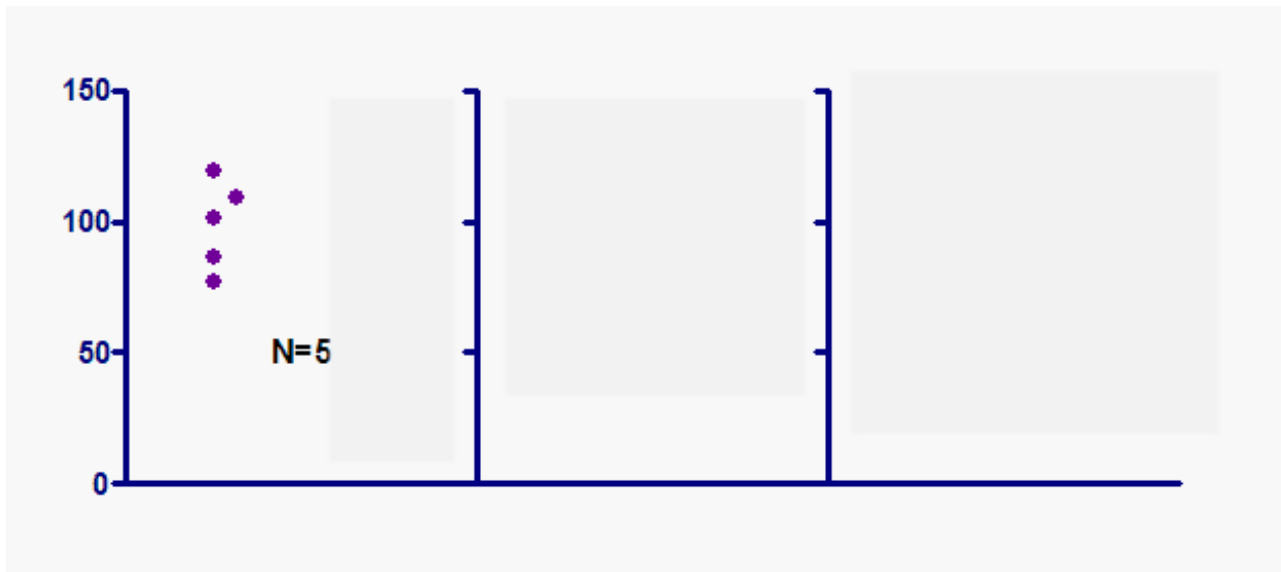


Learning Objectives

- **Describe** *confidence intervals* at the conceptual level
- **Contrast** *confidence intervals* with other measures of variability (e.g., standard deviation)
- **Calculate** *confidence intervals*

