

# Stat 201: Introduction to Statistics

Standard 28: Significance Test - Means

# Confidence Intervals to Testing

- We've seen earlier that we can come up with interesting observations of our confidence intervals
- Next we will learn how to formally test whether or not the population mean is a particular value based off our sample mean

# Hypothesis Test for Means: Step 1

- **State Hypotheses:** it's usually easier to write the alternative hypothesis first
  - **Null hypothesis:** that the population mean equals some  $\mu_0$ 
    - $H_0: \mu \leq \mu_0$  (one sided test)
    - $H_0: \mu \geq \mu_0$  (one sided test)
    - $H_0: \mu = \mu_0$  (two sided test)
  - **Alternative hypothesis:** What we're interested in
    - $H_a: \mu > \mu_0$  (one sided test)
    - $H_a: \mu < \mu_0$  (one sided test)
    - $H_a: \mu \neq \mu_0$  (two sided test)

# Hypothesis Test for Means: Step 2

- Check the assumptions
  - The variable must be quantitative
  - The data are obtained using randomization
  - We're dealing with data from the normal distribution
    - If  $n > 30$
    - If a histogram of the data is approximately normal which indicates that the probability is normal

# Hypothesis Test for Means: Step 3

- When we don't know  $\sigma_x$
- Calculate Test Statistic,  $t^*$ 
  - The test statistic measures how different the sample mean we have is from the null hypothesis
  - We calculate the t-statistic by assuming that  $\mu_0$  is the population mean

$$t^* = \frac{(\bar{x} - \mu_0)}{\frac{s_x}{\sqrt{n}}}$$

# Hypothesis Test for Means: Step 3

- When we know  $\sigma_x$
- Calculate Test Statistic,  $z^*$ 
  - The test statistic measures how different the sample mean we have is from the null hypothesis
  - We calculate the t-statistic by assuming that  $\mu_0$  is the population mean

$$z = \frac{(\bar{x} - \mu_0)}{\frac{\sigma_x}{\sqrt{n}}}$$

# Hypothesis Test for Means: Step 4

- When we don't know  $\sigma_x$
- Determine the P-value
  - The P-value describes how unusual the data would be if  $H_0$  were true.
  - We will use software or your calculator to find this, or I will give it to you.

Alternative Hypothesis	Probability	Formula for the P-value
$H_a: \mu > \mu_0$	Right tail	$P(T > t^*) = 1 - P(T < t^*)$
$H_a: \mu < \mu_0$	Left tail	$P(T < t^*)$
$H_a: \mu \neq \mu_0$	Two-tail	$2 * P(T < - t^* )$

# Hypothesis Test for Proportions: Step 4

- When we know  $\sigma_x$
- **Determine the P-value**
  - The P-value describes how unusual the sample data would be if  $H_0$  were true, which is what we're assuming ( $\mu = \mu_0$ ).
  - $z^*$  is the test statistic from step 3

Alternative Hypothesis	Probability	Formula for the P-value
$H_a: \mu > \mu_0$	Right tail	$P(Z > z^*) = 1 - P(Z < z^*)$
$H_a: \mu < \mu_0$	Left tail	$P(Z < z^*)$
$H_a: \mu \neq \mu_0$	Two-tail	$2 * P(Z < - z^* )$



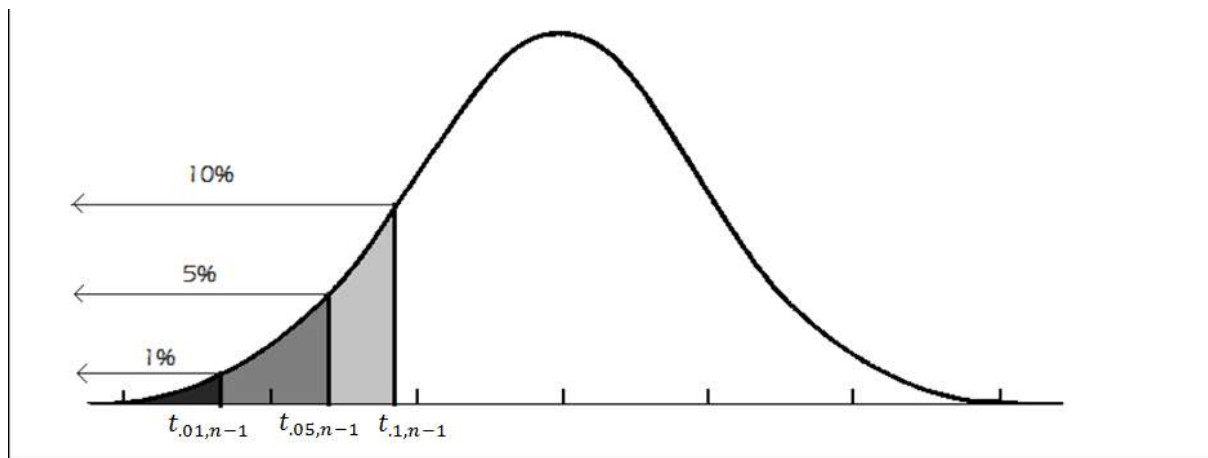
# Hypothesis Test for Means: Step 5

- Summarize the test by reporting and interpreting the P-value
  - Smaller p-values give stronger evidence against  $H_0$
- If  $p\text{-value} \leq (1 - \textit{confidence}) = \alpha$ 
  - Reject  $H_0$ , with a p-value = \_\_\_\_\_, we have sufficient evidence that the alternative hypothesis might be true
- If  $p\text{-value} > (1 - \textit{confidence}) = \alpha$ 
  - Fail to reject  $H_0$ , with a p-value = \_\_\_\_\_, we do not have sufficient evidence that the alternative hypothesis might be true

