

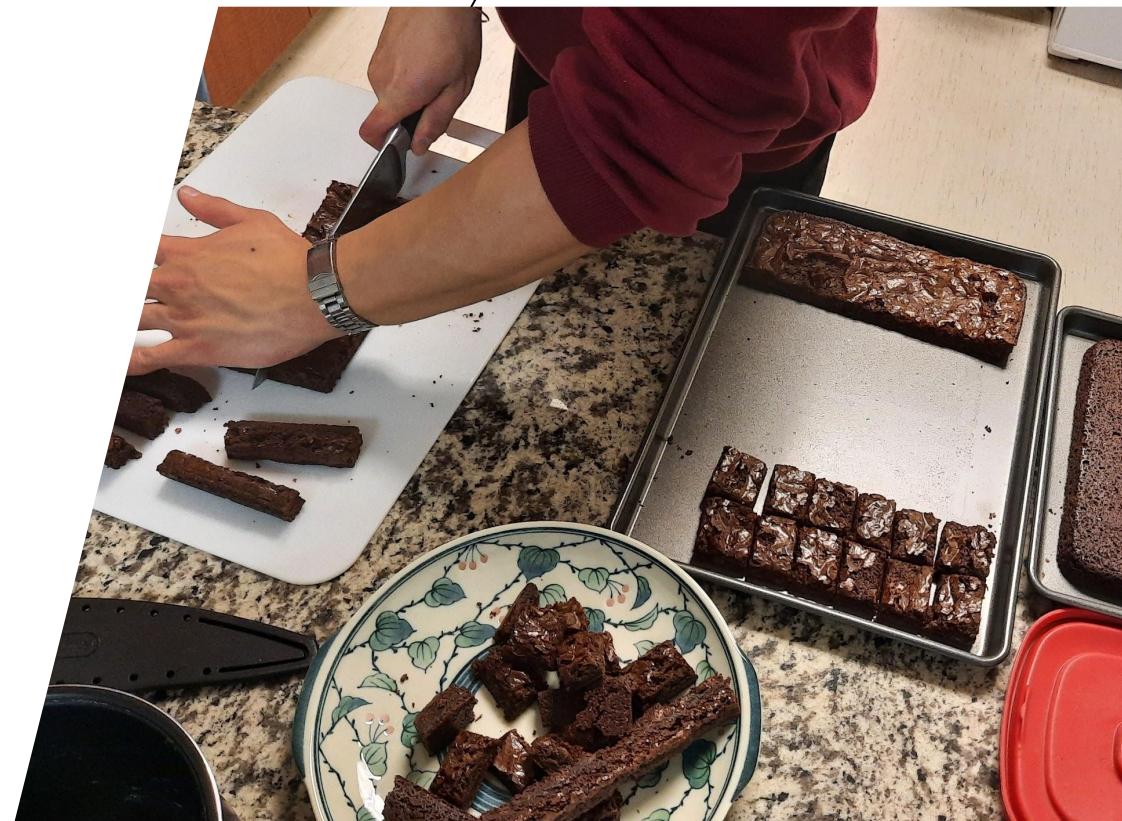
THE BEST BROWNIES

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INTRODUCTION

The price of brownie mix can significantly vary, but it remains unclear if it impacts the taste or overall satisfaction of the baked product. For this study, we will examine eight treatments, including male versus female participant, two types of brownie mix, and the time spent in the oven. The aim of this study is to determine the effect of these factors on the taste and overall satisfaction of the brownies.



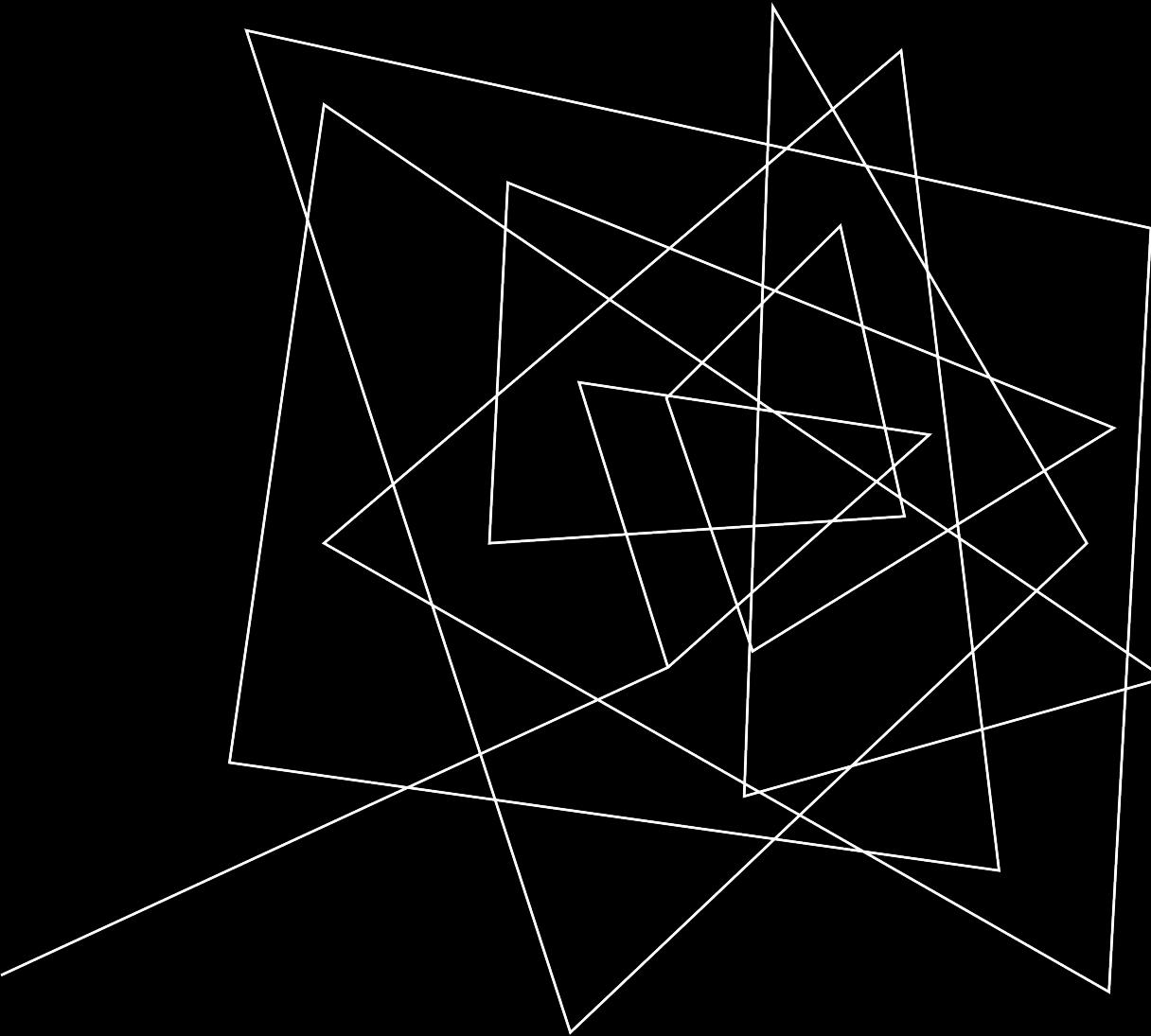
RESEARCH QUESTIONS



Brownie mix varies wildly in price, but does it really affect taste or satisfaction?

Do people have consistent preferences when it comes to leaving brownies in the oven longer?

Does a person's gender affect which brownie they prefer?



HYPOTHESIS

Null Hypothesis: There is no difference in satisfaction from a brownie with varying levels of gender, time in the oven, or brownie mix used.

Alternative Hypothesis: At least one of the test means for these varying treatments is different.



