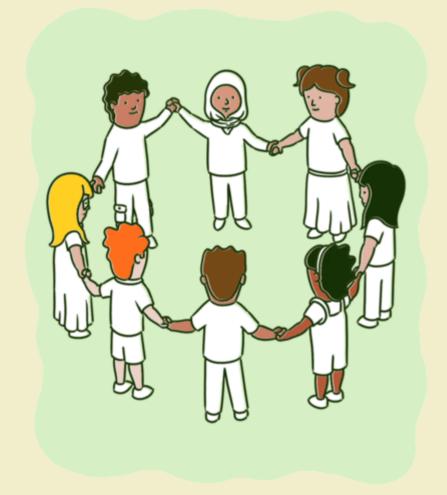
Review of Statistical Inference

Stat 230

March 30 2022





- Look for your group association in moodle
- Find your group mates, say hi and try to seat together
- Go to this google form
- Discuss your answer to the question posed in the next 4 slides, one by one.
- One member from a group enter the answer to questions

What is the difference between a population and a sample? What is the difference between a parameter and a statistic?

• Give an example of a parameter and a statistic, using "common" notation for each.

Suppose you are told a 95% confidence interval for the proportion of registered voters for Trump just before the 2020 election is 0.38 to 0.44.

- What does this mean?
- What is the margin of error for this CI?
- What does "95% confidence" mean?

A study found that participants who spent money on others instead of themselves had significantly lower blood pressure (two-sided p-value=0.012).

- What are the hypotheses for this test?
- What does the p-value mean?

Inference methods rely on understanding the **sampling distribution** of the statistic that we used to estimate our unknown parameter.

- What does the sampling distribution of the sample mean look like?
- What does the standard error of the sample mean measure?
- How is the sampling distribution used in a hypothesis test?
- How is it used with confidence intervals?

Data Analysis Plan

- 1. Data: load and clean/manipulate (if needed)
- 2. EDA: exploratory data analysis
- 3. Inference: run tests/CIs and check assumptions!
- 4. Conclusion: interpret results, avoid lots of stats jargon!

Real-life Data: agstrat

Data from a stratified random sample of size 300 from the U.S. 1992 Census of Agriculture.

Туре	Variable Description
Sampling	counties
Strata	region (North Central, Northeast, South, West)
response	number of farms in 1992 (farms92)

Do the average number of farms per county differ in the western and north central regions in 1992?

Read and inspect the data

```
agstrat <- read.csv("https://raw.githubusercontent.com/deepbas/statdatasets/main/agstrat.csv")</pre>
```

```
names(agstrat)
[1] "county" "state" "acres92" "acres87" "acres82" "farms92"
[7] "farms87" "farms82" "largef92" "largef87" "largef82" "smallf92"
[13] "smallf87" "smallf82" "region" "rn" "weight"
```

```
table(agstrat$region)

NC NE S W
103 21 135 41
```

Faceted Histograms

Summary stats by region

```
# std. deviation
tapply(agstrat2$farms92, agstrat2$region, sd)
          NC      W
358.0873 558.1294
```

Summary stats by region

We can also pipe (%>%) together dplyr commands to get the mean and sd of farms92 grouped by region

%>% passes result on left into first argument of function on right

Your Turn 2

Please download the class activity 2 .Rmd file from moodle and save it into your course project folder. Work in a group to answer the following questions.

Filter the West and North Central data

Make a boxplot and faceted histograms as you were shown in the previous slides.

Your Turn 3

Discuss how would you answer these questions with your group.

- What county has the unusually high number of farms in the western region?
- How many farms do they have?
- Any ideas why it is so large?
- How do the summary stats change when this county is omitted?

Outlier

```
which(agstrat2$farms92 > 3000)
[1] 118
```

```
agstrat2[118,] # see row 118

county state acres92 acres87 acres82 farms92 farms87 farms82

118 HAWAII COUNTY HI 926607 1007287 1172448 3157 2810 2539

largef92 largef87 largef82 smallf92 smallf87 smallf82 region rn weig

118 55 60 58 1960 1602 1468 W 142 10.292
```

Redo without outlier

```
agstrat2_noOutlier <- agstrat2 %>%
slice(-118) # remove row 118
```

```
agstrat2_noOutlier %>%
 group_by(region) %>% # group by region
 summarize(mean(farms92), sd(farms92), median(farms92), IQR(farms92))
# A tibble: 2 × 5
 region `mean(farms92)` `sd(farms92)` `median(farms92)` `IQR(farms92)`
 <chr>
                 <dbl> <dbl>
                                              <dbl>
                                                           <dbl>
1 NC
                 751.
                          358.
                                               738
                                                             479
2 W
                  538. 385.
                                              492.
                                                             464
```

Statistical Inference

After understanding the dataset better, we would like to infer useful information from the data using **statistical tests**.

- Use a relatively small sample of data to infer about the population
- May have to do extensive data cleaning and outlier detection and removal on a case-by-case basis

Statistical Inference: t-test

```
t.test(farms92 ~ region, data = agstrat2)
   Welch Two Sample t-test
data: farms92 by region
t = 1.5775, df = 53.618, p-value = 0.1206
alternative hypothesis: true difference in means between group NC and group
95 percent confidence interval:
-40.22256 336.89885
sample estimates:
mean in group NC mean in group W
       750.6796 602.3415
```

Statistical Inference

```
t.test(farms92 ~ region, data = agstrat2_noOutlier)
   Welch Two Sample t-test
data: farms92 by region
t = 3.0179, df = 66.775, p-value = 0.003601
alternative hypothesis: true difference in means between group NC and group
95 percent confidence interval:
 71.8464 352.5628
sample estimates:
mean in group NC mean in group W
       750.6796 538.4750
```

Your Turn 4

Run the t-test with and without the outlier and answer the given questions

Do the average number of farms per county differ in the western and north central regions in 1992?