



Lecture 6
Sampling Distributions and
Confidence Intervals

Topics

- The Central Limit Theorem.
- Large sample confidence intervals for μ .
- Small sample confidence intervals for μ .

Section 6.5

Distribution of the Sample Mean: The Central Limit Theorem

Definitions

- A *point estimate* is a single number calculated from sample data. It is used to estimate a parameter of the population.
- A *point estimator* is the formula or rule that is used to calculate the point estimate for a particular set of data.

Definitions (cont.)

- The probability distribution of a point estimator or sample statistic is called a *sampling distribution*.
- The *standard error* is the standard deviation of the sampling distribution of a point estimator.

Goals

- To introduce point estimators.
- To introduce sampling distributions for point estimators.
- To use the Central Limit Theorem to compute sampling distributions of the sample mean for large samples.

Example – Home Sales

- The population of interest consists of all houses that are for sale in this area.
- The variable of interest is the number of days the house sat on the market until sold.
- The parameter of interest is:
 μ = mean number of days on the market

Example – Home Sales

- Based on a sample of five homes that sold recently consider the following point estimator for μ :

$$\bar{X} = \frac{1}{5} \sum_{i=1}^5 x_i$$

- Suppose that a sample of five homes produced the following data: 76, 48, 42, 9 and 27 days.

Example – Home Sales

- Then the point estimate for the mean number of days on the market would be:

$$\bar{x} = \frac{1}{5}(76 + 48 + 42 + 9 + 27) = 40.4$$

- Suppose a second sample of five homes produced the following data: 140, 10, 75, 24 and 4 days.

Example – Home Sales

- Then the point estimate for the mean number of days on the market would be:

$$\bar{x} = \frac{1}{5}(140 + 10 + 75 + 24 + 4) = 50.6$$

- Same point estimator but different point estimates.
- So the point estimator, \bar{X} , is a random variable with a probability distribution, a mean and a variance.

Central Limit Theorem (CLT)

- In random sampling from a population with mean μ and standard deviation σ , when n is large enough (≥ 30), the distribution of \bar{X} is:
 - approximately normal
 - $\mu_{\bar{X}}$ = mean of the sampling distribution of $\bar{X} = \mu$
 - $\sigma_{\bar{X}}$ = standard error of $\bar{X} = \frac{\sigma}{\sqrt{n}}$

Special Exception

- If the population we are sampling from has a normal distribution (or at is least approximately normal) then the CLT applies for all sample sizes.

Section 6.6

The Central Limit Theorem – A More Detailed Look

Goal

- A better appreciation of the CLT.

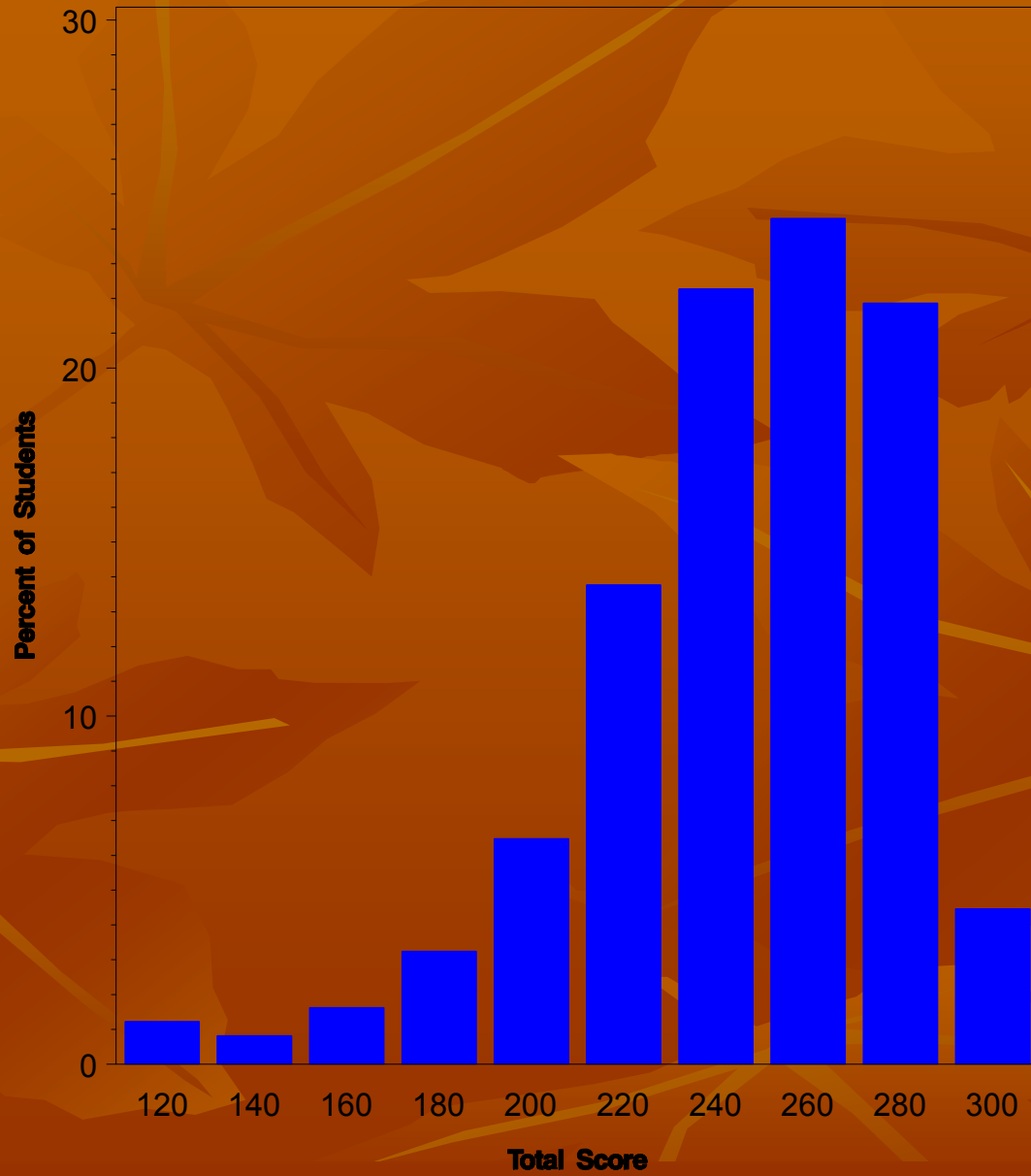
Shape of the Sampling Distribution of \bar{X}