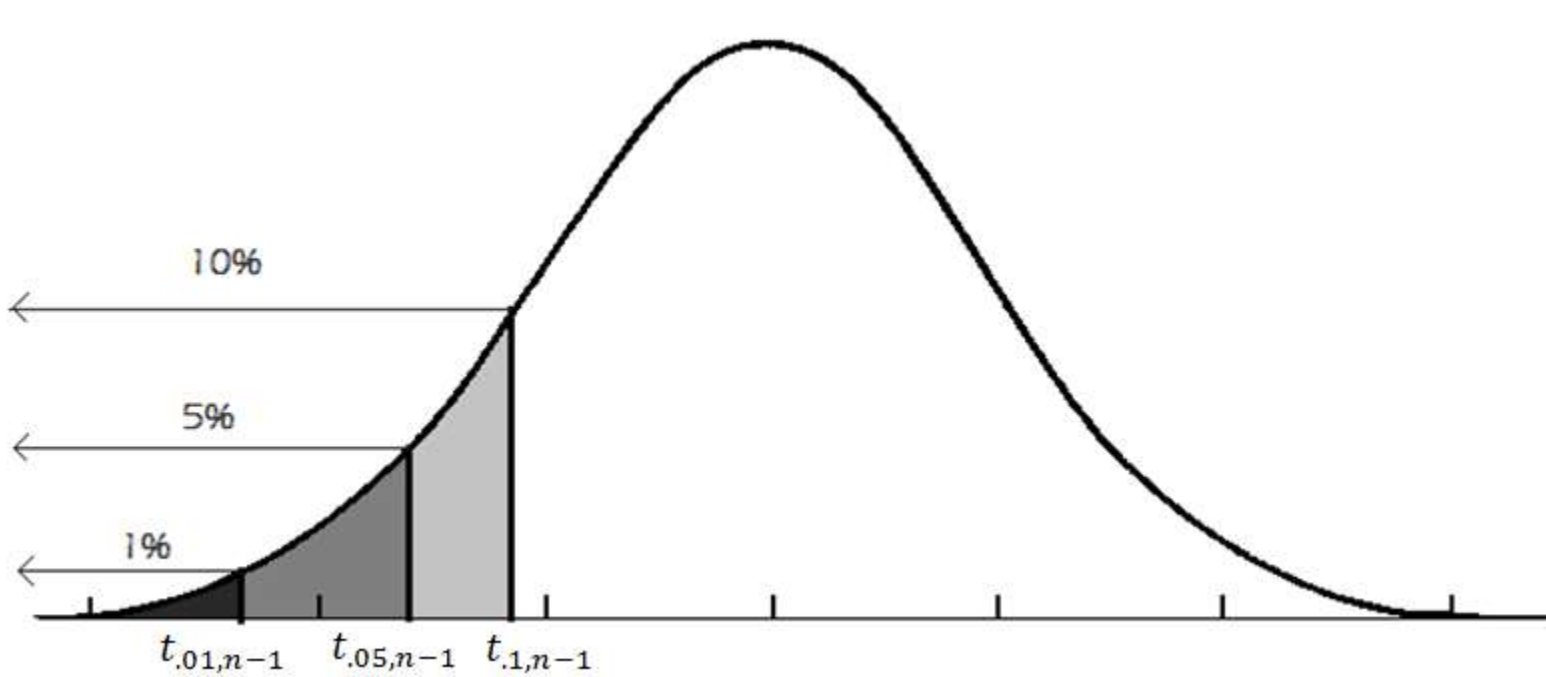
# Hypothesis Test for Means– Step Five with Pictures

- For a left tailed test:  $H_a: \mu < \mu_o \rightarrow$  We have rejection regions for  $H_o$  are as follows

Confidence	Reject (test stat)	Reject (p-value)
0.90	Test-stat $< -t_{.10, n-1}$	P-value $< .1$
0.95	Test-stat $< -t_{.05, n-1}$	P-value $< .05$
0.99	Test-stat $< -t_{.01, n-1}$	P-value $< .01$



