

Lecture 26: Chapter 10, Section 2 Inference for Quantitative Variable Confidence Interval with t

- t Confidence Interval for Population Mean
- Comparing z and t Confidence Intervals
- When neither z nor t Applies
- Other Levels of Confidence
- t Test vs. Confidence Interval

Looking Back: Review

- 4 Stages of Statistics**
 - Data Production (discussed in Lectures 1-4)
 - Displaying and Summarizing (Lectures 5-12)
 - Probability (discussed in Lectures 13-20)
 - Statistical Inference
 - 1 categorical (discussed in Lectures 21-23)
 - 1 quantitative: z CI, z test, t CI, t test
 - categorical and quantitative
 - 2 categorical
 - 2 quantitative

Behavior of Sample Mean (Review)

For random sample of size n from population with mean μ , standard deviation σ , sample mean \bar{X} has

- mean μ
- standard deviation $\frac{\sigma}{\sqrt{n}}$
- shape approximately normal for large enough n

Sample Mean Standardizing to z (Review)

→ If σ is **known**, standardized \bar{X} follows z (standard normal) distribution:

$$z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$$

If σ is **unknown** but n is large enough (20 or 30), then $s \approx \sigma$ and

$$\frac{\bar{x} - \mu}{s / \sqrt{n}} \approx z$$

Sample mean standardizing to t (Review)

For σ unknown and n small, $\frac{\bar{x} - \mu}{s/\sqrt{n}} = t$

- t (like z) centered at 0 since \bar{X} centered at μ
 - t (like z) symmetric and bell-shaped if \bar{X} normal
 - t more spread than z (s.d. > 1) [s gives less info]
- t has “ $n-1$ degrees of freedom” (spread depends on n)

Inference by Hand or with Software: z or t ?

	σ known	σ unknown
small sample ($n < 30$)	$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = z$	$\frac{\bar{x} - \mu}{s/\sqrt{n}} = t$
large sample ($n \geq 30$)	$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = z$	$\frac{\bar{x} - \mu}{s/\sqrt{n}} \approx z$

By Hand:

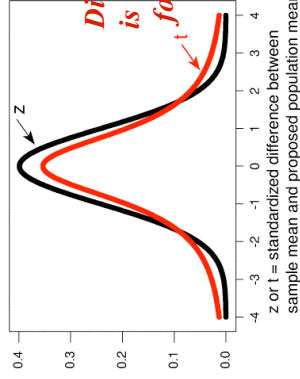
- z used if σ is known **or** n is large
- t used if σ is unknown **and** n is small

With Software:

- z used if σ is known
- t used if σ is unknown

Inference Based on z or t

	σ known	σ unknown
small sample ($n < 30$)	$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = z$	$\frac{\bar{x} - \mu}{s/\sqrt{n}} = t$
large sample ($n \geq 30$)	$\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = z$	$\frac{\bar{x} - \mu}{s/\sqrt{n}} \approx z$



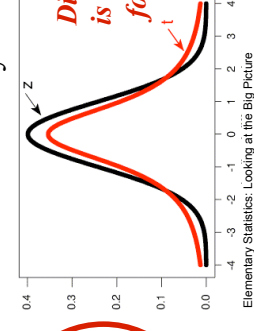
Confidence Interval for Mean (Review)

95% confidence interval for μ (σ known) is

$$\bar{x} \pm 2 \frac{\sigma}{\sqrt{n}}$$

- multiplier 2 is from z distribution (95% of normal values within 2 s.d.s of mean)
- For n small, σ unknown can't say 95% C.I. is

$$\bar{x} \pm 2 \frac{s}{\sqrt{n}}$$



Confidence Interval for Mean: σ Unknown

95% confidence interval for μ is

$$\bar{x} \pm \text{multiplier} \left(\frac{s}{\sqrt{n}} \right)$$

- multiplier from t distribution with $n-1$ *degrees of freedom* (df)
- multiplier at least 2, closer to 3 for very small n

Degrees of Freedom

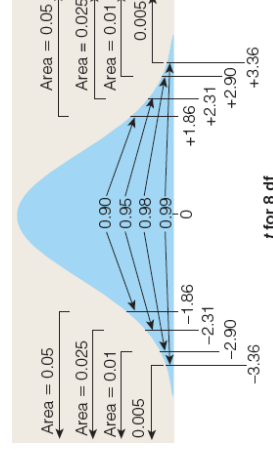
- **Mathematical** explanation of df: not needed for elementary statistics
- **Practical** explanation of df: several useful distributions like t , F , chi-square are *families* of similar curves; df tells us which one applies (depends on sample size n).

z or t : Which to Concentrate On?