- No matter what the shape of the distribution of the population, if $n \geq 30$, the shape of the distribution of \bar{X} for repeated sampling is normal.

Example – Total Scores

- Suppose the population of interest is the 247 total scores to date (Exams 1-2, Quizzes 1-4 and Minute Papers 1-3).

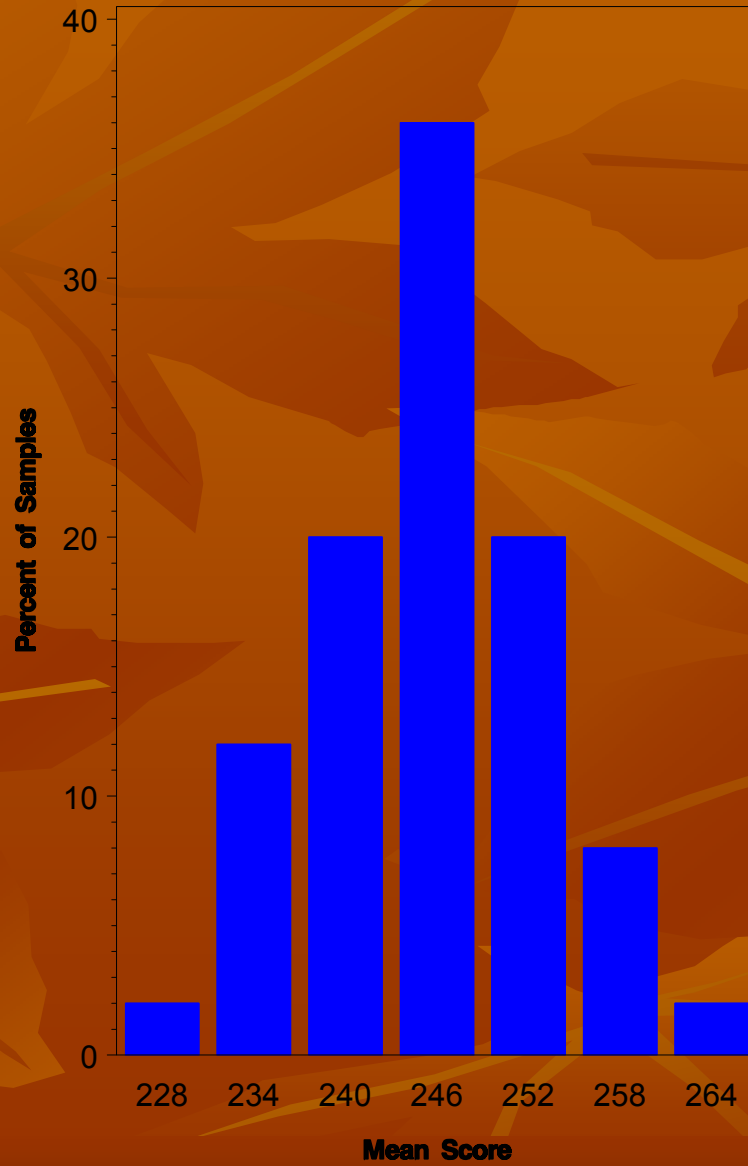
Total Score



Example – Total Scores

- Clearly the distribution of total scores is not normally distributed.
- Suppose you take a sample of size $n = 30$ scores and compute the average total score.
- Repeat this process for a total of 50 times.
- This would yield 50 values for \bar{X} .

Mean Score - n = 30



Mean of Sampling Distribution \bar{X} of

- If $n \geq 30$, then the mean of the sampling distribution of \bar{X} is the mean of the population.

Example – Total Scores

- The mean total score for the 247 students in the population was

$$\mu = 245.7$$

- The mean of the sample mean total scores for the 50 samples of size $n = 30$ was

$$\mu_{\bar{x}} = 245.6$$

Standard Error of the Sampling Distribution of \bar{X}

- If $n \geq 30$, the standard error of \bar{X} is the standard deviation of the population divided by the square root of n .