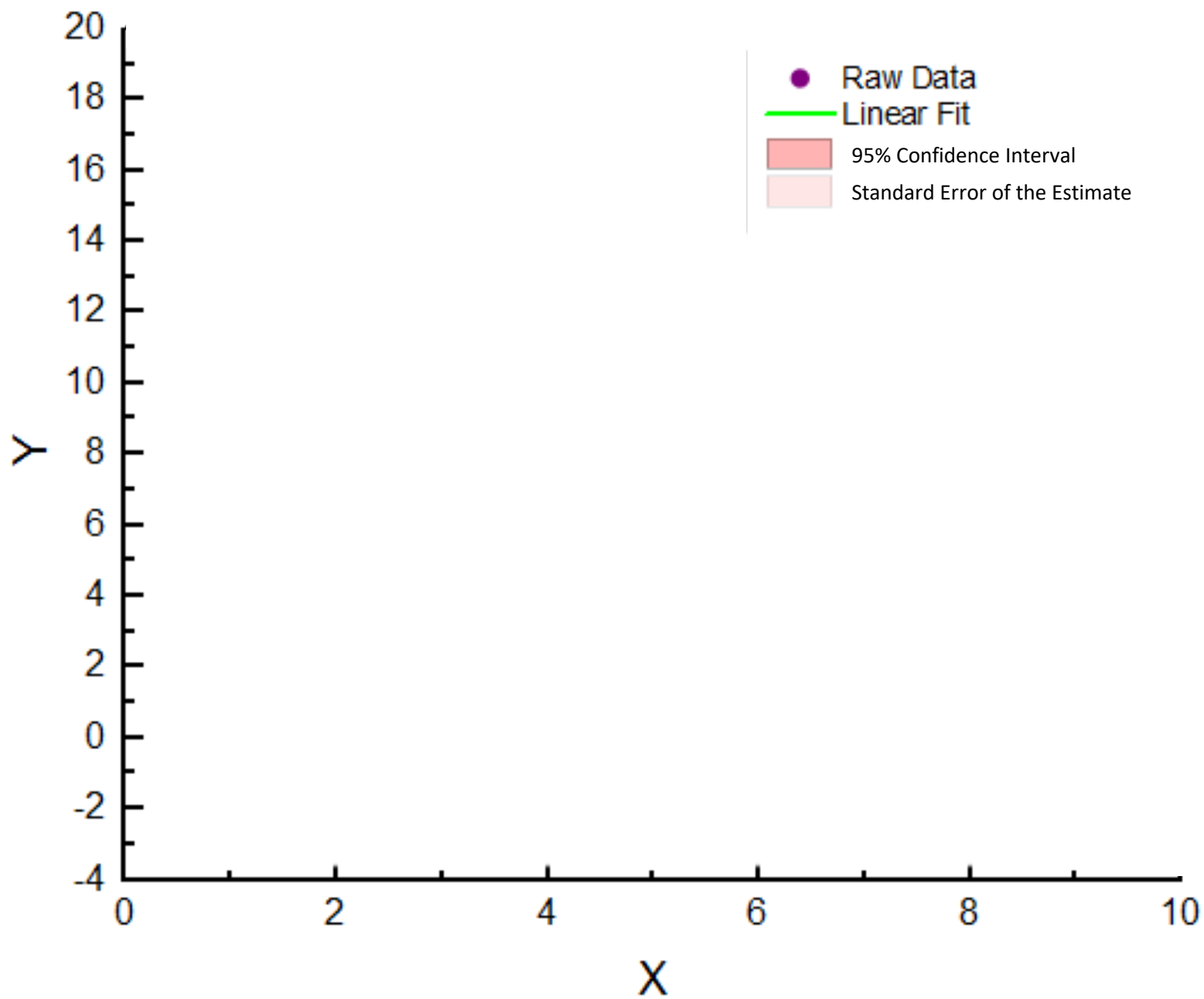
Confidence Intervals

- Characterize our confidence in **parameter** estimates
 - *CI*'s are a range we believe contains the parameter
 - Akin to “margin of error”



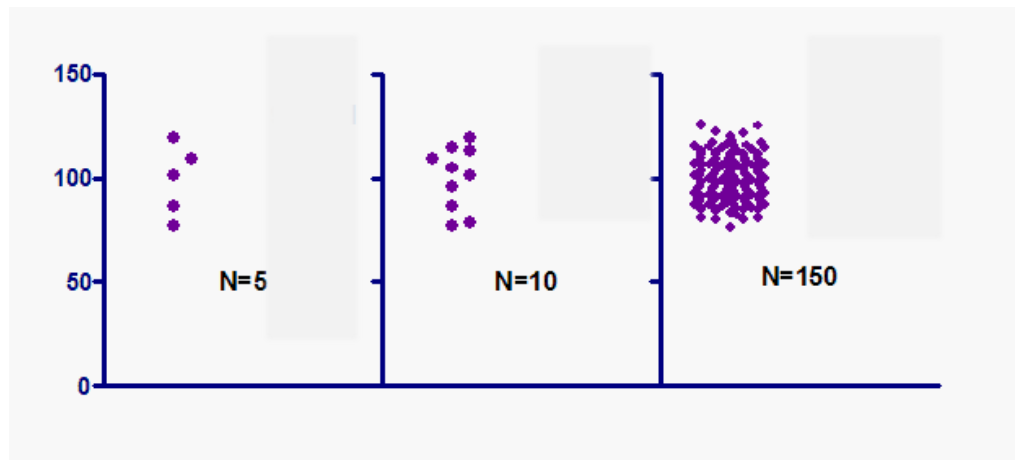
s vs. CI

- Standard deviation describes the variability of observations around a statistical model
 - Ex. Variability of X 's around \bar{X}
 - *How much do observations differ from the sample mean or a regression line?*
- Confidence intervals describe the variability of the statistical model itself
 - Ex. Variability of \bar{X} 's around $\mu_{\bar{X}}$
 - *How different might the sample mean or regression line be if I collected a new sample?*



Confidence Intervals

- Like α , we choose our confidence!
 - Larger *CI's more certainly* contain population parameters
 - But are *less practically useful*; & difficult to falsify
- Imagine a 100% Confidence Interval...



