

# 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  $\mu$ , standard deviation  $\sigma$ , sample mean  $\bar{X}$  has

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

## Sample Mean Standardizing to $z$ (*Review*)

---

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

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

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

- $t$  (like  $z$ ) centered at 0 since  $\bar{X}$  centered at  $\mu$
  - $t$  (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 $t$ ?

	$\sigma$ known	$\sigma$ 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:

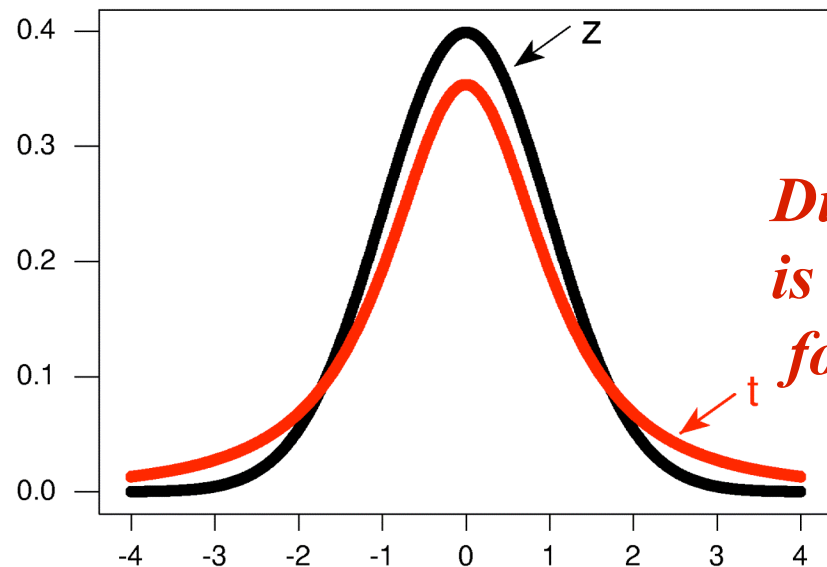
- $z$  used if  $\sigma$  is known **or**  $n$  is large
- $t$  used if  $\sigma$  is unknown **and**  $n$  is small

With Software:

- $z$  used if  $\sigma$  is known
- $t$  used if  $\sigma$  is unknown

# Inference Based on $z$ or $t$

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



*Distribution of  $t$  is “heavy tailed” for small  $n$ .*

$z$  or  $t$  = standardized difference between sample mean and proposed population mean

# Confidence Interval for Mean (*Review*)

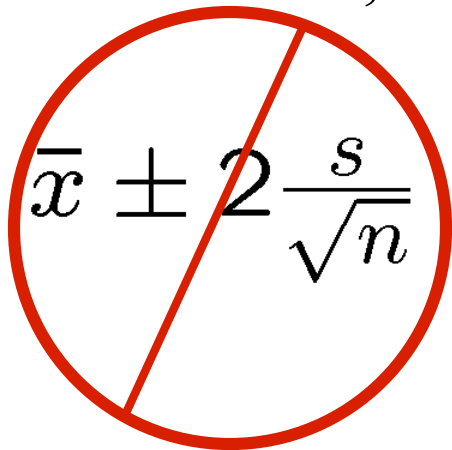
95% confidence interval for  $\mu$  ( $\sigma$  known) is

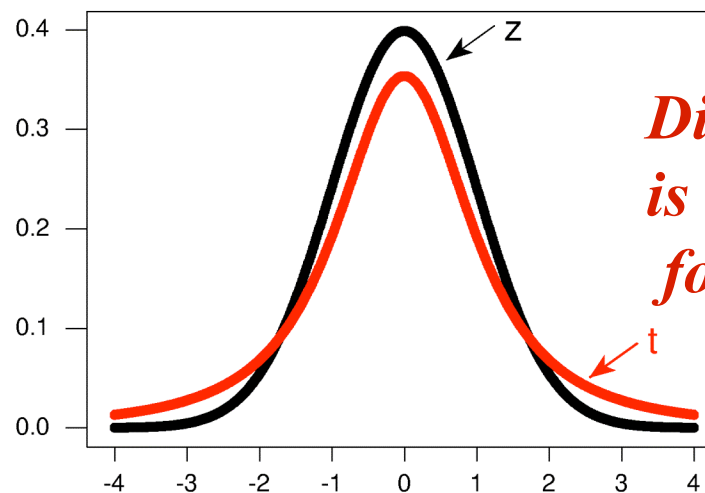
$$\bar{x} \pm 2 \frac{\sigma}{\sqrt{n}}$$