Example – Total Scores

- The standard deviation of the total scores for the 247 students in the population was

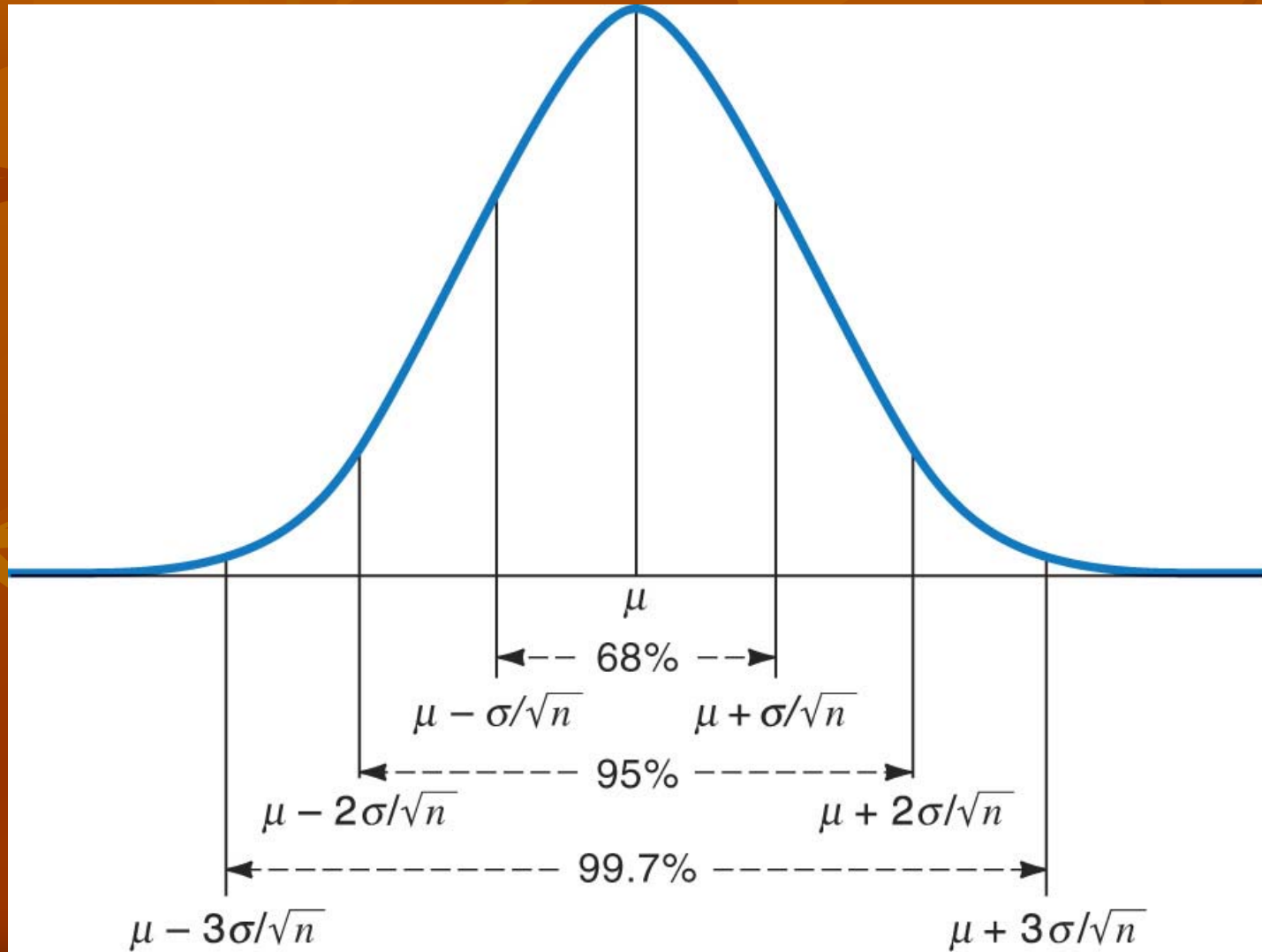
$$\sigma = 34.8$$

- The standard deviation of the sample total mean scores for the 50 samples was

$$\sigma_{\bar{x}} = 7.4$$

- $\frac{34.8}{\sqrt{30}} = 6.4 \cong 7.4$

Summary of CLT



Special Bonus

- As n increases, the standard error of \bar{X} decreases.
- As n increases, \bar{X} becomes a better (more precise) point estimator of μ .
- More is better.