- For purpose of **learning**, start with z (know what to expect from 68-95-99.7 Rule, etc.) (**only one z distribution**)
 - For **practical** purposes, t **more realistic** (usually don't know population s.d. σ)
- Software** automatically uses appropriate t distribution with $n-1$ df: just enter data.

Example: Confidence Interval with t Curve

- **Background:** Random sample of shoe sizes for 9 college males: 11.5, 12.0, 11.0, 15.0, 11.5, 10.0, 9.0, 10.0, 11.0
- **Question:** What is 95% C.I. for population mean?
- **Response:** Mean 11.222, $s=1.698$, $n=9$, multiplier 2.31:



t for 8 df

Example: *t* Confidence Interval with Software

- **Background:** Random sample of shoe sizes for 9 college males: 11.5, 12.0, 11.0, 15.0, 11.5, 10.0, 9.0, 10.0, 11.0
- **Question:** How do we find a 95% C.I. for the population mean, using software?
- **Response:**

One-Sample T: Shoe					
Variable	N	Mean	StDev	SE Mean	95.0% CI
Shoe	9	11.222	1.698	0.566	(9.917, 12.527)

Example: Compare *t* and *z* Confidence Intervals

- **Background:** Random sample of shoe sizes for 9 college males: 11.5, 12.0, 11.0, 15.0, 11.5, 10.0, 9.0, 10.0, 11.0
We produced 95% *t* confidence interval:
 $11.222 \pm 2.31 \left(\frac{1.698}{\sqrt{9}} \right) = 11.222 \pm 1.307 = (9.92, 12.53)$
If 1.698 had been population s.d., would get *z* C.I.:
 $11.222 \pm 1.96 \left(\frac{1.698}{\sqrt{9}} \right) = 11.222 \pm 1.109 = (10.11, 12.33)$
- **Question:** How do the *t* and *z* intervals differ?
- **Response:** *t* multiplier is 2.31, *z* multiplier is 1.96:
t interval width about _____
z interval width about _____

σ known \rightarrow _____ info \rightarrow _____ interval

Example: *t* vs. *z* Confidence Intervals, Large *n*

- **Background:** Earnings for sample of 446 students at a university averaged \$3,776, with s.d. \$6,500.
The *t* multiplier for 95% confidence and 445 df is 1.9653.
- **Question:** How different are the *t* and *z* intervals?
- **Response:** The intervals will be _____, whether we use
 - *t* multiplier 1.9653
 - precise *z* multiplier 1.96
 - approximate *z* multiplier 2Interval approximately

Behavior of Sample Mean (Review)

For random sample of size *n* from population with mean μ , standard deviation σ , sample mean \bar{X} has

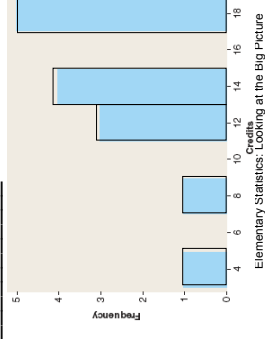
- mean μ
 - standard deviation $\frac{\sigma}{\sqrt{n}}$
 - shape approx. normal for large enough *n*
- \rightarrow If σ is unknown and *n* small, $\frac{\bar{x} - \mu}{s/\sqrt{n}} = t$

Guidelines for \bar{X} Approx. Normal (Review)

- Can assume shape of \bar{X} for random samples of size n is approximately normal if
 - Graph of sample data appears normal; or
 - Sample data fairly symmetric, n at least 15; or
 - Sample data moderately skewed, n at least 30; or
 - Sample data very skewed, n much larger than 30
- If \bar{X} is not normal, $\frac{\bar{x} - \mu}{s/\sqrt{n}}$ is not t .

Example: Small, Skewed Data Set

- Background:** Credits taken by 14 non-traditional students: 4, 7, 11, 11, 12, 13, 13, 14, 14, 17, 17, 17, 17, 18
- Question:** What is a 95% confidence interval for population mean?
- Response:** n small, shape of credits left-skewed \rightarrow



Looking Ahead:
Non-parametric methods can be used for small n , skewed data.

