

Analysis of Factorial Experiment in R

NE Milla, Jr.

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Loading packages

```
library(agricolae)
library(ExpDes)
library(multcomp)
library(tidyverse)
library(emmeans)
```

Two-factor Factorial in CRD

Loading and managing the data

```
options(digits=6)
crd_data <- read.csv("twofactorial_crd.csv")
head(crd_data)
```

```
##   Pesticide Variety Yield
## 1       P1      V1    49
## 2       P1      V1    39
## 3       P1      V2    55
## 4       P1      V2    41
## 5       P1      V3    66
## 6       P1      V3    68
```

```
crd_data %>%
  group_by(Pesticide) %>%
  summarise(M=mean(Yield))
```

```
## # A tibble: 4 x 2
##   Pesticide     M
##   <chr>     <dbl>
## 1 P1         53
## 2 P2        67.8
## 3 P3        51.2
## 4 P4        73.8
```

```
crd_data %>%
  group_by(Variety) %>%
  summarize(M=mean(Yield))
```

```
## # A tibble: 3 x 2
##   Variety      M
##   <chr>    <dbl>
## 1 V1        46.9
## 2 V2        59.2
## 3 V3        78.2
```

ANOVA using aov()

```
out1 <- aov(Yield ~ Pesticide + Variety + Pesticide:Variety,
             data = crd_data)
```

```
summary(out1)
```

```
##                                Df Sum Sq Mean Sq F value Pr(>F)
## Pesticide                  3   2227    742     17.6 0.00011 ***
## Variety                     2   3996    1998     47.2  2e-06 ***
## Pesticide:Variety          6    457     76     1.8 0.18168
## Residuals                   12   508     42
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
TukeyHSD(out1)
```

```
##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = Yield ~ Pesticide + Variety + Pesticide:Variety, data = crd_data)
##
## $Pesticide
##           diff      lwr      upr   p adj
## P2-P1  14.83333  3.68622 25.98045 0.009001
## P3-P1 -1.83333 -12.98045  9.31378 0.960234
## P4-P1  20.83333  9.68622 31.98045 0.000629
## P3-P2 -16.66667 -27.81378 -5.51955 0.003883
## P4-P2   6.00000 -5.14712 17.14712 0.415509
## P4-P3  22.66667 11.51955 33.81378 0.000297
##
## $Variety
##           diff      lwr      upr   p adj
## V2-V1 12.375  3.70017 21.0498 0.006539
## V3-V1 31.375 22.70017 40.0498 0.000001
## V3-V2 19.000 10.32517 27.6748 0.000216
##
## $`Pesticide:Variety`
##           diff      lwr      upr   p adj
```

```

## P2:V1-P1:V1    8.5 -17.318729 34.31873 0.962324
## P3:V1-P1:V1   -3.5 -29.318729 22.31873 0.999979
## P4:V1-P1:V1    6.5 -19.318729 32.31873 0.994460
## P1:V2-P1:V1    4.0 -21.818729 29.81873 0.999924
## P2:V2-P1:V1   18.5 -7.318729 44.31873 0.269500
## P3:V2-P1:V1    3.5 -22.318729 29.31873 0.999979
## P4:V2-P1:V1   35.0  9.181271 60.81873 0.005485
## P1:V3-P1:V1   23.0 -2.818729 48.81873 0.098292
## P2:V3-P1:V1   44.5 18.681271 70.31873 0.000660
## P3:V3-P1:V1   21.5 -4.318729 47.31873 0.139425
## P4:V3-P1:V1   