Section 6.7

Drawing Inferences by Using the Central Limit Theorem

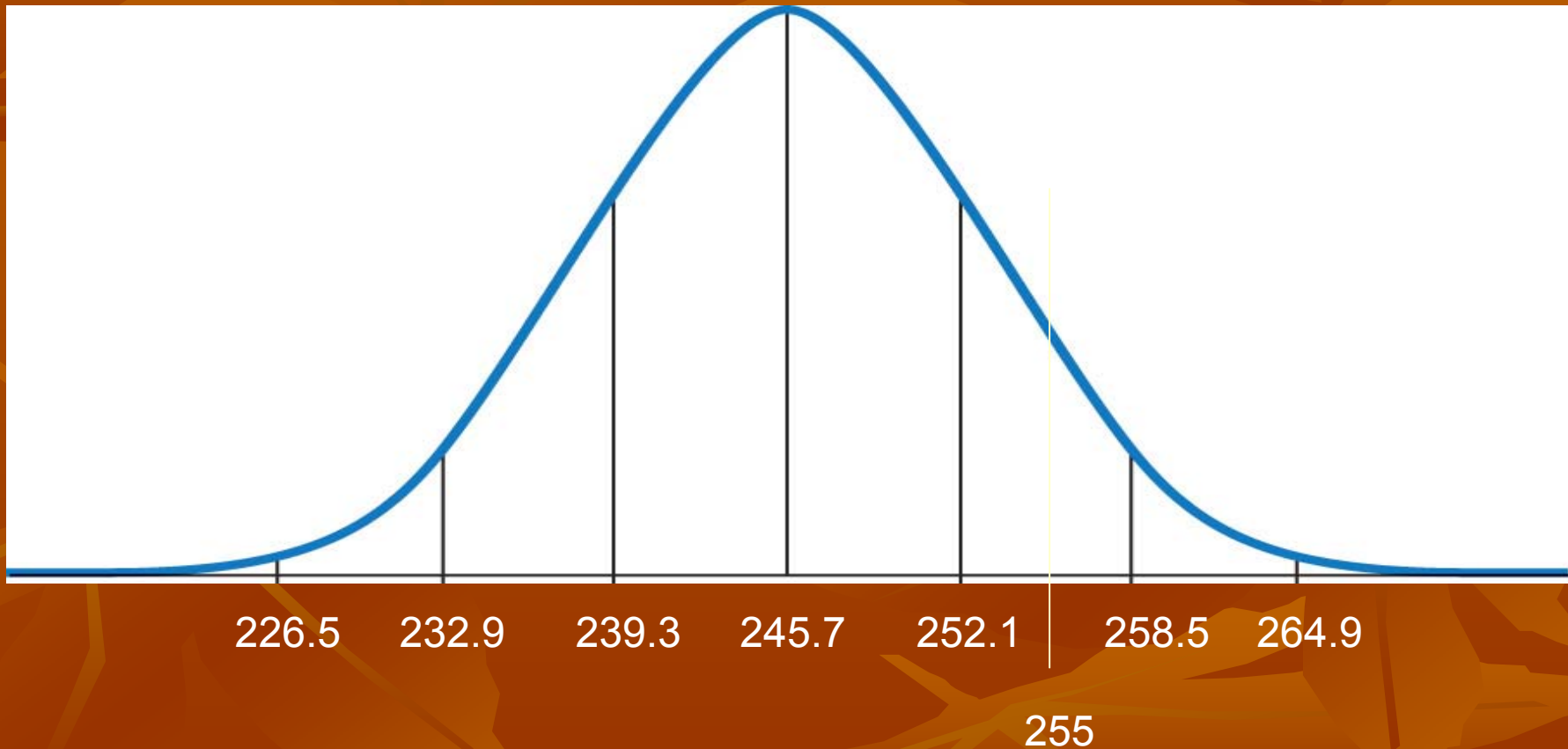
Goal

- To put the CLT to good use.

Example – Total Scores

- Suppose that you and 29 other students in the this class are members of the Nihilists for a Better America (NBA). You believe you are pretty hot and as a group your average total score (255 points) should be well above the class average.
- Assume $\bar{X} \sim N(245.7, 6.4)$

Example – Total Scores



Example – Total Scores

- $$Z = \frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}} = \frac{255 - 245.7}{6.4} = 1.45$$

- $$\begin{aligned} P(\bar{X} > 255) &= P(Z > 1.45) \\ &= 1 - P(Z \leq 1.45) \\ &= 1 - 0.9265 \\ &= 0.0735 \end{aligned}$$

- So you're hot, but not white hot.

σ Unknown

- What if μ is known but σ is unknown?
- Use s in place of σ .

- Replace $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$ with $s_{\bar{x}} = \frac{s}{\sqrt{n}}$.

Example – Car Battery Warranty

- A automobile battery manufacturer guarantees them for 36 months.
- The warranty states that if the battery fails before 36 months it will be replaced with a new one whose cost will be discounted by the fraction of time left on the old one.
- You, as a member of Skeptics for a Better America (SBA), believe it's a scam to keep you hooked on their batteries, i.e., you believe $\mu < 36$ months.

Example – Car Battery Warranty

- Suppose that the life lengths of 9 of these batteries were 35.38, 35.87, 35.49, 34.38, 34.55, 39.99, 36.18, 35.09 and 34.57 months.
- $\bar{x} = 35.72$ months
- $s = 1.71$ months
- Assume that the population of all life lengths is normally distributed.

Example – Car Battery Warranty

- Assume that $\mu = 36$ months.
- If $P(\bar{X} < 35.72)$ is small, assuming that $\mu = 36$, then the sample would suggest $\mu < 36$.

$$\blacksquare Z = \frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}} = \frac{\bar{X} - \mu_{\bar{X}}}{s_{\bar{X}}} = \frac{35.72 - 36}{1.71/\sqrt{9}} = -0.49$$

$$\blacksquare P(\bar{X} < 35.72) \cong P(Z < -0.49) = 0.3121$$

Example – Car Battery Warranty

- The sample does not suggest that $\mu < 36$ months.

Section 6.8

Large-Sample Confidence Intervals for the Mean

Definitions

- A *confidence interval* or an *interval estimate* is a range of values with an associated probability or confidence level $1 - \alpha$.
- The *confidence level* quantifies the chance that, in repeated sampling, the interval contains the true population parameter.

Goal

- To construct a range of values that will contain μ with high probability.

Bits & Pieces of a Confidence Interval

- interval = L to U
 - L = lower bound
 - U = upper bound
- $1 - \alpha$ = confidence level
- $P(L \leq \mu \leq U) = 1 - \alpha$
 - Read this as, “The probability that the interval contains μ is $1 - \alpha$.”

CI Construction: σ known

- Start with a good point estimator of μ : \bar{X}
- If $n \geq 30$ then $\bar{X} \sim N\left(\mu, \sigma/\sqrt{n}\right)$
- $P\left(\mu - 2\sigma/\sqrt{n} \leq \bar{X} \leq \mu + 2\sigma/\sqrt{n}\right) \cong 0.95$
- $P\left(\bar{X} - 2\sigma/\sqrt{n} \leq \mu \leq \bar{X} + 2\sigma/\sqrt{n}\right) \cong 0.95$

CI Construction: σ known

- Therefore, a confidence interval for μ with confidence level 0.95 (if $n \geq 30$ and σ known) is given by:
 - $L = \bar{X} - 2\sigma/\sqrt{n}$
 - $U = \bar{X} + 2\sigma/\sqrt{n}$
- This is also referred to as a 95% confidence interval for μ .

