## **Estimations**

### **EDP 613**

**Week 8**

## **A Note About The Slides**

Currently the equations may not show up properly in Firefox. Other browsers such as Chrome and Safari do appear to render them correctly.

## **A Note About Probability**

We're going to introduce some concepts from Chapter 8 here.

## **From To**

### **Descriptive Statistics**

mathematical techniques for organizing and summarizing a set of numerical data

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**Inferential Statistics**

generalizing from a sample to a population



- **Statistic** Mathematical expression that describes some aspects of a set of scores for a sample
- **Parameter** Describes some aspect of a set of scores for a population

## **First a Brief Intro to Hypothesis Testing**

- Formally Testing an assumption about a population parameter
- In Better Terms An assumption about a particular situation of the world that is testable

## **The Null Hypothesis**

- Represented as  $\bar{H}_0$
- is basically what you expect to happen before you run an experiment
- You have to know what the Null is!

## **The Alternative Hypothesis**

- Represented as  $H_1$  (or  $H_A$ )
- is basically what else could happen if what you expect doesn't occur
- You don't have to know this!

## **Tests of Statistical Significance**

- *Formally* Done to determine whether  $H_0$  or  $\overline{H_1}$  can be rejected
- Better Explanation Test to figure out whether you can reasonably say if your initial assumption won't happen
- Results If the outcomes of a study don't go against what you expected to happen, then you aren't finding anything new or surprising



A **(statistical) estimation** is a sample statistic is used to estimate the value of an unknown population parameter.

## **Idea of Positive and Negative Outcomes**

- The Null hypothesis  $\overline{H}_0$  is typically assuming nothing is going to happen
	- If  $H_0$  turns out to be right, then its called a *negative* outcome because nothing changed.
	- If  $H_1$  turns out to be right, then its called a *positive* outcome because something that you expected to happen didn't happen.
		- Experiment: Over the span of one year, a group of people with ADHD gets an experimental pill that may help them focus better than their current medication
			- $\overline{H}_0$ : Group stays the same (expected)
			- $H_A$ : Group is more focused (what we want to happen)
		- Results: After an assessment
			- if the Group doesn't show greater focus, then we have a **negative** outcome because that's what was expected to happen
			- if the Group shows greater focus, then we have a **positive** outcome because that's NOT what was expected to happen

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# **Notes about**  $H_0$  **and**  $H_A$

 $H_A$  is typically not the only alternative explanation

- What if the Group was found to more focused?
	- As a rule of thumb don't say that  $\bar{H_A}$  is correct unless you absolutely know there are two outcomes (aka *binary* outcomes)
	- Instead write that "we reject  $H_0$ " because you don't know if that's the ONLY alternative hypothesis.
		- $\circ$  It could also be that in other experiments that groups are found to be less focused!
- What if nothing happened to the Group?  $\bullet$

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- You can absolutely say that  $\overline{H}_0$  is correct because that's what you expected
- So you can write that "we accept  $H_0$ "

## **Formal Table of Statistical Error Types**



## **Nutshell Table of Statistical Error Types**

You changed your mind

**You changed your mind BUT the reality is you shouldn't have**

Results in a **False Positive / Type I Error** True Positive *True Positive* 

**Decision Your first thought was right Your first thought was wrong**

You changed your mind AND in reality that was the right decision

You didn't change your mind

You didn't change your mind AND in reality that was the right decision

**You didn't change your mind BUT the reality is that you should have**

Results in a True Negative **False Negative / Type II Error**

### **Example**





### Formally

- rejecting  $\bar{H_0}$  when it is true
- the probability of making a **Type I Error**

#### In Better Terms

- the chance of making the wrong decision when what was initially expected to happen actually occurs
- Given by  $\alpha$
- Ranges from 0-1 like all other probabilities

Typically  $\alpha = 0.05$  but its really context dependent



#### For airplanes

- if they fly people around, then when **analyzing failures**
	- you may want to lower the probability of making a wrong decision
	- use a **smaller**  $\alpha$
- if they're made of paper, then when **analyzing failures**
	- you might be willing accept the higher risk of making the wrong decision
- $\alpha$  and a set of the contract of the contrac

### **Beta**

### Formally

- not rejecting the  $H_0$  when  $H_1$  is true
- the probability of making a **Type II Error**

#### In Better Terms

- the chance of making the wrong decision when an something else actually occurs
- Given by  $\beta$
- Ranges from 0-1 like all other probabilities

### **Power**

- $1-\beta$  is called **statistical power**
- extremely important!
- Formally the probability of NOT making a Type II error
- In Better Terms the chance that you can separate if an outcome is a result of something occurring vs. pure luck!

## **Decision Making**



## **Decision Making**

Null  $\qquad \qquad H_0=\ ^$  "Forecast says its NOT going to rain" Alternative  $\vert H_1 = \vert$  "Something else will happen"



Note: You could have also gotten wet from snow, a flood, etc. so again **the alternative hypothesis generally does not imply the opposite!**

## **Estimation**

- **(Statistical) Estimation** a sample statistic is used to estimate the value of an unknown population parameter
	- **Point estimation** use of sample data to calculate a single value
	- **Interval estimation** use of sample data to calculate a possible range of values

#### Selecting a sample mean



## **Updating Estimation for Sample Means**

- **Point estimation** use of sample data to calculate a single **mean** value
	- $\circ$  Benefit the sample mean will equal the population mean on average
	- $\circ$  Drawback unable to figure out if a sample mean actually equals the population mean
- **Interval estimation** use of sample data to calculate a possible range of **mean** values

## **The Characteristic of Hypothesis Testing and Estimation**



## **Confidence**

- **Confidence Interval** an interval that contains an unknown parameter (e.g.  $\mu$ ) with certain degree of confidence
- **Level of Confidence** probability or likelihood that an interval estimate will contain an unknown population parameter

## **Determining the Confidence Interval**

1. Calculate the standard error of the mean

$$
\sigma_{\overline{Y}} = \frac{\sigma}{\sqrt{N}}
$$

2. Decide on a level of confidence





Again its typical to have a 95% level of confidence thereby making

 $\alpha = 0.05$ 

## **Determining the Confidence Interval (continued)**

3.  $CI=\overline{Y}\pm z\cdot \sigma_{\overline{Y}}$ 

4. Interpret the results

### **Example**

IQ scores in the general healthy population are approximately normally distributed with  $100\pm15$ . In a sample of 100 students a sample mean IQ of 103. Find the 90% confidence interval for this data.

Firstly we have  $N=100$ ,  $\mu=100$ ,  $\sigma=15$ , and  $\overline{Y}=103$ .

1.

$$
\sigma_{\overline{Y}}=\frac{\sigma}{\sqrt{N}}=\frac{15}{\sqrt{100}}=1.50
$$

2. Want to find 90% confidence interval, so choose a 90% level of confidence.

$$
z\cdot \sigma_{\overline{Y}}=1.645\cdot 1.50=2.47
$$

### $90\% CI = 103 \pm 2.47 = (105.47, 100.53)$

4. We are 90% confident that the overall mean IQ is between 100.53 and 105.47.

### **That's it. Take a break before our R session!**