

Stat 207 HW1

Cheng Luo 912466499
Fan Wu 912538518

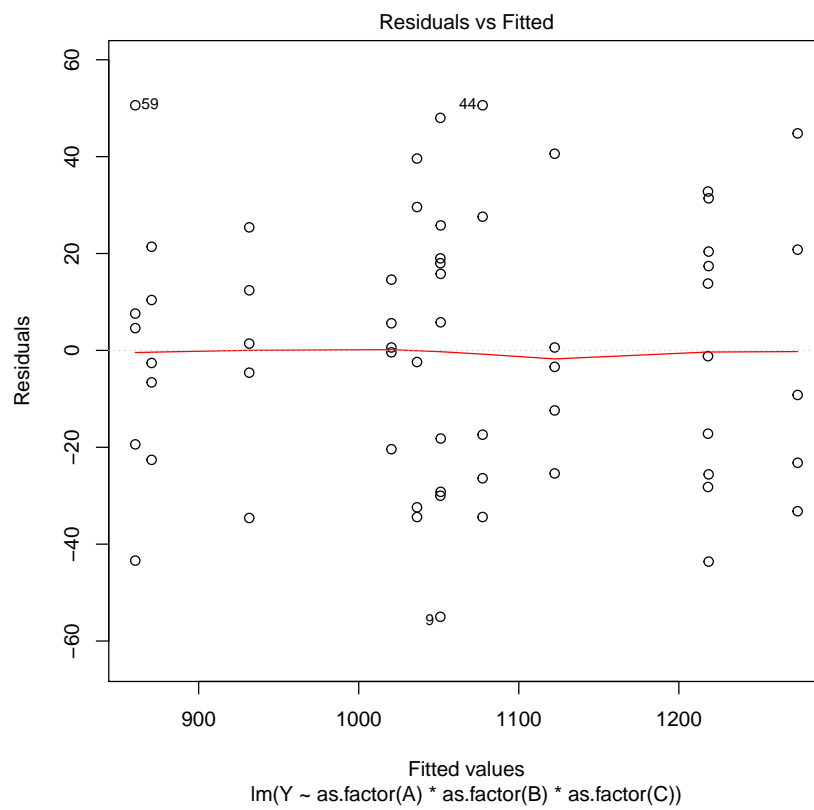
January 15, 2015

1 24.12

```
(a) elec = read.table("CH24PR12.txt")
names(elec) = c("Y", "A", "B", "C", "T")
n = length(unique(elec$T))
a = length(unique(elec$A))
b = length(unique(elec$B))
c = length(unique(elec$C))
fit = lm(Y ~ as.factor(A) * as.factor(B) * as.factor(C) , elec)
res = resid(fit)
res
```