New & Improved 95% CI

- $$\frac{(\mu - 2\sigma/\sqrt{n}) - \mu}{\sigma/\sqrt{n}} = -2$$

- $$\frac{(\mu + 2\sigma/\sqrt{n}) - \mu}{\sigma/\sqrt{n}} = 2$$

- $$P\left(\mu - 2\sigma/\sqrt{n} \leq \bar{X} \leq \mu + 2\sigma/\sqrt{n}\right)$$
$$= P(-2 \leq Z \leq 2) = 0.9772 - 0.0228 = 0.9544$$

New & Improved 95% CI

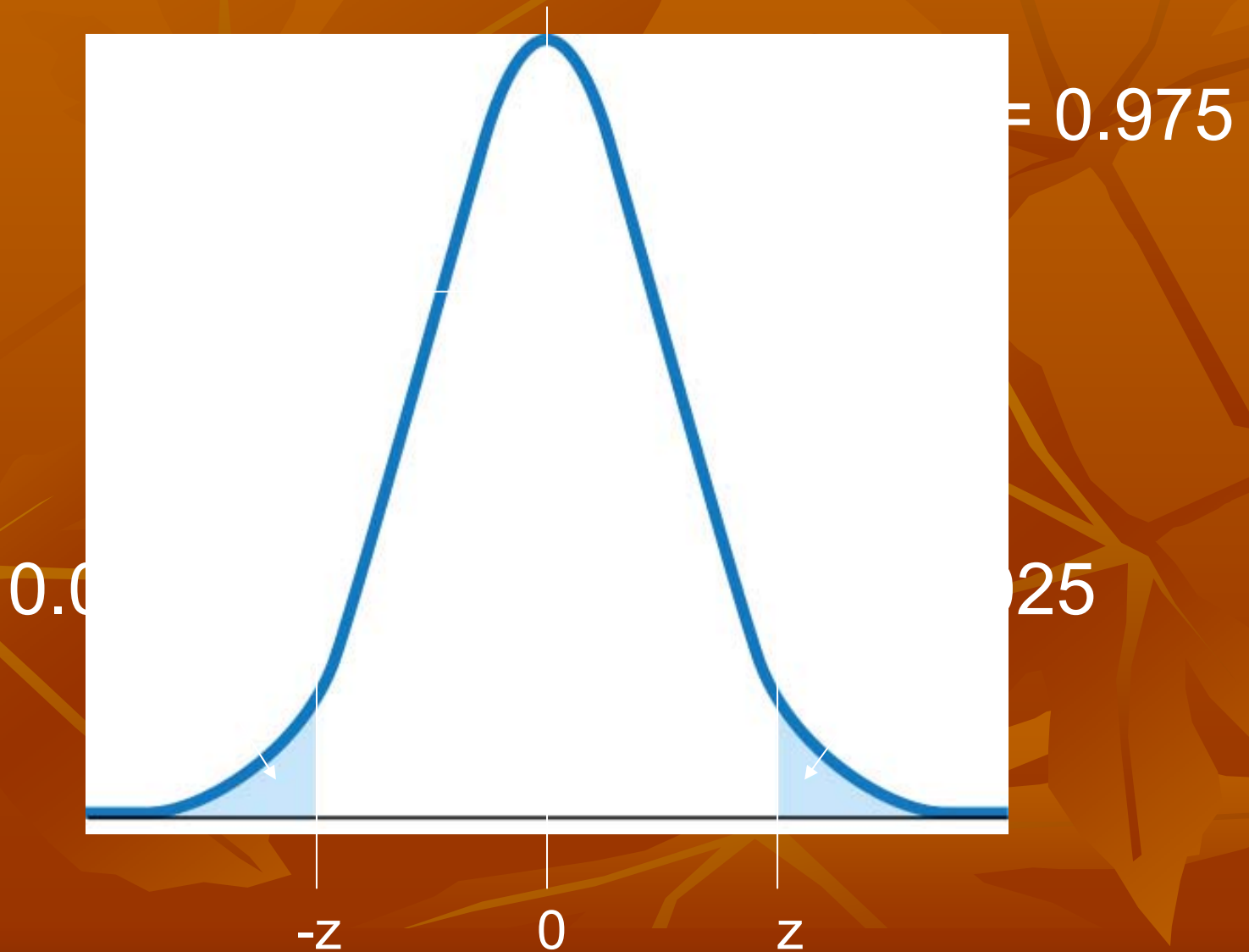
- Need to find z such that

$$P\left(\mu - z\sigma/\sqrt{n} \leq \bar{X} \leq \mu + z\sigma/\sqrt{n}\right)$$

$$= P(-z \leq Z \leq z)$$

$$= 0.95$$

New & Improved 95% CI



New & Improved 95% CI

- Since $P(Z < 1.96) = 0.975$ then $z = 1.96$
- Therefore, a confidence interval for μ with confidence level 0.95 is given by:
 - $L = \bar{X} - 1.96\sigma/\sqrt{n}$
 - $U = \bar{X} + 1.96\sigma/\sqrt{n}$

Notation

- For a 95% CI (confidence level = 0.95) we needed $z_{.025} = 1.96$
- For a $(1 - \alpha)100\%$ CI (confidence level = $1 - \alpha$) we require $z_{\alpha/2}$

Common values of α and $z_{\alpha/2}$

Confidence level ($1 - \alpha$)