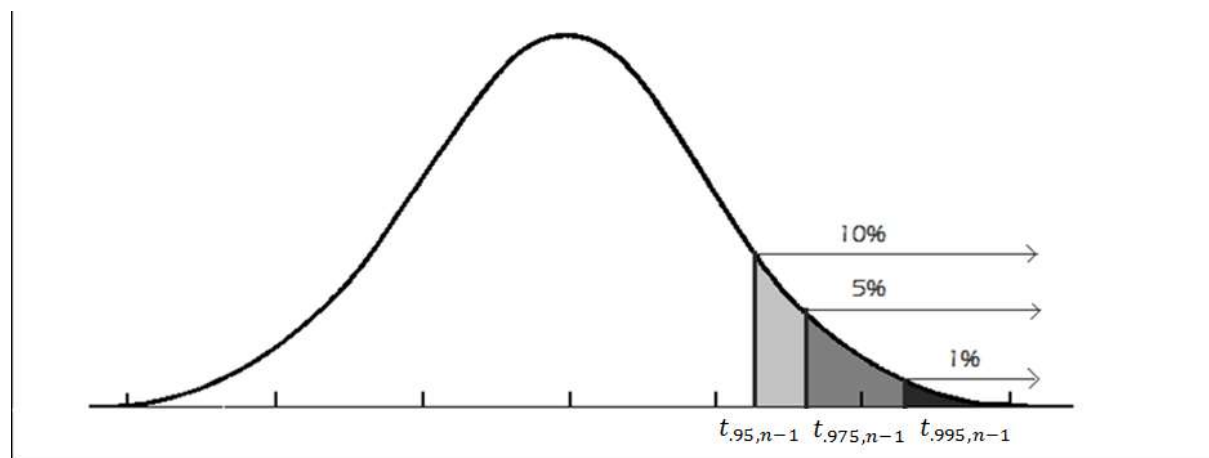
# Zoom In



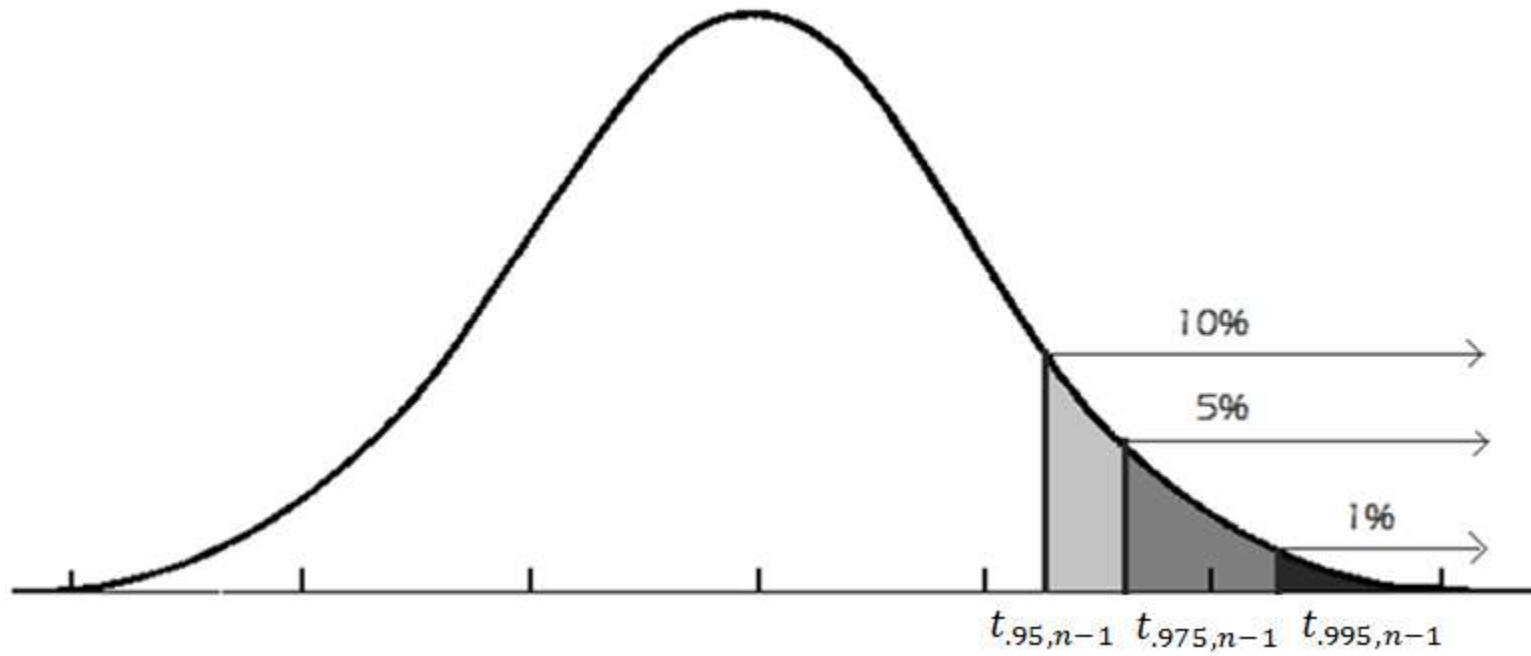
# Hypothesis Test for Means– Step Five with Pictures

- For a left tailed test:  $H_a: \mu > \mu_o \rightarrow$  We have rejection regions for  $H_o$  are as follows

Confidence	Reject (test stat)	Reject (p-value)
0.90	Test-stat $< -t_{.90, n-1}$	P-value $< .1$
0.95	Test-stat $< -t_{.95, n-1}$	P-value $< .05$
0.99	Test-stat $< -t_{.99, n-1}$	P-value $< .01$



