

Stats_101B_HW_5_Charles_Liu (Dis. 3A)

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5/25/2020

Loading Necessary Packages:

```
library(DoE.base)
library(unrepx)
```

Problem 1: (Exercise 7.6)

```
# Set up the problem
response1 <- c(7,9,34,55,16,20,40,60,
              8,10,32,50,18,21,44,61,
              8,12,35,52,15,22,45,65,
              6,10,30,53,15,20,41,63)

k=5
contrast1 <- contr.FrF2(2^(k))
colnames(contrast1) <- c("A", "B", "AB", "C", "AC", "BC", "ABC", "D", "AD",
"BD", "ABD", "CD", "ACD", "BCD", "ABCD", "E", "AE", "BE", "ABE", "CE", "ACE", "BCE",
"ABCE", "DE", "ADE", "BDE", "ABDE", "CDE", "ACDE", "BCDE", "ABCDE")
contrast1
```

##	A	B	AB	C	AC	BC	ABC	D	AD	BD	ABD	CD	ACD	BCD	ABCD	E	AE	BE	ABE	CE	ACE	BCE
## 1	-1	-1	1	-1	1	1	-1	-1	1	1	-1	1	-1	-1	1	-1	1	1	-1	1	-1	-1
## 2	1	-1	-1	-1	-1	1	1	-1	-1	1	1	1	1	-1	-1	-1	-1	1	1	1	1	-1
## 3	-1	1	-1	-1	1	-1	1	-1	1	-1	1	1	-1	1	-1	-1	1	-1	1	1	-1	1
## 4	1	1	1	-1	-1	-1	-1	-1	-1	-1	-1	1	1	1	1	-1	-1	-1	-1	1	1	1
## 5	-1	-1	1	1	-1	-1	1	-1	1	1	-1	-1	1	1	-1	-1	1	1	-1	-1	1	1
## 6	1	-1	-1	1	1	-1	-1	-1	-1	1	1	-1	-1	1	1	-1	-1	1	1	-1	-1	1
## 7	-1	1	-1	1	-1	1	-1	-1	1	-1	1	-1	1	-1	1	-1	1	-1	1	-1	1	-1
## 8	1	1	1	1	1	1	1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1	-1
## 9	-1	-1	1	-1	1	1	-1	1	-1	-1	1	-1	1	1	-1	-1	1	1	-1	1	-1	-1
## 10	1	-1	-1	-1	-1	1	1	1	1	-1	-1	-1	-1	1	1	-1	-1	1	1	1	1	-1
## 11	-1	1	-1	-1	1	-1	1	1	-1	1	-1	-1	1	-1	1	-1	1	-1	1	1	-1	1
## 12	1	1	1	-1	-1	-1	-1	1	1	1	1	1	-1	-1	-1	-1	-1	-1	-1	1	1	1
## 13	-1	-1	1	1	-1	-1	1	1	-1	-1	1	1	-1	-1	1	-1	1	1	-1	-1	1	1
## 14	1	-1	-1	1	1	-1	-1	1	1	-1	-1	1	1	-1	-1	-1	-1	1	1	-1	-1	1
## 15	-1	1	-1	1	-1	1	-1	1	-1	1	-1	1	-1	1	-1	-1	1	-1	1	-1	1	-1
## 16	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	-1	-1	-1	-1	-1	-1
## 17	-1	-1	1	-1	1	1	-1	-1	1	1	-1	1	-1	-1	1	1	-1	-1	1	-1	1	1
## 18	1	-1	-1	-1	-1	1	1	-1	-1	1	1	1	1	-1	-1	1	1	-1	-1	-1	-1	1
## 19	-1	1	-1	-1	1	-1	1	-1	1	-1	1	1	-1	1	-1	1	-1	1	1	-1	1	-1

```
## 20  1  1  1 -1 -1 -1 -1 -1 -1 -1 -1  1  1  1  1  1  1  1 -1 -1 -1
## 21 -1 -1  1  1 -1 -1  1 -1  1  1 -1 -1  1  1 -1  1 -1 -1  1 -1
## 22  1 -1 -1  1  1 -1 -1 -1 -1  1  1 -1 -1  1  1  1  1 -1  1  1 -1
## 23 -1  1 -1  1 -1  1 -1 -1  1 -1  1 -1  1 -1  1  1 -1  1 -1  1
## 24  1  1  1  1  1  1  1  1 -1 -1 -1 -1 -1 -1 -1  1  1  1  1  1
## 25 -1 -1  1 -1  1  1 -1  1 -1 -1  1 -1  1  1 -1  1 -1  1  1  1
## 26  1 -1 -1 -1 -1  1  1  1  1 -1 -1 -1 -1  1  1  1  1 -1 -1  1
## 27 -1  1 -1 -1  1 -1  1  1 -1  1 -1 -1  1 -1  1  1 -1  1 -1
## 28  1  1  1 -1 -1 -1 -1  1  1  1  1 -1 -1 -1 -1 -1  1  1 -1
## 29 -1 -1  1  1 -1 -1  1  1 -1 -1  1  1 -1 -1  1  1 -1  1 -1
## 30  1 -1 -1  1  1 -1 -1  1  1 -1 -1  1  1  1 -1 -1  1  1 -1
## 31 -1  1 -1  1 -1  1 -1  1 -1  1 -1  1  1 -1  1 -1  1 -1  1
## 32  1  1  1  1  1  1  1  1  1  1  1  1  1  1  1  1  1  1  1
```

```
##      ABCE DE ADE BDE ABDE CDE ACDE BCDE ABCDE
```

```