- multiplier 2 is from z distribution

(95% of normal values within 2 s.d.s of mean)

For  $n$  small,  $\sigma$  unknown can't say 95% C.I. is


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



*Distribution of  $t$  is "heavy tailed" for small  $n$ .*



## Confidence Interval for Mean: $\sigma$ Unknown

---

95% confidence interval for  $\mu$  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?

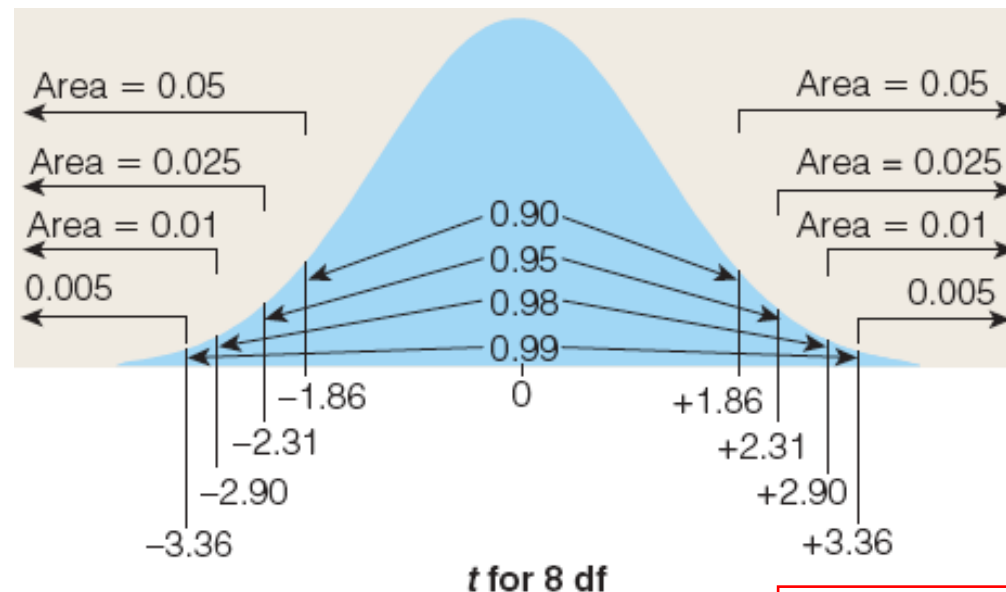
---

- 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.  $\sigma$ )

**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](#):





## 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 \_\_\_\_\_ 

$\sigma$  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  $\mu$ , standard deviation  $\sigma$ , sample mean  $\bar{X}$  has

- mean  $\mu$
- standard deviation  $\frac{\sigma}{\sqrt{n}}$
- **shape** approx. normal for large enough  $n$

→ If  $\sigma$  is unknown and  $n$  small,  $\frac{\bar{x} - \mu}{s/\sqrt{n}} = t$



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

---

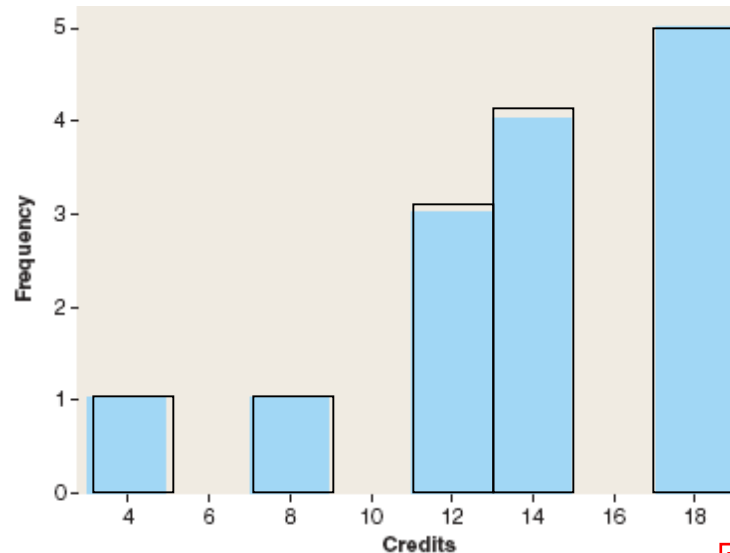
Can assume shape of  $\bar{X}$  for random samples of size  $n$  is approximately normal if