	0.90	0.95	0.98	0.99	0.995
α	0.10	0.05	0.02	0.01	0.005
$z_{\alpha/2}$	1.645	1.96	2.33	2.58	2.81

Example – Total Scores

- Suppose you wish to estimate the mean total score for the 247 students in this class with 95% confidence.
- Assume that you know only that $\sigma = 34.8$ points.
- Suppose that from a sample of $n = 30$ students you obtained $\bar{X} = 250.7$

Example – Total Scores

- Confidence level = 0.95
- $\alpha = 0.05$
- $z_{\alpha/2} = z_{.025} = 1.96$
- $1.96 \frac{\sigma}{\sqrt{n}} = 1.96 \frac{34.8}{\sqrt{30}} = 12.5$
- $L = 250.7 - 12.5 = 238.2$
- $U = 250.7 + 12.5 = 263.2$

Example – Total Scores

- We are 95% confident that the mean total score for all 247 students is between 238.2 and 263.2 points.

CI Construction: σ Unknown

- If σ is unknown and $n \geq 30$ then replace σ by s in the CI calculations.
- If σ is unknown and $n < 30$ then . . .

$$n \geq 30$$

 σ

Population

known

unknown

normal

$$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

$$\bar{X} \pm z_{\alpha/2} \frac{s}{\sqrt{n}}$$

nonnormal

Same as above

Same as above or
Chapter 16

$n < 30$

σ

Population

known

unknown

normal

$$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

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nonnormal

Chapter 16

Chapter 16

“I am 95% Confident . . .” ???

- In the previous example we said we were 95% confident that the mean total score for all 247 students (μ) was between 238.2 to 263.2 points.
- Does this mean that there is a 95% chance that μ is between 238.2 and 263.2 points?
- No! Negative!

“I am 95% Confident . . .” ???

- In the previous example we knew that $\mu = 245.7$ points.
- 245.7 is between 238.2 and 263.2
- So our 95% confidence interval contains μ with probability 1.
- In general we do not know μ .
- Hence, in general, we do not know whether or not the particular interval we've calculated contains μ .

“I am 95% Confident . . .” ???

- What we do know is that in repeated sampling, 95% of the intervals would contain μ .

Twenty 95% CIs for Mean Total Score

\bar{X}	L	U	Covers?	\bar{X}	L	U	Covers?
238.2	225.7	250.6	Yes	250.7	238.2	263.1	Yes
263.3	250.8	275.7	No	234.4	221.9	246.8	Yes
247.8	235.4	260.3	Yes	253.0	240.6	265.5	Yes
246.3	233.8	258.7	Yes	248.9	236.4	261.3	Yes
246.9	234.4	259.3	Yes	233.8	221.3	246.2	Yes
241.7	229.2	254.2	Yes	256.7	244.2	269.1	Yes
248.5	236.1	261.0	Yes	252.8	240.4	265.3	Yes
237.8	225.3	250.2	Yes	257.4	244.9	269.8	Yes
241.3	228.8	253.7	Yes	246.6	234.1	259.0	Yes
252.0	239.5	264.5	Yes	249.4	236.9	261.8	Yes

Twenty 95% CIs for Mean Total Score

- Of the 20 CIs for μ , 19 ($19/20 = 0.95$) contained 245.7, the true mean total score.

Section 6.9

Distribution of the Sample Mean: Small Sample and Unknown σ

Goal

- To become familiar with Student's t distribution.

Student's t Distribution

- By the CLT and Slutsky's Theorem, if $n \geq 30$ and σ unknown then:

- $$Z = \frac{\bar{X} - \mu}{s/\sqrt{n}} \sim N(0,1)$$

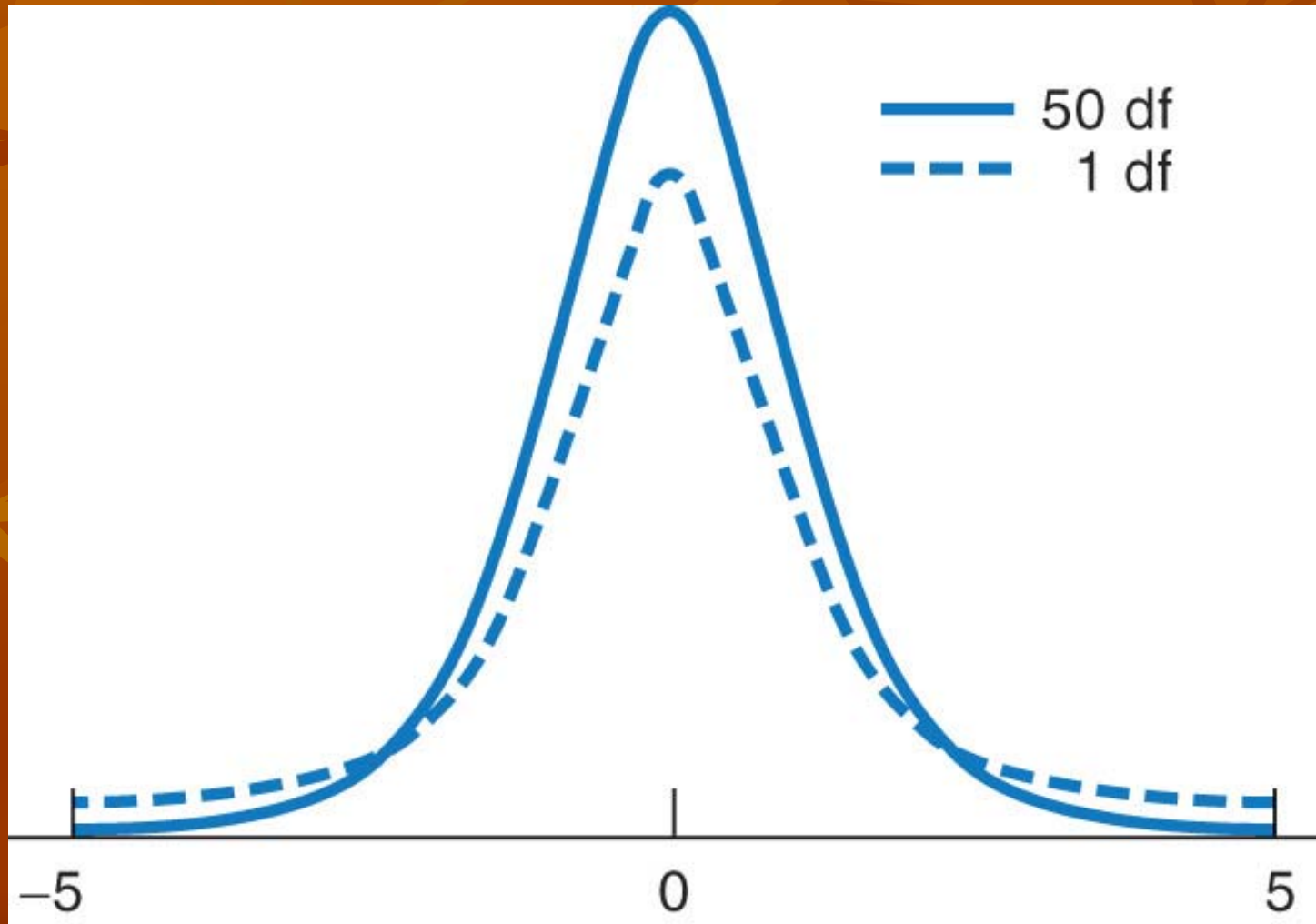
- If $n < 30$, σ unknown and the population is normally distributed then:

- $$t = \frac{\bar{X} - \mu}{s/\sqrt{n}} \sim t_{n-1}$$

Student's t Distribution

- t_{n-1} is Student's t distribution with $(n - 1)$ degrees of freedom (df).
- Due to William S. Gosset of the Guinness Breweries in Ireland.
- Published under the pseudonym *I. Student*.
- Similar to standard normal distribution but more variable (i.e., fatter tails).
- Variability decreases with increasing df .