## 1      1  1 -1 -1  1 -1  1  1 -1
## 2     -1  1  1 -1 -1 -1 -1  1  1
## 3     -1  1 -1  1 -1 -1  1 -1  1
## 4      1  1  1  1  1 -1 -1 -1 -1
## 5     -1  1 -1 -1  1  1 -1 -1  1
## 6      1  1  1 -1 -1  1  1 -1 -1
## 7      1  1 -1  1 -1  1 -1  1 -1
## 8     -1  1  1  1  1  1  1  1  1
## 9      1 -1  1  1 -1  1 -1 -1  1
## 10    -1 -1 -1  1  1  1  1 -1 -1
## 11    -1 -1  1 -1  1  1 -1  1 -1
## 12     1 -1 -1 -1 -1  1  1  1  1
## 13    -1 -1  1  1 -1 -1  1  1 -1
## 14     1 -1 -1  1  1 -1 -1  1  1
## 15     1 -1  1 -1  1 -1  1 -1  1
## 16    -1 -1 -1 -1 -1 -1 -1 -1 -1
## 17    -1 -1  1  1 -1  1 -1 -1  1
## 18     1 -1 -1  1  1  1  1 -1 -1
## 19     1 -1  1 -1  1  1 -1  1 -1
## 20    -1 -1 -1 -1 -1  1  1  1  1
## 21     1 -1  1  1 -1 -1  1  1 -1
## 22    -1 -1 -1  1  1 -1 -1  1  1
## 23    -1 -1  1 -1  1 -1  1 -1  1
## 24     1 -1 -1 -1 -1 -1 -1 -1 -1
## 25    -1  1 -1 -1  1 -1  1  1 -1
## 26     1  1  1 -1 -1 -1 -1  1  1
## 27     1  1 -1  1 -1 -1  1 -1  1
## 28    -1  1  1  1  1 -1 -1 -1 -1
## 29     1  1 -1 -1  1  1 -1 -1  1
## 30    -1  1  1 -1 -1  1  1 -1 -1
## 31    -1  1 -1  1 -1  1 -1  1 -1
## 32     1  1  1  1  1  1  1  1  1
```

```
# Estimated Effects w/out Blocks
```

```
effect1 <- t(response1)%*%contrast1/(16*1)
```

```
names(effect1) <- c("A", "B", "AB", "C", "AC", "BC", "ABC", "D", "AD",
"BD", "ABD", "CD", "ACD", "BCD", "ABCD", "E", "AE", "BE", "ABE", "CE", "ACE", "BCE",
"ABCE", "DE", "ADE", "BDE", "ABDE", "CDE", "ACDE", "BCDE", "ABCDE")
```

```
effect1
```

```
##      A      B      AB      C      AC      BC      ABC      D      AD
```

```
## [1,] 11.8125 33.9375 7.9375 9.6875 0.4375 0.0625 -0.4375 -0.8125 -0.0625
##          BD      ABD      CD      ACD      BCD      ABCD      E      AE      BE      ABE
## [1,] -0.6875 0.3125 0.8125 -0.4375 0.4375 -0.0625 0.4375 0.9375 0.5625 -0.1875
##          CE      ACE      BCE      ABCE      DE      ADE      BDE      ABDE      CDE      ACDE
## [1,] 0.3125 0.3125 0.9375 0.1875 -1.1875 0.8125 0.1875 0.9375 -0.8125 -0.3125
##          BCDE      ABCDE
## [1,] -0.9375 -0.1875
## attr(,"names")
## [1] "A"      "B"      "AB"     "C"      "AC"     "BC"     "ABC"    "D"      "AD"
## [10] "BD"     "ABD"    "CD"     "ACD"    "BCD"    "ABCD"   "E"      "AE"     "BE"
## [19] "ABE"    "CE"     "ACE"    "BCE"    "ABCE"   "DE"     "ADE"    "BDE"    "ABDE"
## [28] "CDE"    "ACDE"   "BCDE"   "ABCDE"

# Adding in Blocks
block1 <- contrast1[,c("ABCDE")]
da1 <- data.frame(contrast1,block1,response1)
model1 <- aov(response1~factor(A)*factor(B)*factor(C)*factor(D)*factor(E)+factor(block1),data=da1)
summary(model1)
```

```
##                                     Df Sum Sq Mean Sq
## factor(A)                          1    1116      1116
## factor(B)                          1    9214      9214
## factor(C)                          1     751       751
## factor(D)                          1        5         5
## factor(E)                          1         2          2
## factor(block1)                     1         0          0
## factor(A):factor(B)                 1     504       504
## factor(A):factor(C)                 1         2          2
## factor(B):factor(C)                 1         0          0
## factor(A):factor(D)                 1         0          0
## factor(B):factor(D)                 1         4          4
## factor(C):factor(D)                 1         5          5
## factor(A):factor(E)                 1         7          7
