

Contrasts

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Read Dataframe

```
filedir<-"https://raw.githubusercontent.com/Hugo-Toledo/Applied-Statistics-R-UNIPD/master/data/lambs.csv"
es1<-read.table(file = filedir,stringsAsFactors = TRUE,header = TRUE, sep = ",")
```

```
head(es1) #See the Data
```

```
##      treatment  breed  matr birth_weight age_beginning sex weight_jul_.08
## 1 fieno + conc. brogna 26157      2.50      112 F      19.0
## 2 fieno + conc. brogna 26158      2.80      107 M      23.2
## 3 fieno + conc. brogna 26159      2.30      106 F      14.4
## 4 fieno + conc. brogna 26163      3.55      93 M      16.8
## 5 fieno + conc.  foza 26167      3.30      76 M      20.8
## 6 fieno + conc.  foza 26170      4.20      73 F      14.8
## weight_agoust_08 weight_agoust_31 weight_sept_29 weight_october_28
## 1      20.0      22      25.0      25.0
## 2      25.5      27      32.0      32.5
## 3      15.0      17      19.0      18.5
## 4      17.5      19      22.5      23.0
## 5      20.5      24      26.5      27.5
## 6      17.0      19      21.5      22.0
## slaughtering_weight
## 1      25.2
## 2      32.6
## 3      18.4
## 4      23.2
## 5      27.2
## 6      21.6
```

```
table(es1$sex,es1$breed, by=es1$treatment) #Frequency table for factors
```

```
## , , by = fieno + conc.
##
##
##      alpagota brogna foza
## F      2      2      2
## M      2      2      2
##
## , , by = fieno + conc. + CLA
##
##
##      alpagota brogna foza
## F      2      2      2
## M      2      2      2
##
## , , by = pascolo
##
##
##      alpagota brogna foza
## F      2      2      2
## M      2      2      2
```

ANOVA test

```
# Fit the linear model
tm<-lm(slaughtering_weight ~ sex + breed + treatment, data=es1)
summary(tm) # Linear Model Summary

##
## Call:
## lm(formula = slaughtering_weight ~ sex + breed + treatment, data = es1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.5917  -3.5806  -0.2958   3.7687  12.4444
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      24.700      2.388  10.345 2.06e-11 ***
## sexM              5.706      1.949   2.927  0.00647 **
## breedbrogna      -3.208      2.388  -1.344  0.18910
## breedfoza         3.050      2.388   1.277  0.21124
## treatmentfieno + conc. + CLA  1.100      2.388   0.461  0.64832
## treatmentpascolo   6.492      2.388   2.719  0.01078 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.848 on 30 degrees of freedom
## Multiple R-squared:  0.4435, Adjusted R-squared:  0.3507
## F-statistic: 4.782 on 5 and 30 DF,  p-value: 0.002482

fm<-aov(slaughtering_weight ~ sex + breed + treatment, data=es1) # Fit the ANOVA
summary(fm) # ANOVA table

##           Df Sum Sq Mean Sq F value   Pr(>F)
## sex         1  293.0    293.0    8.566 0.00647 **
## breed        2  235.1    117.5    3.436 0.04533 *
## treatment    2  289.7    144.8    4.235 0.02399 *
## Residuals   30 1026.1     34.2
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

LSM

```
# LSM
library(lsmmeans)

## Loading required package: estimability

lsmmeans(tm,"sex")           #LSM for sex

##   sex    lsmean      SE df lower.CL upper.CL
##   F    27.17778 1.378471 30 24.36256 29.99299
##   M    32.88333 1.378471 30 30.06812 35.69855
##
## Results are averaged over the levels of: breed, treatment
## Confidence level used: 0.95

lsmmeans(tm,"breed")        #LSM for breed

##   breed    lsmean      SE df lower.CL upper.CL
##   alpagota 30.08333 1.688275 30 26.63542 33.53125
##   brogna   26.87500 1.688275 30 23.42708 30.32292
##   foza     33.13333 1.688275 30 29.68542 36.58125
##
## Results are averaged over the levels of: sex, treatment
## Confidence level used: 0.95

lsmmeans(tm,"treatment")    #LSM for treatment

##   treatment    lsmean      SE df lower.CL upper.CL
##   fieno + conc. 27.50000 1.688275 30 24.05208 30.94792
##   fieno + conc. + CLA 28.60000 1.688275 30 25.15208 32.04792
##   pascolo      33.99167 1.688275 30 30.54375 37.43958
##
## Results are averaged over the levels of: sex, breed
## Confidence level used: 0.95