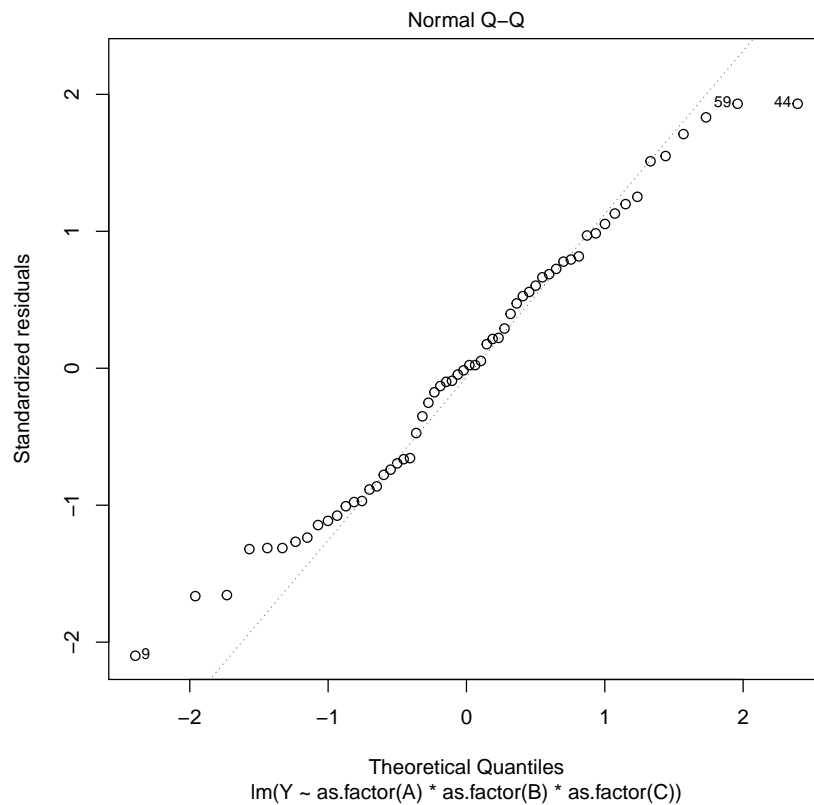
##	1	2	3	4	5	6	7	8	9	10	11	12
##	31.4	-43.6	17.4	20.4	-25.6	-30.0	48.0	18.0	-55.0	19.0	44.8	-23.2
##	13	14	15	16	17	18	19	20	21	22	23	24
##	-33.2	20.8	-9.2	-3.4	-12.4	0.6	-25.4	40.6	-1.2	-28.2	-17.2	13.8
##	25	26	27	28	29	30	31	32	33	34	35	36
##	32.8	-18.2	15.8	5.8	25.8	-29.2	29.6	39.6	-32.4	-34.4	-2.4	-6.6
##	37	38	39	40	41	42	43	44	45	46	47	48
##	-22.6	10.4	21.4	-2.6	27.6	-34.4	-26.4	50.6	-17.4	-4.6	12.4	25.4
##	49	50	51	52	53	54	55	56	57	58	59	60
##	-34.6	1.4	0.6	-0.4	14.6	-20.4	5.6	-19.4	4.6	-43.4	50.6	7.6

```
plot(fit, which = 1)
```



The residuals versus fitted values plots shows no sign for unequal variance.

(b) `plot(fit, which = 2)`



```
cor(sort(res), ppoints(res))

## [1] 0.9930971
```

The normal QQ plot and correlation indicates approximately normal distribution of residuals with slight light tail, so that the normality assumption seems to be reasonable here.

2 24.13

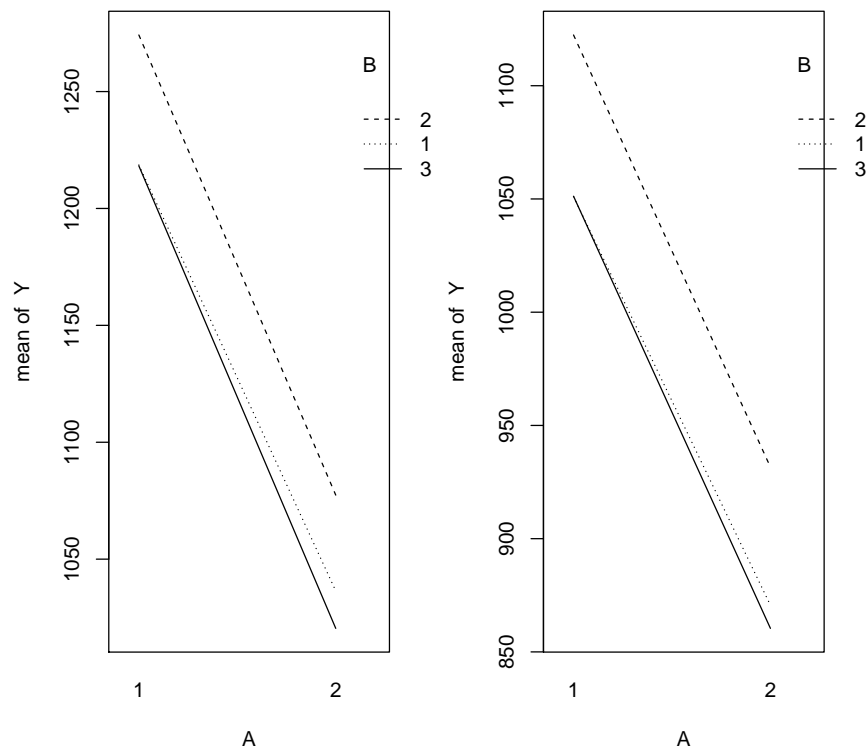
(a) \bar{Y}_{ijk} . are as follow:

```
means = sapply(with(elec, split(Y,list(A,B,C))), mean)
means

## 1.1.1 2.1.1 1.2.1 2.2.1 1.3.1 2.3.1 1.1.2 2.1.2 1.2.2 2.2.2
```

```
## 1218.6 1036.4 1274.2 1077.4 1218.2 1020.4 1051.0 870.6 1122.4 931.6
## 1.3.2 2.3.2
## 1051.2 860.4
```

```
elec.C = split(elec, elec$C)
par(mfrow = c(1,2))
with(elec.C$'1', interaction.plot(A, B, Y))
with(elec.C$'2', interaction.plot(A, B, Y))
```