EXPERIMENTAL DESIGN

- Originally Complete block design: CB[3]
- Three factors
 - Brand - Ghirardelli and Betty Crocker
 - Oven time - Normal and 5 minutes longer
 - Gender
- Switched to BF[2] after factoring out blocking variable (More on this later)
 - Brand
 - Oven time

R Code Used for Randomization

```
1 library(tidyverse)
2
3 factors = c("Brand", "OvenTime", "Gender")
4 Brand = c("BettyCrocker", "Ghirardelli")
5 OvenTime = c("Normal", "Extra")
6 Gender = c("Male", "Female")
7 combinations = expand.grid(Brand, OvenTime, Gender)
8 combinations = combinations %>% setNames(factors)
9 combinations["ID"] = 1:(length(Brand)*length(OvenTime)*length(Gender))
10 combinations = combinations[c(4,1:3)]
11
12 combinations = combinations %>% as_tibble()
13 combinations
14
15 scale = 10; difference = 1.5
16
17 treatments = nrow(combinations)
18 brownie_scores = c(rep(0, treatments-1), difference)
19 #We want to detect a different means significantly by 1.5 points
20 between_var = var(brownie_scores)
21
22 within_var = ((scale-(scale/5))/6)^2
23 #We assume the 99% range is roughly around 8 because
24 #people will give ratings at least 2 or higher
25 n = ceiling(power.anova.test(groups=treatments,
26                               between.var = between_var,
27                               within.var = within_var,
28                               sig.level = 0.05,
29                               power=0.80)$n)
30
31 replications = rep(1:n, max(combinations["ID"]))
32 order = 1:length(replications)
33
34 data = tibble(
35   "Order" = order,
36   "Person" = rep(1:28, each=4),
37   "Treatment" = rep(1:nrow(combinations), n)
38 )
39 data = data[sample(1:nrow(data)),] #Randomize the treatments
40 data = mutate(data,
41               Brand=combinations$Brand[data$Treatment],
42               OvenTime=combinations$OvenTime[data$Treatment],
43               Gender=combinations$Gender[data$Treatment]) %>% arrange(Person)
```

RESULTS (RAW)

Power Analysis - Determining Sample Size

```
15 scale = 10; difference = 1.5
16
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29                               power=0.80)$n)
```

```
Balanced one-way analysis of variance power calculation

groups = 8
      n = 13.86214
between.var = 0.28125
within.var = 1.777778
sig.level = 0.05
power = 0.8
```

NOTE: n is number in each group

```
> 14*8
[1] 112
>
```

```
> combinations
# A tibble: 8 × 4
      ID Brand   OvenTime Gender
    <int> <fct>   <fct>   <fct>
1     1 BettyCrocker Normal  Male
2     2 Ghirardelli  Normal  Male
3     3 BettyCrocker Extra   Male
4     4 Ghirardelli  Extra   Male
5     5 BettyCrocker Normal Female
6     6 Ghirardelli  Normal Female
7     7 BettyCrocker Extra   Female
8     8 Ghirardelli  Extra   Female
```

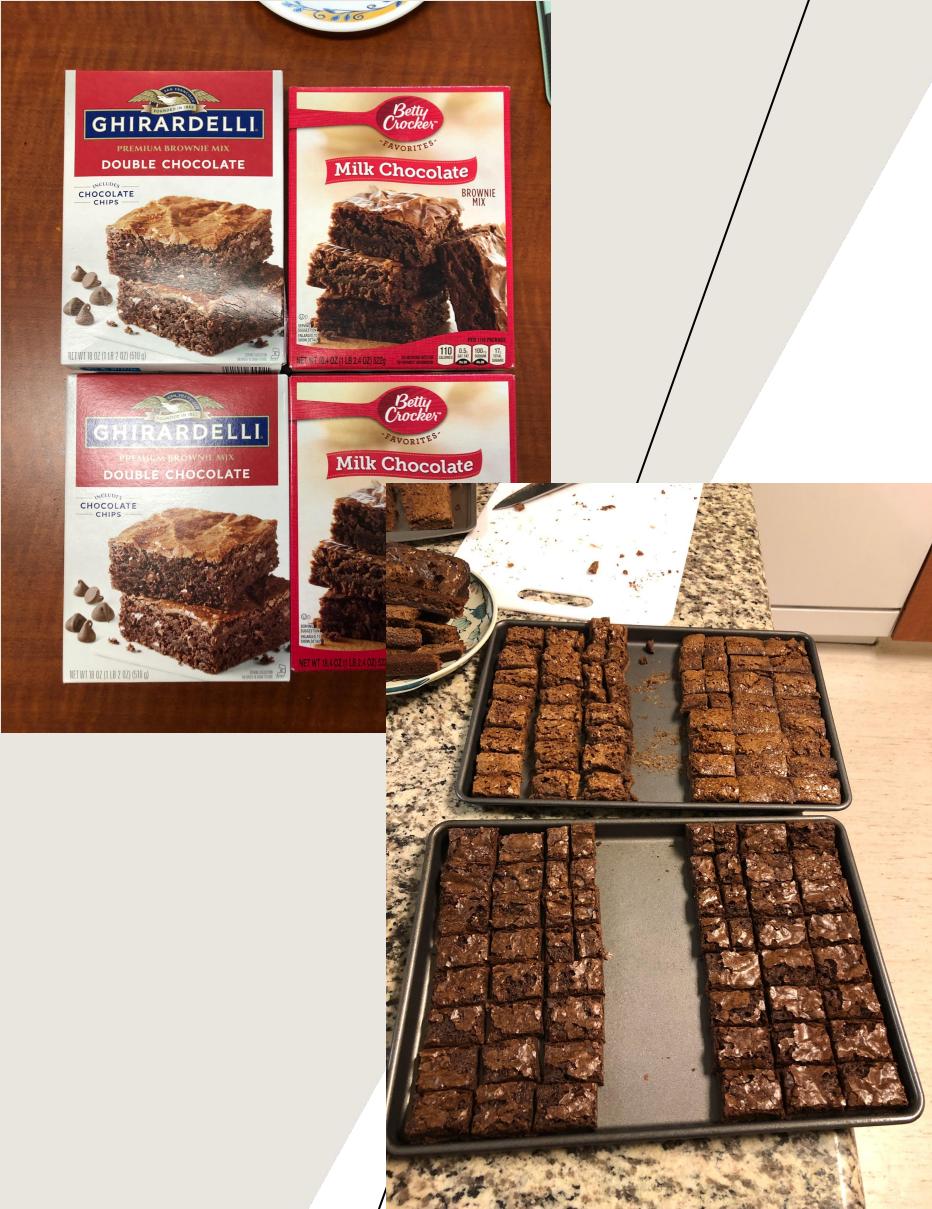
EXPERIMENTAL PROCESS

- Pan size
- Temperature (post cooling)
- Temperature (oven)
- Container and utensils used for mixing
- Size of the brownie samples
- Edge vs. Center
- Cooking spray



EXPERIMENTAL PROCEDURES

- preheating
- mix preparation (brand, container)
- baking (by brand)
- cooling
- trimming
- portioning

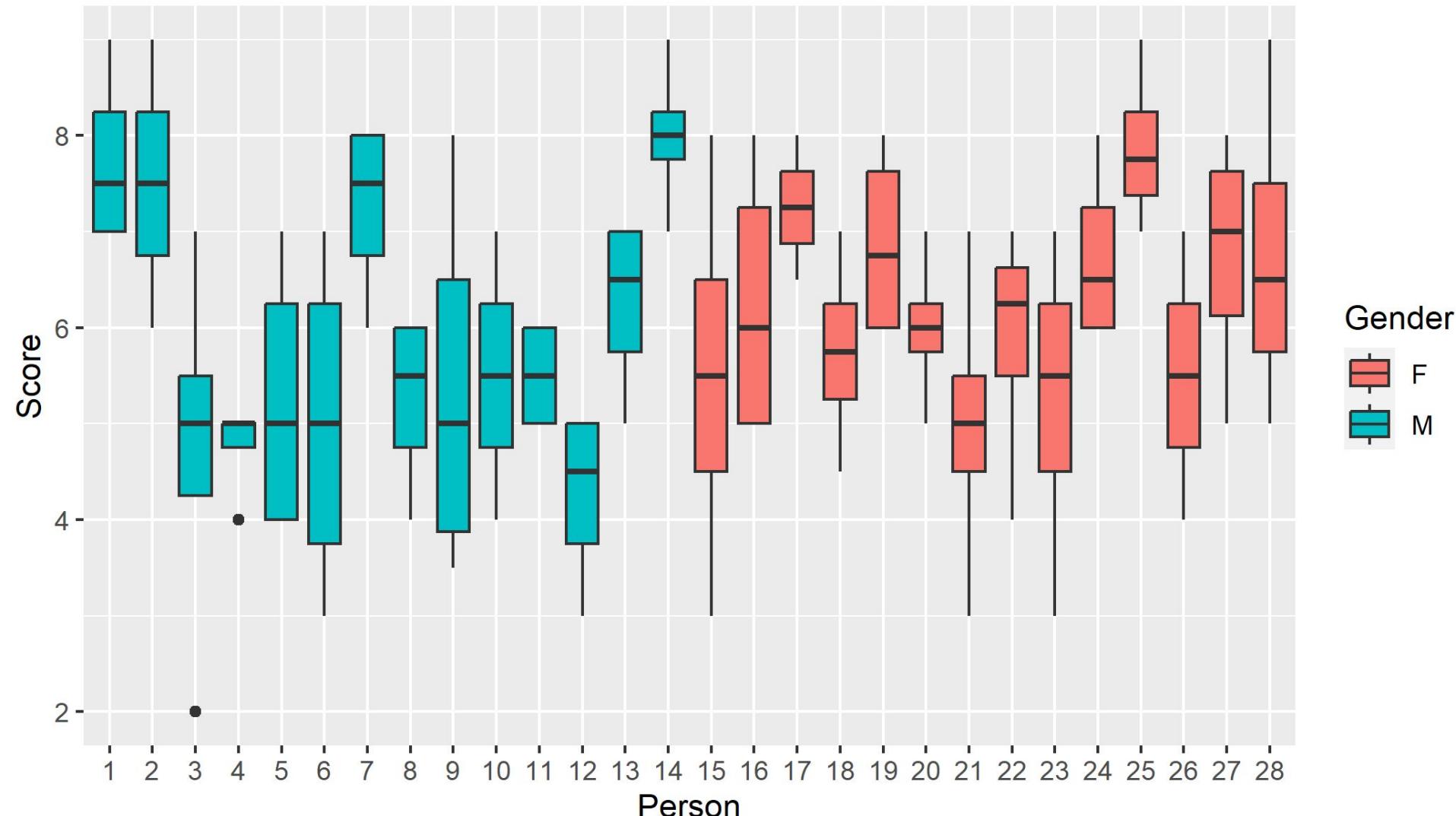


DATA COLLECTION

- 10-point scale for enjoyment
- Randomized trials and treatment orders
- Multiple locations
- Same time period
- Use of a palate cleanser



