

# MA500HW7

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```
library(ggplot2)
```

```
library(car)
```

```
## Loading required package: carData
```

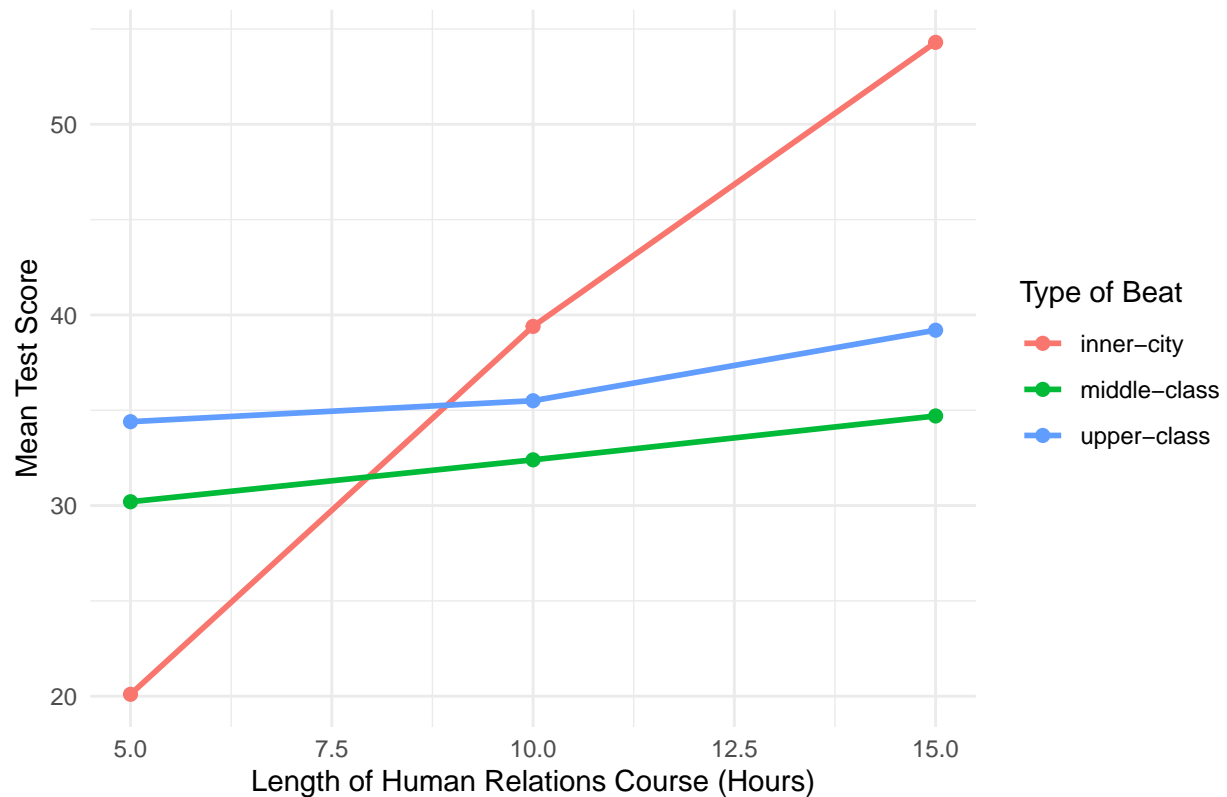
## Question 1

```
# Create the data
df <- data.frame(
  Type_Beat = rep(c("upper-class", "middle-class", "inner-city"), each = 3),
  Course_Hours = rep(c(5, 10, 15), times = 3),
  Score = c(34.4, 35.5, 39.2,
            30.2, 32.4, 34.7,
            20.1, 39.4, 54.3)
)

# Plot interaction graph
ggplot(df, aes(x = Course_Hours, y = Score, color = Type_Beat, group = Type_Beat)) +
  geom_line(size = 1) +
  geom_point(size = 2) +
  labs(title = "Interaction Graph: Type Beat vs. Course Hours",
       x = "Length of Human Relations Course (Hours)",
       y = "Mean Test Score",
       color = "Type of Beat") +
  theme_minimal()
```

