Homework 9

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Here, I did all the Calculations for the ANOVA table by "hand". Note: all R code will be at the end of this document.

First, I display the contrasts matrix

| | Al | Bl | Aq | Bq | AlBl | AlBq | AqBl | AqBq |
|----|----|----|----|----|------|------|------|------|
| 00 | -1 | -1 | 1 | 1 | 1 | -1 | -1 | 1 |
| 01 | -1 | 0 | 1 | -2 | 0 | 2 | 0 | -2 |
| 02 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 |
| 10 | 0 | -1 | -2 | 1 | 0 | 0 | 2 | -2 |
| 11 | 0 | 0 | -2 | -2 | 0 | 0 | 0 | 4 |
| 12 | 0 | 1 | -2 | 1 | 0 | 0 | -2 | -2 |
| 20 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | 1 |
| 21 | 1 | 0 | 1 | -2 | 0 | -2 | 0 | -2 |
| 22 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |

and with the sum of each treatment levels' combinations

| | Al | Bl | Aq | Bq | AlBl | AlBq | AqBl | AqBq | у |
|----|----|----|----|----|------|------|------|------|------|
| 00 | -1 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | 1718 |
| 01 | -1 | 0 | 1 | -2 | 0 | 2 | 0 | -2 | 3262 |
| 02 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 | 4158 |
| 10 | 0 | -1 | -2 | 1 | 0 | 0 | 2 | -2 | 1659 |
| 11 | 0 | 0 | -2 | -2 | 0 | 0 | 0 | 4 | 3105 |
| 12 | 0 | 1 | -2 | 1 | 0 | 0 | -2 | -2 | 3939 |
| 20 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | 1 | 1720 |
| 21 | 1 | 0 | 1 | -2 | 0 | -2 | 0 | -2 | 3164 |
| 22 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 2660 |
| | | | | | | | | | |

Now, we can estimate the effect of each term in our model by multiplying y by each column of this matrix, taking the sum, and dividing by n = 3. The result is

| Al | -531.3333 |
|---------------------|------------|
| Bl | 1886.6667 |
| Aq | -241.3333 |
| Bq | -1069.3333 |
| AlBl | -500.0000 |
| AlBq | -433.3333 |
| AqBl | -393.3333 |
| AqBq | -457.3333 |

Now, for the Sums of Squares for each term in the model we calculate

$$\frac{n(contrasts)^2}{\sum c_i^2}$$

which results in

| Al | 141157.556 |
|------|-------------|
| Bl | 1779755.556 |
| Aq | 9706.963 |
| Bq | 190578.963 |
| AlBl | 187500.000 |
| AlBq | 46944.444 |
| AqBl | 38677.778 |
| AqBq | 17429.481 |

Before I forget, the sum of squares for A, B, and AB is

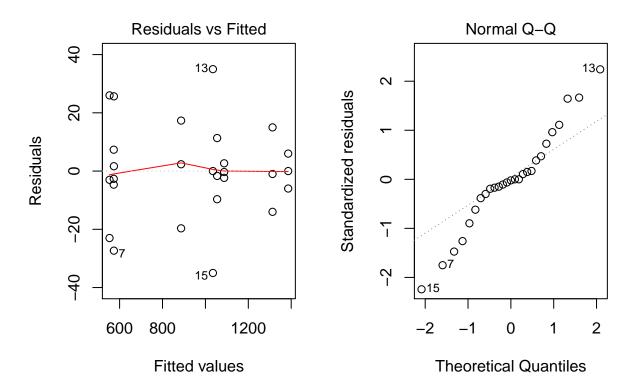
| ssa | ssb | ssab | sse | sst |
|--------|---------|--------|------|---------|
| 150865 | 1970335 | 290552 | 6579 | 2418330 |