From AB plots of the estimated treatment means, the AB curves seem to be parallel, which means there's no interaction between AB, moreover, main effect A and main effect B are present.

(b) `anova(fit)`

```
## Analysis of Variance Table
```

```
##
## Response: Y
##
##              Df Sum Sq Mean Sq F value
## as.factor(A)    1 540361   540361 629.7603
## as.factor(B)    2  49320    24660  28.7396
## as.factor(C)    1 382402   382402 445.6679
## as.factor(A):as.factor(B)    2    543     271   0.3161
## as.factor(A):as.factor(C)    1     91      91   0.1064
## as.factor(B):as.factor(C)    2    911     456   0.5310
## as.factor(A):as.factor(B):as.factor(C)    2     19      10   0.0111
## Residuals      48 41186      858
##
##              Pr(>F)
## as.factor(A)    < 2.2e-16 ***
## as.factor(B)    6.22e-09 ***
## as.factor(C)    < 2.2e-16 ***
## as.factor(A):as.factor(B)    0.7305
## as.factor(A):as.factor(C)    0.7457
## as.factor(B):as.factor(C)    0.5914
## as.factor(A):as.factor(B):as.factor(C)    0.9890
## Residuals
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

(c) For three-factor interaction:

$$H_0: \text{all } (\alpha\beta\gamma)_{ijk} = 0$$

$$\text{VS. } H_1: \text{not all } (\alpha\beta\gamma)_{ijk} = 0$$

$$F^* = \frac{MS_{ABC}}{MSE} = 0.0111$$

we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 3.19$, otherwise reject H_1

reject H_1 because $F^* < F(1 - 0.05; 2, 48) = 3.19$,

therefore, we conclude H_0 at 0.05 level, and there's no three-factor interaction effect, and P-value = 0.9890

(d) For AB interaction:

$$H_0: \text{all } (\alpha\beta)_{ij} = 0$$

$$\text{VS. } H_1: \text{not all } (\alpha\beta)_{ij} = 0$$

$$F^* = \frac{MS_{AB}}{MSE} = 0.3161$$

we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 3.19$, otherwise reject H_1

reject H_1 because $F^* < F(1 - 0.05; 2, 48) = 3.19$,

therefore, we conclude H_0 at 0.05 level, and there's no AB interaction effect, and P-value = 0.7305

For AC interaction:

$H_0: \text{all } (\alpha\gamma)_{ik}=0$
VS. $H_1: \text{not all } (\alpha\gamma)_{ik} = 0$
 $F^* = \frac{MSAC}{MSE} = 0.1064$
we can reject H_0 if $F^* > F(1 - 0.05; 1, 48) = 4.04$, otherwise reject H_1
reject H_1 because $F^* < F(1 - 0.05; 2, 48) = 4.04$,
therefore, we conclude H_0 at 0.05 level, and there's no AC interaction
effect, and P-value = 0.7457

For BC interaction:

$H_0: \text{all } (\beta\gamma)_{jk}=0$
VS. $H_1: \text{not all } (\beta\gamma)_{jk} = 0$
 $F^* = \frac{MSBC}{MSE} = 0.5310$
we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 3.19$, otherwise reject H_1
reject H_1 because $F^* < F(1 - 0.05; 2, 48) = 3.19$,
therefore, we conclude H_0 at 0.05 level, and there's no BC interaction
effect, and P-value = 0.5914

(e) For A main effect:

$H_0: \text{all } (\alpha)_i=0$
VS. $H_1: \text{not all } (\alpha)_i = 0$
 $F^* = \frac{MSA}{MSE} = 629.7603$
we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 4.04$, otherwise reject H_1
reject H_0 because $F^* > F(1 - 0.05; 2, 48) = 4.04$,
therefore, we conclude H_1 at 0.05 level, and there's A main effect, and
P-value = 2.2e-16

For B main effect:

$H_0: \text{all } (\beta)_j=0$
VS. $H_1: \text{not all } (\beta)_j = 0$
 $F^* = \frac{MSB}{MSE} = 28.7396$
we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 3.19$, otherwise reject H_1
reject H_0 because $F^* > F(1 - 0.05; 2, 48) = 3.19$,
therefore, we conclude H_1 at 0.05 level, and there's B main effect, and
P-value = 6.22e-09

For C main effect:

$$H_0: \text{all } (\gamma)_k = 0$$

$$\text{VS. } H_1: \text{not all } (\gamma)_k = 0$$

$$F^* = \frac{MSC}{MSE} = 445.6679$$

we can reject H_0 if $F^* > F(1 - 0.05; 2, 48) = 4.04$, otherwise reject H_1

reject H_0 because $F^* > F(1 - 0.05; 2, 48) = 3.19$,

therefore, we conclude H_1 at 0.05 level, and there's C main effect, and

$$P\text{-value} = 2.2\text{e-}16$$

- (f) A, B and C main effects are present, but there's no any interaction effect. As for upper bound for the family level of significance, using kimball inequality, $\alpha < 1 - (1 - \alpha_1)(1 - \alpha_2) \cdots (1 - \alpha_7) = 1 - (1 - 0.05)^7 = 0.3016627$
- (g) The results in part(f) confirm my graphic analysis in part(a)

3 24.14