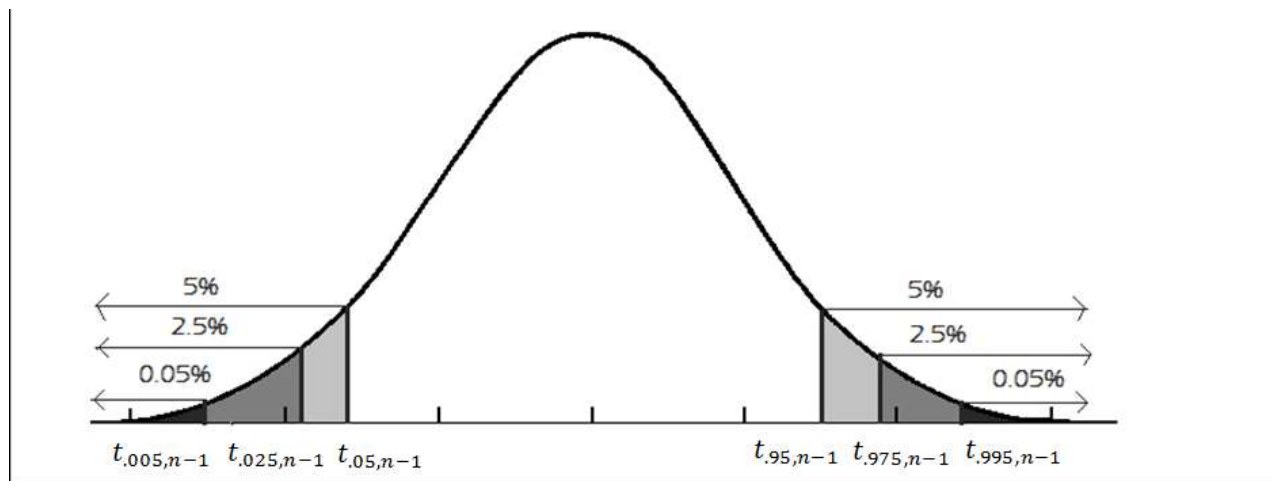
# Zoom In



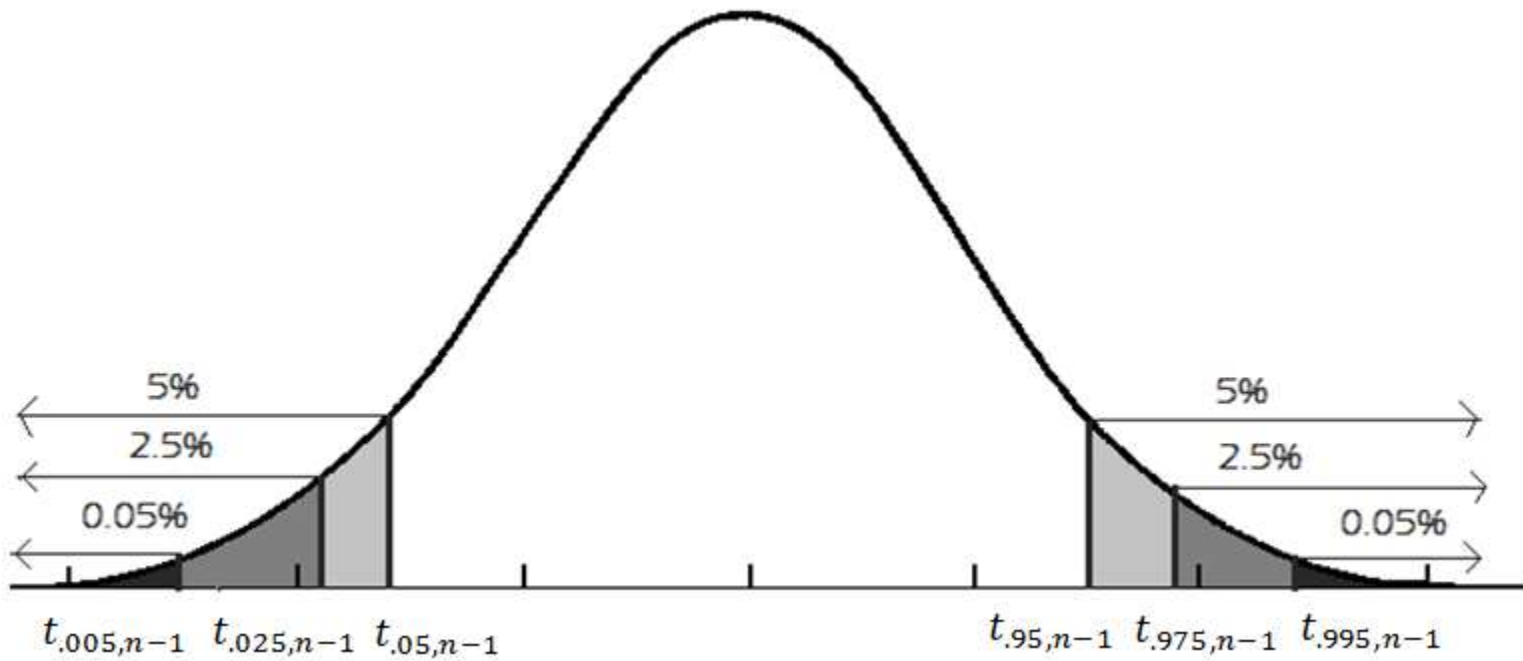
# Hypothesis Test for Means– Step Five with Pictures

- For a two tailed test:  $H_a: \mu \neq \mu_o \rightarrow$  We have rejection regions for  $H_o$  are as follows

Confidence	Reject (test stat)	Reject (p-value)
0.90	$ \text{Test-stat}  < -t_{.90, n-1}$	P-value < .1
0.95	$ \text{Test-stat}  < -t_{.95, n-1}$	P-value < .05
0.99	$ \text{Test-stat}  < -t_{.99, n-1}$	P-value < .01



# Zoom In



# Hypothesis Test for Means– Step Five

- The pictures are the same when we know  $z$  as they are for proportions.
- In almost all feasible cases we will not know  $\sigma_x$  as it is usually unrealistic to know it



# Example 1

- It is often hypothesized that Velociraptors were warm blooded creatures, some scientists guessed their normal blood temperature was 87.5 degrees. Test whether or not the mean differs from 87.5 degrees at a .05 significance level, or 95% confidence.
- A random sample of thirteen Velociraptors during the shooting of Jurassic Park gave the data below  
88.6, 86.4, 87.2, 87.4, 87.2, 87.6, 86.8, 86.1, 87.4,  
87.3, 86.4, 86.6, 87.1

# Example 1 – Step One

- A random sample of thirteen Velociraptors during the shooting of Jurassic Park gave the data below

88.6, 86.4, 87.2, 87.4, 87.2, 87.6, 86.8, 86.1, 87.4,  
87.3, 86.4, 86.6, 87.1

- State the Hypotheses: we are interested in whether or not the mean is **not equal to 87.5 degrees**

