

# STA 570

# Spring 2011

Lecture 19

*Thursday, March 31*

- **Correspondence Between Significance Tests and Confidence Intervals**
- **Small Sample Inference for Means**

# Correspondence Between Tests and Confidence Intervals

- Results of confidence intervals and of two-sided significance tests are consistent:
  - Whenever the hypothesized mean  $\mu_0$  is not in the confidence interval around  $\bar{Y}$ , then the  $p$ -value for testing  $H_0 : \mu = \mu_0$  is smaller than 5% (significance at the 5%-level)
  - In general, a (1-alpha)-confidence interval corresponds to a test at significance level alpha
  - This is true for means as well as proportions

# Example

- The mean score for all high school seniors taking a college entrance exam equals 500.
  - A study is conducted to see whether a different mean applies to those students born in a foreign country.
  - For a random sample of 100 of such students, the mean and standard deviation on this exam equal 508 and 100.
1. Set up hypotheses for a significance test.
  2. Compute the test statistic.
  3. Report the  $P$ -value, and interpret.
  4. Can you conclude that the population mean for students born in a foreign country equals 500? Why or why not?
  5. Make a decision about  $H_0$ , using  $\alpha=0.05$
  - 6. Construct a 95% confidence interval for  $\mu$ .**

# One-Sided Significance Tests

- Example: Usually, students have an average score of 85% on the STA 570 midterm exam
- You want to prove that a certain learning method helps improve the score
- 40 students try out the new method
- Null hypothesis:  $H_0 : \mu = 85\%$
- Alternative hypothesis:  $H_1 : \mu > 85\%$

# One-Sided Significance Tests

- Attention! For one-sided and two-sided tests, the calculation of the  $p$ -value is different!
- For this example, “everything *at least as extreme* as the observed value” is everything ***above*** the observed value

(if  $H_1 : \mu > \mu_0$ )

# One-Sided Significance Tests

- If we want to prove that  $\mu$  is smaller than a particular number  $\mu_0$ , then

$$H_0 : \mu = \mu_0 \quad \text{vs.} \quad H_1 : \mu < \mu_0$$

- The P-value is obtained taking the probability of all Y-values **less** than the observed Y-value

# One-Sided Versus Two-Sided Test

- Two-sided tests are more common
- Look for formulations like
  - “test whether the mean has ***changed***”
  - “test whether the mean has ***increased***”
  - “test whether the mean is ***the same***”
  - “test whether the mean has ***decreased***”

# Significance Test for a Proportion

## ***Assumptions***

- What type of data?
  - *Qualitative*
- Which sampling method has been used?
  - *Random sampling*
- What is the sample size?
  - *$n=20$  if  $p_0$  is between 0.25 and 0.75*
  - *In general (rule of thumb): Choose  $n$  such that*

$$n > 5 / p_0 \quad \text{and} \quad n > 5 / (1 - p_0)$$

# Significance Test for a Proportion

## *Hypotheses*

- Null hypothesis  $H_0 : p = p_0$   
where  $p_0$  is a priori specified
- Alternative hypotheses can be one-sided or two-sided
- Again, two-sided is more common

# Significance Test for a Proportion

$Z_{\text{obs}}$

=  $\frac{\text{estimator of the parameter} - \text{null hypothesis value of the parameter}}{\text{standard error of the estimator}}$

$$= \frac{\hat{p} - p_0}{\sqrt{p_0(1-p_0)/n}}$$

## ***P-Value***

- Calculation is exactly the same as for the test for a mean
- Find one- or two-sided tail probabilities using online tools

# Example

- Let  $p$  denote the proportion of Kentuckians who think that government environmental regulations are too strict
  - Test  $H_0: p=0.5$  against a two-sided alternative using data from a telephone poll of 834 people in which 26.6% said regulations were too strict
1. Calculate the test statistic
  2. Find the  $p$ -value and interpret
  3. Using  $\alpha=0.01$ , can you determine whether a majority or minority think that environmental regulations are too strict, or is it plausible that  $p=0.5$  ?
  4. Construct a 99% confidence interval. Explain the advantage of the confidence interval over the test.

# Summary

## Large Sample Significance Test for a Mean

	One-Sided Tests		Two-Sided Test
Null Hypothesis	$H_0 : \mu = \mu_0$		
Research Hypothesis	$H_1 : \mu < \mu_0$	$H_1 : \mu > \mu_0$	$H_1 : \mu \neq \mu_0$
Test Statistic	$z = \frac{\bar{Y} - \mu_0}{s / \sqrt{n}}$		
<i>p</i> -value	$P(Z < z_{obs})$	$P(Z > z_{obs})$	$2 \cdot P(Z >  z_{obs} )$

# Large Sample Significance Test for a Population Proportion

	One-Sided Tests		Two-Sided Test
Null Hypothesis	$H_0 : p = p_0$		
Research Hypothesis	$H_1 : p < p_0$	$H_1 : p > p_0$	$H_1 : p \neq p_0$
Test Statistic	$z_{obs} = \frac{\hat{p} - p_0}{\sqrt{p_0(1 - p_0) / n}}$		
$p$ -value	$P(Z < z_{obs})$	$P(Z > z_{obs})$	$2 \cdot P(Z >  z_{obs} )$