Formula for Confidence Intervals

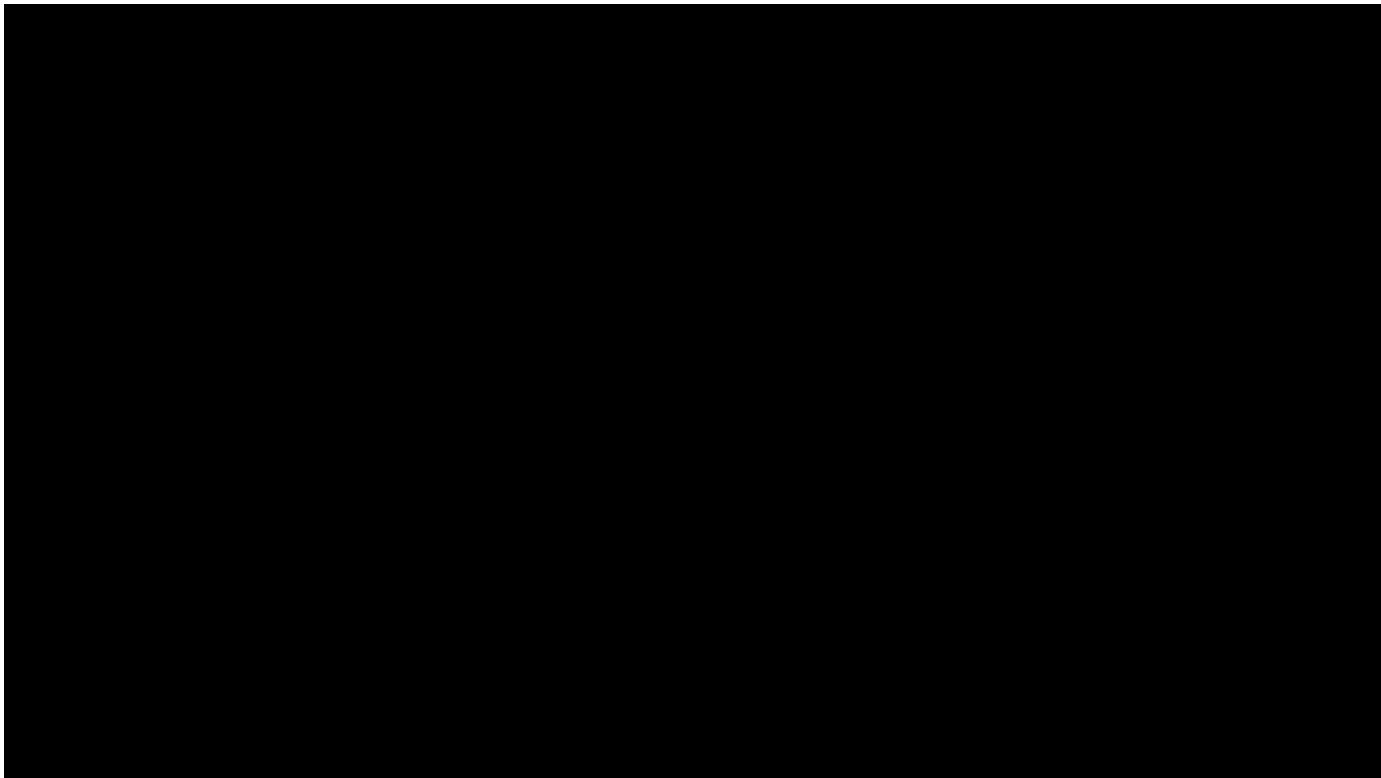
$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.025})$$

– We start by assuming $\bar{X}_{\text{obt}} = \mu$

- We assume that our observed data IS exactly the population parameter!

NHST: Assuming H_0 ***a priori* expectation is true**, here are plausible *a posteriori* observations (given sampling variability)

CI: Assuming ***a posteriori* observation is true**, here are plausible population parameters (given sampling variability)



Formula for Confidence Intervals

$$\begin{aligned}\mu_{\text{lower}} &= \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.025}) \\ \mu_{\text{upper}} &= \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_{.025})\end{aligned}$$

- We start by assuming $\bar{X}_{\text{obt}} = \mu$
- We add or subtract from the best guess (\bar{X}_{obt}) to set upper and lower bounds for plausible population parameters
- Standard error describes the variability of \bar{X} 's given N
 - $s_{\bar{X}} = CI_{68\%}$
- Multiply by value of t that sets this area into each tail of the distribution of sample means
 - Normally this will be $t_{.050}$ (when $CI_{90\%}$), $t_{.025}$ ($CI_{95\%}$), $t_{.005}$ ($CI_{99\%}$)

Formula for Confidence Intervals

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_?)$$

$$\mu_{\text{upper}} = \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_?)$$

What about when $t = 1$?