```
(a) means = with(elec, sapply(list(A, B, C), function(x) by(Y, x, mean)))
      means

## [[1]]
## x: 1
## [1] 1155.933
## -----
## x: 2
## [1] 966.1333
##
## [[2]]
## x: 1
## [1] 1044.15
## -----
## x: 2
## [1] 1101.4
## -----
## x: 3
## [1] 1037.55
##
## [[3]]
## x: 1
## [1] 1140.867
## -----
## x: 2
## [1] 981.2
```



```

D1.hat = means[[1]][1] - means[[1]][2]
D2.hat = means[[2]][1] - means[[2]][2]
D3.hat = means[[2]][1] - means[[2]][3]
D4.hat = means[[2]][2] - means[[2]][3]
D5.hat = means[[3]][1] - means[[3]][2]
mse = anova(fit)["Residuals", 3]
S_D1 = sqrt(mse/(n*b*c)*2)
S_D2 = sqrt(mse/(n*a*c)*2)
S_D3 = sqrt(mse/(n*a*c)*2)
S_D4 = sqrt(mse/(n*a*c)*2)
S_D5 = sqrt(mse/(n*a*b)*2)
B = qt(1-.1/(2*5), 48)
c(D1.hat-B*S_D1, D1.hat+B*S_D1)

##          1          1
## 171.5984 208.0016

c(D2.hat-B*S_D2, D2.hat+B*S_D2)

##          1          1
## -79.54229 -34.95771

c(D3.hat-B*S_D3, D3.hat+B*S_D3)

##          1          1
## -15.69229  28.89229

c(D4.hat-B*S_D4, D4.hat+B*S_D4)

##          2          2
## 41.55771  86.14229

c(D5.hat-B*S_D5, D5.hat+B*S_D5)

##          1          1
## 141.4651 177.8682

```

$$\begin{aligned}
171.5984 &\leq D_1 \leq 208.0016 \\
-79.54229 &\leq D_2 \leq -34.95771 \\
-15.69229 &\leq D_3 \leq 28.89229 \\
41.55771 &\leq D_4 \leq 86.14229 \\
141.4651 &\leq D_5 \leq 177.8682
\end{aligned}$$