48.0 22.181271 73.81873 0.000320
## P3:V1-P2:V1  -12.0 -37.818729 13.81873 0.769940
## P4:V1-P2:V1   -2.0 -27.818729 23.81873 1.000000
## P1:V2-P2:V1   -4.5 -30.318729 21.31873 0.999768
## P2:V2-P2:V1   10.0 -15.818729 35.81873 0.902399
## P3:V2-P2:V1   -5.0 -30.818729 20.81873 0.999394
## P4:V2-P2:V1   26.5  0.681271 52.31873 0.042379
## P1:V3-P2:V1   14.5 -11.318729 40.31873 0.559505
## P2:V3-P2:V1   36.0 10.181271 61.81873 0.004344
## P3:V3-P2:V1   13.0 -12.818729 38.81873 0.688133
## P4:V3-P2:V1   39.5 13.681271 65.31873 0.001956
## P4:V1-P3:V1   10.0 -15.818729 35.81873 0.902399
## P1:V2-P3:V1    7.5 -18.318729 33.31873 0.983812
## P2:V2-P3:V1   22.0 -3.818729 47.81873 0.124226
## P3:V2-P3:V1    7.0 -18.818729 32.81873 0.990217
## P4:V2-P3:V1   38.5 12.681271 64.31873 0.002450
## P1:V3-P3:V1   26.5  0.681271 52.31873 0.042379
## P2:V3-P3:V1   48.0 22.181271 73.81873 0.000320
## P3:V3-P3:V1   25.0 -0.818729 50.81873 0.060949
## P4:V3-P3:V1   51.5 25.681271 77.31873 0.000160
## P1:V2-P4:V1   -2.5 -28.318729 23.31873 0.999999
## P2:V2-P4:V1   12.0 -13.818729 37.81873 0.769940
## P3:V2-P4:V1   -3.0 -28.818729 22.81873 0.999996
## P4:V2-P4:V1   28.5  2.681271 54.31873 0.026050
## P1:V3-P4:V1   16.5 -9.318729 42.31873 0.399164
## P2:V3-P4:V1   38.0 12.181271 63.81873 0.002744
## P3:V3-P4:V1   15.0 -10.818729 40.81873 0.517354
## P4:V3-P4:V1   41.5 15.681271 67.31873 0.001257
## P2:V2-P1:V2   14.5 -11.318729 40.31873 0.559505
## P3:V2-P1:V2   -0.5 -26.318729 25.31873 1.000000
## P4:V2-P1:V2   31.0  5.181271 56.81873 0.014211
## P1:V3-P1:V2   19.0 -6.818729 44.81873 0.242656
## P2:V3-P1:V2   40.5 14.681271 66.31873 0.001566
## P3:V3-P1:V2   17.5 -8.318729 43.31873 0.329912
## P4:V3-P1:V2   44.0 18.181271 69.81873 0.000733
## P3:V2-P2:V2  -15.0 -40.818729 10.81873 0.517354
## P4:V2-P2:V2   16.5 -9.318729 42.31873 0.399164
## P1:V3-P2:V2    4.5 -21.318729 30.31873 0.999768
## P2:V3-P2:V2   26.0  0.181271 51.81873 0.047850
## P3:V3-P2:V2    3.0 -22.818729 28.81873 0.999996
## P4:V3-P2:V2   29.5  3.681271 55.31873 0.020430
## P4:V2-P3:V2   31.5  5.681271 57.31873 0.012598
## P1:V3-P3:V2   19.5 -6.318729 45.31873 0.217998
## P2:V3-P3:V2   41.0 15.181271 66.81873 0.001402

```

```

## P3:V3-P3:V2 18.0 -7.818729 43.81873 0.298580
## P4:V3-P3:V2 44.5 18.681271 70.31873 0.000660
## P1:V3-P4:V2 -12.0 -37.818729 13.81873 0.769940
## P2:V3-P4:V2 9.5 -16.318729 35.31873 0.926373
## P3:V3-P4:V2 -13.5 -39.318729 12.31873 0.645435
## P4:V3-P4:V2 13.0 -12.818729 38.81873 0.688133
## P2:V3-P1:V3 21.5 -4.318729 47.31873 0.139425
## P3:V3-P1:V3 -1.5 -27.318729 24.31873 1.000000
## P4:V3-P1:V3 25.0 -0.818729 50.81873 0.060949
## P3:V3-P2:V3 -23.0 -48.818729 2.81873 0.098292
## P4:V3-P2:V3 3.5 -22.318729 29.31873 0.999979
## P4:V3-P3:V3 26.5 0.681271 52.31873 0.042379