## factor(B):factor(E)                 1         3          3
## factor(C):factor(E)                 1         1          1
## factor(D):factor(E)                 1        11         11
## factor(A):factor(B):factor(C)       1         2          2
## factor(A):factor(B):factor(D)       1         1          1
## factor(A):factor(C):factor(D)       1         2          2
## factor(B):factor(C):factor(D)       1         2          2
## factor(A):factor(B):factor(E)       1         0          0
## factor(A):factor(C):factor(E)       1         1          1
## factor(B):factor(C):factor(E)       1         7          7
## factor(A):factor(D):factor(E)       1         5          5
## factor(B):factor(D):factor(E)       1         0          0
## factor(C):factor(D):factor(E)       1         5          5
## factor(A):factor(B):factor(C):factor(D) 1         0          0
## factor(A):factor(B):factor(C):factor(E) 1         0          0
## factor(A):factor(B):factor(D):factor(E) 1         7          7
## factor(A):factor(C):factor(D):factor(E) 1         1          1
## factor(B):factor(C):factor(D):factor(E) 1         7          7
```

```
# Estimated effects w/ Blocks
effect1_a <- t(response1)%*%cbind(contrast1)/(16*1)
colnames(effect1_a)[colnames(effect1_a) == "ABCDE"] <- "ABCDE+block1"
```

```
effect1_a # -0.1875 is the estimate of (block effect + effect of interaction ABCDE)
```

```
##          A          B          AB          C          AC          BC          ABC          D          AD
## [1,] 11.8125 33.9375 7.9375 9.6875 0.4375 0.0625 -0.4375 -0.8125 -0.0625
##          BD          ABD          CD          ACD          BCD          ABCD          E          AE          BE          ABE
## [1,] -0.6875 0.3125 0.8125 -0.4375 0.4375 -0.0625 0.4375 0.9375 0.5625 -0.1875
##          CE          ACE          BCE          ABCE          DE          ADE          BDE          ABDE          CDE          ACDE
## [1,] 0.3125 0.3125 0.9375 0.1875 -1.1875 0.8125 0.1875 0.9375 -0.8125 -0.3125
##          BCDE ABCDE+block1
## [1,] -0.9375          -0.1875
```

```
effect1_b <- 2*coef(lm(response1~A*B*C*D*E+block1,data=da1))[-1]
effect1_b
```

```
##          A          B          C          D          E          block1          A:B          A:C
## 11.8125 33.9375 9.6875 -0.8125 0.4375 -0.1875 7.9375 0.4375
##          B:C          A:D          B:D          C:D          A:E          B:E          C:E          D:E
## 0.0625 -0.0625 -0.6875 0.8125 0.9375 0.5625 0.3125 -1.1875
##          A:B:C          A:B:D          A:C:D          B:C:D          A:B:E          A:C:E          B:C:E          A:D:E
## -0.4375 0.3125 -0.4375 0.4375 -0.1875 0.3125 0.9375 0.8125
##          B:D:E          C:D:E          A:B:C:D          A:B:C:E          A:B:D:E          A:C:D:E          B:C:D:E          A:B:C:D:E
## 0.1875 -0.8125 -0.0625 0.1875 0.9375 -0.3125 -0.9375          NA
```

```
# Estimates of effects for treatment factors.
```

```
# In other words, remove the highest-order interaction and the blocking factor, as they are confounded.
```

```
effect1_c <- effect1_a[setdiff(names(effect1_a),c("block","A:B:C:D:E"))]
effect1_c
```

```
## numeric(0)
```

