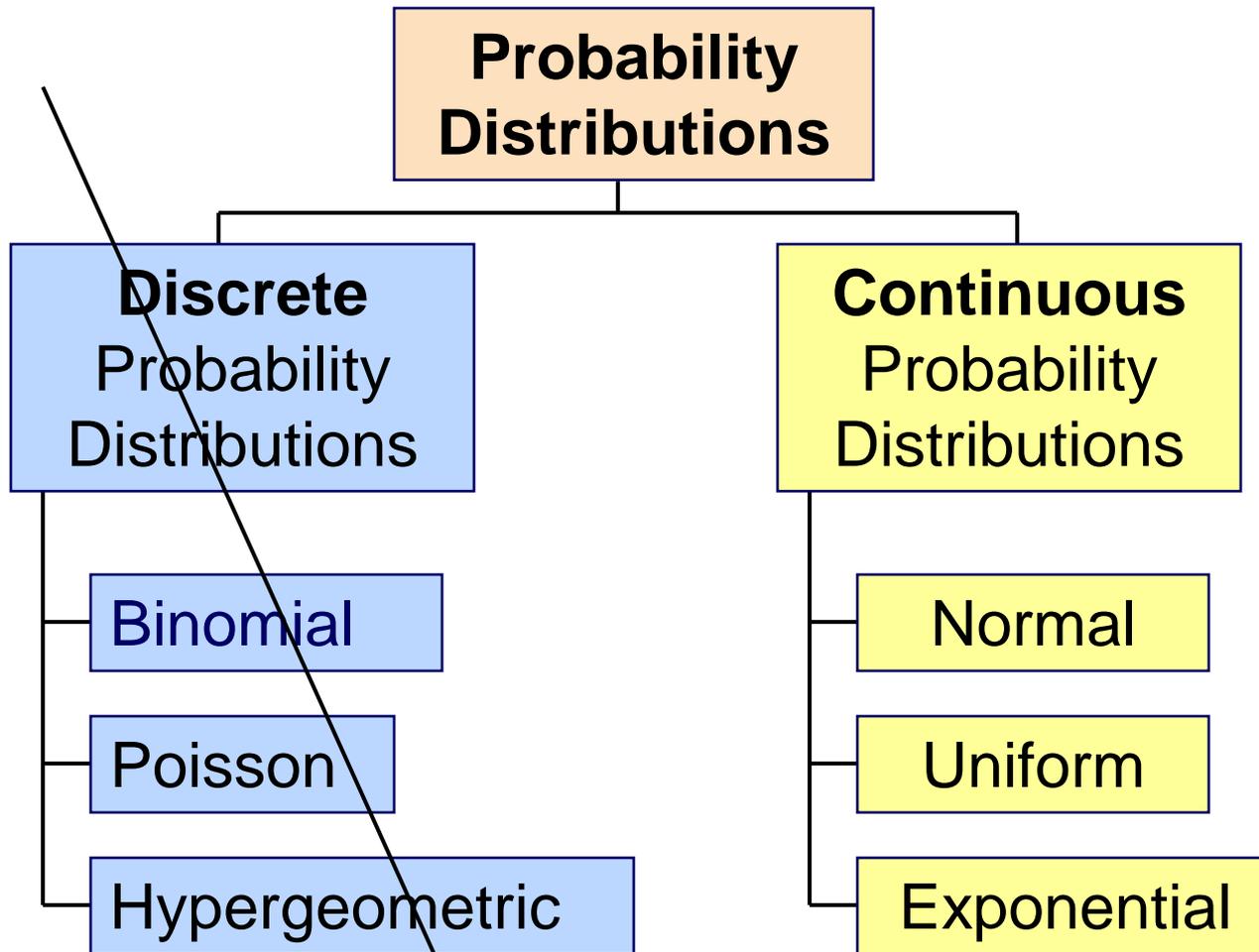
# Continuous Probability Distributions

# Chapter Goals

**After completing this chapter, you should be able to:**

- Apply the binomial distribution to applied problems
- Compute probabilities for the Poisson and hypergeometric distributions
- Find probabilities using a normal distribution table and apply the normal distribution to business problems
- Recognize when to apply the uniform and exponential distributions

