**1. Scope of Inference.** It is very important to understand what conclusions can be drawn from statistical studies. This is called **scope of inference**.

Case 1: If we randomly sample individuals from a population WE CAD MAKE INFORMUSS

ABOUT THE POPULATION BASEN UPON THE SAMPLE.

Example: The US Census Bureau carries out a monthly Current Population Survey of about 60,000 households. Their goal is to use data from these randomly selected households to estimate the percent of unemployed individuals in the population.

What can be inferred from this study? WE CAN INFOR THE 90 OF UNEMPLOYED INDIVIDUALS IN THE POPULATION BASED UPON THE SAMPLE.

Case 2: If we randomly assign subjects in an experiment WE CAN MAKE INFORTACES

A BOOT CAUSE + EFFECT FOR THE POPULATION THAT SUBJECTS

CAME FROM (IF CHOSON RANDOMLY)

<u>Example</u>: Scientists performed an experiment that randomly assigned 21 <u>volunteer</u> subjects to one of two treatments: sleep deprivation for one night or unrestricted sleep. The <u>experimenters</u> hope to show that sleep deprivation causes a decrease in performance two days later.

What can be inferred from the study? WE CAN INFOR CAUSE + EFFECT FOR THIS GADIP OF VOLUNTEURS. WE ARE LIMITORS ON INFORMING ABOUT POPULATION SINCE VOLUNTARY SAMPLE.

Random selection of individuals from the population -----> allows inference from the population.

Random assignment of individuals to groups -----> allows inference about cause and effect.

((((Both random selection and random assignment introduce *chance* variation into a statistical study.

When performing inference, statisticians use the laws of probability to describe this chance variation.))))

		Were Individuals randomly assigned to groups?			
Were individuals randomly selected?		Yes		No	
	Yes	Inference about population:	46 S	Inference about population:	465
		Inference about cause/effect:	YES	Inference about cause/effect:	No
	No	Inference about population:	20	Inference about population:	NO
		Inference about cause/effect:	YES	Inference about cause/effect:	No

Team Work: Many students insist that they study better when listening to music. A teacher doubts this claim and suspects that listening to music actually hurts academic performance. Here are four possible study designs to address this question at your school. In each case the response variable is the students' GPA at the end of the semester. For each design, suppose that the mean GPA for students who listen to music while studying was significantly lower than the mean GPA of students who did not listen to music while studying. What can you conclude from each design?

1. Get all the students in your AP Statistics class to participate in a study. Ask them whether or not they study with music on and divide them into two groups based on their answer to this question.

NO RADDOM SELECTION -> ONLY APPLY TO AP STATS CLASS

11 ASSIGN. -> NO CAUSE/EFFECT.

AP STIDENTS WIND LISTEN TO MUSIC = LOWER GPA'S. WE DO NOT KDOW WIT.

2. Select a random sample of students from your school to participate in a study. Then divide them into groups as in Design 1.

RANDOM SELECTION -> STUDENTS @ SCHOOL! MUSIC => LOWER GPAS NO RANDOM ASSIGNMENT > DO NOT KNOW WHY GPAS ARE LOWER.

3. Get all the students in your AP Statistics class to participate in a study. Randomly assign half of the students to listen to music while studying for the entire semester and have the remaining abstain from listening to music while studying.

NO RADDOM SELECTION -> CANNOT TRADSFOR TO POP.

RADD ASSIEN -> WE CAN SAY CAUSE/EFFEUT BUT ONLY

FOR AP STATS CLASS

4. Select a random sample of students from your school to participate in a study. Randomly assign half of the students to listen to music while studying for the entire semester and have the remaining half abstain from listening to music while studying.

RANDOM SELECTION -> POPULATION /
RANDOM ASSIGNMENT -> CAUSE/EFFECT

1 = TUNCE IS A RELATIONSHIP BETWEEN MUSIC +

GPA @ THIS SCHOOL + THATE IS CAUSE +

EFFECT.

2. The Challenges of Establishing Causation - A well-designed experiment tells us that changes in the explanatory variable cause changes in the response variable. More precisely, it tells us that this happened for specific individuals in the specific environment of this specific experiment.

The serious threat is that the treatments, subjects, or the environment of the experiment is not *realistic*. Lack of realism can limit our ability to apply the conclusions of an experiment to the settings of greatest interest.

## **Examples**

- Determining if texting while driving causes accidents?
- Does smoking cause lung cancer?
- Does not wearing seatbelts cause more fatalities than wearing seatbelts?

= OBS.

What are the criteria for establishing causation when we cannot do an experiment?

- 1. The association is strong
- 2. The association is consistent
- 3. Larger values of the explanatory variable are associated with stronger responses.
- 4. The alleged cause precedes the effect in time.
- 5. The alleged cause is plausible.

HW: 87-94, 97, 99, 101, 103, 104

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