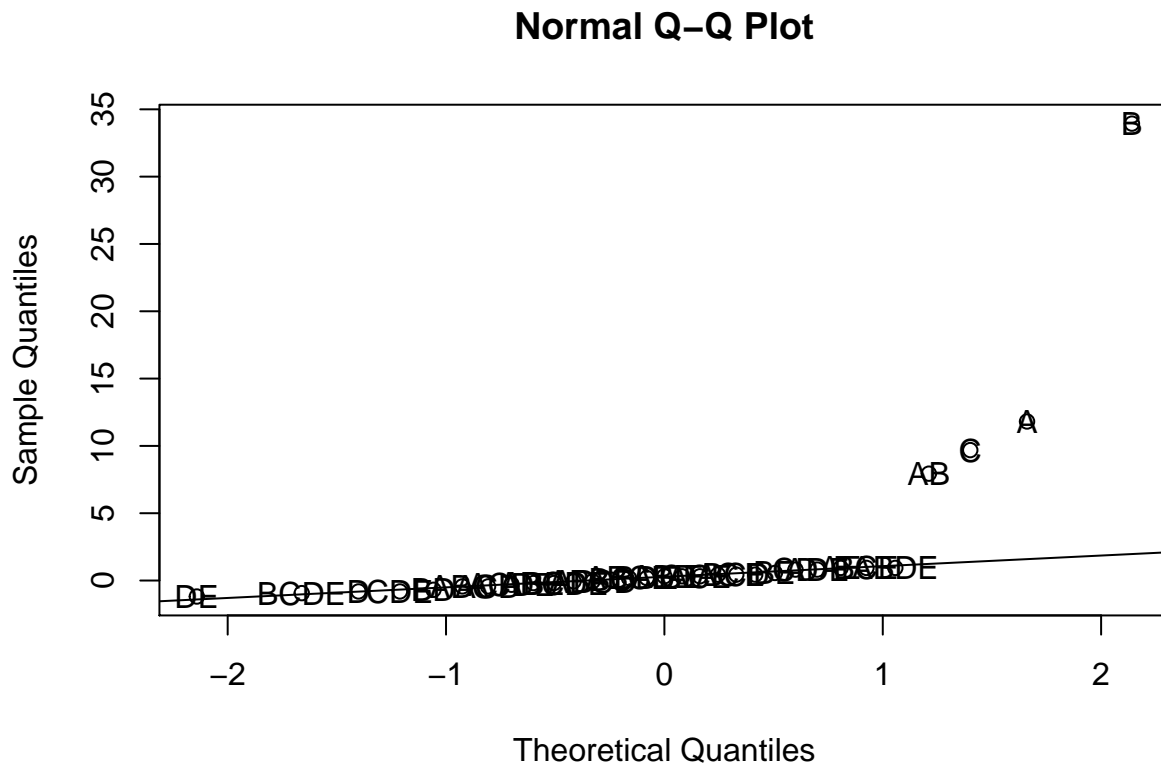
```
# using Lenth's method
```

```
t.tests1 <- eff.test(effect1, method = "Lenth")
t.tests1[, -5]
```

```
##          effect Lenth_PSE t.ratio p.value
## B          33.9375    0.65625 51.714 0.0000
## A          11.8125    0.65625 18.000 0.0000
## C           9.6875    0.65625 14.762 0.0000
## AB          7.9375    0.65625 12.095 0.0000
## DE         -1.1875    0.65625 -1.810 0.0808
## BCDE        -0.9375    0.65625 -1.429 0.1571
## ABDE         0.9375    0.65625 1.429 0.1571
## BCE          0.9375    0.65625 1.429 0.1571
## AE           0.9375    0.65625 1.429 0.1571
## CDE         -0.8125    0.65625 -1.238 0.2124
## ADE          0.8125    0.65625 1.238 0.2124
## CD           0.8125    0.65625 1.238 0.2124
## D          -0.8125    0.65625 -1.238 0.2124
## BD         -0.6875    0.65625 -1.048 0.2864
## BE          0.5625    0.65625 0.857 0.3792
## E           0.4375    0.65625 0.667 0.5212
## BCD          0.4375    0.65625 0.667 0.5212
## ACD         -0.4375    0.65625 -0.667 0.5212
## ABC         -0.4375    0.65625 -0.667 0.5212
## AC           0.4375    0.65625 0.667 0.5212
## ACDE        -0.3125    0.65625 -0.476 0.6492
```

```
## ACE      0.3125    0.65625    0.476    0.6492
## CE       0.3125    0.65625    0.476    0.6492
## ABD      0.3125    0.65625    0.476    0.6492
## ABCDE   -0.1875    0.65625   -0.286    0.7865
## BDE      0.1875    0.65625    0.286    0.7865
## ABCE     0.1875    0.65625    0.286    0.7865
## ABE     -0.1875    0.65625   -0.286    0.7865
## ABCD    -0.0625    0.65625   -0.095    0.9259
## AD      -0.0625    0.65625   -0.095    0.9259
## BC       0.0625    0.65625    0.095    0.9259
```

```
# Using normal probability plot
qqnorm(effect1)
text(qqnorm(effect1)$x,qqnorm(effect1)$y, names(effect1))
qqline(effect1)
```