Distribution of Scores Across People



TRANSFORMATION OF DATA

Can we assume the population variances are equal?

```
85 stds = data %>% group_by(Person) %>%
86   summarize(
87     std = sd(Score)
88   )
89 |
90 stds
91
92 #This is greater than 2
93 max(stds$std)/min(stds$std)

> max(stds$std)/min(stds$std)
[1] 4.163332
```

A log transformation does not work with our data

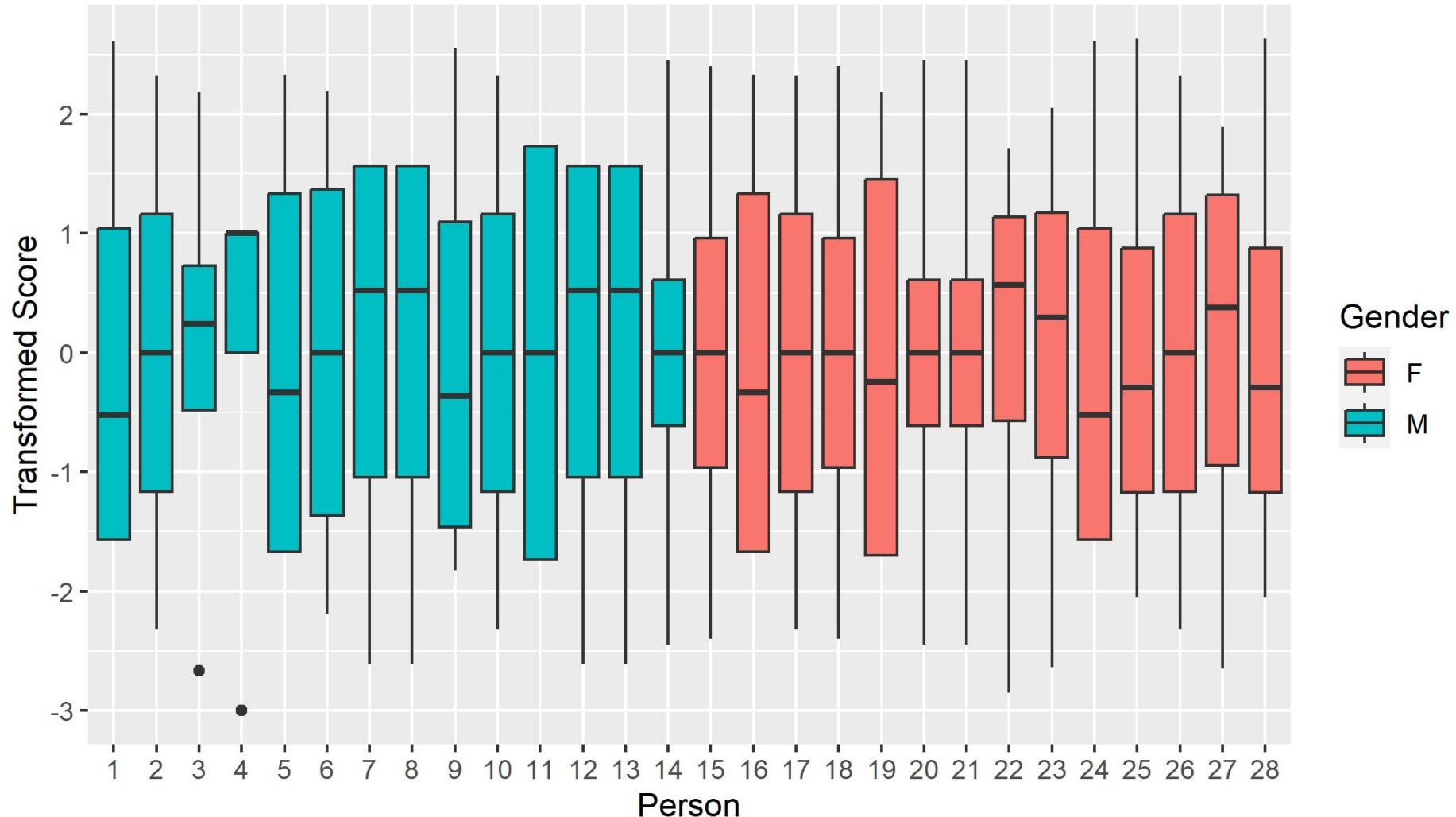
```
#This does not work with our data
log_stds = data %>% group_by(Person) %>%
  summarize(
    std = sd(log(Score))
  )

#This is also greater than 2
max(log_stds$std)/min(log_stds$std)

> #This is also greater than 2
> max(log_stds$std)/min(log_stds$std)
[1] 6.021513
>
```

Z TRANSFORMATION

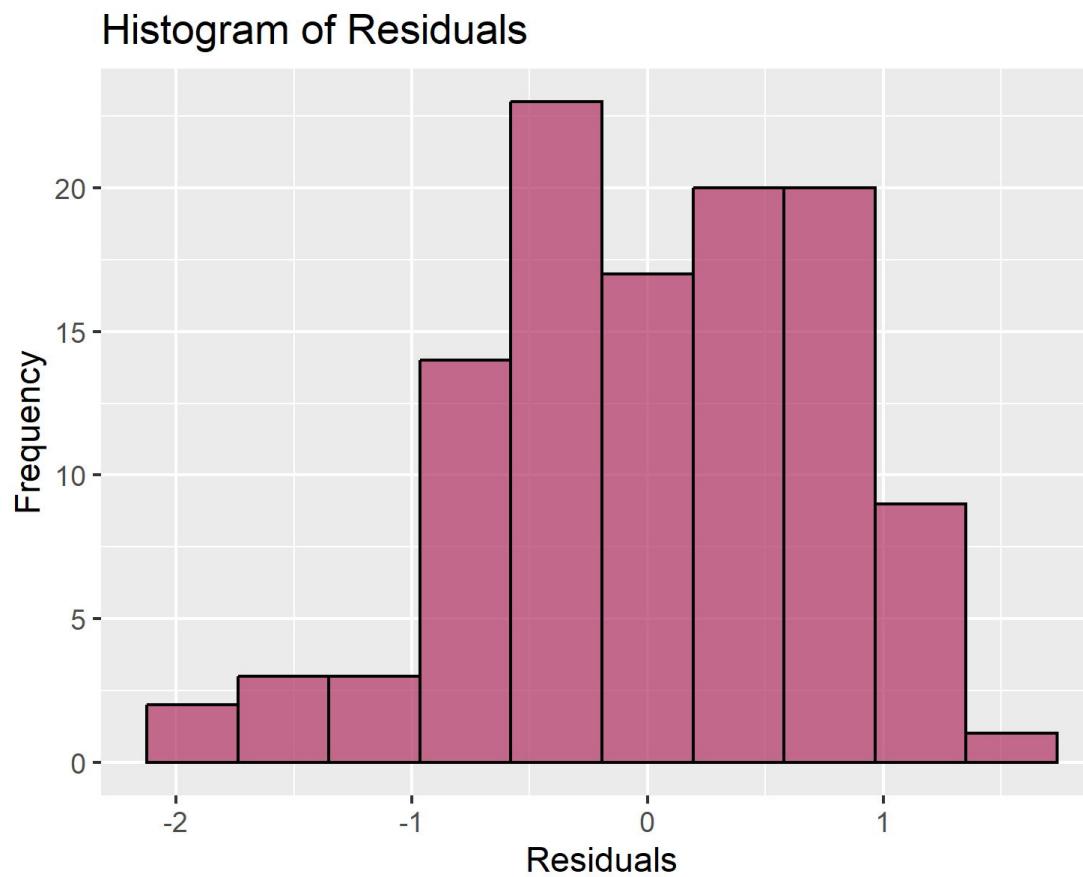
```
105 #Z Transformation  
106 data = data %>% group_by(Person) %>% mutate(  
107   z = (Score-mean(Score))/(sd(Score))  
108 )
```



STATISTICAL MODEL

```
fit <- lm(Z ~ Brand + OvenTime + Brand:OvenTime, data = data)
```