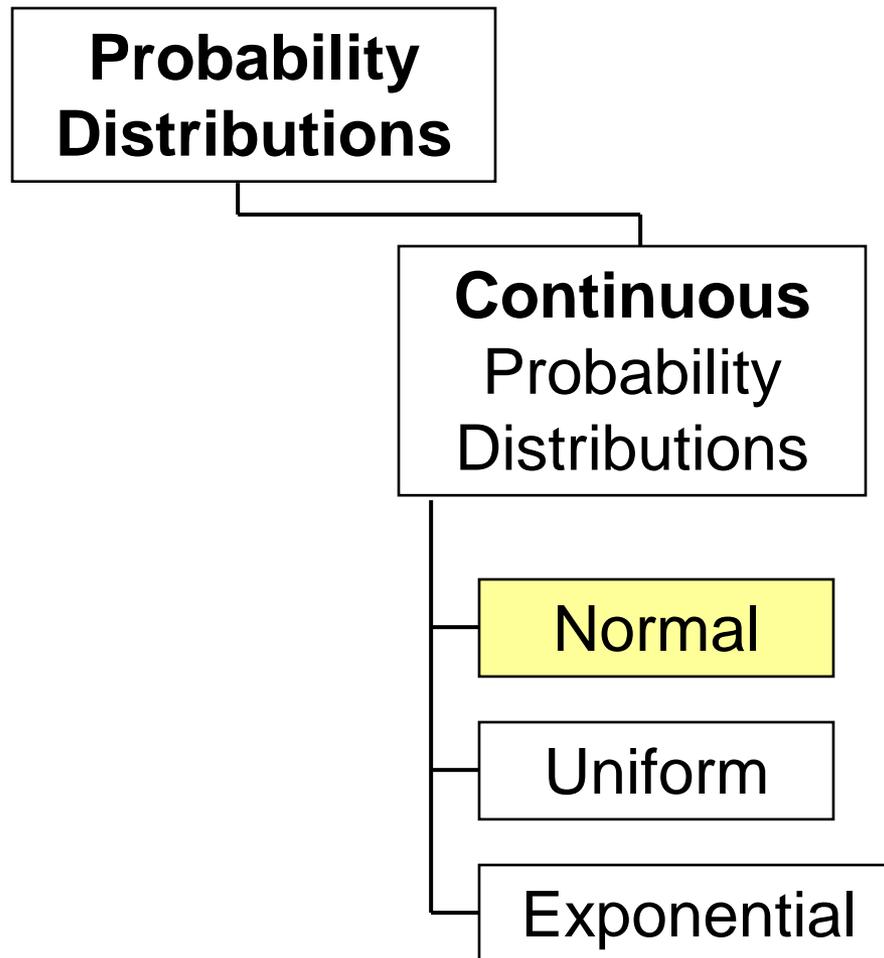
# Probability Distributions



# Continuous Probability Distributions

- A **continuous random variable** is a variable that can assume any value on a continuum (can assume an uncountable number of values)
  - thickness of an item
  - time required to complete a task
  - temperature of a solution
  - height, in inches
- These can potentially take on any value, depending only on the ability to measure accurately.

# The Normal Distribution



# The Normal Distribution

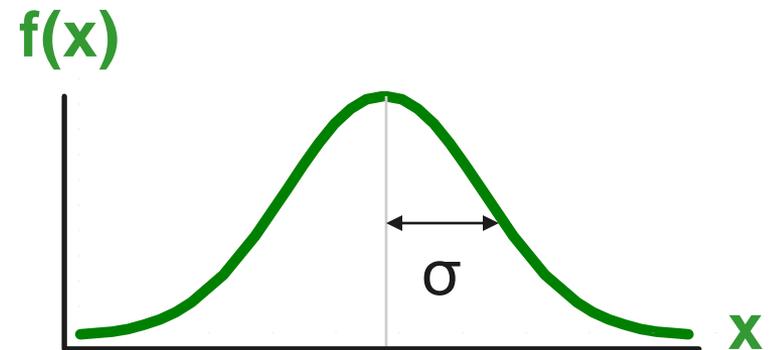
- Bell Shaped
- Symmetrical
- Mean, Median and Mode are Equal

Location is determined by the mean,  $\mu$

Spread is determined by the standard deviation,  $\sigma$

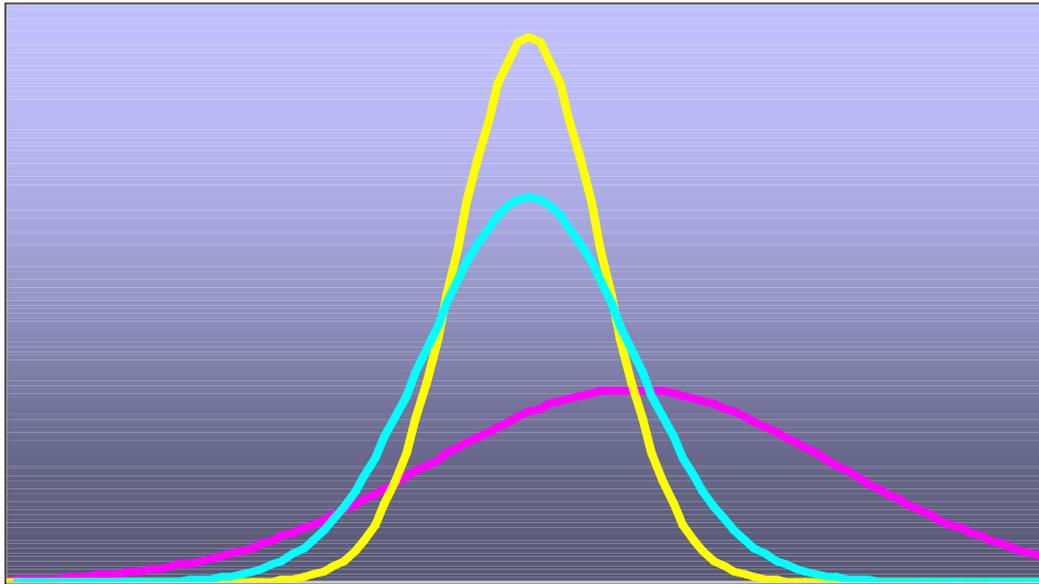
The random variable has an infinite theoretical range:

$+\infty$  to  $-\infty$



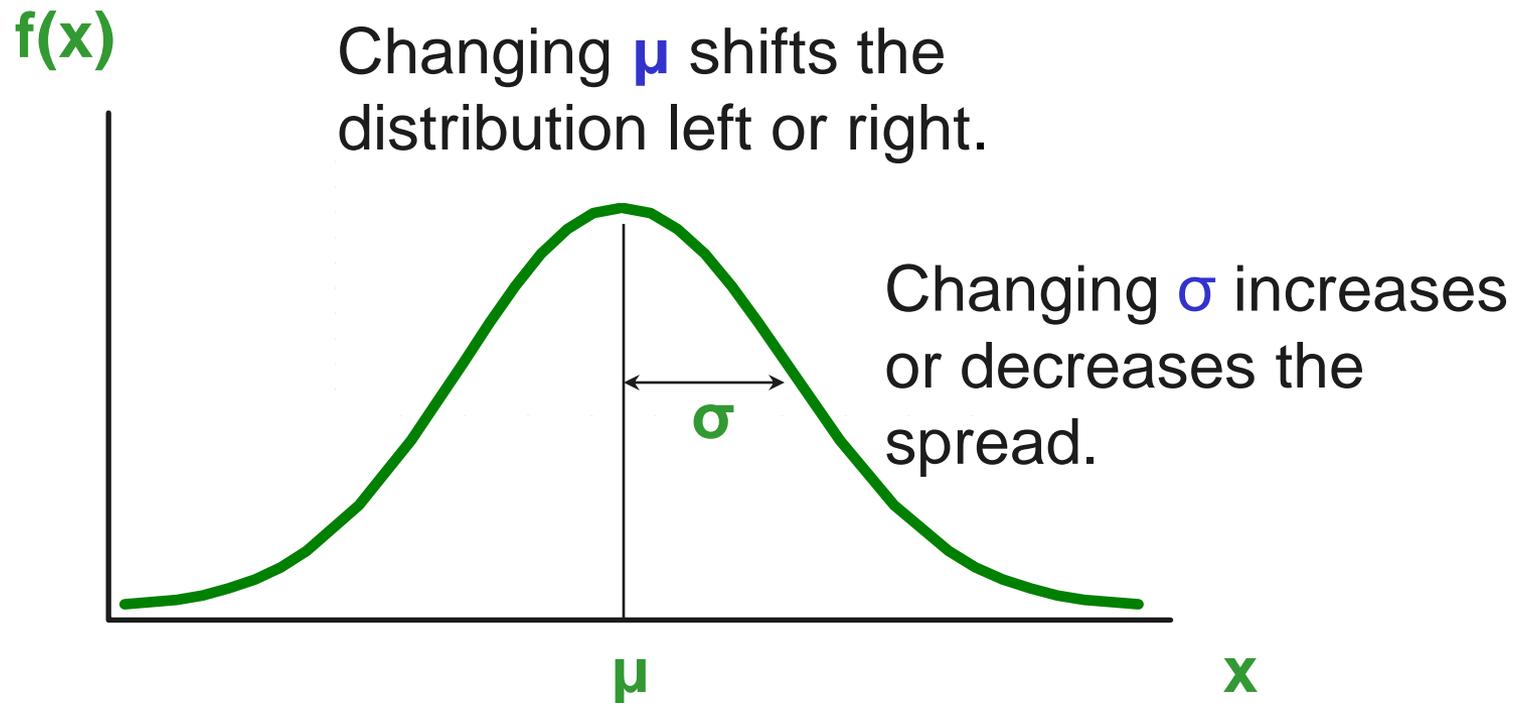
$\mu$   
↑  
Mean  
= Median  
= Mode

# Many Normal Distributions



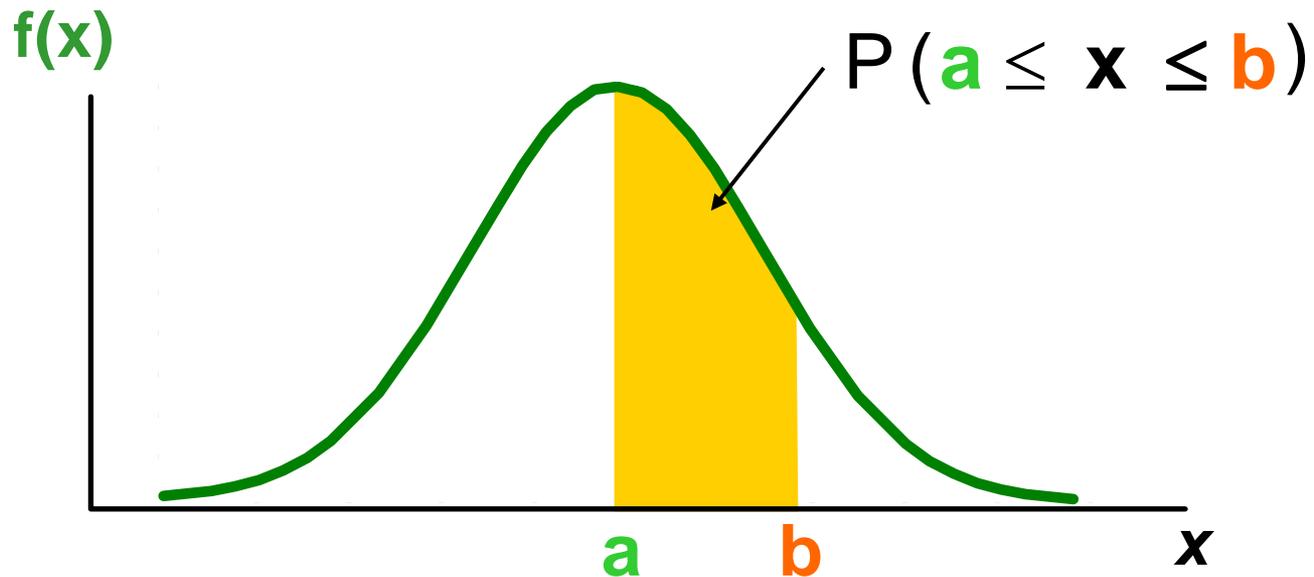
By varying the parameters  $\mu$  and  $\sigma$ , we obtain different normal distributions