```
## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

Interaction Graph: Type Beat vs. Course Hours



## Question 2

```
# (b)
set.seed(123)

df <- expand.grid(Start_Angle = 1:3, Stop_Angle = 1:3)

# Replicate each combination twice
df <- df[rep(1:nrow(df), each = 2), ]

# Add a random order column and shuffle
df$Trial <- sample(1:nrow(df))

# Order by randomized trial number
df <- df[order(df$Trial), ]

# Show the randomized design
df
```

```
##      Start_Angle Stop_Angle Trial
## 6             3           2     1
## 3             3           1     2
## 2             2           1     3
## 5             2           2     4
```

```
## 4.1      1      2      5
## 3.1      3      1      6
## 9.1      3      3      7
## 9        3      3      8
## 8        2      3      9
## 2.1      2      1     10
## 4        1      2     11
## 7.1      1      3     12
## 5.1      2      2     13
## 1.1      1      1     14
## 1        1      1     15
## 7        1      3     16
## 6.1      3      2     17
## 8.1      2      3     18
```

```
# (e)
set.seed(1022)

# Create the 3x3 factorial design with 2 replicates per cell
design <- expand.grid(Start_Angle = 1:3, Stop_Angle = 1:3, Replicate = 1:2)

# Simulate response variable (distance) with some made-up main effects and noise
design$Distance <- with(design,
  50 +                                # base distance
  5 * Start_Angle +                   # effect of start angle
  3 * Stop_Angle +                   # effect of stop angle
  2 * Start_Angle * Stop_Angle +     # interaction effect
  rnorm(nrow(design), mean = 0, sd = sqrt(12)) # random error
)

head(design)
```

```
##   Start_Angle Stop_Angle Replicate Distance
## 1           1           1           1 57.81626
## 2           2           1           1 66.25575
## 3           3           1           1 70.34698
## 4           1           2           1 63.21392
## 5           2           2           1 73.86091
## 6           3           2           1 85.24894
```

```
summary(design)
```

```
##   Start_Angle  Stop_Angle  Replicate    Distance
##  Min.   :1    Min.   :1    Min.   :1.0    Min.   :57.82
## 1st Qu.:1    1st Qu.:1    1st Qu.:1.0    1st Qu.:66.76
##  Median:2    Median :2    Median :1.5    Median :70.67
##  Mean   :2    Mean   :2    Mean   :1.5    Mean   :74.29
## 3rd Qu.:3    3rd Qu.:3    3rd Qu.:2.0    3rd Qu.:84.31
##  Max.   :3    Max.   :3    Max.   :2.0    Max.   :94.01
```

```
# (f)
# Make sure Start_Angle and Stop_Angle are factors
```

```

design$Start_Angle <- as.factor(design$Start_Angle)
design$Stop_Angle <- as.factor(design$Stop_Angle)

# Fit two-way ANOVA model with interaction
anova_model <- aov(Distance ~ Start_Angle * Stop_Angle, data = design)

summary(anova_model)

```

```

##                Df Sum Sq Mean Sq F value    Pr(>F)
## Start_Angle      2  773.7   386.9   28.149 0.000134 ***
## Stop_Angle       2  763.5   381.8   27.778 0.000141 ***
## Start_Angle:Stop_Angle  4  235.2    58.8    4.279 0.032700 *
## Residuals       9  123.7    13.7
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

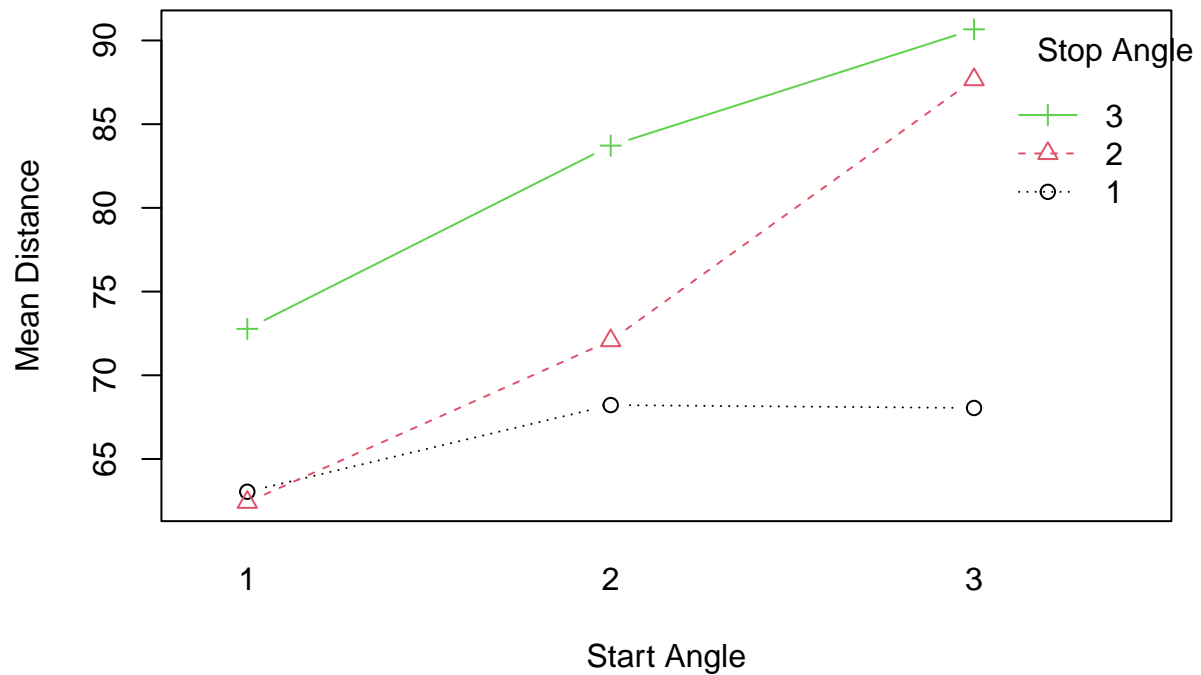
```

```

# (g)
# Interaction Plot
interaction.plot(
  x.factor = design$Start_Angle,
  trace.factor = design$Stop_Angle,
  response = design$Distance,
  fun = mean,
  type = "b", col = 1:3, pch = 1:3,
  xlab = "Start Angle", ylab = "Mean Distance",
  trace.label = "Stop Angle",
  main = "Interaction Plot: Start Angle × Stop Angle"
)