ANOVA does not print the ABCDE interaction, since it is confounded with blocking. Also, the estimated block effect includes ABCDE interaction effect. (-0.1875) is the estimate of (block effect + effect of interaction ABCDE). We see that “A, B, C, AB” are significant factors from our T-test using Lenth’s Method. We create our Normal Q-Q Plot and see that “A, B, C, AB” are significant.

Problem 2: (Exercise 7.24)

```
mx1_b <- matrix(names(effect1_b), nrow=8, ncol=4, byrow=FALSE)
mx1_b
```

```
##      [,1]      [,2]  [,3]      [,4]
## [1,] "A"      "B:C" "A:B:C" "B:D:E"
## [2,] "B"      "A:D" "A:B:D" "C:D:E"
## [3,] "C"      "B:D" "A:C:D" "A:B:C:D"
## [4,] "D"      "C:D" "B:C:D" "A:B:C:E"
## [5,] "E"      "A:E" "A:B:E" "A:B:D:E"
## [6,] "block1" "B:E" "A:C:E" "A:C:D:E"
## [7,] "A:B"    "C:E" "B:C:E" "B:C:D:E"
## [8,] "A:C"    "D:E" "A:D:E" "A:B:C:D:E"
```

The answer is (d) “BE” is in the same block as “ACDE”. We can see that from the 6th row that “B:E” is along the same row as “A:C:D:E”.

Problem 3:

```
# Choice 1: B_1 = {1,2,3}, B_2 = {4,5,6,7}, B_3 = {3,4,5}
# Choice 2: B_1 = {1,2,3,4}, B_2 = {3,4}, B_3 = {5,6,7}
# Choice 3: B_1 = {1,2,5,6}, B_2 = {1,2,3,4}, B_3 = {1,3,5,7}

# C1 --> (B_1)(B_2)(B_3) --> (1,2,3)(4,5,6,7)(3,4,5) = (1,2,6,7)
# C2 --> (B_1)(B_2)(B_3) --> (1,2,3,4)(3,4)(5,6,7) = (1,2,5,6,7)
# C3 --> (B_1)(B_2)(B_3) --> (1,2,5,6)(1,2,3,4)(1,3,5,7) = (4,6,7)

# 3 Block Effects!
# In Conclusion,...

# C3 < C1 < C2
```

Since we are using 3 Block Effects, we can see that Choice 3 (C3) is the worst one of the choices. After that, we see that Choice 1 (C1) is the next worst with a confounding effect of (1,2,6,7). Lastly, we see that Choice 2 (C2) has the best confounding effect of (1,2,5,6,7) out of all the choices given.