– When $t = 1$...then \bar{X}_{obt} is 1 $s_{\bar{X}}$ away from μ

- What is the area under the curve (*AUC*)?

– When $t = 1$; then $CI_{68\%}^*$

**If N is large*

Formula for Confidence Intervals

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.005})$$

$$\mu_{\text{upper}} = \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_{.005})$$

$t_{.005}$ is appropriate for a **99% CI**

- 0.5% of the time, actual parameter will be lower than the interval
- 0.5% of the time, actual parameter will be higher than the interval

Formula for Confidence Intervals

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.025})$$

$$\mu_{\text{upper}} = \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_{.025})$$

$t_{.025}$ is appropriate for a **95% CI**

- 2.5% of the time, actual parameter will be lower than the interval*
- 2.5% of the time, actual parameter will be higher than the interval

Formula for Confidence Intervals

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.X})$$

$$\mu_{\text{upper}} = \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_{.X})$$

Conceptual steps to creating $CI_{X\%}$

1. Determine total **AUC** that we're willing to be wrong ($1 - X$)
2. Split total **AUC** between the two tails ($\frac{\text{AUC}}{2}$)
3. Visualize $\frac{\text{AUC}}{2}$ in lower/upper tail
4. Determine value of t for $\frac{\text{AUC}}{2}$, using appropriate df