# Contrasts
#tm$contrasts$sex
#tm$contrasts$breed
#tm$contrasts$treatment
contrasts(es1$sex)          #Grid for contrasts

##   M
## F 0
## M 1

contrasts(es1$breed)        #Grid for contrasts

##           brogna foza
## alpagota      0    0
## brogna        1    0
## foza          0    1

contrasts(es1$treatment)    #Grid for contrasts

##           fieno + conc. + CLA pascolo
## fieno + conc.              0      0
## fieno + conc. + CLA        1      0
## pascolo                    0      1

fm<-aov(slaughtering_weight ~ sex + breed + treatment + age_beginning, data=es1) # Fit the ANOVA
summary(fm) #ANOVA table

##           Df Sum Sq Mean Sq F value    Pr(>F)
## sex         1  293.0    293.0   13.664 0.000906 ***
```

```
## breed          2  235.1   117.5    5.481 0.009568 **
## treatment      2  289.7   144.8    6.755 0.003904 **
## age_beginning  1  404.3   404.3   18.856 0.000157 ***
## Residuals      29  621.8    21.4
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

tm<-lm(slaughtering_weight ~ sex + breed + treatment + age_beginning, data=es1) # Fit the linear model
summary(tm) # Linear Model Summary

##
## Call:
## lm(formula = slaughtering_weight ~ sex + breed + treatment +
##     age_beginning, data = es1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.3549  -2.4533  -0.3365   2.7244   9.4278
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.66837     5.84818   0.114 0.909799
## sexM              3.76233     1.60706   2.341 0.026309 *
## breedbrogna       2.62136     2.31860   1.131 0.267497
## breedfoza        10.87573     2.61178   4.164 0.000256 ***
## treatmentfieno + conc. + CLA 1.71782     1.89573   0.906 0.372323
## treatmentpascolo    6.26988     1.89107   3.316 0.002466 **
## age_beginning      0.19010     0.04378   4.342 0.000157 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.63 on 29 degrees of freedom
## Multiple R-squared:  0.6628, Adjusted R-squared:  0.593
## F-statistic: 9.499 on 6 and 29 DF,  p-value: 8.57e-06

#contrasts(es1$sex)<-c(1,0)
contrasts(es1$sex)<-contr.SAS #Change contrast grid
#contrasts(es1$breed)<-cbind(c(1,0,0), c(0,1,0))
contrasts(es1$breed)<-contr.SAS #Change contrast grid
#contrasts(es1$treatment)<-cbind(c(1,0,0), c(0,1,0))
contrasts(es1$treatment)<-contr.SAS ##Change contrast grid

fm1<-aov(slaughtering_weight ~ sex + breed + treatment + age_beginning, data=es1) # Fit the ANOVA
summary(fm1) #Anova table

##              Df Sum Sq Mean Sq F value    Pr(>F)
## sex            1  293.0   293.0  13.664 0.000906 ***
## breed          2  235.1   117.5    5.481 0.009568 **
## treatment      2  289.7   144.8    6.755 0.003904 **
## age_beginning  1  404.3   404.3   18.856 0.000157 ***
## Residuals      29  621.8    21.4
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

tm1<-lm(slaughtering_weight ~ sex + breed + treatment + age_beginning, data=es1) # Fit the linear model
summary(tm1) # Linear Model Summary

