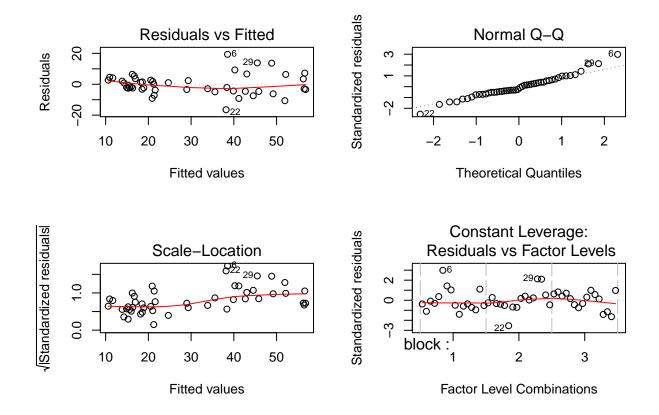
## Poplar Example: Four-way Factorial Analysis

Four way factorial experiment run in a blocked design.

```
library(dplyr)
library(ggplot2)
library(car)
library(emmeans)
Poplar <- read.csv("C:/hess/STAT512/RNotes/ExpDesign2/ED2_Poplar.csv")</pre>
str(Poplar)
                  48 obs. of 6 variables:
## 'data.frame':
## $ block: int 1 1 1 1 1 1 1 1 1 ...
        : int 1 1 1 1 1 1 1 2 2 ...
## $ N
          : int 1111222111...
## $ K
          : int 1 1 2 2 1 1 2 2 1 1 ...
## $ P
         : int 1212121212...
## $ drywt: num 13.9 14.2 14.7 13.6 31.7 57.9 49.5 49.7 15.3 11.8 ...
#Important: Define things as.factor!
Poplar$block <- as.factor(Poplar$block)</pre>
Poplar$L <- as.factor(Poplar$L)</pre>
Poplar$N <- as.factor(Poplar$N)</pre>
Poplar$K <- as.factor(Poplar$K)</pre>
Poplar$P <- as.factor(Poplar$P)</pre>
str(Poplar)
## 'data.frame':
                  48 obs. of 6 variables:
   $ block: Factor w/ 3 levels "1","2","3": 1 1 1 1 1 1 1 1 1 1 ...
        : Factor w/ 2 levels "1", "2": 1 1 1 1 1 1 1 2 2 ...
## $ N
         : Factor w/ 2 levels "1","2": 1 1 1 1 2 2 2 2 1 1 ...
         : Factor w/ 2 levels "1", "2": 1 1 2 2 1 1 2 2 1 1 ...
## $ K
          : Factor w/ 2 levels "1", "2": 1 2 1 2 1 2 1 2 1 2 ...
## $ drywt: num 13.9 14.2 14.7 13.6 31.7 57.9 49.5 49.7 15.3 11.8 ...
SumStats <- summarize(group_by(Poplar, N, L, K, P),</pre>
                  = n(),
              n
              mean = mean(drywt),
              sd = sd(drywt),
                  = sd/sqrt(n))
SumStats
## # A tibble: 16 x 8
              N, L, K [?]
## # Groups:
##
           L
                K
                     Ρ
                                                  SE
                                n mean
                                           sd
     <fct> <fct> <fct> <fct> <fct> <int> <dbl>
##
                                        <dbl>
                                               <dbl>
##
  1 1
                            3 14.7 1.00
                                               0.578
         1 1 1
## 2 1
               1
                      2
                               3 19.7 4.78
                                               2.76
## 3 1
                2
                               3 13.6 0.985 0.569
           1
                      1
           1 2
                      2
                               3 13.9 1.43
## 4 1
                                               0.825
## 5 1
          2 1 1
                               3 16.9 2.21
                                               1.28
          2 1 2
## 6 1
                               3 19.2 6.44
                                               3.72
          2 2 1
## 7 1
                               3 19.9
                                        1.89
                                               1.09
## 8 1
           2 2
                      2
                              3 17.3
                                        2.57
                                               1.48
## 9 2
                    1
                              3 27.7 3.49
                                               2.02
```

```
## 10 2
                      2
           1
                1
                                3 36.9 18.8
                                               10.9
## 11 2
           1
                 2
                      1
                                3 38.6 9.77
                                                5.64
## 12 2
          1
                      2
                                3 41.4 7.27
                                                4.20
## 13 2
           2
                                3 44.2 13.9
                                                8.02
                 1
                      1
## 14 2
           2
                 1
                       2
                                3 47.5 13.3
                                                7.67
## 15 2
           2
                 2
                                3 54.9 12.0
                                                6.92
                       1
## 16 2
                 2
                       2
                                3 55.1 2.92
                                                1.68
#Change contrasts options to get meaningful Type 3 tests!
options(contrasts = c("contr.sum", "contr.poly"))
Model1 <- lm(drywt ~ block + L*N*K*P, data = Poplar)</pre>
Anova(Model1, type = 3)
## Anova Table (Type III tests)
##
## Response: drywt
              Sum Sq Df F value
                                   Pr(>F)
## (Intercept) 43458 1 644.5710 < 2.2e-16 ***
## block
                 215 2
                         1.5978 0.219094
## L
                 885 1 13.1254 0.001064 **
## N
                8350 1 123.8513 3.58e-12 ***
## K
                 147 1
                          2.1855 0.149741
## P
                          1.1840 0.285216
                  80 1
## L:N
                 395 1
                          5.8590 0.021761 *
## L:K
                  22 1
                          0.3304 0.569705
## N:K
                 293 1
                          4.3390 0.045870 *
                          0.5529 0.462924
## L:P
                  37 1
## N:P
                  21 1
                          0.3066 0.583876
## K:P
                  70 1
                          1.0359 0.316918
## L:N:K
                  5 1
                          0.0686 0.795175
## L:N:P
                  1 1
                          0.0213 0.884976
                   2 1
## L:K:P
                          0.0234 0.879476
## N:K:P
                   0 1
                          0.0001 0.993046
                   2 1
                          0.0354 0.852074
## L:N:K:P
## Residuals
                2023 30
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
par(mfrow = c(2, 2))
```

