Example: Do sophomores spend more time studying than freshmen?

Data: Random samples of students who responded to a survey question. Data can be downloaded at: <u>http://www.bradthiessen.com/html5/stats/m300/hourstudy.txt</u>



Interval estimation: Goal = estimate the difference $\mu_{\rm soph}$ - $\mu_{\rm fresh}$ with 90% confidence



Method 1: Bootstrap method

Applet: <u>http://lock5stat.com/statkey/bootstrap_1_quant_1_cat/bootstrap_1_quant_1_cat.html</u>

Result: 90% CI = (0.596, 6.134)

Method #2: Parametric (theory-based) method

Formula:
$$(\overline{X}_{s} - \overline{X}_{f}) \pm (t_{n_{s}+n_{f}-2}) \sqrt{\frac{1}{n_{s}} + \frac{1}{n_{f}}} \sqrt{\frac{(n_{s}-1)s_{s}^{2} + (n_{f}-1)s_{f}^{2}}{n_{s}+n_{f}-2}}$$

Applet for t-distribution: <u>http://lock5stat.com/statkey/theoretical_distribution/theoretical_distribution.html#t</u>



Result: 90% CI = (0.600, 6.102)

Note how similar this is to the CI from the bootstrap method

Null hypothesis significance testing: Goal = determine likelihood of observing data if the null hypothesis were true

Method 1: Randomization/Simulation method

There are two applets that will do this. The first one animates the shuffling of groups. Applet #1: <u>http://www.rossmanchance.com/applets/AnovaShuffle.htm?hideExtras=2</u>



Result: p-value = 0.0253





Result: p-value = 0.028

Method 2: Parametric (theory-based) method = Independent samples t-test

Sampling distribution:



Critical value:

Applet: <u>http://lock5stat.com/statkey/theoretical_distribution/theoretical_distribution.html#t</u> Result: t = 1.657 (same as what I did on page 2 of this example)

I can convert this critical t-value into a difference in averages using the formula:

$$(\mu_{s} - \mu_{f}) + (t_{n_{s}+n_{f}-2})(SE_{pooled}) = (0) + (1.657)(1.66) = 2.75$$

Observed value:

Result: The observed difference in averages is $\overline{X}_s - \overline{X}_f = 3.35$ I can convert this observed difference in averages to a t-score using the formula:

$$t = \frac{\left(\overline{X}_{s} - \overline{X}_{f}\right) - \left(\mu_{s} - \mu_{f}\right)}{SE_{pooled}} = \frac{3.35 - 0}{1.66} = 2.018$$

We can compare our observed and critical values to make a decision regarding the null hypothesis.

	Observed	Critical	
Difference in means	3.35	2.75	<– compare these
t-score	2.018	1.657	<- or compare these

Better yet, we can estimate a p-value:

p = P(observing our data or something more extreme | true null hypothesis)

$$p = P(\overline{X}_{s} - \overline{X}_{f} > 3.35) = P(t_{123} > 2.018) = 0.023 \text{ (see applet listed below to get this p-value)}$$

Applet: <u>http://lock5stat.com/statkey/theoretical_distribution/theoretical_distribution.html#t</u>



Note that the following applet will run a complete independent samples t-test for you, but it does <u>not</u> make the equal variance assumption.

Applet: http://www.rossmanchance.com/applets/TBIA.html



Theory-Based Inference

The results (p = 0.0255 and a CI between 0.5349, 6.1671) are similar to what we obtained with our bootstrap, randomization, and parametric methods, but the degrees of freedom (df = 104.88) are weird.

This is because this applet is actually conducting a Welch-Satterthwaite test (which was mentioned at the end of activity #17).

I don't recommend you use this applet on the test, just because you won't be able to answer any follow-up questions (such as questions related to power). Also, if you're not willing to make the homogeneity of variances assumption, why not just use bootstrap/randomization methods?

Assuming the true (actual) difference in study time is $\mu_s - \mu_f = 3$, estimate the power of our t-test.

Power = P(rejecting the null hypothesis | the null hypothesis is false) Looking at the sampling distribution and our critical value, this probability statement becomes:

$$P(\overline{X}_{s} - \overline{X}_{f} > 2.75 | \mu_{s} - \mu_{f} = 3)$$

= $P(t_{123} > \frac{2.75 - 3}{1.66}) = P(t_{123} > -0.15) = 0.559$ (p-value was obtained with the t-distribution applet)