# The Normal Distribution Shape



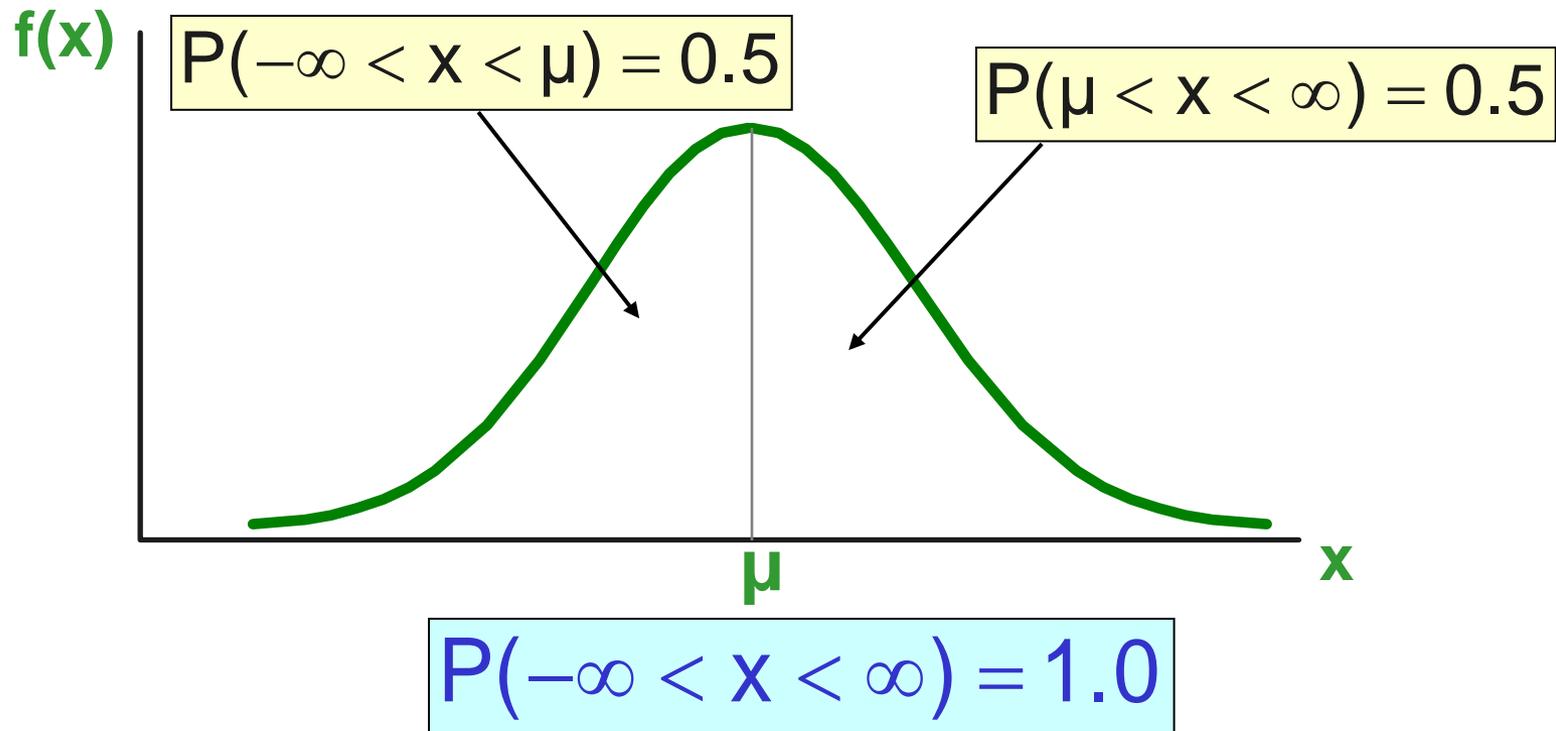
# Finding Normal Probabilities

Probability is measured by the area under the curve



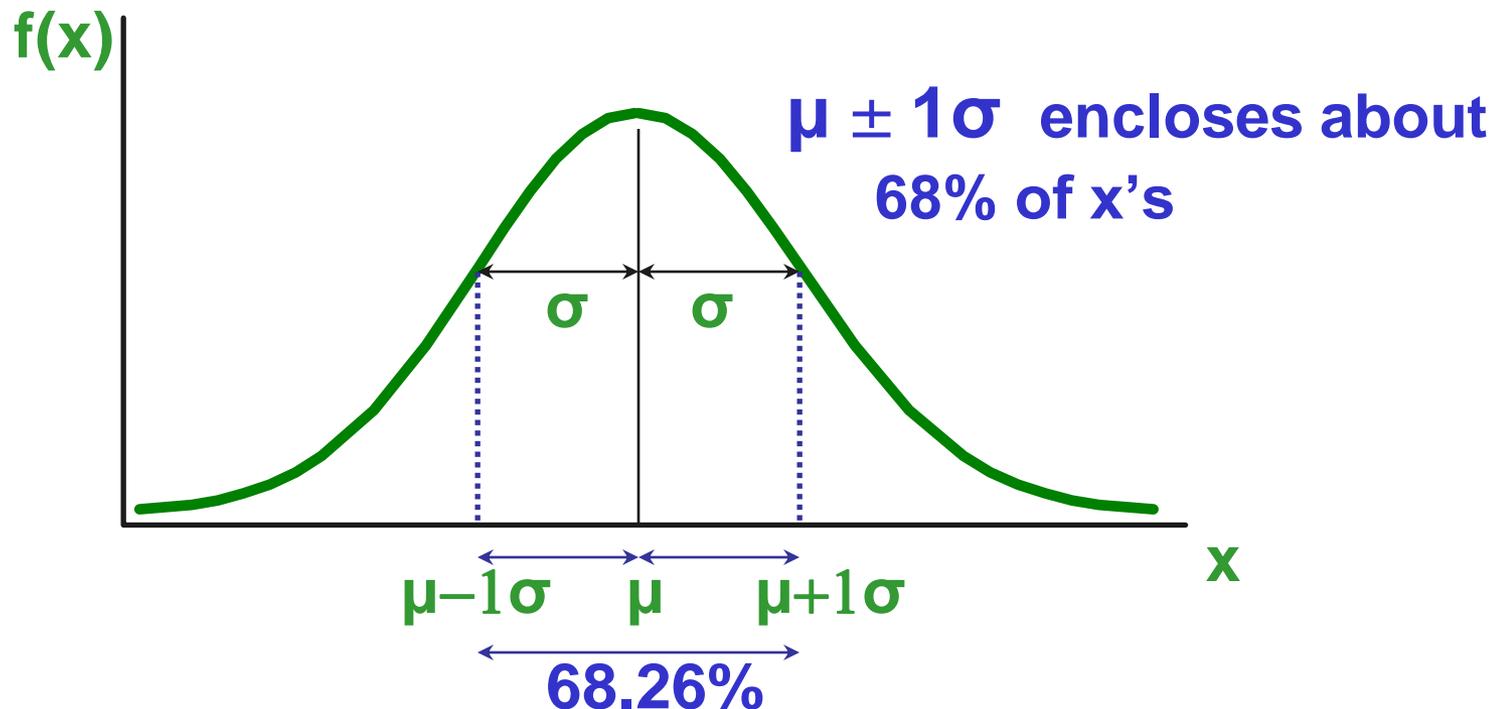
# Probability as Area Under the Curve

The total area under the curve is 