# **Comparing Two or more Means**

**Stat 120** 

May 19 2023

### **Inference tools (Classical methods)**

#### Categorical Response

- 1. One proportion: sample z test/CI
- 2. Difference in 2 props: 2 sample
   z test/CI OR chi-square test
- 3. Association between 2 categorical variables: chi-square test

#### **Quantitative Response**

- 1. One mean: 1 sample t test/CI
- 2. Difference in 2 means: 2
   independent sample t test/CI OR
   Matched pairs
- 3. Compare >2 means: One-way ANOVA

### **Multiple Categories**

So far, we've learned how to do inference for a difference in means IF the categorical variable has only two categories (i.e. compare two groups)

In this section, we'll learn how to do hypothesis tests for a difference in means across multiple categories (i.e. compare more than two groups)

## **Hypotheses**

To test for a difference in true/population means across k groups:

 $H_0: \quad \mu_1=\mu_2=\ldots=\mu_k$ 

 $H_a: ext{ At least one } \mu_i 
eq \mu_j$ 

### **Frisbee Example**

### Does Frisbee grip affect the distance of a throw?

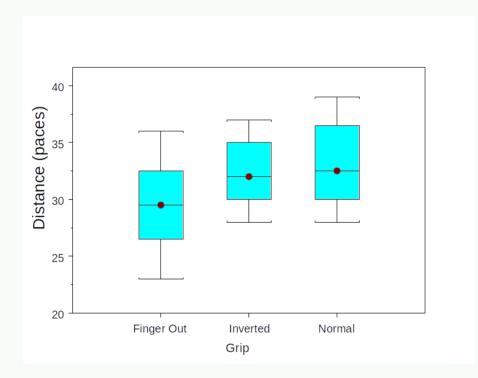
A student performed the following experiment: 3 grips, 8 throws using each grip

- 1. Normal grip
- 2. One finger out grip
- 3. Frisbee inverted grip

A grip type is randomly assigned to each of the 24 throws she plans on making

- Response: measured in paces how far her throw went
- Question: How might you summarize her data?

## **Frisbee Example**



	Finger-out	Inverted	Normal
$\left[\begin{array}{c} n \end{array}\right]$	8	8	8
Mean	29.5	32.375	33.125
$\left[\begin{array}{c}SD\end{array}\right]$	4.175	3.159	3.944

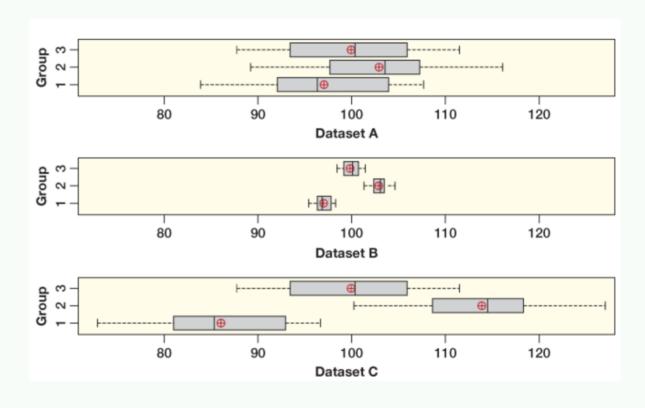
### Question: Is this evidence that grip affects mean distance thrown?

 $H_0: \quad \mu_1 = \mu_2 = \mu_3$ 

 $H_a$ : At least one  $\mu_1, \mu_2, \mu_3$  is not the same

### Why Analyze Variability to Test for a Difference in Means?

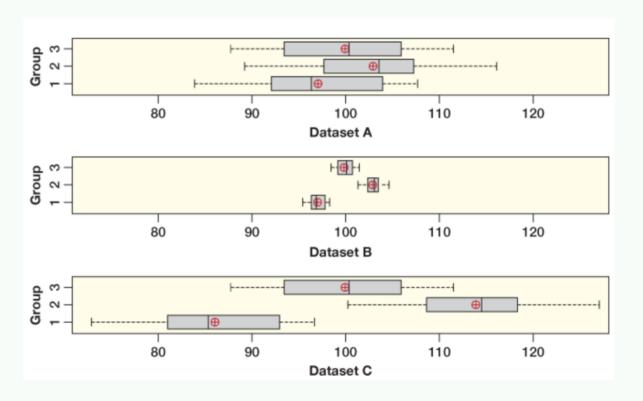
- The group means in Datasets A and B are the same, but the boxes show different spread.
- Datasets A and C have the same spread for the boxes, but different group means.



Which of these graphs appear to give strong visual evidence for a difference in the group means?