```

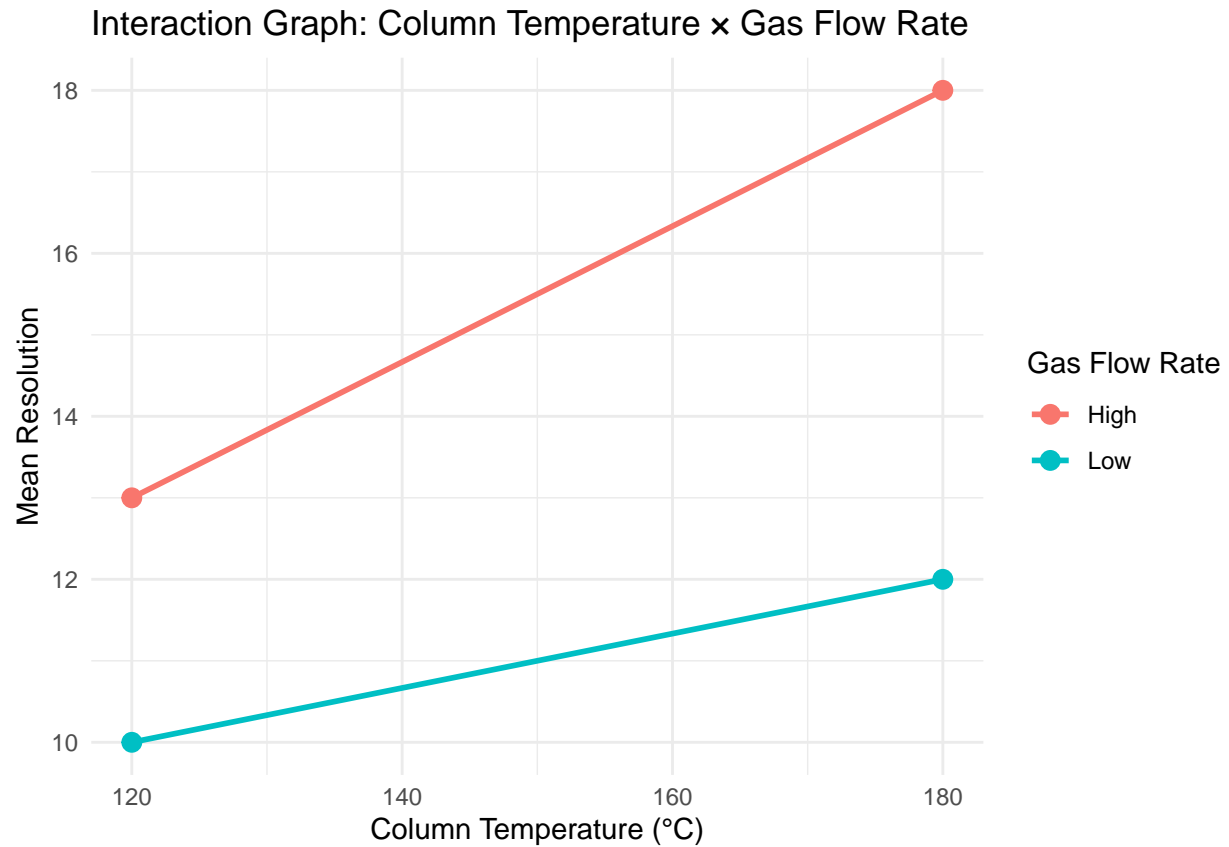
### Interaction Plot: Start Angle × Stop Angle