```

ANOVA using fact2.crd() in ExpDes package

```

with(crd_data, fat2.crd(factor1 = Pesticide,
                        factor2 = Variety,
                        resp = Yield,
                        fac.names = c("Pesticide", "Variety"),
                        quali = c(TRUE, TRUE),
                        mcomp = "tukey"))

## -----
## Legend:
## FACTOR 1: Pesticide
## FACTOR 2: Variety
## -----
## 
## 
## Analysis of Variance Table
## -----
##           DF   SS  MS   Fc  Pr>Fc
## Pesticide     3 2227  4 17.56 0.00011
## Variety       2 3996  2 47.24 0.00000
## Pesticide*Variety 6  457  5  1.80 0.18168
## Residuals     12  508  3
## Total         23 7188  1
## -----
## CV = 10.58 %
## 
## 
## Shapiro-Wilk normality test
## p-value: 0.0654941
## According to Shapiro-Wilk normality test at 5% of significance, residuals can be considered normal.
## 
## 
## No significant interaction: analyzing the simple effect
## -----
## Pesticide
## Tukey's test
## 
## Groups Treatments Means

```

```

## a      P4      73.8333
## a      P2      67.8333
## b      P1      53
## b      P3      51.1667
## -----
## 
## Variety
## Tukey's test
## -----
## Groups Treatments Means
## a      V3      78.25
## b      V2      59.25
## c      V1      46.875
## -----

```

Two-factorial in RCBD

```

rcbd_data <- read.csv("two_factorial_rcbd.csv")
head(rcbd_data)

##   Day A     B   TS A1 B1
## 1 day1 c1 Mylar  9.2  1  1
## 2 day1 c2 Mylar  8.7  2  1
## 3 day1 c3 Mylar  9.1  3  1
## 4 day1 c4 Mylar 12.4  4  1
## 5 day1 c5 Mylar 10.5  5  1
## 6 day2 c1 Mylar  8.2  1  1

```

ANOVA using aov()

```

out2 <- aov(TS ~ Day + A + B + A:B,
            data = rcbd_data)

summary(out2)

##             Df Sum Sq Mean Sq F value    Pr(>F)
## Day          2   18.6   9.29   11.85 0.00019 ***
## A            4    6.3   1.58    2.02 0.11925
## B            2   10.1   5.07    6.47 0.00490 **
## A:B          8   33.1   4.14    5.28 0.00043 ***
## Residuals   28   22.0   0.78
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

#Post hoc analysis on the treatment combinations

Table of means

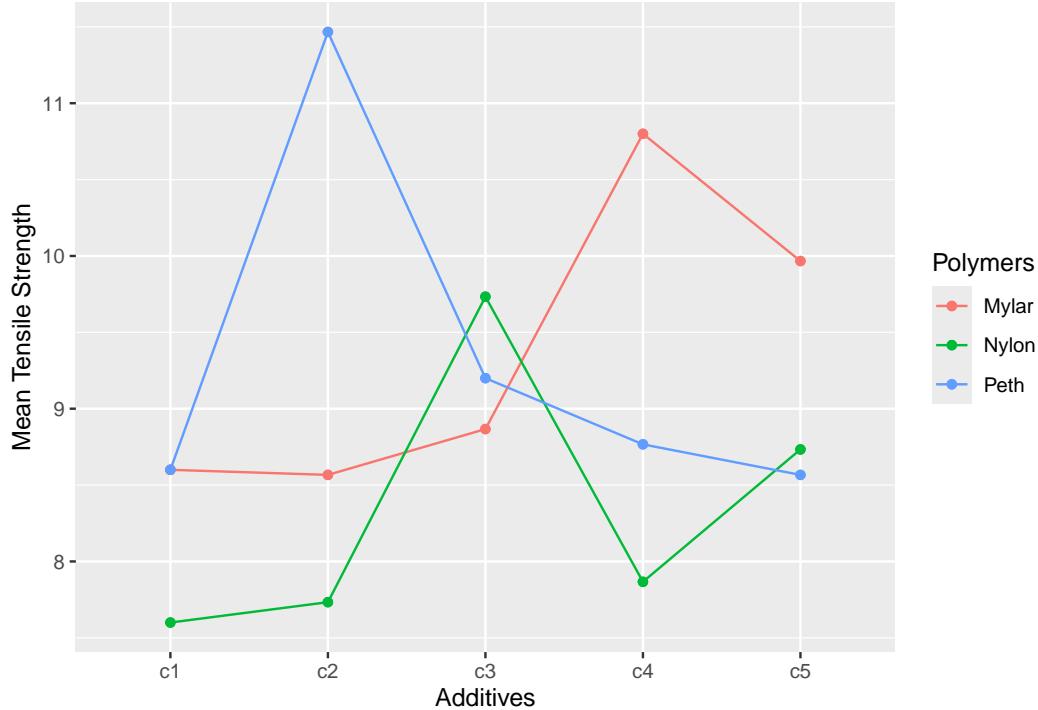
```
rcbd_data %>%
  group_by(A, B) %>%
  summarize(meanTS = round(mean(TS), 2))