t Intervals at Other Levels of Confidence

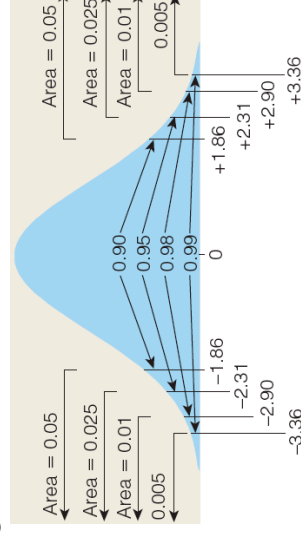
Confidence Level

	90%	95%	98%	99%
z (infinite n)	1.645	1.960 or 2	2.326	2.576
t: $df = 19$ ($n = 20$)	1.73	2.09	2.54	2.86
t: $df = 11$ ($n = 12$)	1.80	2.20	2.72	3.11
t: $df = 3$ ($n = 4$)	2.35	3.18	4.54	5.84

- Lower confidence \rightarrow smaller t multiplier
- Higher confidence \rightarrow larger t multiplier
- Table excerpt \rightarrow at any given level, $t > z$ mult \rightarrow using s not σ gives wider interval (less info)
- t multipliers decrease as df (and n) increase

Example: Intervals at Other Confidence Levels

- Background:** Random sample of shoe sizes for 9 college males: 11.5, 12.0, 11.0, 15.0, 11.5, 10.0, 9.0, 10.0, 11.0



- Question:** What is t multiplier for 99% confidence?

- Response:**

t for 8 df

Example: Intervals at Other Confidence Levels

- **Background:** Random sample of shoe sizes for 9 college males: 11.5, 12.0, 11.0, 15.0, 11.5, 10.0, 9.0, 10.0, 11.0. We can produce 95% confidence interval:

$$11.2 \pm 2.31 \frac{1.7}{\sqrt{9}} = (9.9, 12.5)$$

- **Question:** What would 99% confidence interval be, and how does it compare to 95% interval? (Use the fact that t multiplier for 8 df, 99% confidence is 3.36.)
- **Response:** 99% interval interval is

- Width _____ for 95%
- Width _____ for 99%

Summary of t Confidence Intervals

Confidence interval for μ is $\bar{x} \pm$ multiplier $\left(\frac{s}{\sqrt{n}}\right)$ where **multiplier** depends on

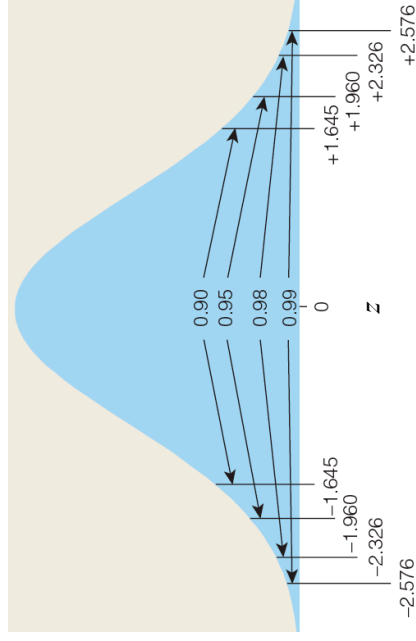
- **df:** smaller for larger n , larger for smaller n
 - **level:** smaller for lower level, larger for higher
- Note: margin of error is larger for larger s .

→ interval **narrower** for

- larger n (via df and \sqrt{n} in denominator)
- lower level of confidence
- smaller s.d. (distribution with less spread)

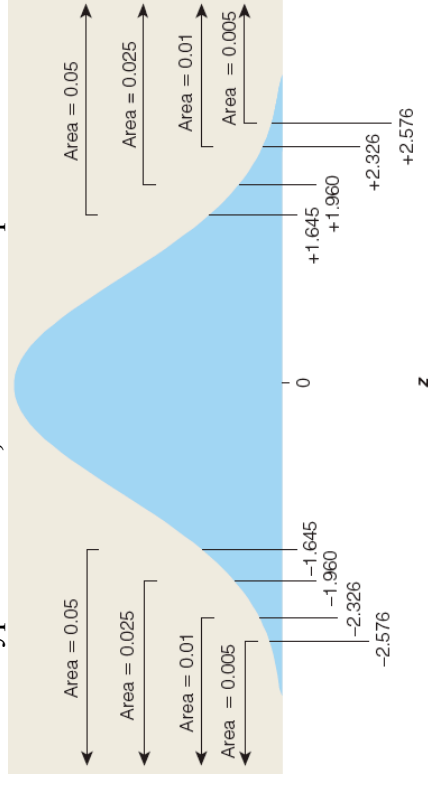
From z Confidence Intervals to Tests (Review)

For confidence intervals, used “inside” probabilities.



From z Confidence Intervals to Tests (Review)

For hypothesis tests, used “outside” probabilities.