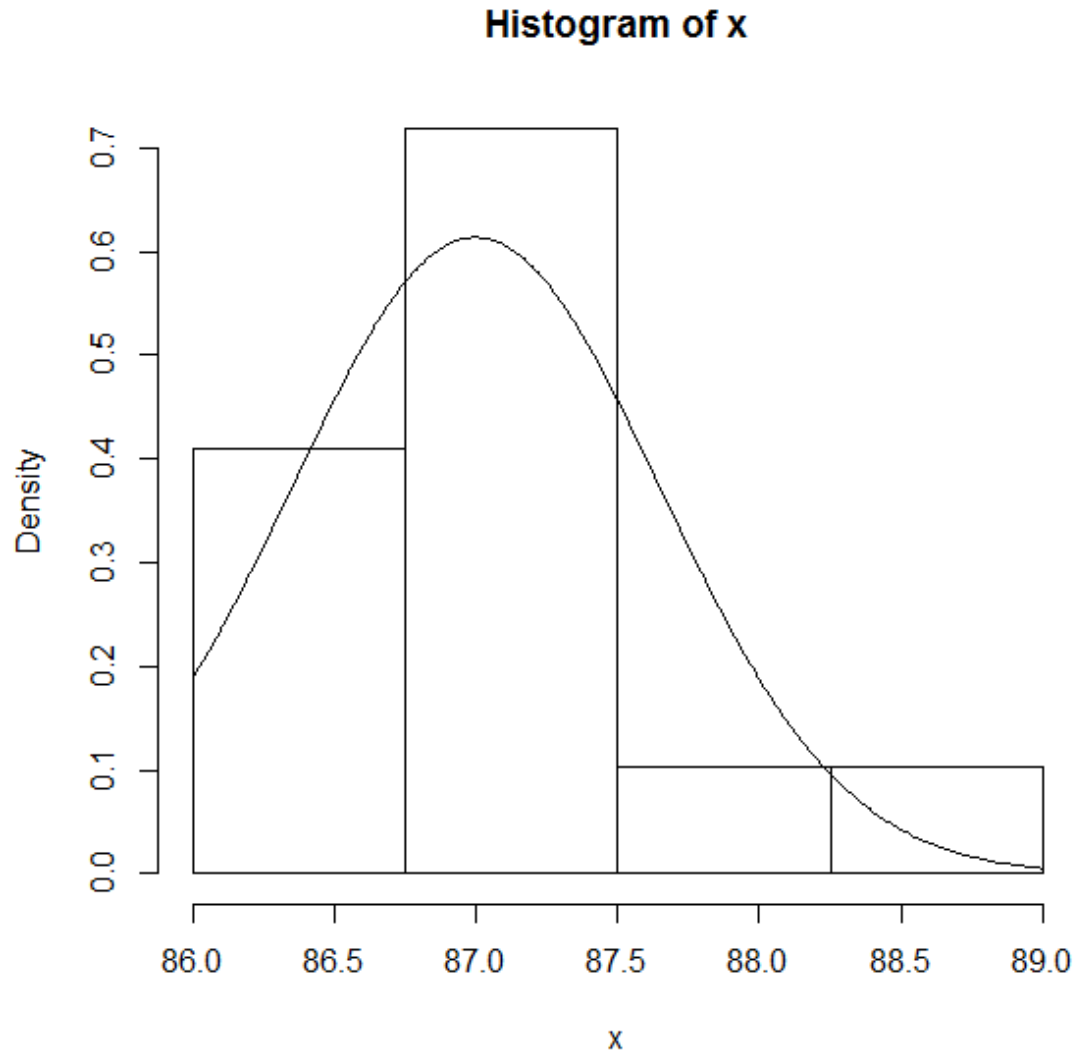
- $H_o: \mu = 87.5$

- $H_a: \mu \neq 87.5$

# Example 1 – Step Two

- A random sample of thirteen Velociraptors during the shooting of Jurassic Park gave the data below  
88.6, 86.4, 87.2, 87.4, 87.2, 87.6, 86.8, 86.1, 87.4,  
87.3, 86.4, 86.6, 87.1
- Check Assumptions:
  - The data is quantitative
  - The sample is randomly selected
  - Even though  $n < 30$ , a histogram of the data shows approximately normal

See, I told you. (close enough for us)



# Example 1 – Step Three

- Calculate Test Statistic

Variable	Sample Mean ( $\bar{x}$ )	Standard Deviation ( $s_x$ )	Standard Error ( $s_{\bar{x}}$ )
Blood Temperature	87.0846	.6492	.1800

$$t = \frac{(\bar{x} - \mu_o)}{\frac{s}{\sqrt{n}}} = \frac{87.08 - 87.5}{\frac{.6492}{\sqrt{13}}} = \frac{.42}{.1800} = -2.\overline{33}$$

# Example 1 – Step Four

- Determine P-value

P-value from software is .0397

# Example 1 – Step Five

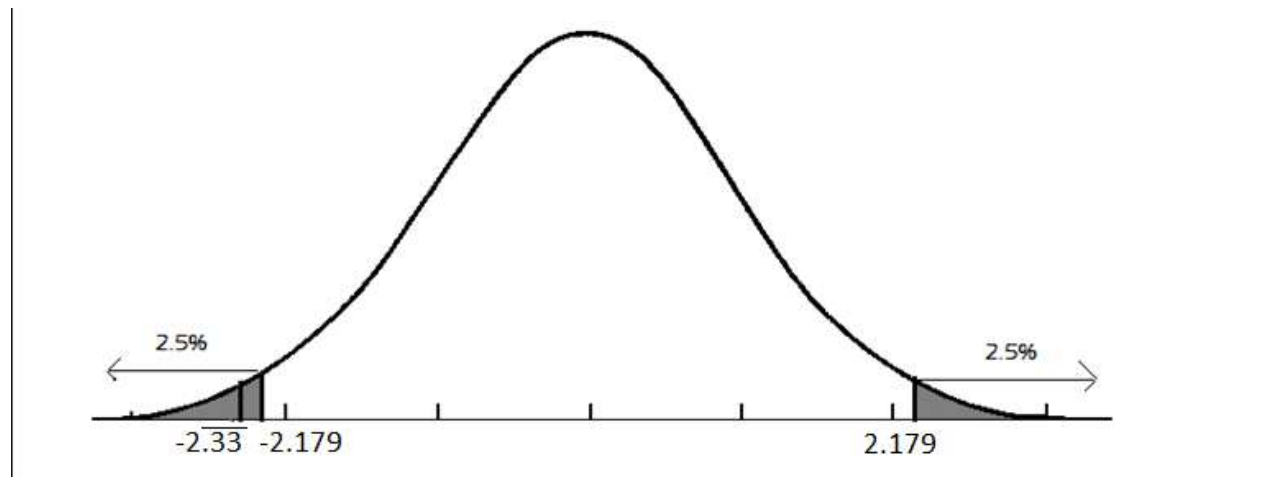
- State Conclusion

- Since  $.0397 < .05$  we reject  $H_0$

- At the .05 level of significance, or 95% confidence level, there is sufficient evidence that the mean blood temperature is different than 87.5.

