## STAT565 Lab

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## Lab5 Two factor Factorial 1

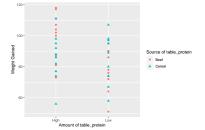
An experiment was designed to study weight gain of rats fed four different diets, where there were two levels of protein (high or low) and two sources of protein (beef or cereal). This gives 2 x 2 treatment combinations: high/beef (HB), high/cereal (HC), low/beef (LB), low/cereal (LC). Ten rats were in each of the four treatment groups. Use  $\alpha = 0.01$ 

- (a) Plot the data and report the plot here (A plot with data and means of treatment combinations). Do not report code here. Describe the observed relationship between two factors.
- (b) Obtain the numerical summary for each treatment combination and factor levels separately. Report them here in a tabular form.
- (c) Fit the two-factor factorial model and report the complete ANOVA table here. Do not report code here. The complete ANOVA table should have a row for each of the following: main effects of each treatment, two-factor interaction effects, error and total.
- (d) Based on the ANOVA table write your conclusion appropriately. Perform all the necessary tests and report the conclusion along with the p-value.
- (e) Provide the plots of residuals here. Do not report code here.
- (f) Based on the residual plots, clearly explain whether assumptions in the model are satisfied or violated.
- (g) Report the code here without output.

```
table_protein <- read_excel("Protein.xlsx")
head(table_protein)</pre>
```

```
## # A tibble: 6 x 3
##
     Source Amount
                     Gain
##
     <chr>>
             <chr>
                     <dbl>
## 1 Beef
                        90
             Low
## 2 Beef
                        76
## 3 Beef
             Low
                        90
## 4 Beef
             Low
                        64
## 5 Beef
                        86
             Low
## 6 Beef
                        51
```

```
# Install and load ggplot2 package before using ggplot function #
ggplot(data = table_protein, aes(x = Amount, y = Gain, colour=Source, group=Source))+
geom_point(aes(shape = Source,color = Source),size = 2) +
labs(y = "Weight Gained", x="Amount of table_protein", color = "Source of table_protein", shape = "Source")
```



```
#Plots the Mean and 1SD error bars for each treatment group #
ggplot(data = table_protein, aes(x = Amount, y = Gain, colour=Source, shape = Source, group=Source)) +
  stat_summary() + labs(y= "Weight Gained", x="Amount of table_protein", color="Source", shape="Source"
## No summary function supplied, defaulting to `mean_se()
# Install and load qqpubr package before using qqline function #
ggline(data = table_protein, x = "Amount", y = "Gain", add = c("mean", "jitter"), shape= "Source", color
#Load mosaic package before using favstats function#
favstats(Gain ~ Source, data=table_protein)
     Source min
                 Q1 median
                              Q3 max mean
                                                sd n missing
                        90 102.5 118 89.6 17.71232 20
      Beef 51 77.5
## 2 Cereal 56 74.0
                        87 95.5 111 84.9 14.99438 20
                                                             0
favstats(Gain ~ Amount , data=table_protein)
     Amount min
                   Q1 median
                                Q3 max mean
                                                   sd n missing
                       93.5 104.75 118 92.95 16.36259 20
      High 56 81.75
                       83.0 91.25 107 81.55 14.63045 20
       Low 51 73.50
favstats(Gain ~ Source Amount, data=table_protein)
##
                       Q1 median
          Amount min
                                     Q3 max
                                              mean
                                                         sd n missing
## 1
      Beef.High 73 90.25 103.0 110.00 118 100.00 15.13642 10
## 2 Cereal.High 56 78.25 87.0 94.25 111 85.90 15.02184 10
                                                                     0
       Beef.Low 51 73.00
                            82.0 90.00 95 79.20 13.88684 10
                            84.5 96.50 107 83.90 15.70881 10
## 4
     Cereal.Low 58 74.00
                                                                     0
## 5
            High 56 81.75
                            93.5 104.75 118 92.95 16.36259 20
                                                                     0
## 6
            Low 51 73.50
                            83.0 91.25 107 81.55 14.63045 20
                                                                      0
#favstats(Gain ~ Source+Amount, data=table_protein)
#Create Categorical variables so that plot of residuals versus each treatment combination can be obtain
table_protein$Trt1 = as.factor(table_protein$Source)
table_protein$Trt2 = as.factor(table_protein$Amount)
model_protein <- aov(Gain ~ Trt1*Trt2, data=table_protein)</pre>
```

```
summary(model_protein)
                Df Sum Sq Mean Sq F value Pr(>F)
##
## Trt1
                             220.9
                                      0.988 0.3269
                      221
## Trt2
                 1
                     1300
                            1299.6
                                      5.812 0.0211 *
## Trt1:Trt2
                 1
                      884
                             883.6
                                      3.952 0.0545 .
## Residuals
                36
                     8049
                             223.6
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
plot(model_protein, pch=16)
                                                                             :
  8
  9
                                                                      1.0
 0
                                                                      9.0
           Fitted values
aov(Gain ~ Trt1 * Trt2)
                                              Theoretical Quantiles
aov(Gain - Trt1 * Trt2)
# Pairwise comparisons using t tests with pooled Standard Deviation #
\# The output gives a matrix of p values for each pair of treatments \#
pairwise.t.test(table_protein$Gain, table_protein$Trt2, p.adj = "none")
##
##
    Pairwise comparisons using t tests with pooled SD
##
## data: table_protein$Gain and table_protein$Trt2
##
##
       High
## Low 0.026
## P value adjustment method: none
# Pairwise comparisons using t tests with pooled Standard Deviation and Bonferroni adjustment #
# The output gives a matrix of p values for each pair of treatments #
pairwise.t.test(table_protein$Gain, table_protein$Trt2, p.adj = "bonf")
##
##
    Pairwise comparisons using t tests with pooled SD
##
## data: table_protein$Gain and table_protein$Trt2
##
##
       High
## Low 0.026
## P value adjustment method: bonferroni
```

