Unit 3: Foundations for inference

3. Hypothesis tests

Sta 101 - Fall 2015

Duke University, Department of Statistical Science

2. Main ideas

- 1. Use hypothesis tests to make decisions about population parameters
- Hypothesis tests and confidence intervals at equivalent significance/confidence levels should agree
- Results that are statistically significant are not necessarily practically significant
 - 4. Hypothesis tests are prone to decision errors

- Peer eval feedback:
 - Lateness / absence, mental presence, communication, commitment
 - Tips: Move seats around in class and lab, give your best, and don't assume the worst
 - Instructions: Give scores that reflect your verbal comments
- Midterm 1: Monday, Oct 5
 - Come to class with questions on Wednesday

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1. Use hypothesis tests to make decisions about population parameters

Hypothesis testing framework:

- 1. Set the hypotheses.
- 2. Check assumptions and conditions.
- 3. Calculate a test statistic and a p-value.
- 4. Make a decision, and interpret it in context of the research question.

1. Set the hypotheses

- $H_0: \mu = null\ value$
- $H_A: \mu < \text{or} > \text{or} \neq null\ value}$

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- Sample size / skew: $n \ge 30$ (or larger if sample is skewed), no extreme skew

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, where $SE = \frac{s}{\sqrt{n}}$

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- 4. Make a decision, and interpret it in context of the research question
 - If p-value $< \alpha$, reject H_0 , data provide evidence for H_A
 - If p-value $> \alpha$, do not reject H_0 , data do not provide evidence for H_A

Application exercise: 3.2 Hypothesis testing for a single mean

See course website for details.

Which of the following is the correct interpretation of the p-value from App Ex 3.2?

- (a) The probability that average GPA of Duke students has changed since 2001.
- (b) The probability that average GPA of Duke students has not changed since 2001.
- (c) The probability that average GPA of Duke students has not changed since 2001, if in fact a random sample of 63 Duke students this year have an average GPA of 3.58 or higher.
- (d) The probability that a random sample of 63 Duke students have an average GPA of 3.58 or higher, if in fact the average GPA has not changed since 2001.
- (e) The probability that a random sample of 63 Duke students have an average GPA of 3.58 or higher or 3.16 or lower, if in fact the average GPA has not changed since 2001.

Common misconceptions about hypothesis testing

P-value is the probability that the null hypothesis is true
 A p-value is the probability of getting a sample that results
 in a test statistic as or more extreme than what you
 actually observed (and in favor of the null hypothesis) if in
 fact the null hypothesis is correct. It is a conditional
 probability, conditioned on the null hypothesis being
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 A high p-value means the data do not provide convincing evidence for the alternative hypothesis and hence that the null hypothesis can't be rejected.

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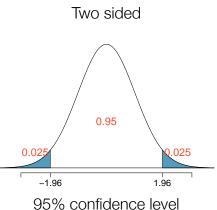
 A high p-value means the data do not provide convincing evidence for the alternative hypothesis and hence that the null hypothesis can't be rejected.
- 3. A low p-value confirms the alternative hypothesis.

 A low p-value means the data provide convincing evidence for the alternative hypothesis, but not necessarily that it is confirmed.

2. Main ideas

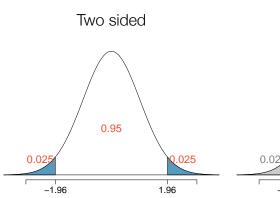
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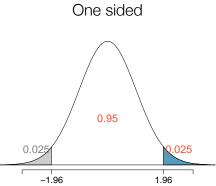


95% confidence level is equivalent to two sided HT with $\alpha=0.05$

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95% confidence level is equivalent to two sided HT with $\alpha=0.05$



95% confidence level is equivalent to one sided HT with $\alpha=0.025$

What is the confidence level for a confidence interval that is equivalent to a two-sided hypothesis test at the 1% significance level? *Hint: Draw a picture and mark the confidence level in the center.*

- (a) 0.80
- **(b)** 0.90
- (c) 0.95
- (d) 0.98
- **(e)** 0.99

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A 95% confidence interval for the average normal body temperature of humans is found to be (98.1 F, 98.4 F). Which of the following is <u>true</u>?

- (a) The hypothesis H_0 : $\mu = 98.2$ would be rejected at $\alpha = 0.05$ in favor of H_A : $\mu \neq 98.2$.
- (b) The hypothesis $H_0: \mu = 98.2$ would be rejected at $\alpha = 0.025$ in favor of $H_A: \mu > 98.2$.
- (c) The hypothesis H_0 : $\mu = 98$ would be rejected using a 90% confidence interval.
- (d) The hypothesis H_0 : $\mu=98.2$ would be rejected using a 99% confidence interval.

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Clicker question

All else held equal, will p-value be lower if n=100 or n=10,000?

- (a) n = 100
- (b) n = 10,000

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- ▶ A *Type 1 Error* is rejecting the null hypothesis when H_0 is true: α
 - For those cases where H_0 is actually true, we do not want to incorrectly reject it more than 5% of those times
 - Increasing α increases the Type 1 error rate, hence we prefer to small values of α

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- ▶ A *Type 2 Error* is failing to reject the null hypothesis when H_A is true: β
- ▶ *Power* is the probability of correctly rejecting H_0 , and hence the complement of the probability of a Type 2 Error: 1β

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Summary of main ideas

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