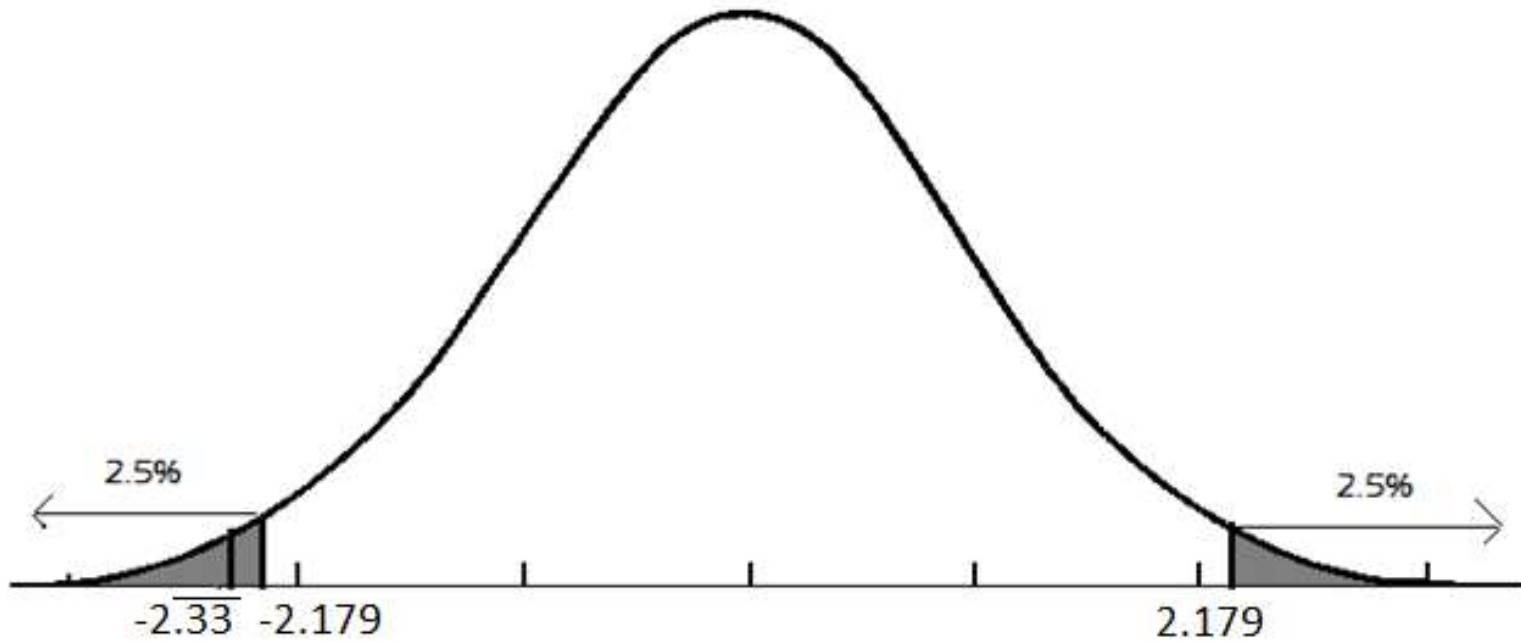
# Example 1 – Step Five with pictures

- State Conclusion
  - Anything with a p-value < .05 or a  $|t\text{-value}| > t_{1-\frac{\alpha}{2}, n-1} = t_{.975, 12} = 2.179$  will be in the rejection region
  - Since  $.2932 > .05$  we fail to reject  $H_0$





# Zoom In



# Example

- Suppose a random sample of 38 yearly average temperature measures in New Haven, CT. Among the sampled years the **sample mean temperature was 51.0474** degrees Fahrenheit with a **sample standard deviation of 1.3112**.
- Test whether or not the population mean differs from 50 degrees at a .05 significance level, or 95% confidence.

# Example – Step One

- State the Hypotheses: we are interested in whether or not the mean is **not equal to 50 degrees**
  - $H_o: \mu = 50$
  - $H_a: \mu \neq 50$

# Example – Step Two

- Check Assumptions:
  - The data is quantitative
  - The sample is randomly selected
  - $n > 30$  so it is safe to assume the sampling distribution for the sample mean is normal

# Example – Step Three

- Calculate Test Statistic

Variable	Sample Mean ( $\bar{x}$ )	Standard Deviation ( $s_x$ )	Standard Error ( $s_{\bar{x}}$ )
Yearly Temperature	51.0474	1.3112	.2127047

$$t = \frac{(\bar{x} - \mu_o)}{\frac{s}{\sqrt{n}}} = \frac{51.0474 - 50}{\frac{1.3112}{\sqrt{38}}} = \frac{1.0474}{.2127047} = 4.924198$$

# Example – Step Four

- Determine P-value

$$\begin{aligned} P \text{ value} &= 2 * P(T < -|t^*|) \\ &= 2 * P(T < -|4.924198|) \\ &= 2 * P(T < -4.924198) \\ &= .00001782519 \end{aligned}$$