- 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  
→



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

## *t* Intervals at Other Levels of Confidence

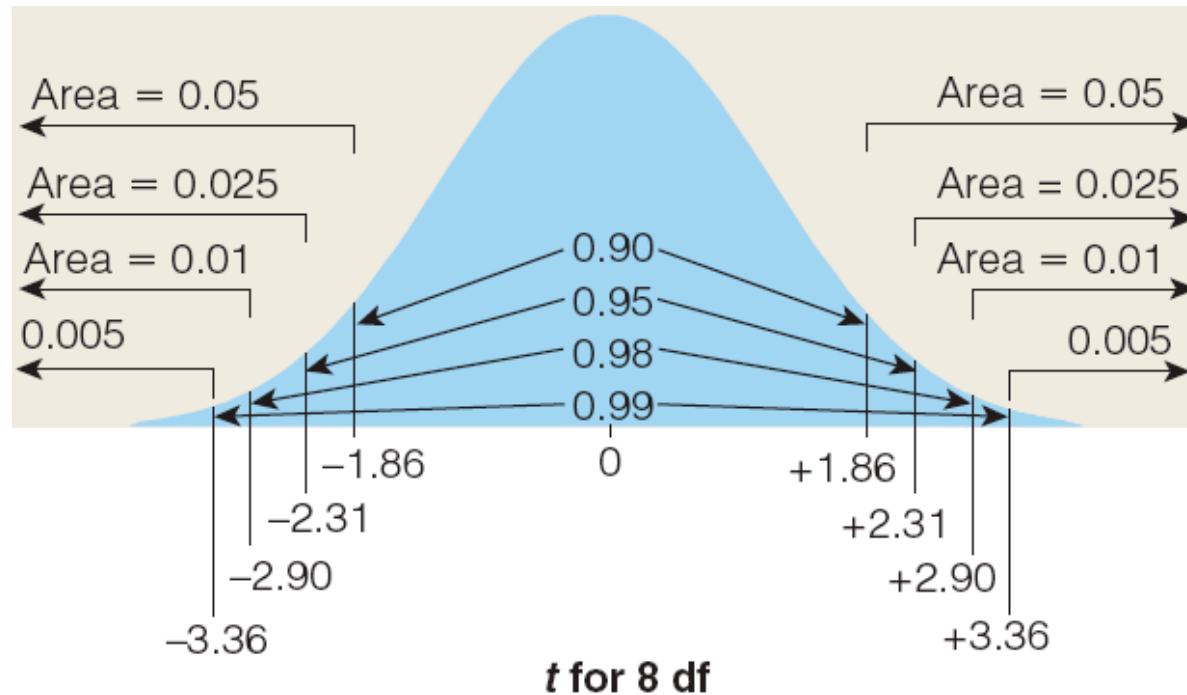
### Confidence Level

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

- Lower confidence → smaller *t* multiplier
- Higher confidence → larger *t* multiplier
- Table excerpt → at any given level,  $t > z$  mult → using *s* not  $\sigma$  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:**

## 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  $\mu$  is  $\bar{x} \pm \text{multiplier} \left( \frac{s}{\sqrt{n}} \right)$   
where **multiplier** depends on

- **df**: smaller for larger  $n$ , larger for smaller  $n$
- **level**: smaller for lower level, larger for higher