# Sample Size: Test for a Mean

	One-Sided Tests		Two-Sided Test
Null Hypothesis	$H_0 : \mu = \mu_0$		
Research Hypothesis	$H_1 : \mu < \mu_0$	$H_1 : \mu > \mu_0$	$H_1 : \mu \neq \mu_0$
Sample Size	$\left( \sigma \frac{z_{1-\beta} + z_{1-\alpha}}{\mu_0 - \mu_1} \right)^2$		$\left( \sigma \frac{z_{1-\beta} + z_{1-\alpha/2}}{\mu_0 - \mu_1} \right)^2$

$z_p = p$ th quantile of standard normal

# Sample Size: Test for a Proportion

	One-Sided Tests		Two-Sided Test
Null Hypothesis	$H_0 : p = p_0$		
Research Hypothesis	$H_1 : p < p_0$	$H_1 : p > p_0$	$H_1 : p \neq p_0$
Sample Size	$\left( \frac{z_{1-\beta}S_1 + z_{1-\alpha}S_0}{p_0 - p_1} \right)^2$		$\left( \frac{z_{1-\beta}S_1 + z_{1-\alpha/2}S_0}{p_0 - p_1} \right)^2$

$z_p = p$ th quantile of standard normal

# Multiple Choice Question I

- A 95% confidence interval for  $\mu$  is (96, 110). Which of the following statements about significance tests for the same data are correct?
  - a) In testing the null hypothesis  $\mu=100$  (two-sided),  $P>0.05$
  - b) In testing the null hypothesis  $\mu=100$  (two-sided),  $P<0.05$
  - c) In testing the null hypothesis  $\mu=x$  (two-sided),  $P>0.05$  if  $x$  is any of the numbers inside the confidence interval
  - d) In testing the null hypothesis  $\mu=x$  (two-sided),  $P<0.05$  if  $x$  is any of the numbers outside the confidence interval

# Multiple Choice Question II

- The P-value for testing the null hypothesis  $\mu=100$  (two-sided) is  $P=.001$ . This indicates
  - a) There is strong evidence that  $\mu = 100$
  - b) There is strong evidence that  $\mu$  does not equal 100
  - c) There is strong evidence that  $\mu > 100$
  - d) There is strong evidence that  $\mu < 100$
  - e) If  $\mu$  were equal to 100, it would be unusual to obtain data such as those observed

# Multiple Choice Question II

- The P-value for testing the null hypothesis  $\mu=100$  (two-sided) is  $P=.001$ . Suppose that in addition you know that the z score of the test statistic was  $z=3.29$ . Then
  - a) There is strong evidence that  $\mu = 100$
  - b) There is strong evidence that  $\mu > 100$
  - c) There is strong evidence that  $\mu < 100$

## Small Sample Confidence Interval for a Mean

- What if we want to make inference about the population mean, but our sample size is not big enough to meet the minimal sample size requirement  $n > 25$  to apply the central limit theorem?
- Confidence intervals are constructed in the same way as before, but now we are using **t-values instead of z-values**
- For a random sample **from a normal distribution**, a 95% confidence interval for  $\mu$  is

$$\bar{Y} \pm t_{0.025} \frac{s}{\sqrt{n}}$$

- where  $t_{0.025}$  is a t-score (instead of z-score) from a site like <http://stattrek.com/Tables/T.aspx>
- degrees of freedom are  $df = n - 1$

# Small Sample Hypothesis Test for a Mean

- Assumptions
  - Quantitative variable, random sampling, **population distribution is normal, any sample size**
- Hypotheses
  - Same as in the large sample test for the mean

$$H_0 : \mu = \mu_0 \text{ vs. } H_1 : \mu \neq \mu_0$$

$$\text{or } H_0 : \mu = \mu_0 \text{ vs. } H_1 : \mu > \mu_0$$

$$\text{or } H_0 : \mu = \mu_0 \text{ vs. } H_1 : \mu < \mu_0$$

# Small Sample Hypothesis Test for a Mean

- Test statistic
  - Exactly the same as for the large sample test

$$t_{obs} = \frac{\bar{Y} - \mu_0}{s / \sqrt{n}}$$

- *p*-Value
  - Same as for the large sample test (one-or two-sided), but using an online tool for the *t* distribution
- Conclusion
  - Report *p*-value and make formal decision

# Small Sample Hypothesis Test for a Mean: Example

- A study was conducted of the effects of a special class designed to improve children/s verbal skills
- Each child took a verbal skills test twice, both before and after a three-week period in the class
- $Y = 2^{\text{nd}} \text{ exam score} - 1^{\text{st}} \text{ exam score}$
- If the population mean for  $Y$ ,  $E(Y) = \mu$  equals 0, the class has no effect
- Test the null hypothesis of no effect against the alternative hypothesis that the effect is positive
- Sample ( $n=4$ ): 3,7,3,3