## # A tibble: 15 x 3
## # Groups:   A [5]
##   A     B     meanTS
##   <chr> <chr>  <dbl>
## 1 c1    Mylar   8.6 
## 2 c1    Nylon   7.6 
## 3 c1    Peth    8.6 
## 4 c2    Mylar   8.57
## 5 c2    Nylon   7.73
## 6 c2    Peth    11.5 
## 7 c3    Mylar   8.87
## 8 c3    Nylon   9.73
## 9 c3    Peth    9.2  
## 10 c4   Mylar   10.8 
## 11 c4   Nylon   7.87
## 12 c4   Peth    8.77
## 13 c5   Mylar   9.97
## 14 c5   Nylon   8.73
## 15 c5   Peth    8.57
```

Interaction Plot

```
emmmeans::emmpip(out2, B ~ A,
                  xlab = "Additives",
                  ylab = "Mean Tensile Strength",
                  tlab = "Polymers")

## Warning: `aes_()`' was deprecated in ggplot2 3.0.0.
## i Please use tidy evaluation idioms with `aes()`
## i The deprecated feature was likely used in the emmeans package.
##   Please report the issue at <https://github.com/rvlenth/emmeans/issues>.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```



Post hoc analysis of significant interaction

We can use the `emmeans()` function in the **emmeans** package to perform post hoc analysis when interaction effect is significant. A sample code chunk is given below.

```
emmeans(out2, list(pairwise ~ B | A),
        adjust = "tukey",
        lmer.df = "satterthwaite")

## $`emmeans of B | A`
## A = c1:
##   B      emmean     SE df lower.CL upper.CL
##   Mylar    8.60 0.511 28     7.55    9.65
##   Nylon    7.60 0.511 28     6.55    8.65
##   Peth     8.60 0.511 28     7.55    9.65
##
## A = c2:
##   B      emmean     SE df lower.CL upper.CL
##   Mylar    8.57 0.511 28     7.52    9.61
##   Nylon    7.73 0.511 28     6.69    8.78
##   Peth   11.47 0.511 28    10.42   12.51
##
## A = c3:
##   B      emmean     SE df lower.CL upper.CL
##   Mylar    8.87 0.511 28     7.82    9.91
##   Nylon    9.73 0.511 28     8.69   10.78
##   Peth    9.20 0.511 28     8.15   10.25
##
## A = c4:
```

```

##   B      emmean     SE df lower.CL upper.CL
## Mylar  10.80 0.511 28     9.75    11.85
## Nylon  7.87 0.511 28     6.82     8.91
## Peth   8.77 0.511 28     7.72     9.81
##
## A = c5:
##   B      emmean     SE df lower.CL upper.CL
## Mylar  9.97 0.511 28     8.92    11.01
## Nylon  8.73 0.511 28     7.69     9.78
## Peth   8.57 0.511 28     7.52     9.61
##
## Results are averaged over the levels of: Day
## Confidence level used: 0.95
##
## $'pairwise differences of B | A'
## A = c1:
##   2      estimate     SE df t.ratio p.value
## Mylar - Nylon  1.000 0.723 28   1.383  0.3630
## Mylar - Peth   0.000 0.723 28   0.000  1.0000
## Nylon - Peth  -1.000 0.723 28  -1.383  0.3630
##
## A = c2:
##   2      estimate     SE df t.ratio p.value
## Mylar - Nylon  0.833 0.723 28   1.153  0.4906
## Mylar - Peth   -2.900 0.723 28  -4.011  0.0012
## Nylon - Peth  -3.733 0.723 28  -5.164  0.0001
##
## A = c3:
##   2      estimate     SE df t.ratio p.value
## Mylar - Nylon -0.867 0.723 28  -1.199  0.4637
## Mylar - Peth   -0.333 0.723 28  -0.461  0.8899
## Nylon - Peth   0.533 0.723 28   0.738  0.7434
##
## A = c4:
##   2      estimate     SE df t.ratio p.value
## Mylar - Nylon  2.933 0.723 28   4.058  0.0010
## Mylar - Peth   2.033 0.723 28   2.813  0.0234
## Nylon - Peth  -0.900 0.723 28  -1.245  0.4374
##
## A = c5:
##   2      estimate     SE df t.ratio p.value
## Mylar - Nylon  1.233 0.723 28   1.706  0.2207
## Mylar - Peth   1.400 0.723 28   1.937  0.1472
## Nylon - Peth   0.167 0.723 28   0.231  0.9712
##
## Results are averaged over the levels of: Day
## P value adjustment: tukey method for comparing a family of 3 estimates