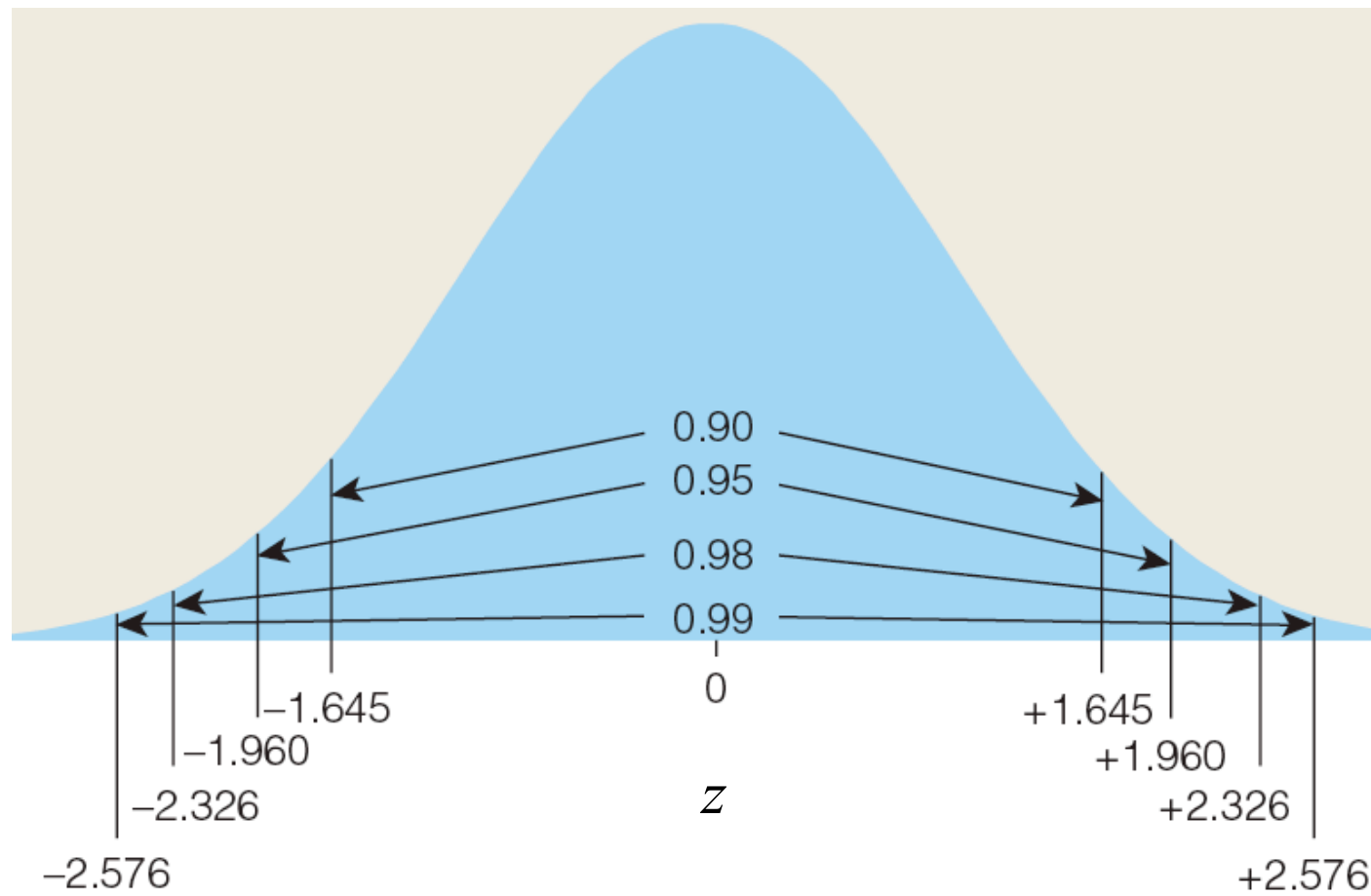
Note: margin of error is larger for larger  $s$ .