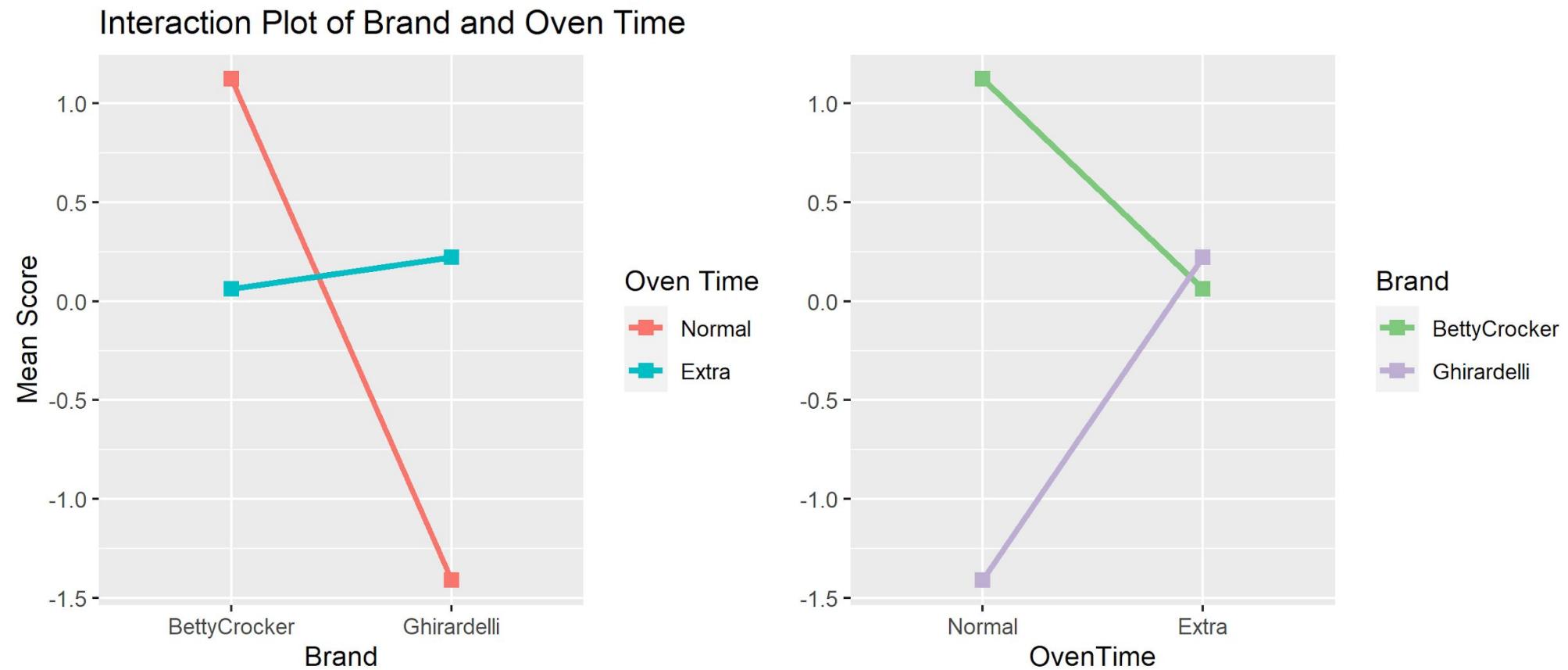
```
results = anova(fit)  
results
```



ANOVA TABLE

	Df	Sum of Squares	Mean Sq	F value	P(>F)
Brand	1.0000	9.8658	9.8658	17.5067	0.0001
OvenTime	1.0000	0.5674	0.5674	1.0068	0.3179
Brand:OvenTime	1.0000	12.7044	12.7044	22.5440	0.0000
Residuals	108.0000	60.8624	0.5635		

INTERACTION PLOTS

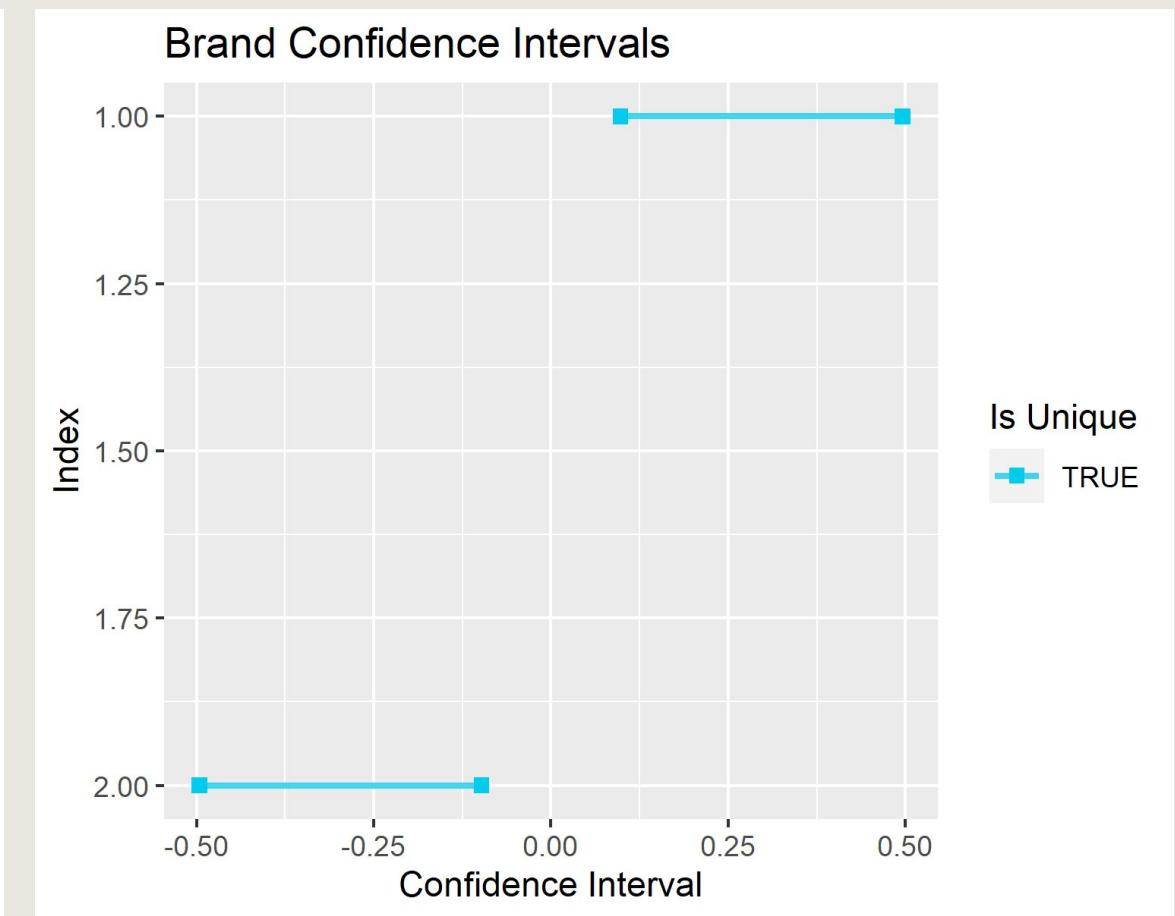
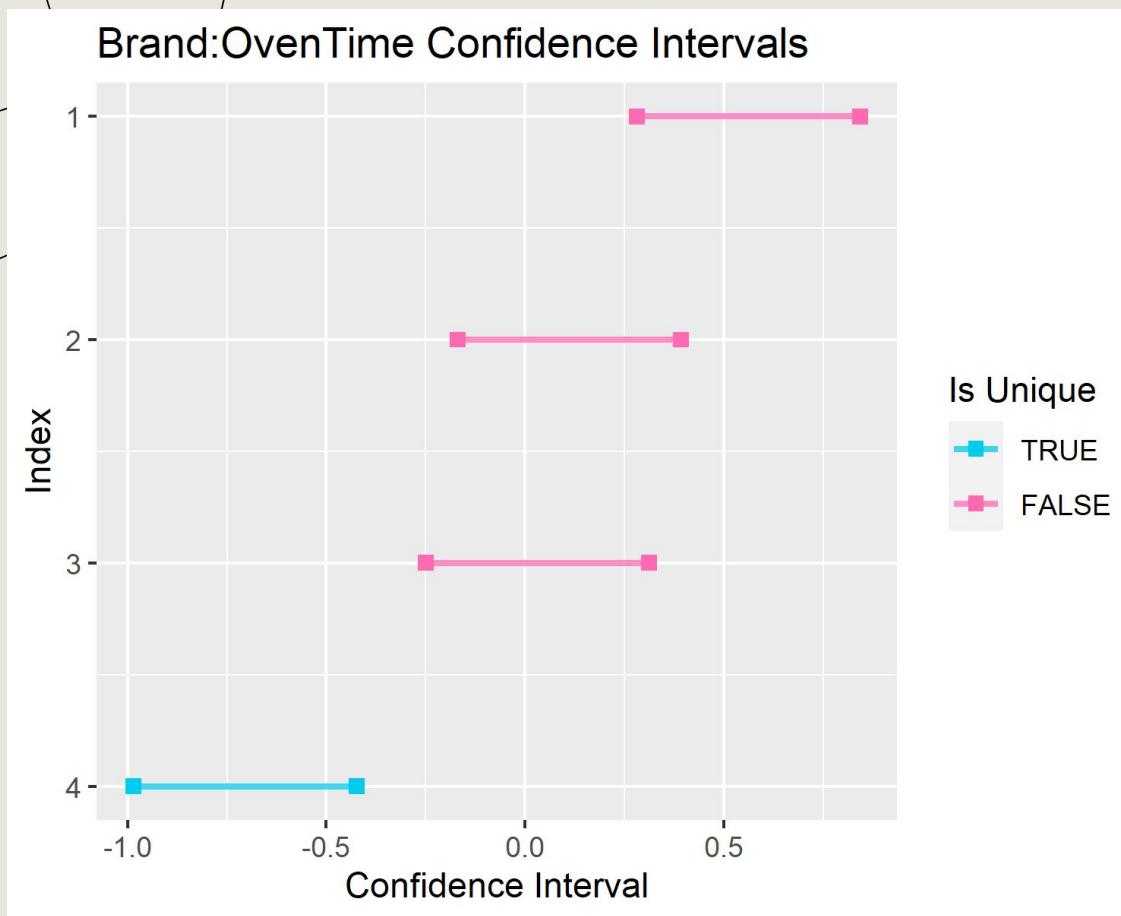