t distributions with $df = 1$ and 50



Section 6.10

Small-Sample Confidence Intervals for the Mean

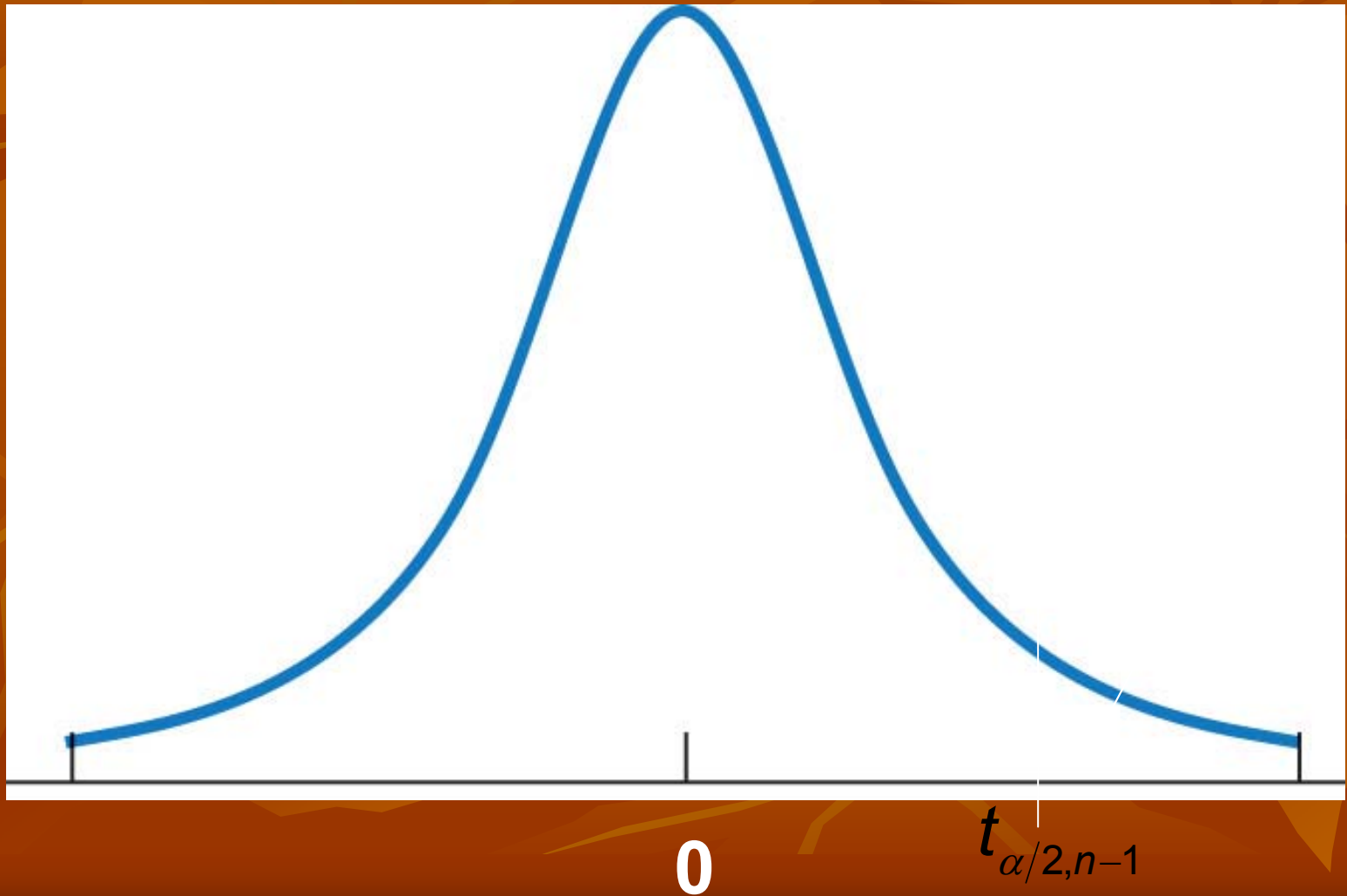
Goal

- To construct a range of values that will contain μ with high probability for $n < 30$.

$(1 - \alpha)100\%$ CI for μ : $n < 30$ & σ Unknown

- Assume the population we are sampling from is normally distributed.
- $L = \bar{X} - t_{\alpha/2, n-1} s / \sqrt{n}$
- $U = \bar{X} + t_{\alpha/2, n-1} s / \sqrt{n}$
- $t_{\alpha/2, n-1} = (1 - \alpha/2)100^{\text{th}}$ percentile of the t distribution with $(n - 1)$ df .

Percentiles of the t Distribution



Example – Gas Mileage

- You are interested in estimating the average mpg for a Honda Civic with 95% confidence.
- You have available to you 25 measurements on the mpg from 25 fillups.
- The mean of the 25 measurements was 28.99 mpg with a standard deviation of 3.22 mpg.
- Assume the population of mpg is normally distributed.

Example – Gas Mileage

- $\alpha = 0.05$
- $(n - 1) = 25 - 1 = 24$
- $t_{\alpha/2, n-1} = t_{0.025, 24} = 2.064$

t CRITICAL VALUES

<i>df</i>	0.15	0.10	0.05	0.025	0.015	0.01	0.005	0.001	0.0005
4	1.190	1.533	2.132	2.776	3.298	3.747	4.604	7.173	8.610
5	1.156	1.476	2.015	2.571	3.003	3.365	4.032	5.893	6.869
6	1.134	1.440	1.943	2.447	2.829	3.143	3.707	5.208	5.959
7	1.119	1.415	1.895	2.365	2.715	2.998	3.499	4.785	5.408
8	1.108	1.397	1.860	2.306	2.634	2.896	3.355	4.501	5.041
9	1.100	1.383	1.833	2.262	2.574	2.821	3.250	4.297	4.781
10	1.093	1.372	1.812	2.228	2.527	2.764	3.169	4.144	4.587
11	1.088	1.363	1.796	2.201	2.491	2.718	3.106	4.025	4.437
12	1.083	1.356	1.782	2.179	2.461	2.681	3.055	3.930	4.318
13	1.079	1.350	1.771	2.160	2.436	2.650	3.012	3.852	4.221
14	1.076	1.345	1.761	2.145	2.415	2.624	2.977	3.787	4.140
15	1.074	1.341	1.753	2.131	2.397	2.602	2.947	3.733	4.073
16	1.071	1.337	1.746	2.120	2.382	2.583	2.921	3.686	4.015
17	1.069	1.333	1.740	2.110	2.368	2.567	2.898	3.646	3.965
18	1.067	1.330	1.734	2.101	2.356	2.552	2.878	3.610	3.922
19	1.066	1.328	1.729	2.093	2.346	2.539	2.861	3.579	3.883
20	1.064	1.325	1.725	2.086	2.336	2.528	2.845	3.552	3.850
21	1.063	1.323	1.721	2.080	2.328	2.518	2.831	3.527	3.819
22	1.061	1.321	1.717	2.074	2.320	2.508	2.819	3.505	3.792
23	1.060	1.319	1.714	2.069	2.313	2.500	2.807	3.485	3.768
24	1.059	1.318	1.711	2.064	2.307	2.492	2.797	3.467	3.745
25	1.058	1.316	1.708	2.060	2.301	2.485	2.787	3.450	3.725
26	1.058	1.315	1.706	2.056	2.296	2.479	2.779	3.435	3.707
27	1.057	1.314	1.703	2.052	2.291	2.473	2.771	3.421	3.690
28	1.056	1.313	1.701	2.048	2.286	2.467	2.763	3.408	3.674
29	1.055	1.311	1.699	2.045	2.282	2.462	2.756	3.396	3.659
30	1.055	1.310	1.697	2.042	2.278	2.457	2.750	3.385	3.646
40	1.050	1.303	1.684	2.021	2.250	2.423	2.704	3.307	3.551
50	1.047	1.299	1.676	2.009	2.234	2.403	2.678	3.261	3.496
60	1.045	1.296	1.671	2.000	2.223	2.390	2.660	3.232	3.460
120	1.041	1.289	1.658	1.980	2.196	2.358	2.617	3.160	3.373

Example – Gas Mileage

- $t_{\alpha/2, n-1} \frac{s}{\sqrt{n}} = 2.064 \frac{3.22}{\sqrt{25}} = 1.33$

- $L = 28.99 - 1.33 = 27.66$

- $U = 28.99 + 1.33 = 30.32$

Example – Gas Mileage

- With 95% confidence, the true mean mpg for a Honda Civic is between 27.66 and 30.32 mpg.