1.0, and the curve is symmetric, so half is above the mean, half is below



# Empirical Rules

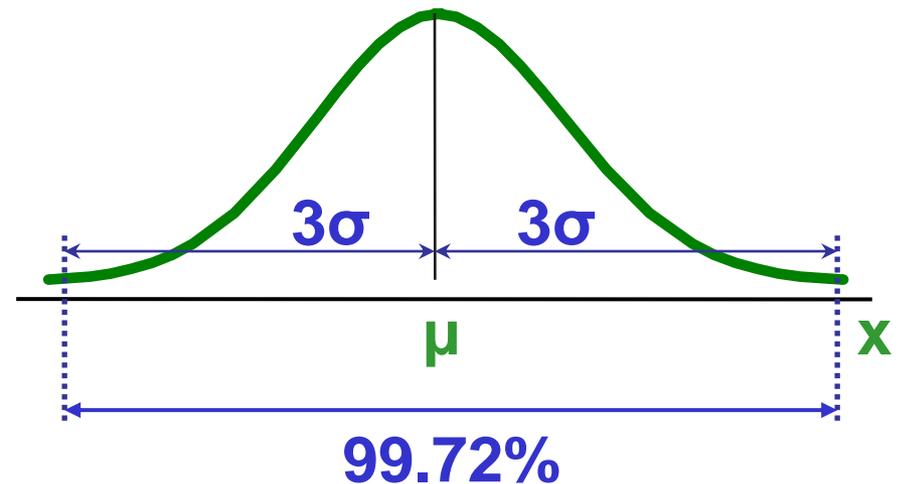
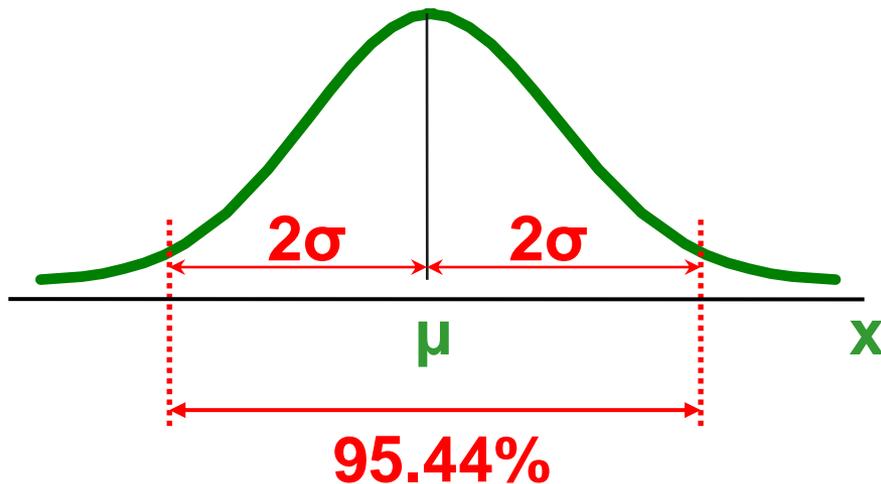
What can we say about the distribution of values around the mean? There are some general rules:



