Portfolio Task 3 - Multifactorial Experimental Design

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Load and Prepare Data

```
# Load dataset
# The dataset is stored in the Downloads > GRA 4158 > Task3 folder

df <- read.csv("~/Downloads/GRA 4158/Task3/Multifactorial_design_example.csv", sep = ";")
# Convert categorical variables to factors
df$messagetype <- as.factor(df$messagetype)
df$pricelevel <- as.factor(df$pricelevel)
# View first few rows
head(df)</pre>
## response messagetype pricelevel
```

```
## 1
     194
         1
     228
              2
## 2
                      1
## 3
     236
              1
                     2
     251
              2
                     2
## 4
## 5
      241
              1
                      3
## 6
      307
              2
                      3
```

Summary Statistics

```
# Calculate group-wise means, standard deviations, and sample sizes
group_summary <- df %>%
  group_by(messagetype, pricelevel) %>%
  summarise(
   mean_response = mean(response),
   sd_response = sd(response),
   count = n()
)
```

```
\#\# `summarise()` has grouped output by 'messagetype'. You can override using the \#\# `.groups` argument.
```

Display the summary group_summary ## # A tibble: 6 x 5 ## # Groups: messagetype [2] messagetype pricelevel mean_response sd_response count ## <fct> <dbl> <dbl> <int> ## 1 1 201. 9.18 200 1 200 ## 2 1 2 221. 10.3 ## 3 1 3 240. 9.78 200

230.

250.

290.

Two-Way ANOVA

1

2

3

4 2

5 2

6 2

```
# Fit a two-way ANOVA model with interaction term
anova_model <- aov(response ~ messagetype * pricelevel, data = df)

# View summary of the ANOVA
summary(anova_model)</pre>
```

10.5

10.3

9.72

200

200

200

Post-hoc Test: Tukey HSD

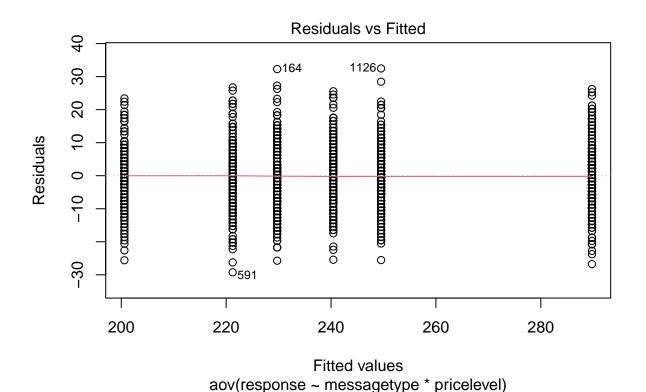
```
# Conduct Tukey HSD post-hoc comparisons
TukeyHSD(anova_model)
```

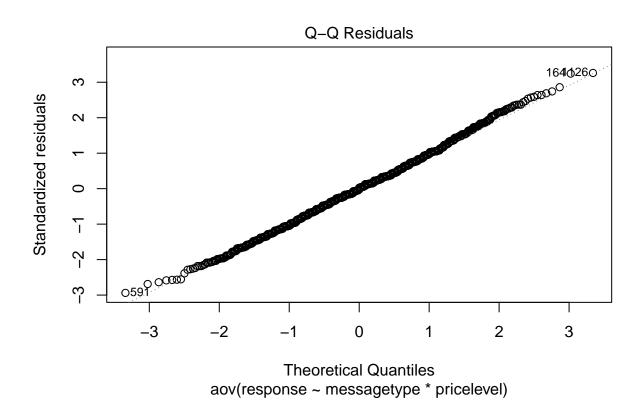
```
Tukey multiple comparisons of means
##
##
       95% family-wise confidence level
##
## Fit: aov(formula = response ~ messagetype * pricelevel, data = df)
##
## $messagetype
##
        diff
                 lwr
                         upr p adj
## 2-1 35.58 34.4499 36.7101
##
## $pricelevel
##
          diff
                    lwr
                             upr p adj
## 2-1 20.2300 18.57454 21.88546
## 3-1 49.9325 48.27704 51.58796
```

```
## 3-2 29.7025 28.04704 31.35796
##
## $`messagetype:pricelevel`
             {\tt diff}
##
                          lwr
                                    upr p adj
## 2:1-1:1 29.125
                   26.277297 31.972703
## 1:2-1:1 20.645
                   17.797297 23.492703
                                            0
## 2:2-1:1 48.940
                   46.092297 51.787703
                                            0
## 1:3-1:1 39.835
                   36.987297 42.682703
                                            0
## 2:3-1:1 89.155
                   86.307297 92.002703
                                            0
## 1:2-2:1 -8.480 -11.327703 -5.632297
                                            0
## 2:2-2:1 19.815
                   16.967297 22.662703
                                            0
## 1:3-2:1 10.710
                    7.862297 13.557703
                                            0
## 2:3-2:1 60.030
                   57.182297 62.877703
                                            0
## 2:2-1:2 28.295
                   25.447297 31.142703
                                            0
## 1:3-1:2 19.190
                   16.342297 22.037703
                                            0
## 2:3-1:2 68.510
                   65.662297 71.357703
                                            0
## 1:3-2:2 -9.105 -11.952703 -6.257297
                                            0
## 2:3-2:2 40.215
                   37.367297 43.062703
                                            0
                   46.472297 52.167703
## 2:3-1:3 49.320
```

Diagnostic Plots

```
# Residuals vs Fitted
plot(anova_model, which = 1)
```

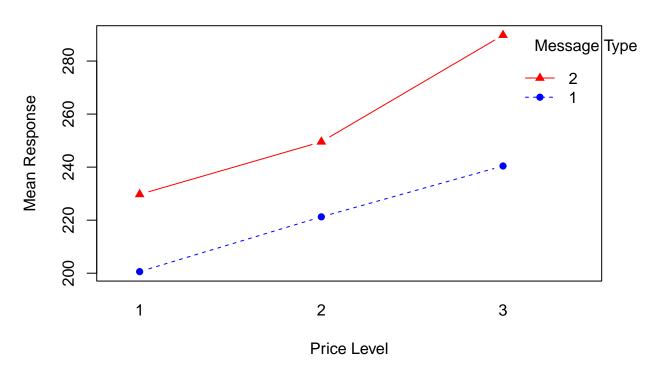




Interaction Plot

```
# Plot the interaction between message type and price level
interaction.plot(
    x.factor = df$pricelevel,
    trace.factor = df$messagetype,
    response = df$response,
    fun = mean,
    type = "b",
    col = c("blue", "red"),
    pch = c(16, 17),
    xlab = "Price Level",
    ylab = "Mean Response",
    trace.label = "Message Type",
    main = "Interaction Plot: Message Type × Price Level"
)
```

Interaction Plot: Message Type x Price Level



Conclusion

```
cat("Both messagetype and pricelevel significantly affect the response.\n")
```

Both messagetype and pricelevel significantly affect the response.

```
\mathtt{cat}("\mathtt{There}\ \mathtt{is}\ \mathtt{also}\ \mathtt{a}\ \mathtt{significant}\ \mathtt{interaction}\ \mathtt{between}\ \mathtt{them}. \setminus \mathtt{n}")
```

There is also a significant interaction between them.

cat("This suggests marketers should carefully pair messages with the right price strategy.\n")

This suggests marketers should carefully pair messages with the right price strategy.