```
(b) dat.231 = elec[with(elec, A == 2 & B == 3 & C == 1), ]
means = mean(dat.231$Y)
s = sqrt(mse/n)
t. = qt(1 - .05/2, (n - 1)*a*b*c)
c(means - t.*s, means + t.*s)

## [1] 994.0608 1046.7392
```

Therefore, μ_{231} with a 95 percent confidence interval is $994.0608 \leq \mu_{231} \leq 1046.7392$

4 24.16

```
(a) dat.reg = elec[-c(19, 40, 43), ]
x1 = dat.reg$A
x1 = replace(x1, which(dat.reg$A == 2), -1)
x2 = dat.reg$B
x2 = replace(x2, which(dat.reg$B == 3), -1)
x2 = replace(x2, which(dat.reg$B == 2), 0)
x3 = dat.reg$B
x3 = replace(x3, which(dat.reg$B == 3), -1)
x3 = replace(x3, which(dat.reg$B == 1), 0)
x3 = replace(x3, which(dat.reg$B == 2), 1)
x4 = dat.reg$C
x4 = replace(x4, which(dat.reg$C == 2), -1)
model.full = lm(dat.reg$Y ~ x1*(x2+x3)*x4)
sse.full = sum(resid(model.full)^2)
```

For full model:

$$\begin{aligned}
Y_{ijkm} = & \mu... + \alpha_1 X_{ijkm1} + \beta_1 X_{ijkm2} + \beta_2 X_{ijkm3} + \gamma X_{ijkm4} \\
& + (\alpha\beta)_{11} X_{ijkm1} X_{ijkm2} + (\alpha\beta)_{12} X_{ijkm1} X_{ijkm3} + (\alpha\gamma)_{11} X_{ijkm1} X_{ijkm4} \\
& + (\beta\gamma)_{11} X_{ijkm2} X_{ijkm4} + (\beta\gamma)_{21} X_{ijkm3} X_{ijkm4} + (\alpha\beta\gamma)_{111} X_{ijkm1} X_{ijkm2} X_{ijkm4} \\
& + (\alpha\beta\gamma)_{121} X_{ijkm1} X_{ijkm3} X_{ijkm4} + \epsilon_{ijkm}
\end{aligned}$$

$$X_{ijkm1} = \begin{cases} 1 & \text{if factor A is level 1} \\ -1 & \text{if factor A is level 2} \end{cases} \quad (1)$$

$$X_{ijkm2} = \begin{cases} 1 & \text{if factor B is level 1} \\ -1 & \text{if factor B is level 3} \\ 0 & \text{else} \end{cases} \quad (2)$$

$$X_{ijkm3} = \begin{cases} 1 & \text{if factor B is level 2} \\ -1 & \text{if factor B is level 3} \\ 0 & \text{else} \end{cases} \quad (3)$$

$$X_{ijkm4} = \begin{cases} 1 & \text{if factor C is level 1} \\ -1 & \text{if factor C is level 2} \end{cases} \quad (4)$$

```
(b) model.red = lm(dat.reg$Y ~ x1*(x2+x3)*x4 -x4)
      sse.red = sum(resid(model.red)^2)
```

For reduced model:

$$\begin{aligned} Y_{ijkm} = & \mu... + \alpha_1 X_{ijkm1} + \beta_1 X_{ijkm2} + \beta_2 X_{ijkm3} \\ & + (\alpha\beta)_{11} X_{ijkm1} X_{ijkm2} + (\alpha\beta)_{12} X_{ijkm1} X_{ijkm3} + (\alpha\gamma)_{11} X_{ijkm1} X_{ijkm4} \\ & + (\beta\gamma)_{11} X_{ijkm2} X_{ijkm4} + (\beta\gamma)_{21} X_{ijkm3} X_{ijkm4} + (\alpha\beta\gamma)_{111} X_{ijkm1} X_{ijkm2} X_{ijkm4} \\ & + (\alpha\beta\gamma)_{121} X_{ijkm1} X_{ijkm3} X_{ijkm4} + \epsilon_{ijkm} \end{aligned}$$