CONFIDENCE INTERVALS

	Brand	Mean	SE	df	lower.CL	upper.CL
1	BettyCrocker	0.2968	0.1003	108	0.0980	0.4956
2	Ghirardelli	-0.2968	0.1003	108	-0.4956	-0.0980

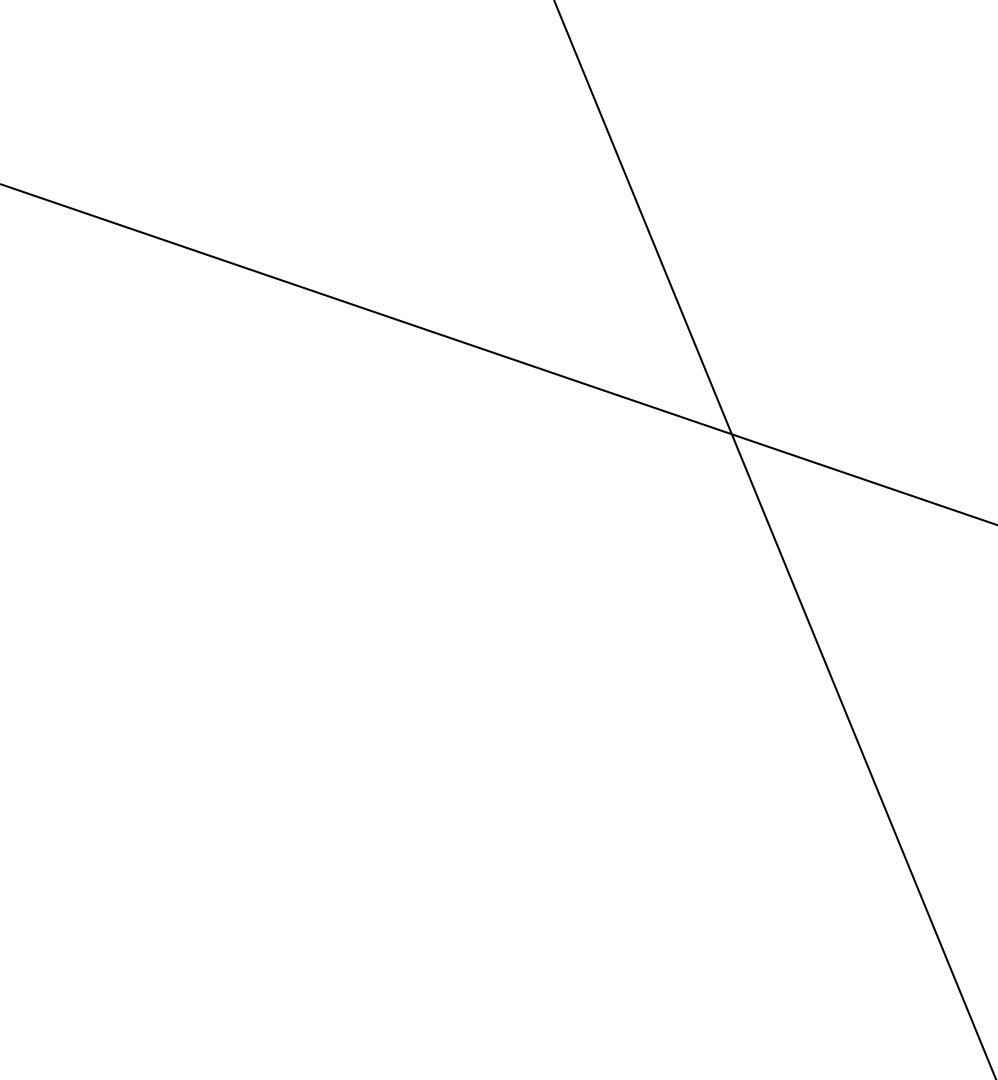
	Brand	OvenTime	Mean	SE	df	lower.CL	upper.CL
1	BettyCrocker	Normal	0.5624	0.1419	108	0.2812	0.8436
2	Ghirardelli	Extra	0.1112	0.1419	108	-0.1700	0.3924
3	BettyCrocker	Extra	0.0312	0.1419	108	-0.2500	0.3124
4	Ghirardelli	Normal	-0.7048	0.1419	108	-0.9860	-0.4236

CONFIDENCE INTERVALS



INTERPRETATION

- Failed to reject the null hypothesis regarding oven time. We did not observe a significant difference in the preference for baking time in the oven.
- Rejected the null hypothesis regarding brand. We found significant evidence that there is a difference in the preference for Betty Crocker brownies and Ghirardelli.
- Rejected the null hypothesis regarding the interaction between oven time and brand. We found significant evidence to conclude that at least one group (Ghirardelli brownies baked at normal time in this case) differs from the rest.



CONCLUSION

The findings may be useful for consumers and bakers who want to make a decision about which brownie mix to purchase and how long to bake their brownies.

FUTURE STUDIES

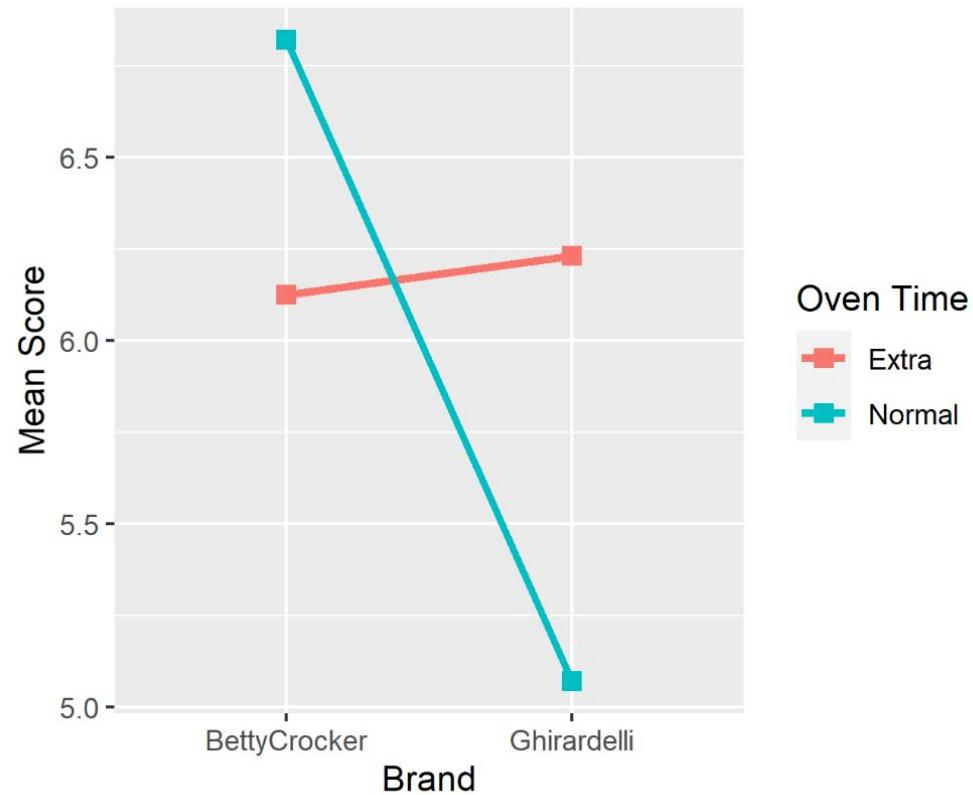
- More types of brownie mix
- More levels for oven time
- Different types of palate cleansers
- Different pan sizes / material (glass)
- Different time period (morning, afternoon, evening)

APPENDIX

Had we done the analysis without transforming the data...
This relies on the critical assumption the population variances are equal notwithstanding the deviation that we observed between the scores.