```

We can manually assign the letter designations, based on the above results.

But we can use the `cld()` function in the **multcomp** package to make letter assignments automatic. This is illustrated in the code chunk below.

```

emm <- emmeans(out2, list(pairwise ~ B | A),
               adjust = "tukey",
               lmer.df = "satterthwaite")
cld(emm,
    alpha=0.05,
    Letters=letters,
    adjust="tukey",
    reversed = T)

## A = c1:
##   B      emmean     SE df lower.CL upper.CL .group
##   Mylar    8.60 0.511 28     7.30     9.90  a
##   Peth     8.60 0.511 28     7.30     9.90  a
##   Nylon    7.60 0.511 28     6.30     8.90  a
##
## A = c2:
##   B      emmean     SE df lower.CL upper.CL .group
##   Peth    11.47 0.511 28    10.17    12.76  a
##   Mylar    8.57 0.511 28     7.27     9.86  b
##   Nylon    7.73 0.511 28     6.44     9.03  b
##
## A = c3:
##   B      emmean     SE df lower.CL upper.CL .group
##   Nylon    9.73 0.511 28     8.44    11.03  a
##   Peth     9.20 0.511 28     7.90    10.50  a
##   Mylar    8.87 0.511 28     7.57    10.16  a
##
## A = c4:
##   B      emmean     SE df lower.CL upper.CL .group
##   Mylar   10.80 0.511 28     9.50    12.10  a
##   Peth     8.77 0.511 28     7.47    10.06  b
##   Nylon    7.87 0.511 28     6.57    9.16  b
##
## A = c5:
##   B      emmean     SE df lower.CL upper.CL .group
##   Mylar   9.97 0.511 28     8.67    11.26  a
##   Nylon    8.73 0.511 28     7.44    10.03  a
##   Peth     8.57 0.511 28     7.27     9.86  a
##
## Results are averaged over the levels of: Day
## Confidence level used: 0.95
## Conf-level adjustment: sidak method for 3 estimates
## P value adjustment: tukey method for comparing a family of 3 estimates
## significance level used: alpha = 0.05
## NOTE: If two or more means share the same grouping symbol,
##       then we cannot show them to be different.
##       But we also did not show them to be the same.