# Normality Assumption

- An assumption for the  $t$ -test is that the population distribution is normal
- In practice, it is impossible to be 100% sure if the population distribution is normal
- It is useful to look at histogram or stem-and-leaf plot (or normal probability plot) to check whether the normality assumption is reasonable

# Normality Assumption

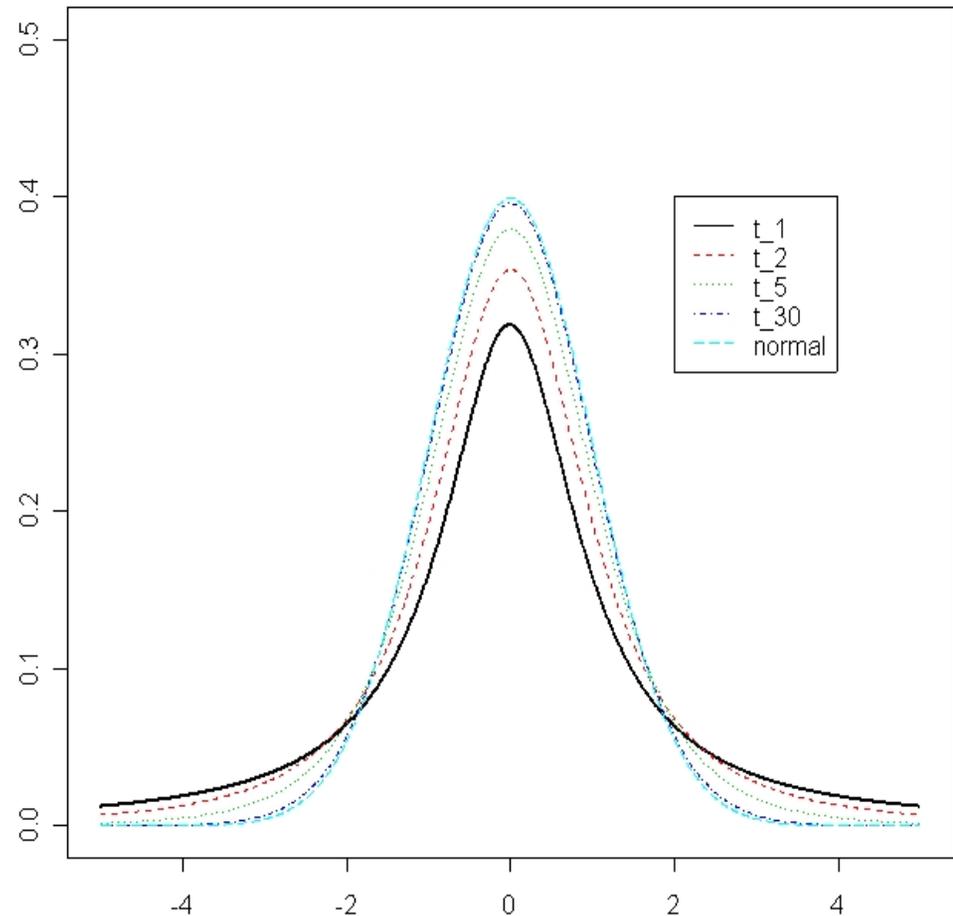
- Good news: The  $t$ -test is relatively ***robust*** against violations of the assumption that the population distribution is normal
- Unless the population distribution is highly skewed, the  $p$ -values and confidence intervals are fairly accurate
- However: The random sampling assumption must never be violated, otherwise the test results are completely invalid

Summary: Small Sample... Significance Test for a Mean  
*(Assumption: Population distribution is normal)*

	One-Sided Tests		Two-Sided Test
Null Hypothesis	$H_0 : \mu = \mu_0$		
Research Hypothesis	$H_1 : \mu < \mu_0$	$H_1 : \mu > \mu_0$	$H_1 : \mu \neq \mu_0$
Test Statistic	$t_{obs} = \frac{\bar{Y} - \mu_0}{s / \sqrt{n}}$ , degrees of freedom = $n - 1$		
<i>p</i> -value	$P(T_{n-1} < t_{obs})$	$P(T_{n-1} > t_{obs})$	$2 \cdot P(T_{n-1} >  t_{obs} )$

# t-Distributions

- The t-distributions are bell-shaped and symmetric around 0
- The smaller the degrees of freedom, the more spread out is the distribution
- $t$ -distributions look almost like a normal distribution
- In fact, the limit of the  $t$ -distributions is a normal distribution when  $n$  gets larger



# Statistical Methods for One Sample Summary I

- Testing the Mean
  - Large sample size (*30 or more*):  
Use the large sample test for the mean  
(z-Scores, normal distribution)
  - Small sample size:  
Check whether the data is not very skewed  
Use the *t* test for the mean  
(*t*-Scores, *t* distribution)

# Statistical Methods for One Sample Summary II

- Testing the Proportion
  - Large sample size ( $np > 5$ ,  $n(1-p) > 5$ ):  
Use the large sample test for the proportion  
(z-Scores, normal distribution)
  - Small sample size:  
Binomial distribution  
*(not covered in class)*