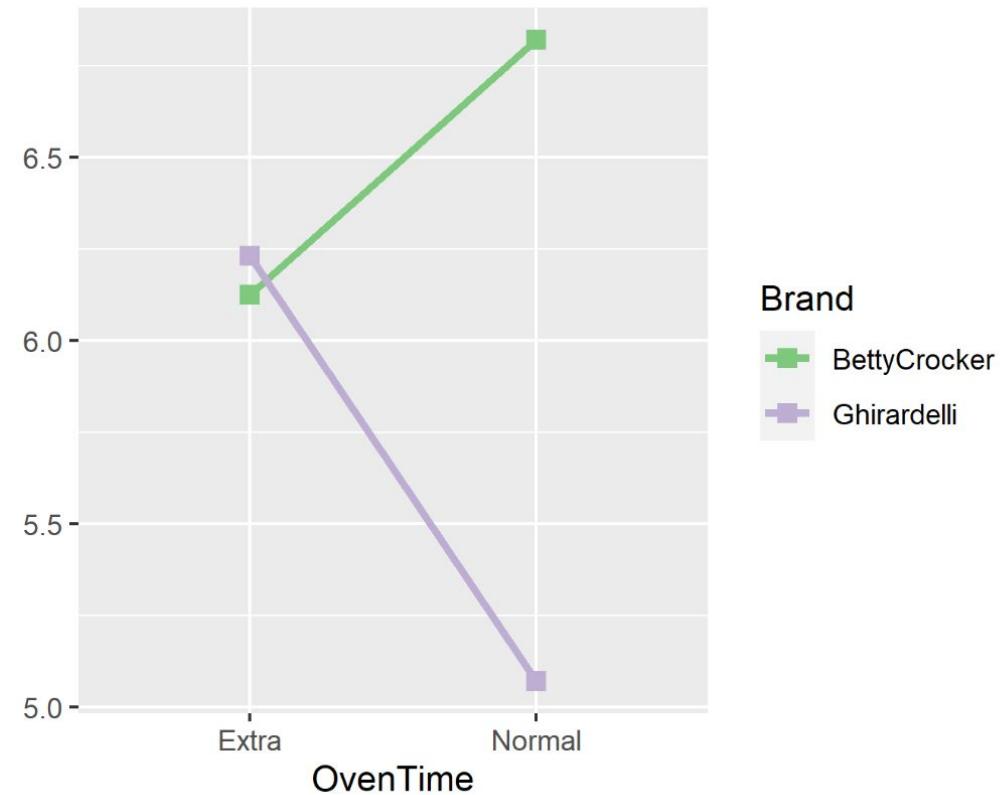
INTERACTION PLOTS

Interaction Plot of Brand and Oven Time



Oven Time

- Extra
- Normal

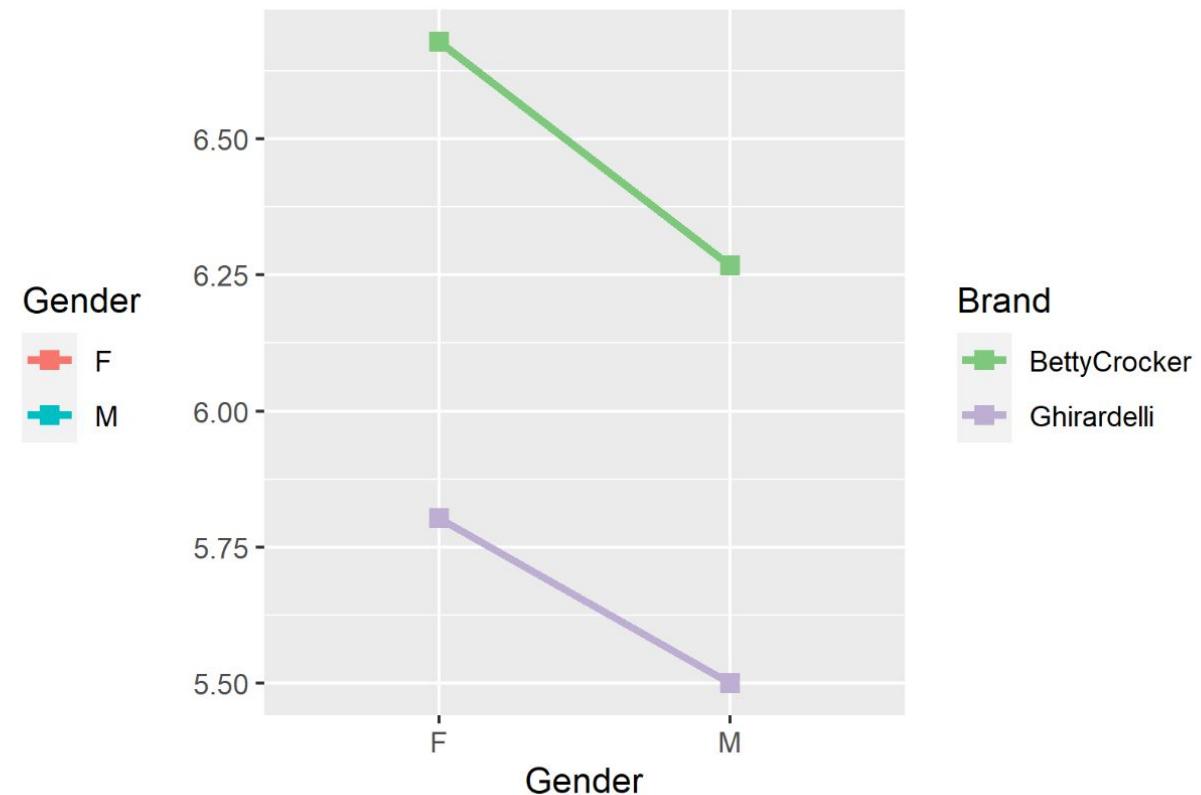
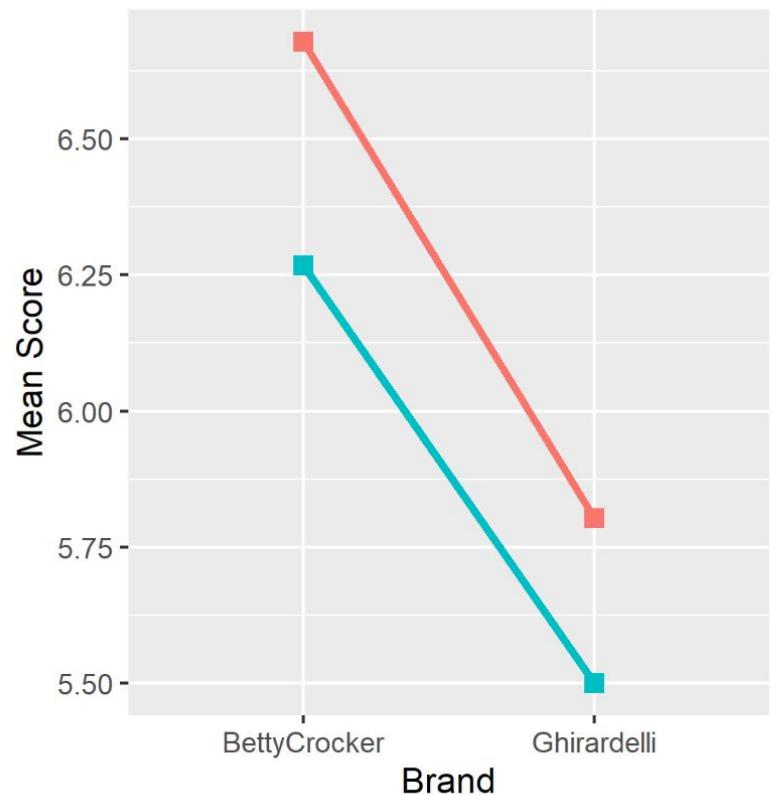