→ interval **narrower** for

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

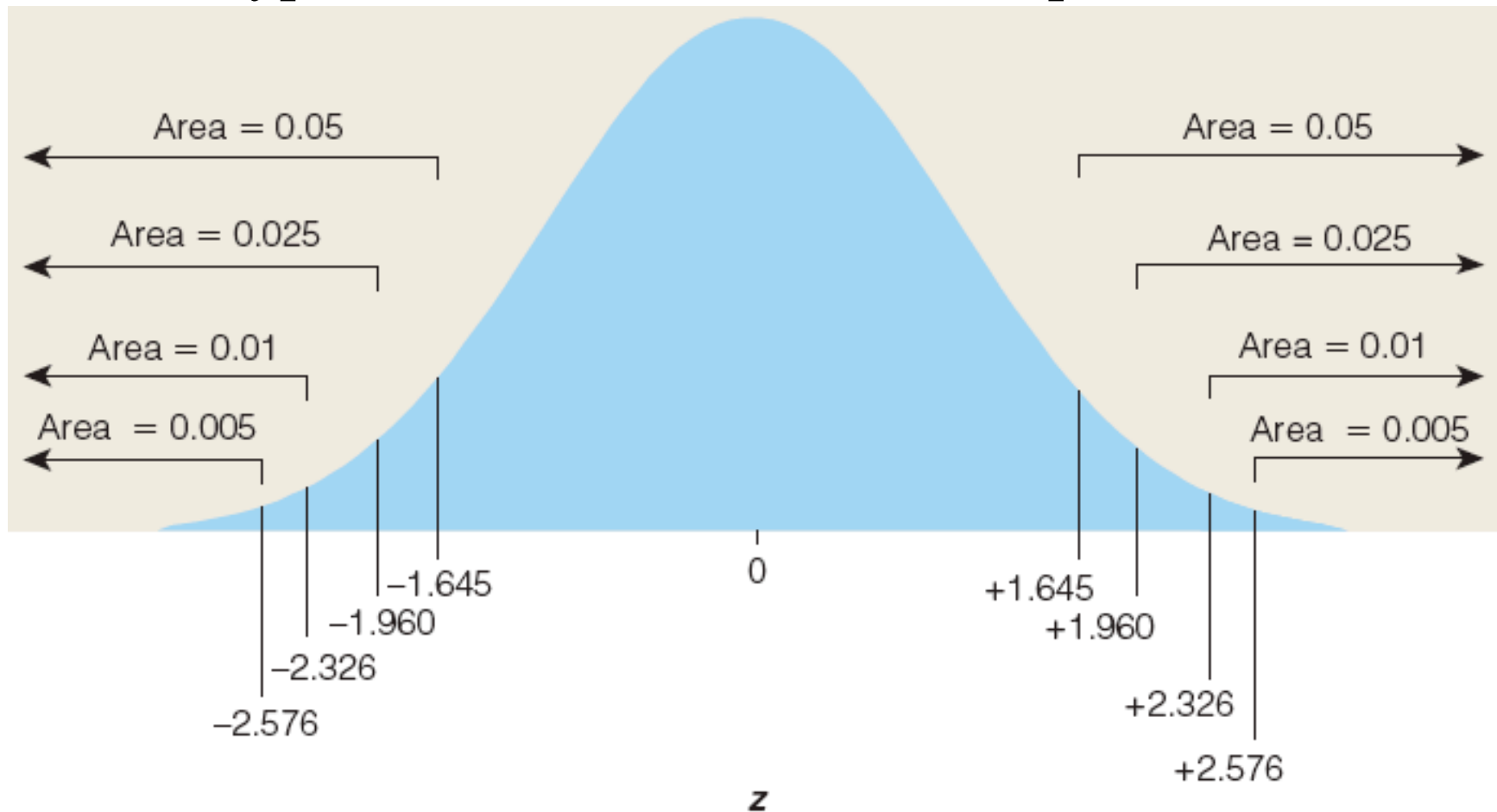
# From $z$ Confidence Intervals to Tests (*Review*)