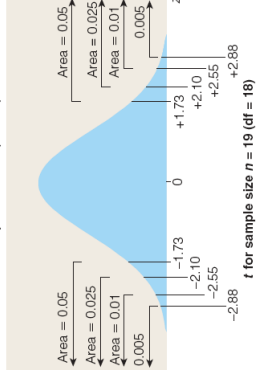


From t Confidence Intervals to Tests

Confidence interval: use multiplier for t dist, $n-1$ df
 Hypothesis test: P -value based on tail of t dist, $n-1$ df

Example: Hypothesis Test: t vs. z

- Background: Suppose one test with very large n has $z = 2$; another test with $n = 19$ (18 df) has $t = 2$.
 t tail probabilities (df = 18)

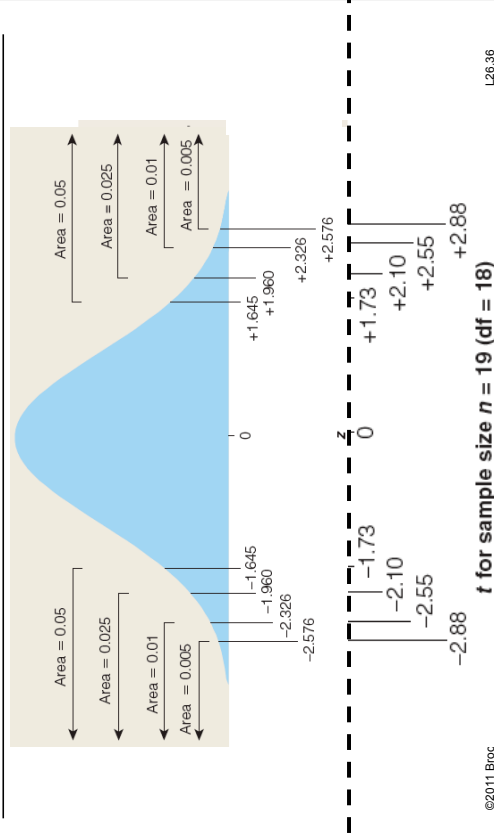


- Question: How do P -values compare for z and t ? (Assume alternative is “greater than”.)
- Response: 90-95-98-99 Rule $\rightarrow z$ P -value _____.
- Response: t curve for 18 df $\rightarrow t$ P -value _____.

From t Confidence Intervals to Tests

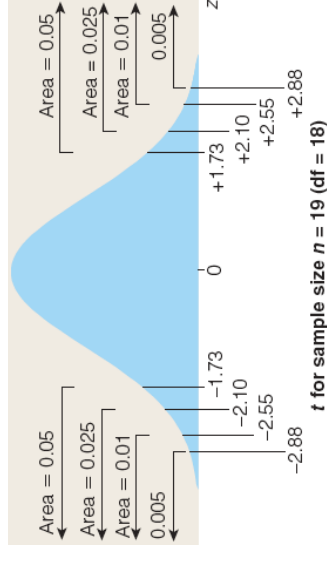
Confidence interval: use multiplier for t dist, $n-1$ df
 Hypothesis test: P -value based on tail of t dist, $n-1$ df

Comparing Critical Values, z with t for 18 df



Example: Hypothesis Test: t vs. z

- Background: Consider t curve for 18 df.



- Question: Would a value of $t = 3$ be considered extreme?
- Response: _____; $|t|$ for 18 df almost never exceeds _____.

Example: t Test (by Hand)

- Background:** Wts. of 19 female college students:
110 110 112 120 120 120 125 125 130 130 132 133 134 135 135 135 145 148 159
- Question:** Is pop. mean 141.7 reported by NCHS plausible, or is there evidence that we've sampled from pop. with lower mean (or that there is bias due to under-reporting)?
- Response:**
 1. Pop. $\geq 10(19)$; shape of weights close to normal $\rightarrow n=19$ OK
 2. $\bar{x} = 129.36, s = 12.82, t =$.
 3. P -value = _____ small because $|t|$ more extreme than 3 can be considered unusual for most n ; in particular, for 18 df, $P(t < -2.88)$ is less than 0.005.
 4. Reject H_0 ? _____ Conclude?

Lecture Summary (Inference for Means: t Confidence Intervals)

- t confidence interval for population mean
 - Multiplier from t distribution with $n-1$ df
 - When to perform inference with z or t
 - Constructing t CI by hand or with software
- Comparing z and t confidence intervals
- When neither z nor t applies
- Other levels of confidence
- from confidence interval to hypothesis test
- t test by hand