# Example – Step Five

- State Conclusion

- Since  $.00001782519 < .05$  we reject  $H_0$

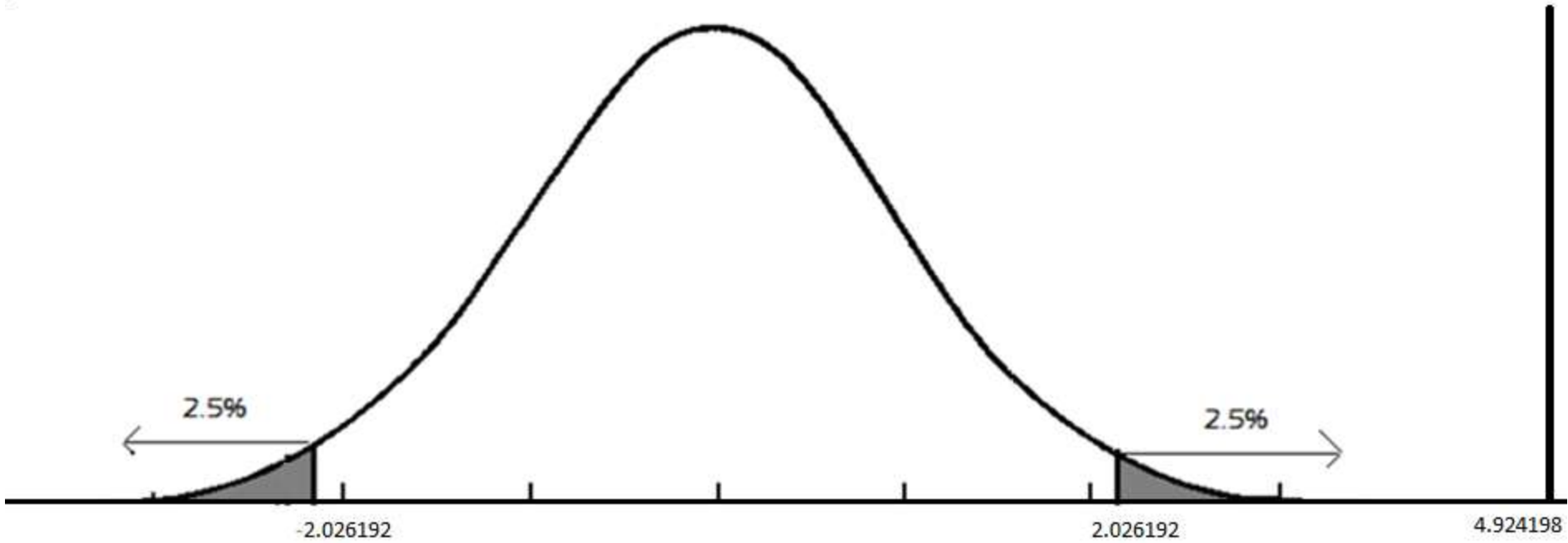
- At the .05 level of significance, or 95% confidence level, there is sufficient evidence that the mean yearly temperature is different than 50 degrees.

# Example – Step Five with pictures

- State Conclusion
  - Anything with a p-value < .05 or a  $|t\text{-value}| > t_{1-\frac{\alpha}{2}, n-1} = t_{.975, 37} = 2.026192$  will be in the rejection region
  - By P-value:
    - Since .00001782519 > .05 we reject  $H_0$
  - By T-statistic:
    - Since  $|4.924198| > 2.026192$  we reject  $H_0$



# Zoom In



# Hypothesis Testing for Means in Your TI Calculator

- Hypothesis testing for means
  - [https://www.youtube.com/watch?v=StpX5\\_AHKSs](https://www.youtube.com/watch?v=StpX5_AHKSs)
  - <https://www.youtube.com/watch?v=31fFfsSmuK8>
  - [https://www.youtube.com/watch?v=dyj0Mjvu\\_mQ](https://www.youtube.com/watch?v=dyj0Mjvu_mQ)

# Hypothesis Testing for Means in Your TI Calculator

- **When we don't know  $\sigma_x$ , with data**
- **INPUT:**
  1. Press STAT
  2. Press  $\rightarrow$  to TESTS
  3. Highlight '2: T-Test' and Press ENTER
  4. **With Data**
    1. Enter the we're interested in next to ' $\mu_0$ :'
    2. You should have your data table entered in L1
      - If you forgot: Press STAT, Press ENTER with 'Edit' highlighted, Enter the data into the L1 col.
    3. Next to 'List:' press 2<sup>nd</sup> then press 1
    4. Set 'Frequency' to 1
    5. Select the appropriate alternative hypothesis on the ' $\mu$ :' line by highlighting the correct inequality and pressing ENTER
    6. Highlight Calculate and press ENTER

# Hypothesis Testing for Means in Your TI Calculator

- **When we don't know  $\sigma_x$ , with data**
- **Output:**
  - Confirm the first line shows the hypothesis you would like to test
  - $t$  = the test statistic for our hypothesis test
  - $p$  = the p-value for this test
    - We make our decision based on this
  - $\bar{x}$  is the sample mean for the problem and should match the number you entered
  - $s_x$  is the sample standard deviation for the problem
  - $n$  is the sample size and should match the number you entered

# Hypothesis Testing for Means in Your TI Calculator

- **When we don't know  $\sigma_x$ , with stats**

- **INPUT:**

1. Press STAT

2. Press  $\rightarrow$  to TESTS

3. Highlight '2: T-Test' and Press ENTER

4. **With Stats**

1. Enter the we're interested in next to ' $\mu_0$ :'

2. Put the sample mean next to ' $\bar{x}$ :'

3. Enter the sample standard deviation next to ' $s_x$ :'

4. Put the sample size next to 'n:'

5. Select the appropriate alternative hypothesis on the ' $\mu$ :' line by highlighting the correct inequality and press ENTER

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  - $s_x$  is the sample standard deviation for the problem
  - $n$  is the sample size and should match the number you entered