Brand

- Betty Crocker
- Ghirardelli

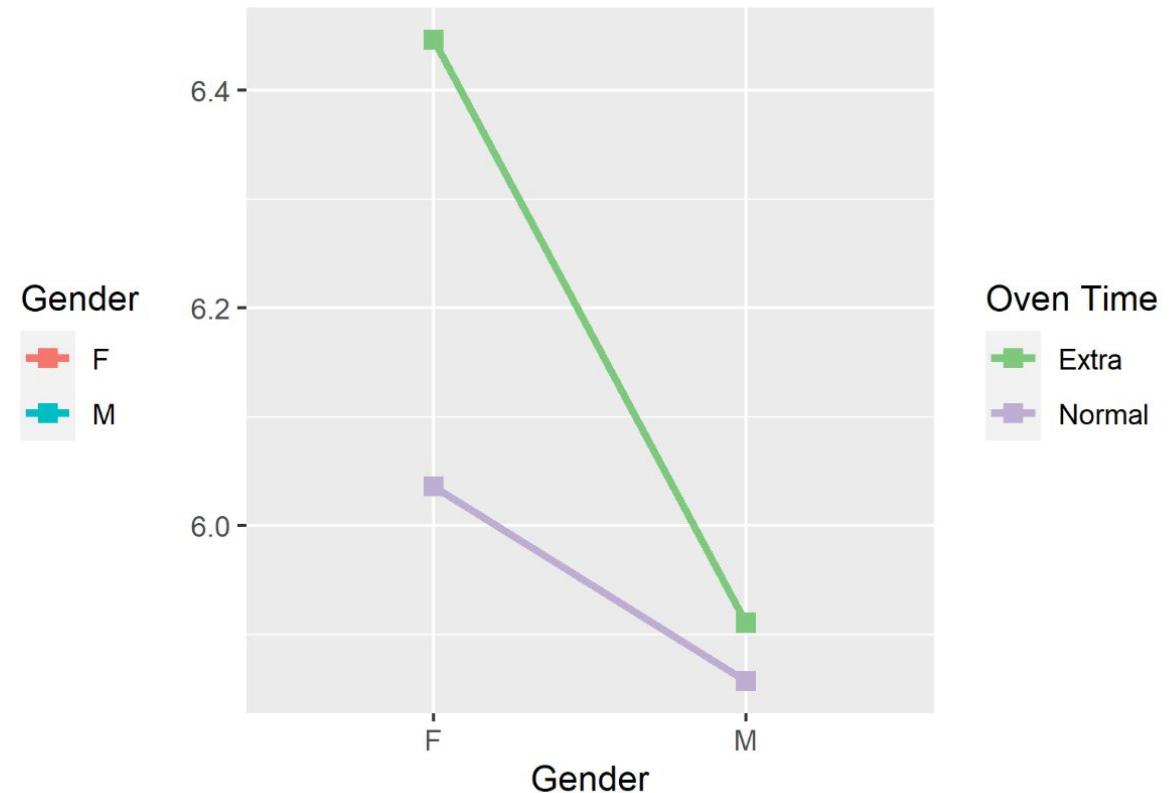
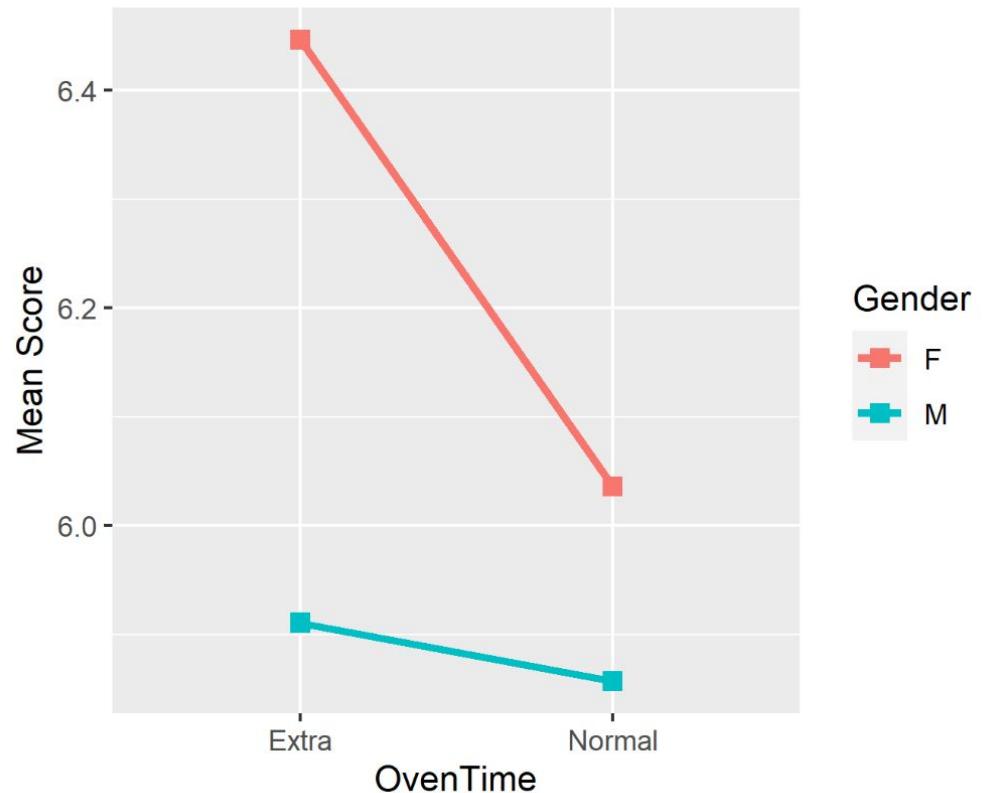
INTERACTION PLOTS