Formula for Confidence Intervals

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.05})$$

$$\mu_{\text{upper}} = \bar{X}_{\text{obt}} + (s_{\bar{X}} * t_{.05})$$

Practical steps to create **CI**_{90%}

1. Subtract **.90** from 1

$$1 - X = .10$$

2. Divide **.10** by 2

$$\frac{AUC}{2} = .05$$

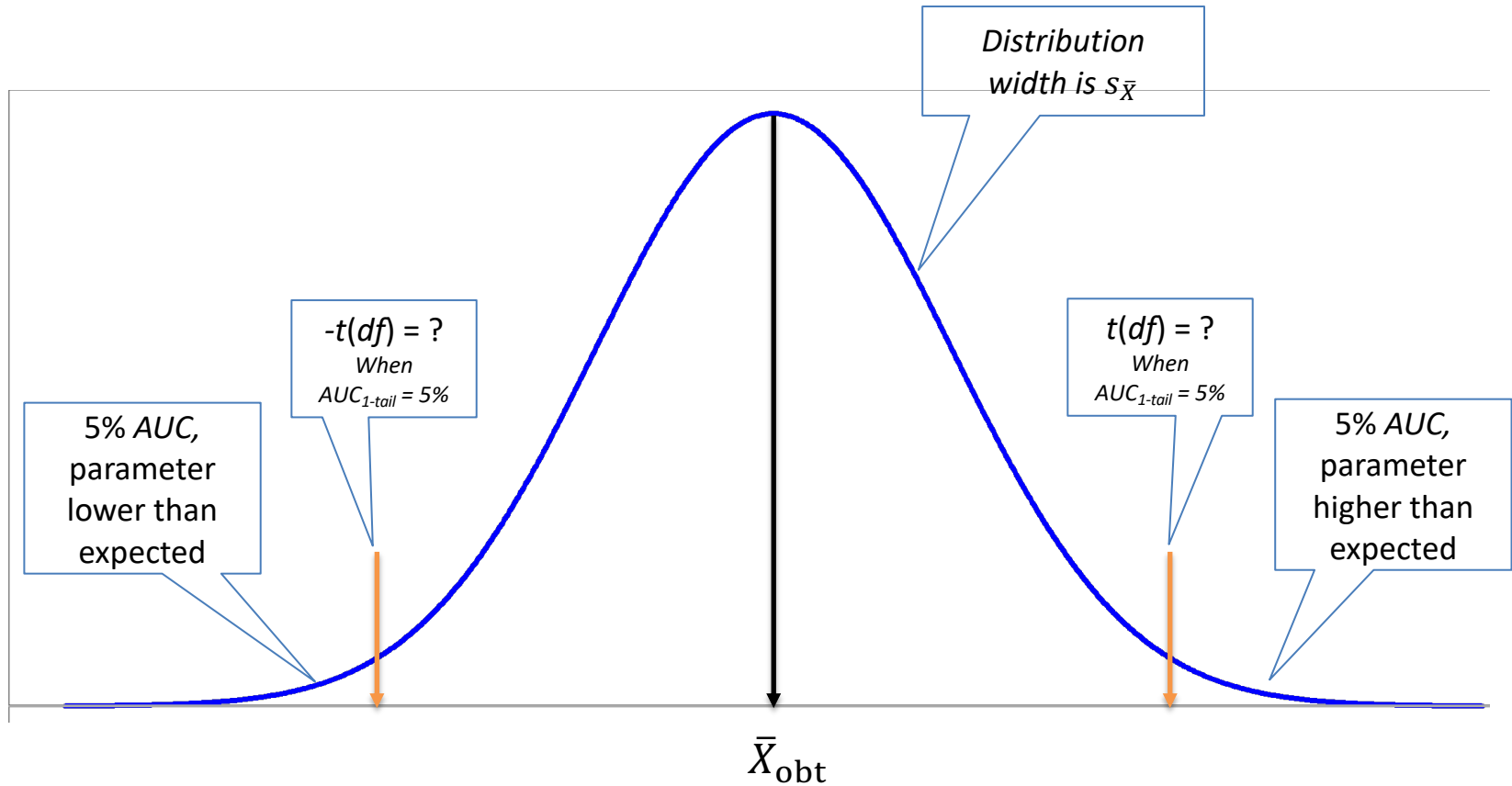
3. Place result in lower/upper tail

4. Find value of t that puts .05 in one tail

$$t_{.05}(28) = 1.701$$

(let's assume $N = 30$ & independent groups)

Visualize $CI_{90\%}$



Example: Practice Problem 13.4 (pp. 345)

- Manufacturer samples from 200 light bulbs and records how many hours it takes for them to burn out. The sample has a mean life of 215hrs ($s = 8$ hrs). Construct the 99% CI using the closest value for df

$$\bar{X}_{\text{obt}} = 215$$

$$s = 8$$

$$N = 200$$

$$s_{\bar{X}} = ???$$

$$t_{???} = ???$$

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_?)$$

Determine t :

- $1 - .99 = .01$ (total AUC)
- $\frac{.01}{2} = .005$ (in each tail)
- $t_{.005}(199) = 2.617^*$

*Approximate using closest df
(always defer to lower df value)

Example: Practice Problem 13.4 (pp. 345)

- Manufacturer samples from 200 light bulbs and records how many hours it takes for them to burn out. The sample has a mean life of 215hrs ($s = 8$ hrs). Construct the 99% CI using the closest value for df

$$\bar{X}_{\text{obt}} = 215$$

$$s = 8$$

$$N = 200$$

$$s_{\bar{X}} =$$

$$t_{.005} = 2.617$$

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.005})$$

$$s_{\bar{X}} = \frac{s}{\sqrt{N}}$$

Example: Practice Problem 13.4 (pp. 345)

- Manufacturer samples from 200 light bulbs and records how many hours it takes for them to burn out. The sample has a mean life of 215hrs ($s = 8$ hrs). Construct the 99% CI using the closest value for df

$$\bar{X}_{\text{obt}} = 215$$

$$s = 8$$

$$N = 200$$

$$s_{\bar{X}} = .5657$$

$$t_{.005} = 2.617$$

$$\mu_{\text{lower}} = \bar{X}_{\text{obt}} - (s_{\bar{X}} * t_{.005})$$

$$s_{\bar{X}} = \frac{s}{\sqrt{N}} = \frac{8}{\sqrt{200}} = 0.5657$$

$$s_{\bar{X}} * t_{.005} =$$

$$\text{Lower} = 215 - 1.48 = 213.52$$

$$\text{Upper} = 215 + 1.48 = 216.48$$

Example: Practice Problem 13.4 (pp. 345)

- Reporting with confidence intervals:

“Our lightbulbs had a relatively long life, $M = 215$ hours ($s = 8$), $CI_{99\%} [213.52, 216.48]$.”

Inference:

“Our lightbulbs last at least 213 hours on average, (with greater than 99% confidence) $CI_{99\%} [213.52, 216.48]$.”