# The Empirical Rule

*(continued)*

- $\mu \pm 2\sigma$  covers about **95.44%** of  $x$ 's
- $\mu \pm 3\sigma$  covers about **99.72%** of  $x$ 's

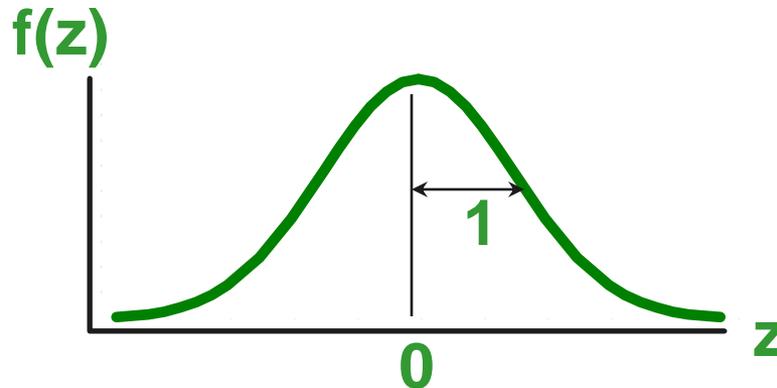


# Importance of the Rule

- If a value is about **2 or more** standard deviations away from the mean in a normal distribution, then it is **far** from the mean
- The chance that a value that far or farther away from the mean is **highly unlikely**, given that particular mean and standard deviation

# The Standard Normal Distribution

- Also known as the “z” distribution
- Mean is defined to be 0
- Standard Deviation is 1



Values above the mean have **positive** z-values,  
values below the mean have **negative** z-values

# The Standard Normal

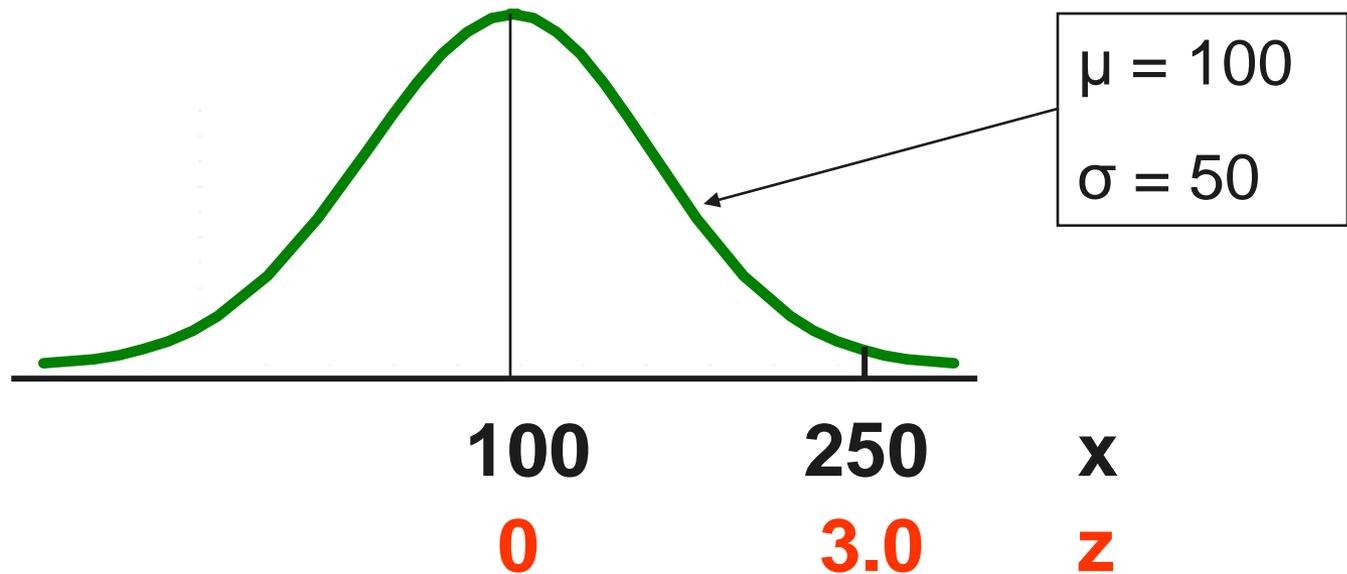
- Any normal distribution (with any mean and standard deviation combination) can be transformed into the standard normal distribution ( $z$ )
- Need to transform  $x$  units into  $z$  units

# Translation to the Standard Normal Distribution

- Translate from  $x$  to the standard normal (the “ $z$ ” distribution) by subtracting the mean of  $x$  and dividing by its standard deviation:

$$z = \frac{x - \mu}{\sigma}$$

# Comparing x and z units

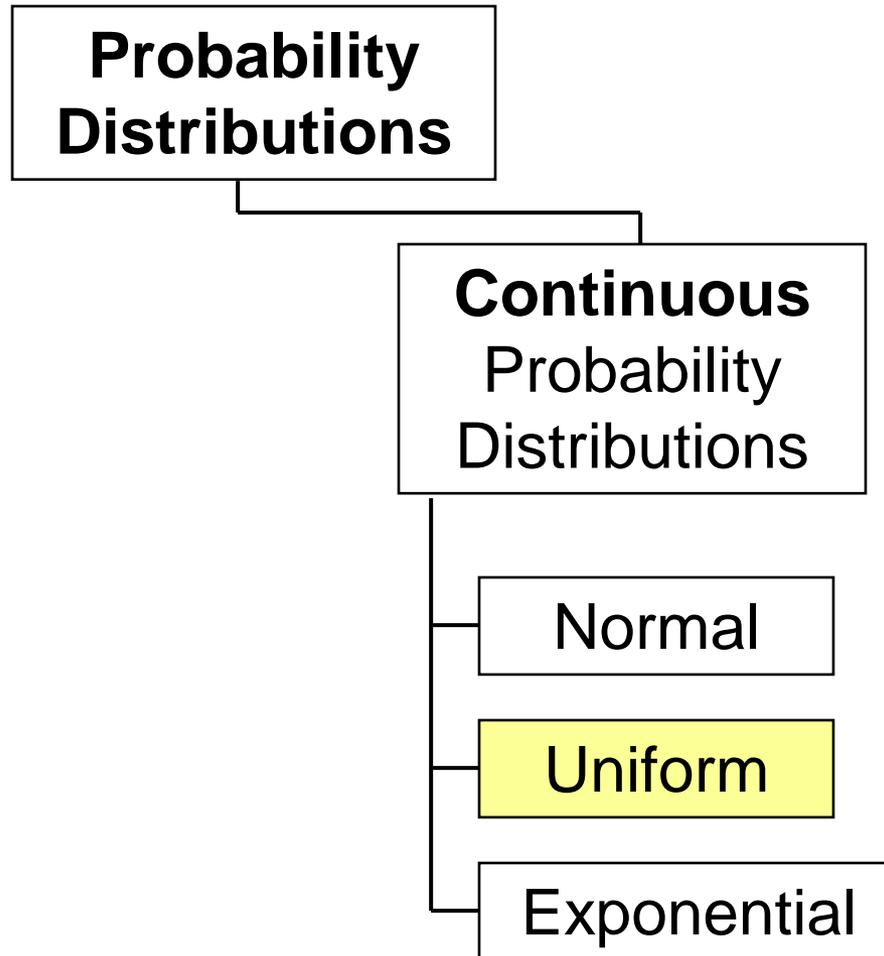


**Note that the distribution is the same, only the scale has changed. We can express the problem in original units (x) or in standardized units (z)**

# Excel Based Examples

**Please practice the sample problems that are posted.**

# The Uniform Distribution



# The Uniform Distribution

- The uniform distribution is a probability distribution that has equal probabilities for all possible outcomes of the random variable

# The Uniform Distribution

*(continued)*

The Continuous Uniform Distribution:

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

where

$f(x)$  = value of the density function at any  $x$  value

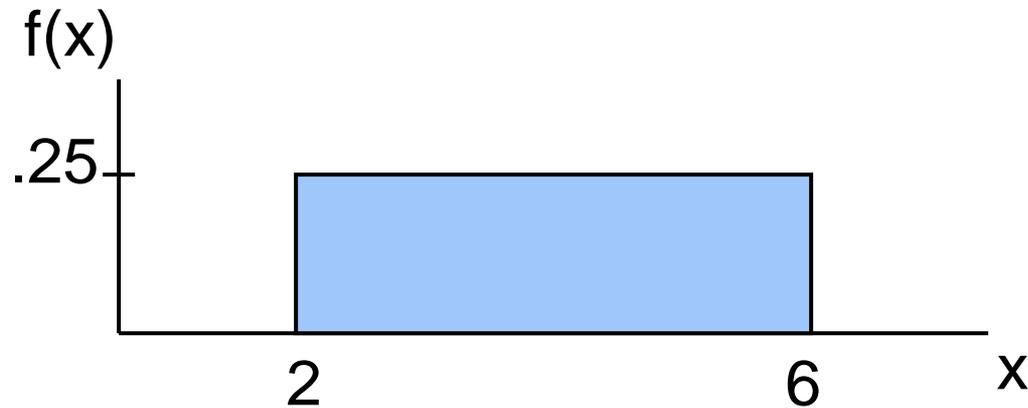
$a$  = lower limit of the interval

$b$  = upper limit of the interval

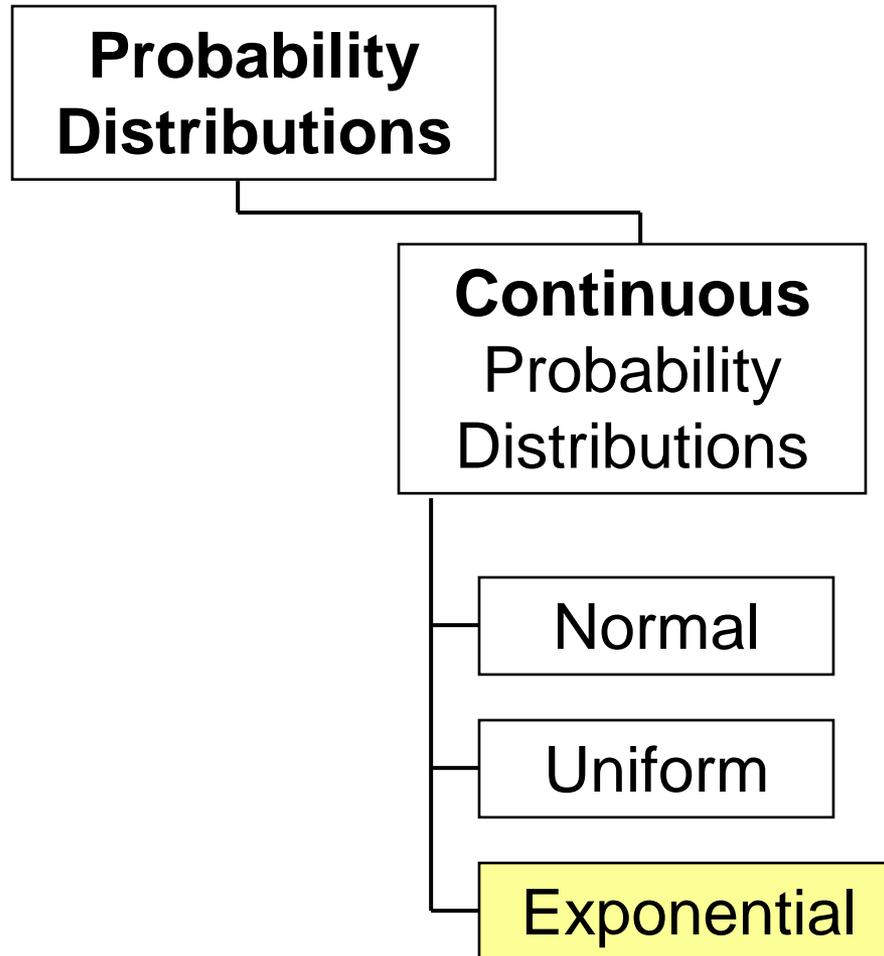
# Uniform Distribution

**Example:** Uniform Probability Distribution  
Over the range  $2 \leq x \leq 6$ :

$$f(x) = \frac{1}{6 - 2} = .25 \quad \text{for } 2 \leq x \leq 6$$



# The Exponential Distribution



# The Exponential Distribution

- Used to measure the **time that elapses between two occurrences** of an event (the time between arrivals)
  - Examples:
    - Time between trucks arriving at an unloading dock
    - Time between transactions at an ATM Machine
    - Time between phone calls to the main operator

# The Exponential Distribution

- The probability that an arrival time is equal to or less than some specified time  $a$  is

$$P(0 \leq x \leq a) = 1 - e^{-\lambda a}$$

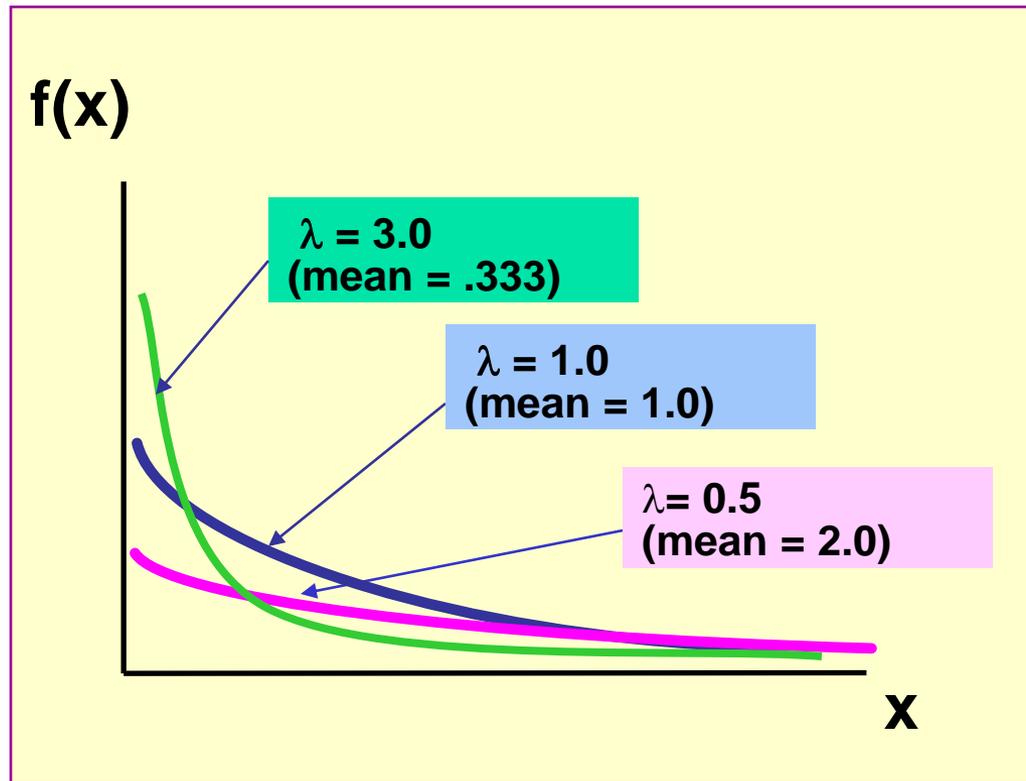
where  $1/\lambda$  is the mean time between events

Note that if the **number of occurrences per time period** is Poisson with mean  $\lambda$ , then the **time between occurrences** is exponential with mean time  $1/\lambda$

# Exponential Distribution

*(continued)*

- Shape of the exponential distribution



# Example

**Example:** Customers arrive at the claims counter at the rate of 15 per hour (Poisson distributed). **What is the probability that the arrival time between consecutive customers is less than five minutes?**

- Time between arrivals is exponentially distributed with mean time between arrivals of 4 minutes (15 per 60 minutes, on average)
- $1/\lambda = 4.0$ , so  $\lambda = .25$
- $P(x < 5) = 1 - e^{-\lambda a} = 1 - e^{-(.25)(5)} = .7135$

# Module Summary

- Reviewed key continuous distributions
  - normal, uniform, exponential
  - Examples using Excel (sample problems)