#Install and load the agricolae package before running the LSD.test function below #
#p.adj option in the LSD.test function can be used to apply different adjustments to control error rate
plot(LSD.test(model\_protein, trt = "Trt2", alpha = 0.05))

```
Groups and Range
```

```
#The treatments sharing the same letter on the plot are not different#
(LSD.test(model_protein, trt = "Trt2", alpha = 0.05)) # Using outer parentheses prints the output#
## $statistics
##
     MSerror Df Mean
                             CV t.value LSD
##
     223.5944 36 87.25 17.13819 2.028094 9.59
##
## $parameters
##
           test p.ajusted name.t ntr alpha
##
     Fisher-LSD
                            Trt2
                                   2 0.05
                     none
##
## $means
                               LCL
##
         Gain
                   std r
                                        UCL Min Max
                                                      Q25 Q50
## High 92.95 16.36259 20 86.16885 99.73115 56 118 81.75 93.5 104.75
## Low 81.55 14.63045 20 74.76885 88.33115 51 107 73.50 83.0 91.25
## $comparison
## NULL
##
## $groups
##
         Gain groups
## High 92.95
## Low 81.55
##
## attr(,"class")
## [1] "group"
#Tukey's test to get observed difference in means, CI and p value#
TukeyHSD(model_protein, conf.level = 0.95)
##
     Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = Gain ~ Trt1 * Trt2, data = table_protein)
##
## $Trt1
##
               diff
                       lwr upr
                                    p adj
## Cereal-Beef -4.7 -14.29 4.89 0.3268783
## $Trt2
##
             diff
                     lwr
                           upr
                                   p adj
```

```
## Low-High -11.4 -20.99 -1.81 0.0211449
##
## $`Trt1:Trt2`
##
                                     lwr
                           diff
                                               upr
                                                       p adj
## Cereal:High-Beef:High -14.1 -32.1102 3.910198 0.1697711
## Beef:Low-Beef:High
                         -20.8 -38.8102 -2.789802 0.0182745
## Cereal:Low-Beef:High
                        -16.1 -34.1102 1.910198 0.0936982
                        -6.7 -24.7102 11.310198 0.7492577
## Beef:Low-Cereal:High
## Cereal:Low-Cereal:High -2.0 -20.0102 16.010198 0.9905411
## Cereal:Low-Beef:Low
                           4.7 -13.3102 22.710198 0.8952934
# Scheffe's test to get observed difference in means, CI and p value #
# Install and load the DescTools package before using the ScheffeTest function #
ScheffeTest(model_protein, conf.level = 0.95)
##
##
     Posthoc multiple comparisons of means : Scheffe Test
       95% family-wise confidence level
##
##
## $Trt1
##
               diff
                       lwr.ci
                                upr.ci
## Cereal-Beef -4.7 -18.56594 9.165941 0.8042
##
## $Trt2
             diff
                     lwr.ci
                             upr.ci
## Low-High -11.4 -25.26594 2.465941 0.1410
## $`Trt1:Trt2`
                           diff
                                  lwr.ci
                                            upr.ci
                                                    pval
## Cereal: High-Beef: High -14.1 -33.7094 5.509402 0.2358
## Beef:Low-Beef:High
                          -20.8 -40.4094 -1.190598 0.0338 *
## Cereal:Low-Beef:High
                        -16.1 -35.7094 3.509402 0.1418
## Beef:Low-Cereal:High
                          -6.7 -26.3094 12.909402 0.8004
## Cereal:Low-Cereal:High -2.0 -21.6094 17.609402 0.9929
## Cereal:Low-Beef:Low
                           4.7 -14.9094 24.309402 0.9195
##
## ---
```

## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' ' 1