The remaining ANOVA table is similar to previous methods used in this class, where mean squares are $\frac{SS}{df}$ and the F statistic is $\frac{MS_i}{MS_E}$. Here, I finish filling out the ANOVA table

| Source | Sum Sq | Df | Mean Sq | F value | Pr(>F) |
|-------------------|---------------------------|----|-------------------------|--------------|---------|
| glass | 1.4115756×10^5 | 1 | 1.4115756×10^5 | 386.1844158 | < 0.000 |
| temp | 1.7797556×10^6 | 1 | 1.7797556×10^6 | 4869.1255446 | < 0.000 |
| $glass^2$ | 9706.962963 | 1 | 9706.962963 | 26.5566927 | < 0.000 |
| $temp^2$ | 1.9057896×10^5 | 1 | 1.9057896×10^5 | 521.3934543 | < 0.000 |
| temp:glass | 1.875×10^{5} | 1 | 1.875×10^{5} | 512.9699058 | < 0.000 |
| $glass: temp^2$ | 4.6944444×10^4 | 1 | 4.6944444×10^4 | 128.4324653 | < 0.000 |
| $glass^2: temp$ | 3.8677778×10^4 | 1 | 3.8677778×10^4 | 105.8161921 | < 0.000 |
| $glass^2: temp^2$ | 1.7429481×10^4 | 1 | 1.7429481×10^4 | 47.6842639 | < 0.000 |
| Error | 6579.3333333 | 18 | 365.5185185 | | |
| Total | 2.4183301×10^{6} | 26 | | | |

All effects have a significant effect on *light output* and are statistically significant, with **glass** having the most significant effect. There is also some *curvature* to the main effects **glass** and **temp**

Now, I take a look at the residual plots of the model



The normality assumption may not be valid. We may need to reexamine the model or possibly perform a transformation.

R code:

```
###### Exploring 3^2 full factorial models using R #######
df <- read.csv("~/Documents/STAT4100/data/hw9.csv")</pre>
df.c <- df
library(dplyr)
y \leftarrow df y
y.i <- df.c %>%
  group_by(glass) %>%
  summarise_each(funs(sum), y)
y.j <- df.c %>%
  group_by(temp) %>%
  summarise_each(funs(sum), y)
yij. <- df.c %>%
  group_by(glass, temp) %>%
  summarise_each(funs(sum), y)
y... \leftarrow sum(y)
a \leftarrow 3; b \leftarrow 3; n \leftarrow 3; N \leftarrow a*b*n
ssa <- (1/(b*n))*sum(y.i$y^2) - y..^2/N
ssb \leftarrow (1/(a*n))*sum(y.j$y^2) - y..^2/N
sssub <- (1/n)*sum(yij.$y^2) - y..^2/N
ssab <- sssub - ssa - ssb
sst <-sum(y^2) - y..^2/N
sse <- sst - ssab - ssa - ssb
#round(c(ssa,ssb,ssab,sse,sst))
# now to add squared terms
```

```
row.sums <- df %>%
 group_by(glass,temp) %>%
  summarise_each(funs(sum), y)
y <- row.sums$y
id <- c("00", "01", "02", "10", "11", "12", "20", "21", "22")
#data.frame(id, y) # these are our summaries..
#... now to develop the contrasts matrix
Al \leftarrow c(rep(-1,3), rep(0,3), rep(1,3))
Bl <- rep(c(-1,0,1),3)
Aq \leftarrow c(rep(1,3), rep(-2,3), rep(1,3))
Bq \leftarrow rep(c(1,-2,1), 3)
AlBl <- Al*Bl; AlBq <- Al*Bq; AqBl <- Aq*Bl; AqBq <- Aq*Bq
dm <- cbind(Al,Bl,Aq,Bq,AlBl,AlBq,AqBl,AqBq)</pre>
rownames(dm) <- id</pre>
m.effects <- apply(dm,2, FUN = function(d) sum(d*y)/3) # effects
# now for the sum of squares...
C <- apply(dm, 2, function(x) sum(x^2)) #contrasts</pre>
ss <- 3*(m.effects^2)/C # sum of squares
ss2 <- round(cbind(ssa,ssb,ssab,sse,sst))</pre>
sse <- sst - sum(ss); mse <- sse/18
par(mfrow=c(1,2))
plot(fit.full, which = c(1,2))
```