```
(c) summary(model.full)

##
## Call:
## lm(formula = dat.reg$Y ~ x1 * (x2 + x3) * x4)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -55.0   -23.2    0.6   20.4   50.6
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1062.1667    3.9426 269.409 < 2e-16 ***
## x1           94.8250    3.9426 24.052 < 2e-16 ***
## x2          -17.8542    5.5757 -3.202 0.0025 **
## x3           42.4708    5.6571  7.508 1.81e-09 ***
## x4           79.8000    3.9426 20.241 < 2e-16 ***
## x1:x2        -4.3375    5.5757 -0.778 0.4407
## x1:x3         2.0125    5.6571  0.356 0.7237
## x1:x4         0.2083    3.9426  0.053 0.9581
## x2:x4         3.3875    5.5757  0.608 0.5465
## x3:x4        -5.3375    5.6571 -0.944 0.3505
## x1:x2:x4      0.4042    5.5757  0.072 0.9425
```

```
## x1:x3:x4      -1.9458      5.6571  -0.344   0.7325
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 29.63 on 45 degrees of freedom
## Multiple R-squared:  0.9595, Adjusted R-squared:  0.9496
## F-statistic: 96.96 on 11 and 45 DF,  p-value: < 2.2e-16

summary(model.red)

##
## Call:
## lm(formula = dat.reg$Y ~ x1 * (x2 + x3) * x4 - x4)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -130.11  -73.71   31.51   74.71  137.88
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1063.731      12.393  85.834 < 2e-16 ***
## x1           96.390      12.393   7.778 6.3e-10 ***
## x2          -14.725      17.523  -0.840  0.405
## x3           40.906      17.784   2.300  0.026 *
## x1:x2        -10.596      17.503  -0.605  0.548
## x1:x3          9.836      17.744   0.554  0.582
## x1:x4          1.773      12.393   0.143  0.887
## x2:x4          3.388      17.529   0.193  0.848
## x3:x4         -10.032      17.770  -0.565  0.575
## x1:x2:x4       3.534      17.523   0.202  0.841
## x1:x3:x4      -3.511      17.784  -0.197  0.844
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 93.15 on 46 degrees of freedom
## Multiple R-squared:  0.5909, Adjusted R-squared:  0.502
## F-statistic: 6.645 on 10 and 46 DF,  p-value: 2.836e-06

sse.full

## [1] 39499.9

sse.red

## [1] 399106.9
```

SSE(F)=39499.9, df(F)=45
 SSE(R)=399106.9, df(R)=46

$$H_0: \text{all } (\gamma)_1 = 0$$

$$\text{VS. } H_1: \text{not all } (\gamma)_1 = 0$$

$$F^* = \frac{(SSE(R) - SSE(F)) / (df(R) - df(F))}{SSE(F) / df(F)} = \frac{359607/1}{39499.9/45} = 409.6799$$

we can reject H_0 if $F^* > F(1 - 0.05; 1, 45) = 4.056612$, otherwise reject H_1

reject H_0 because $F^* > F(1 - 0.05; 1, 45) = 3.19$,

therefore, we conclude H_1 at 0.05 level, and there's C main effect, and
 P-value is 3.114909e-24, very close to zero.

```
f.star = (sse.red - sse.full) / (sse.full/45)
qf(1 - .05, 1, 45)

## [1] 4.056612

pf(f.star, 1, 45, lower = FALSE)

## [1] 3.114909e-24
```

(d)

```
D.hat = 159.6
n.s = with(dat.reg, by(Y, list(A, B, C), length))
s = sqrt(sse.full/(nrow(dat.reg) - a*b*c) / (a^2*b^2) * sum(1/n.s))
s

## [1] 7.885161

t. = qt(1 - .05/2, nrow(dat.reg) - a*b*c)
c(D.hat - s*t., D.hat + s*t.)

