

Analysis of Factorial Experiment in R

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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) %>%
  summarize(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

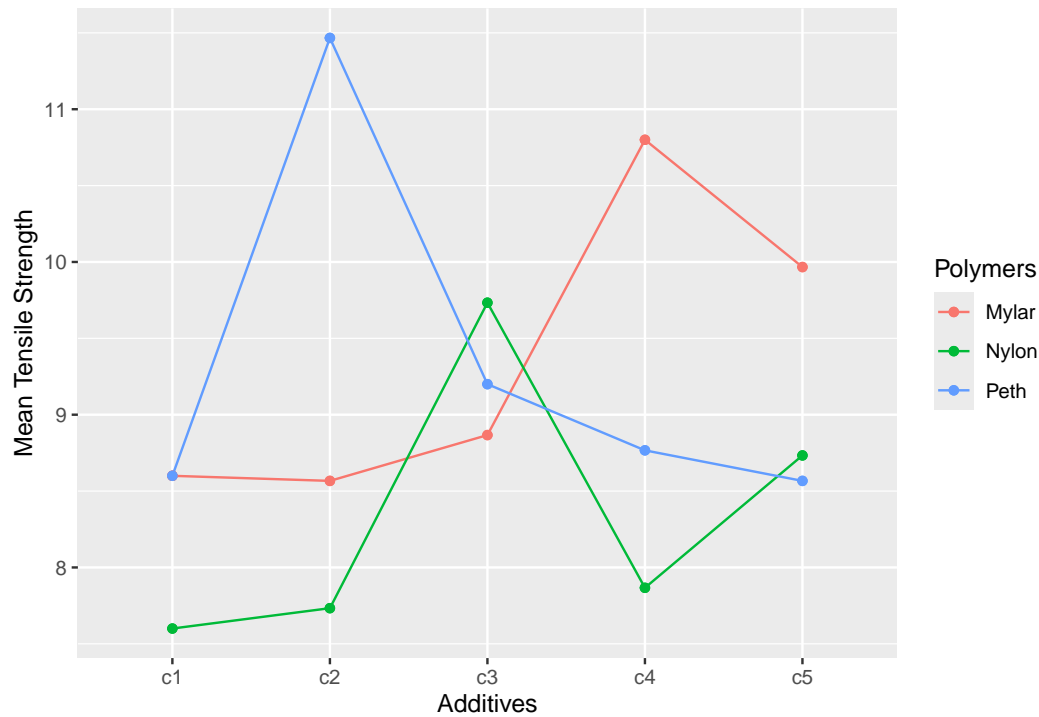
```
rcbd_data %>%  
  group_by(A, B) %>%  
  summarize(meanTS = round(mean(TS),2))
```

```
## # 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

```
emmeans::emmip(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 `clld()` 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      11.4667
## 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
## -----
```