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

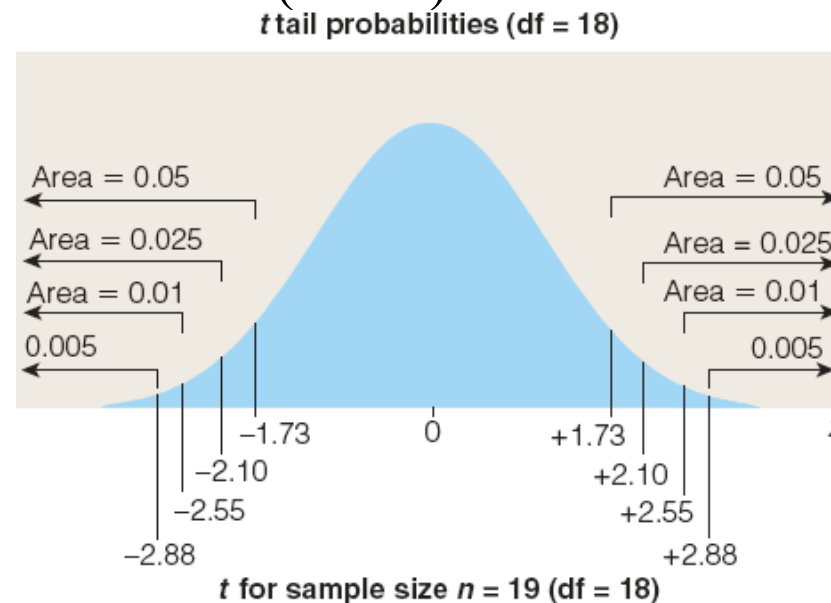
---

Confidence interval: use **multiplier** for  $t$  dist,  $n-1$  df

Hypothesis test:  $P$ -value based on **tail** of  $t$  dist,  $n-1$  df

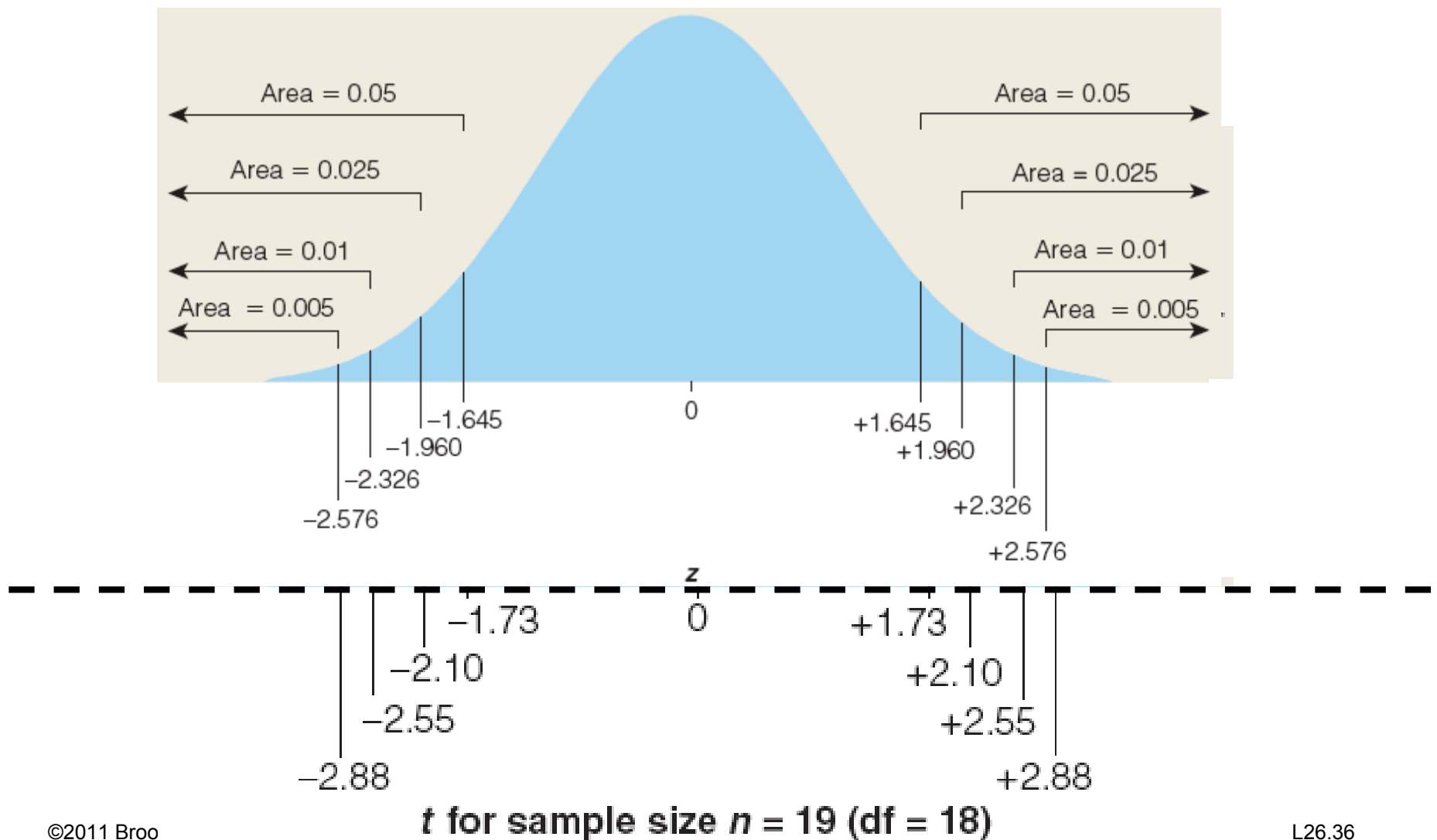
# Example: Hypothesis Test: $t$ vs. $z$

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



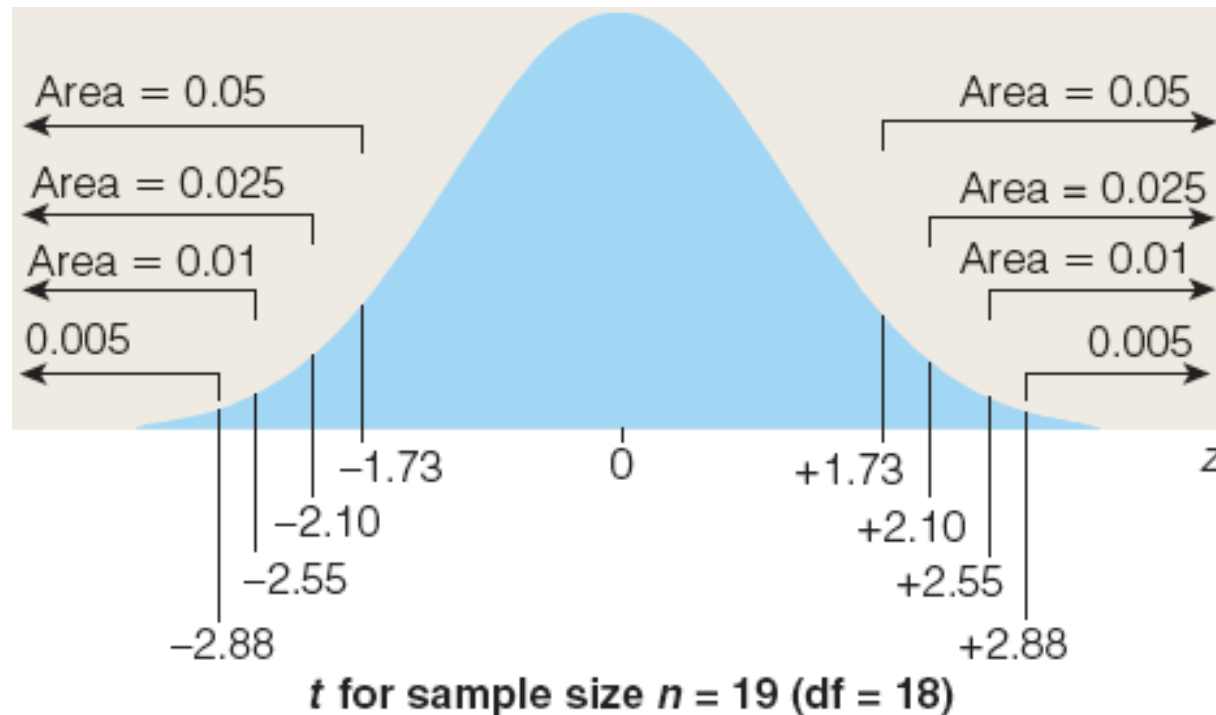
- **Question:** How do  $P$ -values compare for  $z$  and  $t$ ? (Assume alternative is “greater than”.)
- **Response:** 90-95-98-99 Rule  $\rightarrow z$   $P$ -value \_\_\_\_\_.  
 $t$  curve for 18 df  $\rightarrow t$   $P$ -value \_\_\_\_\_.

# 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.  $\bar{x} = 129.36, s = 12.82, t = \cdot$
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