plot(Model1)



After looking at the ANOVA table, follow up with some pairwise comparisons and graphs.

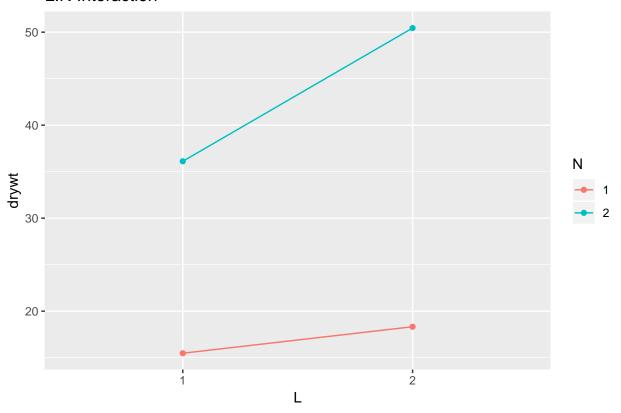
```
## NOTE: Results may be misleading due to involvement in interactions
```

```
## $emmeans
  N = 1:
##
               SE df lower.CL upper.CL
    L emmean
        15.5 2.37 30
##
                          10.6
                                   20.3
        18.3 2.37 30
                          13.5
                                   23.2
##
##
##
  N = 2:
               SE df lower.CL upper.CL
        36.1 2.37 30
                          31.3
##
                                   41.0
                          45.6
                                   55.3
##
        50.4 2.37 30
##
## Results are averaged over the levels of: block, K, P
##
  Confidence level used: 0.95
## $contrasts
    contrast estimate
                         SE df t.ratio p.value
##
                -2.85 3.35 30 -0.850 0.4020
##
## N = 2:
                         SE df t.ratio p.value
```

emmeans(Model1, pairwise ~ L|N)

contrast estimate

## L:N Interaction



```
## NOTE: Results may be misleading due to involvement in interactions
## $emmeans
## N = 1:
## K emmean SE df lower.CL upper.CL
## 1 17.6 2.37 30 12.8 22.5
## 2 16.2 2.37 30 11.3 21.0
##
```

emmeans(Model1, pairwise ~ K|N)

```
## 2 47.5 2.37 30 42.7
                                 52.3
##
## Results are averaged over the levels of: block, L, P
## Confidence level used: 0.95
## $contrasts
## N = 1:
## contrast estimate SE df t.ratio p.value
           1.43 3.35 30 0.428 0.6720
##
## N = 2:
## contrast estimate SE df t.ratio p.value
              -8.44 3.35 30 -2.518 0.0174
##
\mbox{\tt \#\#} Results are averaged over the levels of: block, L, P
SSKN <- aggregate(drywt ~ K + N, data = Poplar, FUN = mean)
SSKN
##
    K N
           drywt
## 1 1 1 17.61667
## 2 2 1 16.18333
## 3 1 2 39.05833
## 4 2 2 47.50000
qplot(x = K, y = drywt, colour = N, group = N, data = SSKN) +
geom_line() + ggtitle("K:N Interaction")
```

## **K:N Interaction**