Interaction Plot of Brand and Gender



INTERACTION PLOTS

Interaction Plot of Gender and Oven Time

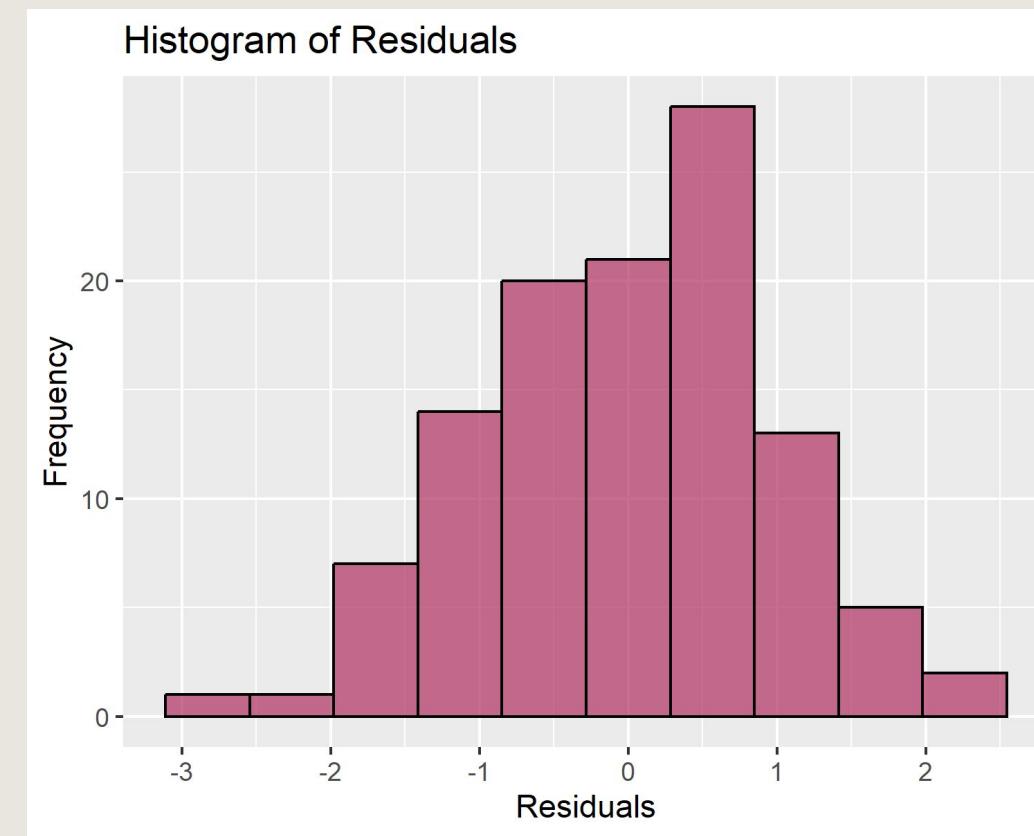


STATISTICAL MODEL

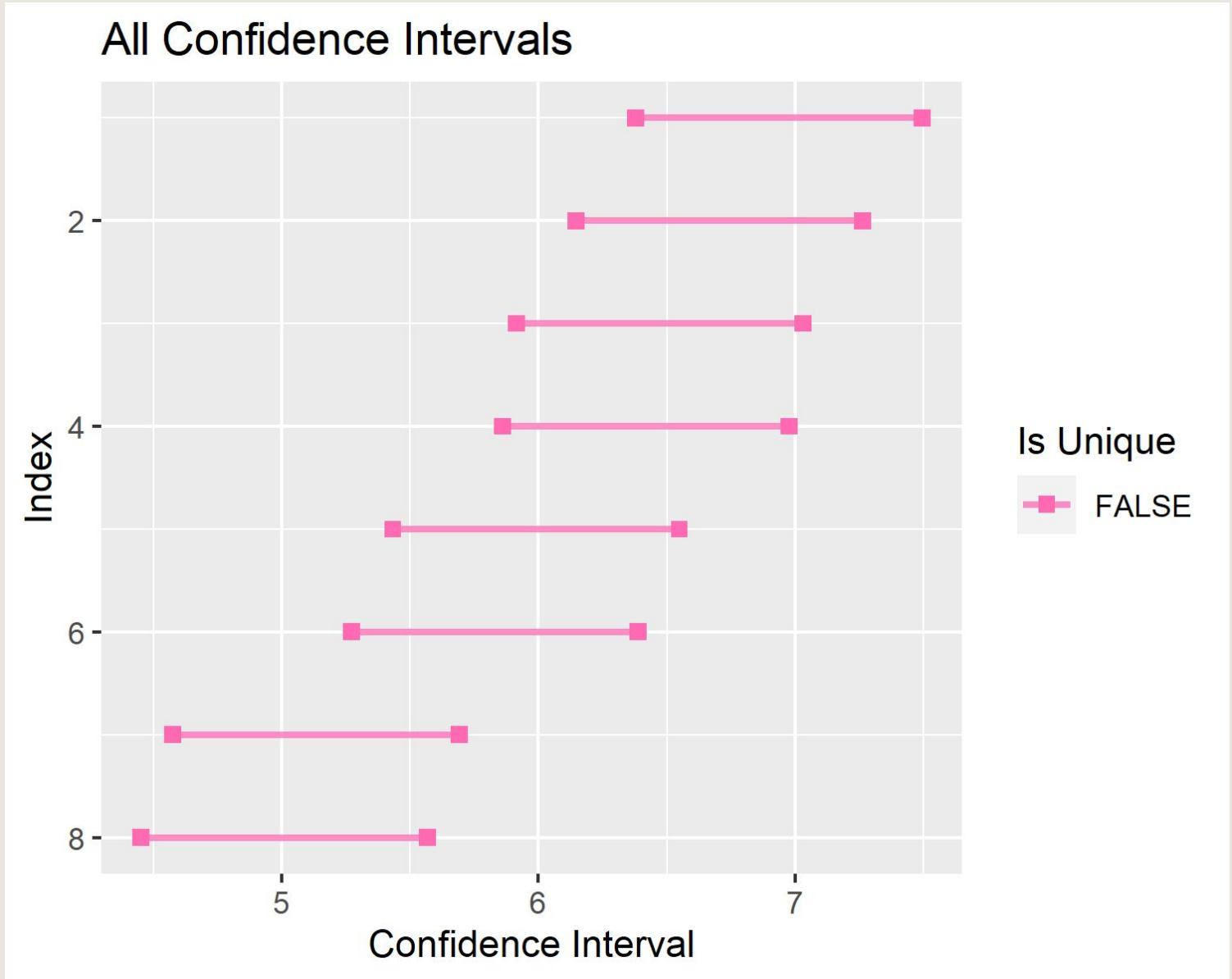
```
fit = lm(Score ~ Brand + OvenTime + Gender + as.factor(Person) +
         Brand:OvenTime + Gender:Brand + Gender:OvenTime, data = data)
```

```
results = anova(fit)
```

```
results
```



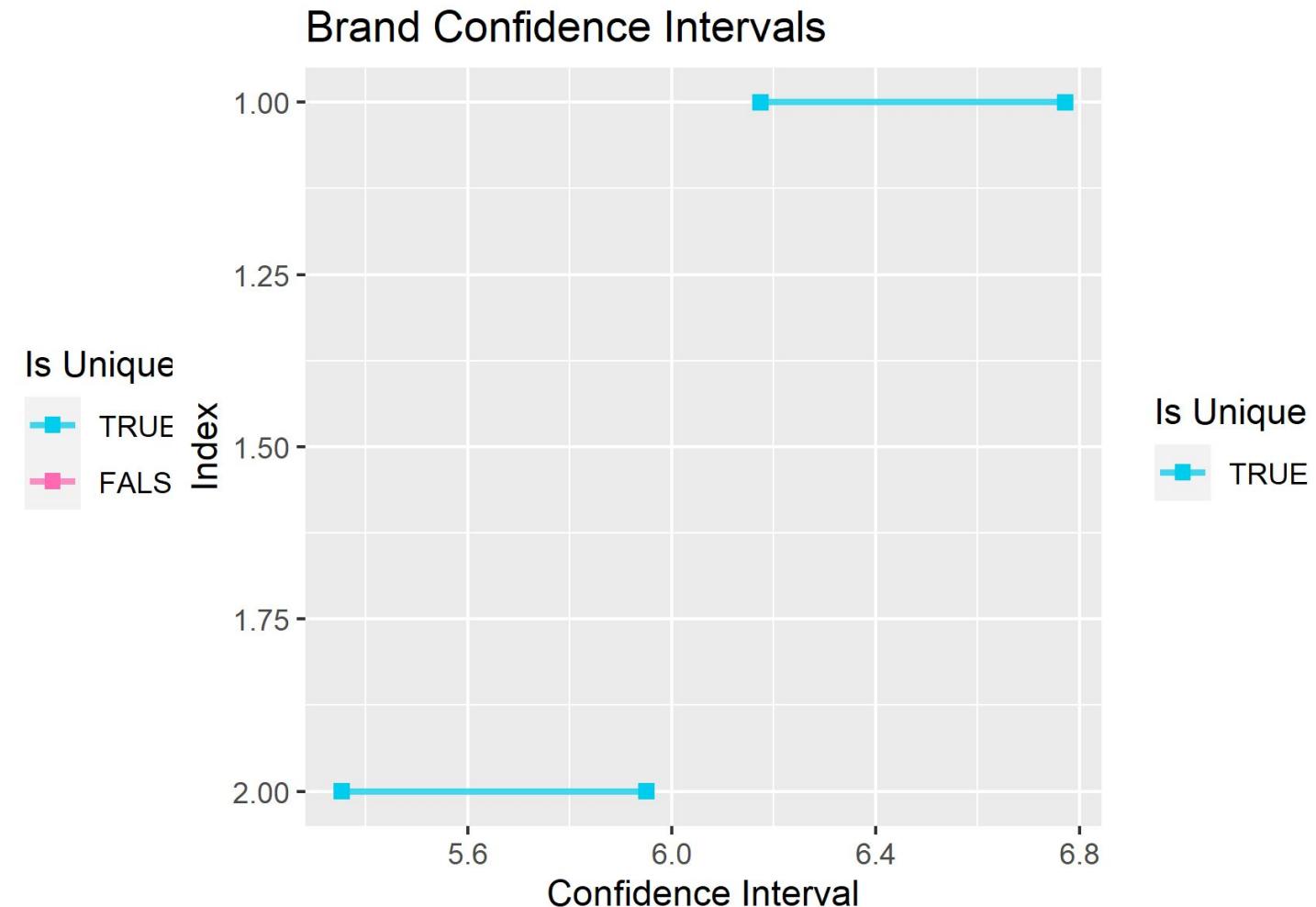
CONFIDENCE INTERVALS



ALL CONFIDENCE INTERVALS

	Brand	OvenTime	Gender	Mean	SE	df	lower.CL	upper.CL
1	BettyCrocker	Normal	F	6.9375	0.2805	79	6.3791	7.4959
2	BettyCrocker	Normal	M	6.7054	0.2805	79	6.1470	7.2638
3	Ghirardelli	Extra	F	6.4732	0.2805	79	5.9148	7.0316
4	BettyCrocker	Extra	F	6.4196	0.2805	79	5.8612	6.9780
5	Ghirardelli	Extra	M	5.9911	0.2805	79	5.4327	6.5495
6	BettyCrocker	Extra	M	5.8304	0.2805	79	5.2720	6.3888
7	Ghirardelli	Normal	F	5.1339	0.2805	79	4.5755	5.6923
8	Ghirardelli	Normal	M	5.0089	0.2805	79	4.4505	5.5673

SIGNIFICANT CONFIDENCE INTERVALS



SIGNIFICANT CONFIDENCE INTERVALS

	Brand	Mean	SE	df	lower.CL	upper.CL
1	BettyCrocker	6.4732	0.1500	79	6.1747	6.7717
2	Ghirardelli	5.6518	0.1500	79	5.3533	5.9503

	Brand	OvenTime	Mean	SE	df	lower.CL	upper.CL
1	BettyCrocker	Normal	6.8214	0.2121	79	6.3993	7.2435
2	Ghirardelli	Extra	6.2321	0.2121	79	5.8100	6.6543
3	BettyCrocker	Extra	6.1250	0.2121	79	5.7029	6.5471
4	Ghirardelli	Normal	5.0714	0.2121	79	4.6493	5.4935

ANOVA TABLE

	Df	Sum of Squares	Mean Sq	F value	P(>F)
Brand	1	18.8929	18.8929	15.0031	0.0002
Oven Time	1	1.5089	1.5089	1.1983	0.2770
Gender	1	3.5714	3.5714	2.8361	0.0961
Person	26	115.4911	4.4420	3.5274	0.0000
Brand:OvenTime	1	24.1429	24.1429	19.1721	0.0000
Brand:Gender	1	0.0804	0.0804	0.0638	0.8012
OvenTime:Gender	1	0.8929	0.8929	0.7090	0.4023
Residuals	79	99.4821	1.2593		