```

ANOVA using fat2.rbd() in the ExpDes package

```

with(rcbd_data, fat2.rbd(factor1 = A,
                         factor2 = B,
                         block = Day,
                         resp = TS,
                         quali = c(TRUE, TRUE),
                         fac.names = c("Additives", "Polymers"),
                         mcomp = "tukey"))

## -----
## Legend:
## FACTOR 1: Additives
## FACTOR 2: Polymers
## -----
## 
## 
## Analysis of Variance Table
## -----
##           DF   SS  MS   Fc  Pr>Fc
## Block      2 18.58 6 11.848 0.00019
## Additives   4  6.33 3  2.017 0.11925
## Polymers    2 10.15 5  6.471 0.00490
## Additives*Polymers 8 33.10 4  5.278 0.00043
## Residuals   28 21.95 2
## Total       44 90.10 1
## -----
## CV = 9.83 %
## 
## 
## Shapiro-Wilk normality test
## p-value: 0.0984005
## According to Shapiro-Wilk normality test at 5% of significance, residuals can be considered normal.
## -----
## 
## 
## Significant interaction: analyzing the interaction
## 
## Analyzing Additives inside of each level of Polymers
## 
## 
## Analysis of Variance Table
## -----
##           DF   SS  MS   Fc  Pr.Fc
## Block      2 18.57644 9.28822 11.8482  2e-04
## Polymers   2 10.14578 5.07289  6.471  0.0049
## Additives:Polymers Mylar 4 11.67600  2.919  3.7235 0.0149
## Additives:Polymers Nylon 4  9.70667 2.42667  3.0955 0.0314
## Additives:Polymers Peth  4 18.04400  4.511  5.7543 0.0016
## Residuals   28 21.95022 0.78394
## Total       44 90.09911
## 
## 

```

```

## 
## 
##   Additives   inside of the level Mylar   of   Polymers
## -----
## Tukey's test
## -----
## Groups Treatments Means
## a      c4      10.8
## ab     c5      9.96667
## ab     c3      8.86667
## b      c1      8.6
## b      c2      8.56667
## -----
## 
## 
## 
##   Additives   inside of the level Nylon   of   Polymers
## -----
## Tukey's test
## -----
## Groups Treatments Means
## a      c3      9.73333
## ab    c5      8.73333
## ab    c4      7.86667
## ab    c2      7.73333
## b      c1      7.6
## -----
## 
## 
## 
##   Additives   inside of the level Peth   of   Polymers
## -----
## Tukey's test
## -----
## Groups Treatments Means
## a      c2      11.46667
## b      c3      9.2
## b      c4      8.76667
## b      c1      8.6
## b      c5      8.56667
## -----
## 
## 
## 
## 
## Analyzing Polymers   inside of each level of   Additives
## -----
## 
## Analysis of Variance Table
## -----
##           DF      SS      MS      Fc  Pr.Fc
## Block          2 18.57644  9.28822 11.8482  2e-04
## Additives       4  6.32578  1.58144  2.0173  0.1192
## Polymers:Additives c1  2  2.00000      1  1.2756  0.295
## Polymers:Additives c2  2 23.04222 11.52111 14.6965      0
## Polymers:Additives c3  2  1.14667  0.57333  0.7314  0.4902
## Polymers:Additives c4  2 13.54889  6.77444  8.6416  0.0012

```

```

## Polymers:Additives c5 2 3.50889 1.75444 2.238 0.1254
## Residuals           28 21.95022 0.78394
## Total              44 90.09911
## -----
## 
## 
## 
## 
## Polymers inside of the level c1 of Additives
## 
## According to the F test, the means of this factor are statistical equal.
## -----
##      Levels     Means
## 1    Mylar      8.6
## 2    Nylon      7.6
## 3    Peth       8.6
## -----
## 
## 
## 
## Polymers inside of the level c2 of Additives
## 
## Tukey's test
## -----
## Groups Treatments Means
## a    Peth     11.4667
## b    Mylar    8.56667
## b    Nylon    7.73333
## -----
## 
## 
## 
## Polymers inside of the level c3 of Additives
## 
## According to the F test, the means of this factor are statistical equal.
## -----
##      Levels     Means
## 1    Mylar     8.86667
## 2    Nylon     9.73333
## 3    Peth      9.20000
## -----
## 
## 
## 
## Polymers inside of the level c4 of Additives
## 
## Tukey's test
## -----
## Groups Treatments Means
## a    Mylar     10.8
## b    Peth      8.76667
## b    Nylon     7.86667
## -----
## 
## 
## 
## Polymers inside of the level c5 of Additives
## 
## According to the F test, the means of this factor are statistical equal.

```

```
## -----  
##      Levels     Means  
## 1    Mylar    9.96667  
## 2    Nylon    8.73333  
## 3    Peth    8.56667  
## -----
```