### Question 3

```
# Dataframe
df <- data.frame(
  Temperature = rep(c(120, 180), 2),
  FlowRate = factor(rep(c("Low", "High"), each = 2)),
  Resolution = c(10, 12, 13, 18)
)

ggplot(df, aes(x = Temperature, y = Resolution, color = FlowRate, group = FlowRate)) +
  geom_line(size = 1) +
  geom_point(size = 3) +
  labs(title = "Interaction Graph: Column Temperature × Gas Flow Rate",
       x = "Column Temperature (°C)",
       y = "Mean Resolution",
       color = "Gas Flow Rate") +
  theme_minimal()
```



#### Question 4

```
# (c)
set.seed(1022)

design_1 <- expand.grid(
  Brand = c("A", "B"),
  Power = c("Low", "Medium", "High"),
  Time = c(2, 2.5, 3, 3.5),
  Rep = 1:8
)

# Randomize order
design_1$Trial <- sample(1:nrow(design_1))
design_1 <- design_1[order(design_1$Trial), ]
head(design_1)
```

```
##      Brand Power Time Rep Trial
## 129     A Medium  2.5   6     1
##  13     A   Low   3.0   1     2
## 132     B  High   2.5   6     3
##  61     A   Low   3.0   3     4
## 138     B  High   3.0   6     5
##  42     B  High   3.0   2     6
```

```
summary(design_1)
```

```
## Brand      Power      Time      Rep      Trial
## A:96 Low :64 Min. :2.000 Min. :1.00 Min. : 1.00
## B:96 Medium:64 1st Qu.:2.375 1st Qu.:2.75 1st Qu.: 48.75
## High :64 Median :2.750 Median :4.50 Median : 96.50
## Mean :2.750 Mean :4.50 Mean : 96.50
## 3rd Qu.:3.125 3rd Qu.:6.25 3rd Qu.:144.25
## Max. :3.500 Max. :8.00 Max. :192.00
```

```
# (e)
# Genarate edible as response variable
set.seed(1022)

design_1$Edible <- with(design_1,
  70 + # base % edible
  ifelse(Brand == "B", 2, 0) + # Brand B slightly better
  ifelse(Power == "Low", -3,
    ifelse(Power == "Medium", 0, 3)) + # increasing trend in power
  -1.5 * (as.numeric(Time) - 2.75)^2 + # optimal time at 2.75
  rnorm(nrow(design_1), mean = 0, sd = 2) # random error
)

# Polynomial contrast analysis
design_1$Power <- factor(design_1$Power, levels = c("Low", "Medium", "High"), ordered = TRUE)

model <- aov(Edible ~ Brand + poly(as.numeric(Power), 2) + poly(Time, 3), data = design_1)
summary(model)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Brand              1  295.0   295.0   67.61 3.42e-14 ***
## poly(as.numeric(Power), 2)  2 1260.8   630.4  144.51 < 2e-16 ***
## poly(Time, 3)              3   33.8    11.3    2.58  0.055 .
## Residuals            185  807.0     4.4
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## Question 7

```
# (b)
set.seed(1022) # For reproducibility of design

# Step 1: Create full factorial design
factors <- expand.grid(
  Flame = factor(c("Low", "High")),
  PanSize = factor(c("Small", "Large")),
  Cover = factor(c("None", "Glass")),
  Salt = factor(c("No", "Yes"))
)
```

```

# Step 2: Replicate each combination 6 times
design_new <- factors[rep(1:nrow(factors), each = 6), ]
design_new$Replicate <- rep(1:6, times = nrow(factors))