### Why Analyze Variability to Test for a Difference in Means?

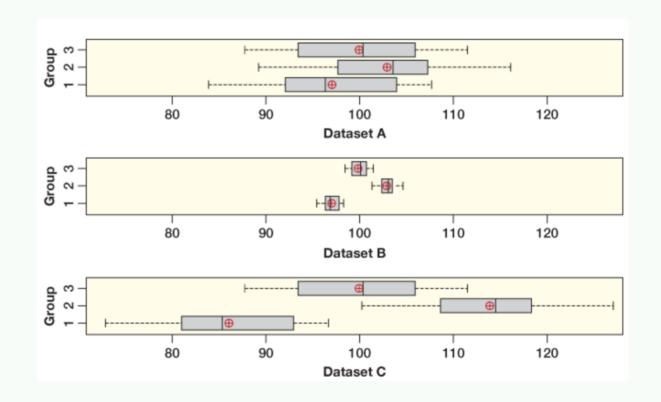
- Dataset A = weakest evidence for a difference in means.
- Datasets B and C = strong evidence for a difference in means.



### Why Analyze Variability to Test for a Difference in Means?

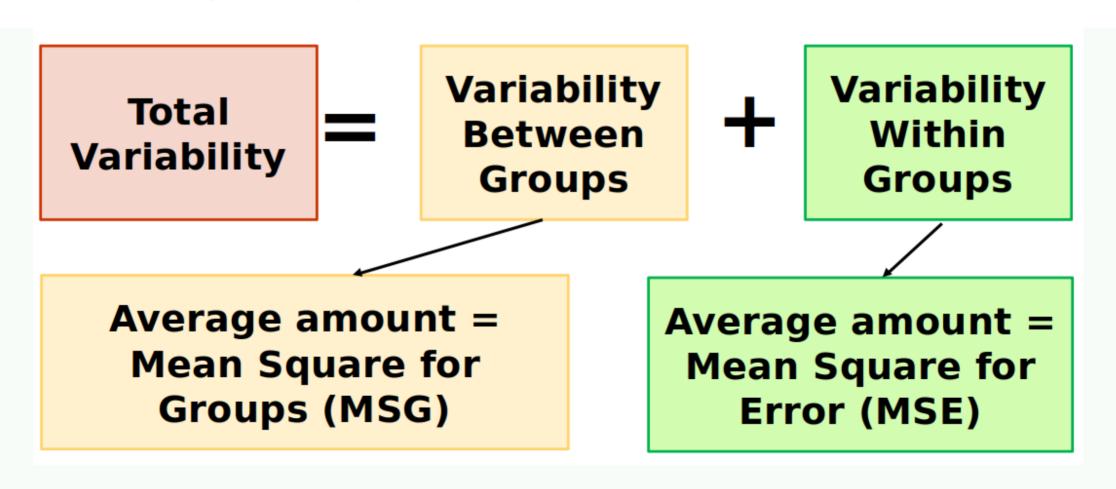
Conclusion: An assessment of the difference in means between several groups depends on two kinds of variability:

- 1. How different the means are between each groups
- The amount of variability within each groups



## **Analysis of Variance**

Analysis of Variance (ANOVA) compares the variability between groups to the variability within groups



#### **F-Statistic**

#### The F-statistic is a ratio:

$$F = \frac{MSG}{MSE} = \frac{\text{average between group variability}}{\text{average within group variability}}$$

If there really is a difference between the groups  $(H_A \ {
m true})$ , we would expect the F-statistic to be

- a) Large positive
- b) Large negative
- c) Close to 0
- ► Click for answer

### **Frisbee Example**

```
frisbee <- read.csv("https://raw.githubusercontent.com/deepbas/statdatasets/main/Frisbee.csv")
frisbee.anova <- aov(Distance ~ Grip, data = frisbee) # fit an ANOVA model</pre>
```

```
      summary(frisbee.anova)

      Df Sum Sq Mean Sq F value Pr(>F)

      Grip 2 58.58 29.29 2.045 0.154

      Residuals 21 300.75 14.32
```

F-test statistic: 2.045

P-value: 0.154

### How to determine significance?

We have a test statistic. What else do we need to perform the hypothesis test?

A distribution of the test statistic assuming  $H_0$  is true.

How do we get this? Two options:

- 1. Simulation
- 2. Theory

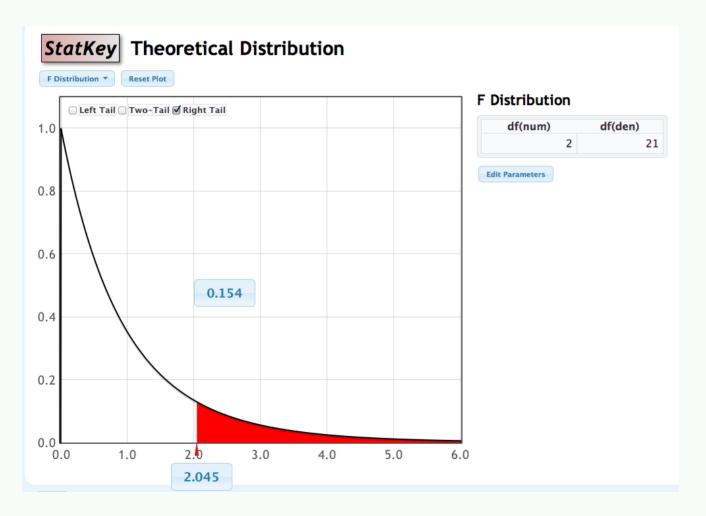
#### **F-Distribution**

We can use the F-distribution to generate a p-value if:

- 1. Sample sizes in each group are large (each  $n_i \geq 30$  ) OR the data within each group are relatively normally distributed
- 2. Variability is **similar** in all groups

- The F-distribution has two degrees of freedom, one for the numerator of the ratio (k-1) and one for the denominator (n-k)
- For F-statistics, the p-value (the area as extreme or more extreme) is always the right tail

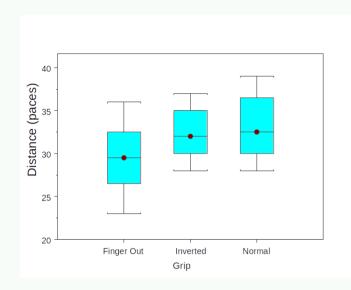
#### **F-distribution**



- An F-statistic as large as 2.045 would occur by chance about 16% of the time if the means were all equal.
- Our results are inconclusive and do not support the claim that grips affects average distance.

```
1 - pf(2.045,2,21) # p-value
[1] 0.1543639
```

### **Check assumptions: large sample size (or normality)**



```
table(frisbee$Grip) # check n's
Finger Out Inverted Normal
8 8 8
```

Small  $n_i$  but all groups are roughly normal

```
# checking normality with qq-plots
ggplot(frisbee, aes(sample = Distance)) +
  geom_qq() + geom_qq_line() + facet_wrap(~Grip) +
  theme(axis.text.x = element_text(size = 4))
```

### **Check Assumptions: Equal Variance**

The F-distribution assumes equal within group variability for each group. This is also an assumption when using the randomization distribution.

 As a rough rule of thumb, this assumption is violated if the largest group standard deviation is more than double the smallest group standard deviation

tapply(frisbee\$Distance, frisbee\$Grip, sd)

Finger Out Inverted Normal 4.174754 3.159453 3.943802

Check if:

$$\frac{\text{largest } s}{\text{smallest } s} > 2$$

## **Frisbee Example: Inference**

Question: Is this evidence that grip affects mean distance thrown?

 $H_0: \ \ \mu_1 = \mu_2 = \mu_3$ 

 $H_a$ : At least one  $\mu_1, \mu_2, \mu_3$  is not the same

 $\mu_{
m i}$  is the true mean distance thrown using grip i .

$$F = 2.05(\mathrm{df} = 2, 21), \; ext{P-value} \; = 0.1543$$

Conclusion: Do not reject the Null hypothesis. The difference in observed means is not statistically significant.

About 15% of the time we would see the grip differences like those observed, or even bigger, when there is actually no difference between the true mean distances thrown with different grips.

# **Picturing the variation**

Green: Variation within groups

Blue: Variation between groups

## **ANOVA Table for Frisbee data**

F test stat = 
$$29.29/14.32 = 2.045$$

Source	df	Sum of Squares	Mean Square
Groups	#groups -1	SSG	SSG/df
	<b>3-1 = 2</b>	<b>58.583</b>	58.583/2 = 29.29
Error (residual)	n - #groups	SSE	SSE/df
	<b>24-3 = 21</b>	<b>300.750</b>	300.75/21= 14.32
Total	n-1 <b>24-1 = 23</b>	SSTotal <b>359.333</b>	

### Frisbee Example: ANOVA table in R

frisbee.anova <- aov(Distance ~ Grip, data = frisbee)</pre>

```
summary(frisbee.anova)

Df Sum Sq Mean Sq F value Pr(>F)

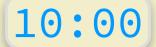
Grip 2 58.58 29.29 2.045 0.154

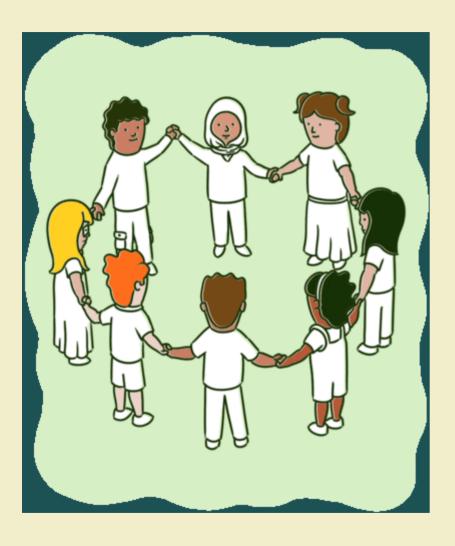
Residuals 21 300.75 14.32
```

```
library(broom)
knitr::kable(tidy(frisbee.anova)) # nicer summary tables
```

term	df	sumsq	meansq	statistic	p.value
(Grip	$\begin{bmatrix} 2 \end{bmatrix}$	58.58333	29.29167	2.045303	0.1543247
Residuals	21	300.75000	14.32143	NA	NA







- Go over to the in class activity file
- Complete the activity as much as possible