## [1] 143.7185 175.4815
```

$$\hat{D} = \hat{\mu}_{..1} - \hat{\mu}_{..2} = \hat{\gamma}_1 - \hat{\gamma}_2 = 2\hat{\gamma}_1 = 2 * 79.8 = 159.6$$

$$MSE = \frac{SSE(F)}{df(F)} = 877.7756$$

$$Var(\bar{Y}_{ijk.}) = \frac{MSE}{n_{ijk}}$$

$$Var(\hat{\mu}_{..1}) = Var\left(\frac{\sum_{i=1}^a \sum_{j=1}^b \bar{Y}_{ij1.}}{ab}\right) = \frac{MSE}{a^2 b^2} \sum_{i=1}^a \sum_{j=1}^b \frac{1}{n_{ij1}}$$

$$S(\hat{D}) = \sqrt{Var(\hat{\mu}_{..1}) + Var(\hat{\mu}_{..2})} = \sqrt{\frac{MSE}{a^2 b^2} \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^c \frac{1}{n_{ijk}}} = 7.885161$$

$$t = t(1 - 0.05/2; 45) = 2.014103$$

Therefore, $\hat{D} \pm S(\hat{D}) * t$, which means $143.7185 \leq D \leq 175.4815$

5 24.18

```

n = 5
s = sqrt(29^2/(n*b*c)*2)
B = qt(1 - .1/(2*5), n*a*b*c - a*b*c)
B*s

## [1] 18.01992

n = 7
s = sqrt(29^2/(n*a*c)*2)
B = qt(1 - .1/(2*5), n*a*b*c - a*b*c)
B*s

## [1] 18.44065

n = 5
s = sqrt(29^2/(n*a*b)*2)
B = qt(1 - .1/(2*5), n*a*b*c - a*b*c)
B*s

## [1] 18.01992

```

We find that

- for L_1 , if the precision of each of the estimates should not exceed, the smallest sample size is 5
- for L_2, L_3 and L_4 , if the precision of each of the estimates should not exceed, the smallest sample size is 7
- for L_1 , if the precision of each of the estimates should not exceed, the smallest sample size is 5.

Therefore, the required sample size should be $n \geq 7$.

6 24.19

$$\begin{aligned}
 \sum_i (\alpha\beta\gamma)_{ijk} &= \sum_i (\mu_{ijk} - \mu_{ij\cdot} - \mu_{i\cdot k} - \mu_{\cdot jk} + \mu_{i\cdot\cdot} + \mu_{\cdot j\cdot} + \mu_{\cdot\cdot k} - \mu_{\cdot\cdot\cdot}) \\
 &= a\mu_{\cdot jk} - a\mu_{\cdot j\cdot} - a\mu_{\cdot\cdot k} - a\mu_{\cdot jk} + a\mu_{\cdot\cdot\cdot} + a\mu_{\cdot j\cdot} + a\mu_{\cdot\cdot k} - a\mu_{\cdot\cdot\cdot} \\
 &= 0
 \end{aligned}$$

7 24.20

The model without three-factor interaction is:

$$Y_{ijk} = \mu_{\cdot\cdot\cdot} + \alpha_i + \beta_j + \gamma_k + (\alpha\beta)_{ij} + (\alpha\gamma)_{ik} + (\beta\gamma)_{jk} + \epsilon_{ijk}$$

	SS	d.f.	MS
A	SSA	a-1	MSA
B	SSB	b-1	MSB
C	SSC	c-1	MSC
AB	SSAB	(a-1)(b-1)	MSAB
AC	SSAC	(a-1)(c-1)	MSAC
BC	SSBC	(b-1)(c-1)	MSBC
Residual	SSE	(a-1)(b-1)(c-1)	MSE
Total	SSTO	abc-1	

8 24.21

$$\begin{aligned}
Var(\hat{L}) &= Var(\sum \sum c_{ij} \bar{Y}_{ij..}) \\
&= \sum \sum c_{ij}^2 Var(\bar{Y}_{ij..}) \\
&= \sum \sum c_{ij}^2 \frac{\sigma^2}{cn} \\
&= \frac{\sigma^2}{cn} \sum \sum c_{ij}^2
\end{aligned}$$