# Step 4: Randomize run order
design_new$Trial <- sample(1:nrow(design_new))
design_new <- design_new[order(design_new$Trial), ]

design_new

```

##	Flame	PanSize	Cover	Salt	Replicate	Trial
## 10.2	High	Small	None	Yes	3	1
## 3.1	Low	Large	None	No	2	2
## 9.3	Low	Small	None	Yes	4	3
## 11.3	Low	Large	None	Yes	4	4
## 5.1	Low	Small	Glass	No	2	5
## 7.1	Low	Large	Glass	No	2	6
## 12.3	High	Large	None	Yes	4	7
## 5.5	Low	Small	Glass	No	6	8
## 16.3	High	Large	Glass	Yes	4	9
## 14.3	High	Small	Glass	Yes	4	10
## 12.5	High	Large	None	Yes	6	11
## 6.2	High	Small	Glass	No	3	12
## 5	Low	Small	Glass	No	1	13
## 7.5	Low	Large	Glass	No	6	14
## 13.4	Low	Small	Glass	Yes	5	15
## 11	Low	Large	None	Yes	1	16
## 10.1	High	Small	None	Yes	2	17
## 15	Low	Large	Glass	Yes	1	18
## 16.1	High	Large	Glass	Yes	2	19
## 8.1	High	Large	Glass	No	2	20
## 11.1	Low	Large	None	Yes	2	21
## 1.4	Low	Small	None	No	5	22
## 11.4	Low	Large	None	Yes	5	23
## 15.1	Low	Large	Glass	Yes	2	24
## 2.4	High	Small	None	No	5	25
## 16.2	High	Large	Glass	Yes	3	26
## 11.2	Low	Large	None	Yes	3	27
## 9.4	Low	Small	None	Yes	5	28
## 8	High	Large	Glass	No	1	29
## 13.3	Low	Small	Glass	Yes	4	30
## 4.4	High	Large	None	No	5	31
## 2.2	High	Small	None	No	3	32
## 3.4	Low	Large	None	No	5	33
## 15.3	Low	Large	Glass	Yes	4	34
## 9.1	Low	Small	None	Yes	2	35
## 1	Low	Small	None	No	1	36
## 6.3	High	Small	Glass	No	4	37
## 13	Low	Small	Glass	Yes	1	38
## 2.1	High	Small	None	No	2	39
## 8.4	High	Large	Glass	No	5	40
## 6.1	High	Small	Glass	No	2	41
## 1.5	Low	Small	None	No	6	42



## 5.4	Low	Small	Glass	No	5	43
## 2	High	Small	None	No	1	44
## 10	High	Small	None	Yes	1	45
## 5.2	Low	Small	Glass	No	3	46
## 16.5	High	Large	Glass	Yes	6	47
## 3.2	Low	Large	None	No	3	48
## 10.4	High	Small	None	Yes	5	49
## 5.3	Low	Small	Glass	No	4	50
## 12.4	High	Large	None	Yes	5	51
## 3.3	Low	Large	None	No	4	52
## 4.3	High	Large	None	No	4	53
## 14.5	High	Small	Glass	Yes	6	54
## 1.1	Low	Small	None	No	2	55
## 14.1	High	Small	Glass	Yes	2	56
## 4.2	High	Large	None	No	3	57
## 1.2	Low	Small	None	No	3	58
## 3.5	Low	Large	None	No	6	59
## 6.5	High	Small	Glass	No	6	60
## 16.4	High	Large	Glass	Yes	5	61
## 8.5	High	Large	Glass	No	6	62
## 13.2	Low	Small	Glass	Yes	3	63
## 7.3	Low	Large	Glass	No	4	64
## 8.2	High	Large	Glass	No	3	65
## 14.2	High	Small	Glass	Yes	3	66
## 14	High	Small	Glass	Yes	1	67
## 9	Low	Small	None	Yes	1	68
## 8.3	High	Large	Glass	No	4	69
## 14.4	High	Small	Glass	Yes	5	70
## 9.2	Low	Small	None	Yes	3	71
## 2.3	High	Small	None	No	4	72
## 13.1	Low	Small	Glass	Yes	2	73
## 12.2	High	Large	None	Yes	3	74
## 4.5	High	Large	None	No	6	75
## 15.5	Low	Large	Glass	Yes	6	76
## 11.5	Low	Large	None	Yes	6	77
## 2.5	High	Small	None	No	6	78
## 6.4	High	Small	Glass	No	5	79
## 7	Low	Large	Glass	No	1	80
## 16	High	Large	Glass	Yes	1	81
## 13.5	Low	Small	Glass	Yes	6	82
## 15.4	Low	Large	Glass	Yes	5	83
## 3	Low	Large	None	No	1	84
## 1.3	Low	Small	None	No	4	85
## 6	High	Small	Glass	No	1	86
## 4	High	Large	None	No	1	87
## 12	High	Large	None	Yes	1	88
## 7.2	Low	Large	Glass	No	3	89
## 4.1	High	Large	None	No	2	90
## 7.4	Low	Large	Glass	No	5	91
## 10.3	High	Small	None	Yes	4	92
## 10.5	High	Small	None	Yes	6	93
## 9.5	Low	Small	None	Yes	6	94
## 15.2	Low	Large	Glass	Yes	3	95
## 12.1	High	Large	None	Yes	2	96