# Hypothesis Testing for Means in Your TI Calculator

- **When we know  $\sigma_x$ , with data**
- **INPUT:**
  1. Press STAT
  2. Press  $\rightarrow$  to TESTS
  3. Highlight '1: Z-Test' and Press ENTER
  4. **With Data**
    1. Enter the we're interested in next to ' $\mu_0$ :'
    2. Enter the population standard deviation next to ' $\sigma$ :'
    3. You should have your data table entered in L1
      - If you forgot: Press STAT, Press ENTER with 'Edit' highlighted, Enter the data into the L1 col.
    4. Next to 'List:' press 2<sup>nd</sup> then press 1
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# Hypothesis Testing for Means in Your TI Calculator

- **When we know  $\sigma_x$ , with data**
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  - Confirm the first line shows the hypothesis you would like to test
  - $z$  = the test statistic for our hypothesis test
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# Hypothesis Testing for Means in Your TI Calculator

- **When we know  $\sigma_x$ , with stats**
- **INPUT:**
  1. Press STAT
  2. Press  $\rightarrow$  to TESTS
  3. Highlight '2: T-Test' and Press ENTER
  4. **With Stats**
    1. Enter the we're interested in next to ' $\mu_0$ :'
    2. Enter the population standard deviation next to ' $\sigma$ :'
    3. Put the sample mean next to ' $\bar{x}$ :'
    4. Put the sample size next to 'n:'
    5. Select the appropriate alternative hypothesis on the ' $\mu$ :' line by highlighting the correct inequality and press ENTER
    6. Highlight Calculate and press ENTER

# Hypothesis Testing for Means in Your TI Calculator

- **When we know  $\sigma_x$ , with data**
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  - $z$  = the test statistic for our hypothesis test
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    - We make our decision based on this
  - $\bar{x}$  is the sample mean for the problem and should match the number you entered
  - $n$  is the sample size and should match the number you entered

# Confidence Intervals for Means

- **StatCrunch Commands w/ data**

- Stat → T Stats → One Sample  
→ with data (if you have the a list of data) → Choose the column → type the success value into the success box → choose hypothesis → enter the correct hypothesis → Compute

- **StatCrunch Commands w/ summaries**

- Stat → T Stats → One Sample  
→ with summary (if you have the count) → enter the number of success and total observations → enter the correct hypothesis → Compute

# Confidence Intervals known $\sigma_x$

Assumptions	Point Estimate	Margin of Error	Margin of Error
<ol style="list-style-type: none"> <li>1. <i>Random Sample</i></li> <li>2. <math>n &gt; 30</math> OR the population is bell shaped</li> </ol>	$\bar{x}$	$\sigma_{\bar{x}} = \frac{\sigma_x}{\sqrt{n}}$	$\bar{x} \pm z_{\frac{\alpha}{2}} \left( \frac{\sigma_x}{\sqrt{n}} \right)$

- We are --% confident that the true population mean lays on the confidence interval.

# Confidence Intervals unknown $\sigma_x$

Assumptions	Point Estimate	Margin of Error	Margin of Error
<ol style="list-style-type: none"> <li><i>Random Sample</i></li> <li><math>n &gt; 30</math> OR the population is bell shaped</li> </ol>	$\bar{x}$	$\sigma_{\bar{x}} = \frac{S_x}{\sqrt{n}}$	$\bar{x} \pm t_{1-\frac{\alpha}{2}, n-1} \left( \frac{S_x}{\sqrt{n}} \right)$

- We are --% confident that the true population mean lays on the confidence interval.

# Hypothesis Testing known $\sigma_x$

Step One:	<ol style="list-style-type: none"> <li>1. <math>H_0: \mu = \mu_0</math> &amp; <math>H_a: \mu \neq \mu_0</math></li> <li>1. <math>H_0: \mu \geq \mu_0</math> &amp; <math>H_a: \mu &lt; \mu_0</math></li> <li>2. <math>H_0: \mu \leq \mu_0</math> &amp; <math>H_a: \mu &gt; \mu_0</math></li> </ol>
Step Two:	<ol style="list-style-type: none"> <li>1. Quantitative</li> <li>2. <i>Random Sample</i></li> <li>3. <math>n &gt; 30</math> OR the population is bell shaped</li> </ol>
Step Three:	$z^* = \frac{(\bar{x} - \mu_0)}{\frac{\sigma_x}{\sqrt{n}}}$
Step Four:	$H_a: \mu \neq \mu_0 \rightarrow \text{p-value} = 2 * P(Z < - z^* )$ $H_a: \mu < \mu_0 \rightarrow \text{p-value} = P(Z < z^*)$ $H_a: \mu > \mu_0 \rightarrow \text{p-value} = P(Z > z^*) = 1 - P(Z < z^*)$
Step Five:	<p>If p-value <math>\leq (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Reject <math>H_0</math></p> <p>If p-value <math>&gt; (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Fail to Reject <math>H_0</math></p>

# Hypothesis Testing unknown $\sigma_x$

Step One:	<ol style="list-style-type: none"> <li>1. <math>H_0: \mu = \mu_0</math> &amp; <math>H_a: \mu \neq \mu_0</math></li> <li>1. <math>H_0: \mu \geq \mu_0</math> &amp; <math>H_a: \mu &lt; \mu_0</math></li> <li>2. <math>H_0: \mu \leq \mu_0</math> &amp; <math>H_a: \mu &gt; \mu_0</math></li> </ol>
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Step Three:	$t^* = \frac{(\bar{x} - \mu_0)}{\frac{s_x}{\sqrt{n}}}$
Step Four:	$H_a: \mu \neq \mu_0 \rightarrow \text{p-value} = 2 * P(T < - t^* )$ $H_a: \mu < \mu_0 \rightarrow \text{p-value} = P(T < t^*)$ $H_a: \mu > \mu_0 \rightarrow \text{p-value} = P(T > t^*) = 1 - P(T < t^*)$
Step Five:	<p>If p-value <math>\leq (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Reject <math>H_0</math></p> <p>If p-value <math>&gt; (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Fail to Reject <math>H_0</math></p>