##
## Call:
## lm(formula = slaughtering_weight ~ sex + breed + treatment +
##     age_beginning, data = es1)
##
```

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.3549  -2.4533  -0.3365   2.7244   9.4278
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    21.57630     4.63376   4.656 6.59e-05 ***
## sex1           -3.76233     1.60706  -2.341 0.026309 *
## breed1        -10.87573     2.61178  -4.164 0.000256 ***
## breed2         -8.25437     1.94547  -4.243 0.000206 ***
## treatment1     -6.26988     1.89107  -3.316 0.002466 **
## treatment2     -4.55206     1.90025  -2.396 0.023275 *
## age_beginning   0.19010     0.04378   4.342 0.000157 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.63 on 29 degrees of freedom
## Multiple R-squared:  0.6628, Adjusted R-squared:  0.593
## F-statistic: 9.499 on 6 and 29 DF,  p-value: 8.57e-06
```

```
car::Anova(tm1, type = "III") #Anova table with SS III
```

```
## Anova Table (Type III tests)
##
## Response: slaughtering_weight
##              Sum Sq Df F value    Pr(>F)
## (Intercept)  464.88  1 21.6814 6.587e-05 ***
## sex          117.52  1  5.4809 0.0263086 *
## breed        500.38  2 11.6685 0.0001914 ***
## treatment    251.26  2  5.8593 0.0072910 **
## age_beginning 404.30  1 18.8561 0.0001568 ***
## Residuals    621.80 29
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
lsmeans(tm1,"sex") #LSM for sex
```

```
## sex  lsmean      SE df lower.CL upper.CL
## F    28.14939 1.114114 29 25.87077 30.42801
## M    31.91172 1.114114 29 29.63310 34.19034
##
## Results are averaged over the levels of: breed, treatment
## Confidence level used: 0.95
```

```
lsmeans(tm1,"breed") #LSM for breed
```

```
## breed  lsmean      SE df lower.CL upper.CL
## alpagota 25.53153 1.698696 29 22.05730 29.00575
## brogna   28.15289 1.368714 29 25.35355 30.95222
## foza     36.40725 1.534671 29 33.26850 39.54601
##
## Results are averaged over the levels of: sex, treatment
## Confidence level used: 0.95
```

```
lsmeans(tm1,"treatment") #LSM for treatment
```

```
## treatment  lsmean      SE df lower.CL upper.CL
## fieno + conc. 27.36799 1.337049 29 24.63341 30.10256
## fieno + conc. + CLA 29.08581 1.341377 29 26.34238 31.82923
## pascolo      33.63787 1.339184 29 30.89893 36.37681
##
## Results are averaged over the levels of: sex, breed
## Confidence level used: 0.95
```

Orthogonal Contrasts

```
#c1<-c(-1,-1,2)
#c2<-c(1,-1, 0)
#mat1<-cbind(c1,c2)
mat1<-matrix(c(-1,-1,2,
               1,-1,0),ncol=2) # Create the contrast for treatment
mat2<-matrix(c(1,-1,0),ncol = 1) # Create the contrast for breed

#contrasts(es1$treatment)<-mat1
#contrasts(es1$breed)<-mat2
modell1<-aov(slaughtering_weight ~ sex + breed + treatment + age_beginning,
            contrasts = list(treatment=mat1, breed=mat2), data=es1) # Fit the linear model
#ANOVA table with the effects splited for each contrast
summary.aov(modell1,split=list(treatment=list("Secco vs Pascolo"=1,"CLA vs noCLA"=2),
                               breed=list("Broghna vs Alpagata"=1)))
```

```
##                               Df Sum Sq Mean Sq F value    Pr(>F)
## sex                          1   293.0   292.98    7.867 0.00875 **
## breed                        1    61.8    61.76    1.658 0.20767
##   breed: Broghna vs Alpagata  1    61.8    61.76    1.658 0.20767
## treatment                    2   289.7   144.84    3.889 0.03149 *
##   treatment: Secco vs Pascolo 1   282.4   282.43    7.583 0.00991 **
##   treatment: CLA vs noCLA     1     7.3     7.26    0.195 0.66200
## age_beginning                1    82.1    82.11    2.205 0.14801
## Residuals                   30  1117.3    37.24
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Polynomial Contrasts#
mat3<-matrix(c(-1,0,1,
               1,-2,1),ncol=2) #Create the contrast for treatment Linear and Quadratic
#contrasts(es1$treatment)<-mat3
#contrasts(es1$breed)<-contr.SAS
modell2<-aov(slaughtering_weight ~ sex + breed + treatment + age_beginning,
            contrasts=list(treatment=mat3),data=es1) # Fit the linear model
#ANOVA table with the effects splited for the contrast
summary.aov(modell2,split=list(treatment=list("Linear"=1,"Quadratic"=2)))
```

```
##                               Df Sum Sq Mean Sq F value    Pr(>F)
## sex                          1   293.0   293.0   13.664 0.000906 ***
## breed                        2   235.1   117.5    5.481 0.009568 **
## treatment                    2   289.7   144.8    6.755 0.003904 **
##   treatment: Linear          1   252.9   252.9   11.793 0.001812 **
##   treatment: Quadratic       1    36.8    36.8    1.718 0.200233
## age_beginning                1   404.3   404.3   18.856 0.000157 ***
## Residuals                   29   621.8    21.4
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#attributes(modell2$qr$qr) # See the model grid
```

```
#~~~~~
# NOTE: Another way to obtain the constrast
#library(multcomp)
#contr<-rbind("Linear"=c(-1,0,1),
#            "Quadratic"=c(1,-2,1))
#modell3<-glht(fm1,linfct=mcp(treatment=contr))
#summary(modell4)
#~~~~~
#####
```