```
# (d)
boiling_data <- read.csv("Boiling_Time_Data.csv")

boiling_model <- aov(Time ~ Flame * PanSize * Cover * Salt, data = boiling_data)
boiling_model
```

```
## Call:
## aov(formula = Time ~ Flame * PanSize * Cover * Salt, data = boiling_data)
##
## Terms:
##              Flame  PanSize    Cover    Salt Flame:PanSize Flame:Cover
## Sum of Squares 60.07670  2.09098 22.33514  1.27674         0.01465         0.00937
## Deg. of Freedom      1        1      1        1             1             1
##              PanSize:Cover Flame:Salt PanSize:Salt Cover:Salt
## Sum of Squares      0.05451      0.00029      0.06094      0.09131
## Deg. of Freedom            1            1            1            1
##              Flame:PanSize:Cover Flame:PanSize:Salt Flame:Cover:Salt
## Sum of Squares              0.00455              0.02250              0.03672
## Deg. of Freedom                1                1                1
##              PanSize:Cover:Salt Flame:PanSize:Cover:Salt Residuals
## Sum of Squares              0.01903              0.00590      2.72407
## Deg. of Freedom                1                1            80
##
## Residual standard error: 0.1845287
## Estimated effects may be unbalanced
```

```
summary(boiling_model)
```

```
##              Df Sum Sq Mean Sq F value Pr(>F)
## Flame          1  60.08   60.08 1764.324 < 2e-16 ***
## PanSize         1   2.09    2.09   61.408 1.69e-11 ***
## Cover           1  22.34   22.34  655.935 < 2e-16 ***
## Salt            1   1.28    1.28   37.495 3.23e-08 ***
## Flame:PanSize   1   0.01    0.01    0.430   0.514
## Flame:Cover     1   0.01    0.01    0.275   0.601
## PanSize:Cover   1   0.05    0.05    1.601   0.209
## Flame:Salt      1   0.00    0.00    0.008   0.927
## PanSize:Salt    1   0.06    0.06    1.790   0.185
## Cover:Salt      1   0.09    0.09    2.682   0.105
## Flame:PanSize:Cover 1   0.00    0.00    0.134   0.716
## Flame:PanSize:Salt 1   0.02    0.02    0.661   0.419
## Flame:Cover:Salt  1   0.04    0.04    1.078   0.302
## PanSize:Cover:Salt 1   0.02    0.02    0.559   0.457
## Flame:PanSize:Cover:Salt 1 0.01    0.01    0.173   0.678
## Residuals      80   2.72    0.03
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Since R gave the warning "Estimated effects may be unbalanced", we should consider using Type III Sum
Anova(boiling_model, type = 3)
```

```
## Anova Table (Type III tests)
```

```
##
## Response: Time
##
```

	Sum Sq	Df	F value	Pr(>F)	
## (Intercept)	200.627	1	5891.9861	< 2.2e-16	***
## Flame	6.673	1	195.9609	< 2.2e-16	***
## PanSize	0.481	1	14.1261	0.0003236	***
## Cover	2.531	1	74.3366	4.864e-13	***
## Salt	0.106	1	3.1188	0.0812095	.
## Flame:PanSize	0.008	1	0.2203	0.6400544	
## Flame:Cover	0.047	1	1.3745	0.2445174	
## PanSize:Cover	0.014	1	0.4222	0.5177273	
## Flame:Salt	0.040	1	1.1833	0.2799490	
## PanSize:Salt	0.028	1	0.8275	0.3657264	
## Cover:Salt	0.007	1	0.2164	0.6430538	
## Flame:PanSize:Cover	0.010	1	0.3056	0.5819344	
## Flame:PanSize:Salt	0.026	1	0.7553	0.3873916	
## Flame:Cover:Salt	0.036	1	1.0580	0.3067606	
## PanSize:Cover:Salt	0.002	1	0.0549	0.8153513	
## Flame:PanSize:Cover:Salt	0.006	1	0.1732	0.6783956	
## Residuals	2.724	80			

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```