

Initiation à la statistique avec R, code et compléments

chapitre 10

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```
#Chapitre 10  
require(BioStatR)
```

```
## Loading required package: BioStatR
```

```
#page 402  
foret<-rep(1:3,c(10,10,10))  
hauteur<-c(23.4,24.4,24.6,24.9,25,26.2,26.1,24.8,25.5,25.8,18.9,21.1,21.1,  
22.1,22.5,23.5,22.7,21.3,22.2,21.7,22.5,22.9,23.7,24,24,24.5,24.3,24.2,  
23.4,23.9)  
foret<-factor(foret)  
arbre<-data.frame(foret,hauteur)  
rm(foret)  
rm(hauteur)  
arbre
```

```
##      foret hauteur  
## 1         1    23.4  
## 2         1    24.4  
## 3         1    24.6  
## 4         1    24.9  
## 5         1    25.0  
## 6         1    26.2  
## 7         1    26.1  
## 8         1    24.8  
## 9         1    25.5  
## 10        1    25.8  
## 11        2    18.9  
## 12        2    21.1  
## 13        2    21.1  
## 14        2    22.1  
## 15        2    22.5  
## 16        2    23.5  
## 17        2    22.7  
## 18        2    21.3  
## 19        2    22.2  
## 20        2    21.7  
## 21        3    22.5  
## 22        3    22.9  
## 23        3    23.7  
## 24        3    24.0  
## 25        3    24.0  
## 26        3    24.5  
## 27        3    24.3  
## 28        3    24.2  
## 29        3    23.4
```

```
## 30      3      23.9
moyennes<-tapply(arbre$hauteur,arbre$foret,mean)
moyennes
```

```
##      1      2      3
## 25.07 21.71 23.74
```

```
#page 403
variances<-tapply(arbre$hauteur,arbre$foret,var)
variances
```

```
##      1      2      3
## 0.7356667 1.5565556 0.4026667
```

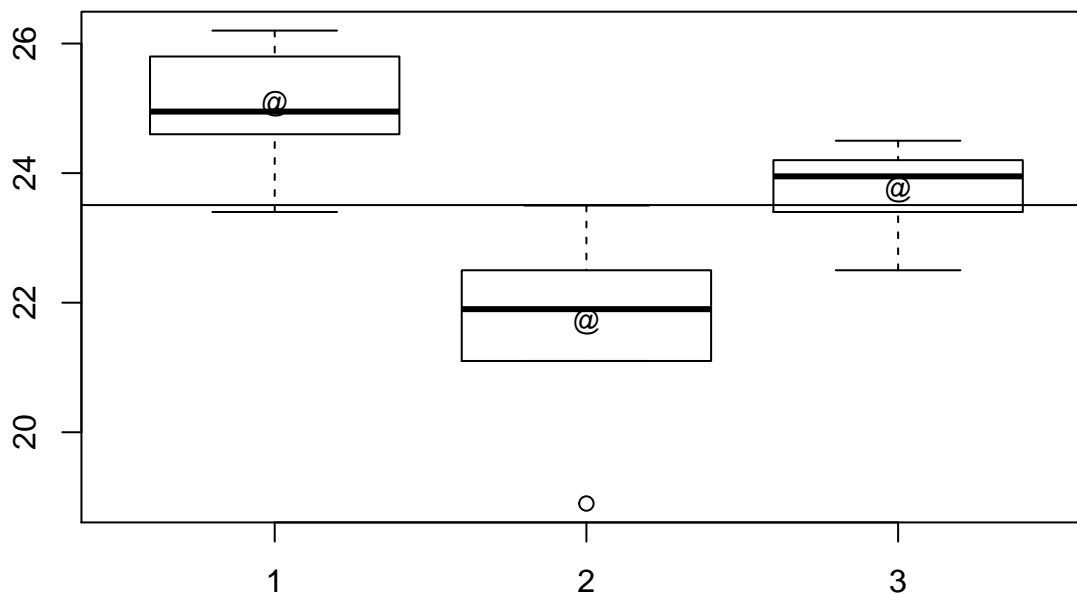
```
#page 404
moy.g<-mean(arbre$hauteur)
moy.g
```

```
## [1] 23.50667
```

```
mean(moyennes)
```

```
## [1] 23.50667
```

```
#page 405
plot(arbre$foret,arbre$hauteur)
points(1:3,moyennes,pch="@")
abline(h=moy.g)
```



```
pdf("ch11fig101.pdf")
plot(arbre$foret,arbre$hauteur)
points(1:3,moyennes,pch="@")
abline(h=moy.g)
dev.off()
```

```
## pdf
## 2
```

```

#page 408
options(contrasts=c("contr.sum","contr.poly"))
modele1<-lm(hauteur~foret,data=arbre)
anova(modele1)

## Analysis of Variance Table
##
## Response: hauteur
##           Df Sum Sq Mean Sq F value    Pr(>F)
## foret      2  57.265  28.6323   31.874 7.809e-08 ***
## Residuals 27  24.254   0.8983
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

modele1_aov<-aov(hauteur~foret,data=arbre)
summary(modele1_aov)

##           Df Sum Sq Mean Sq F value    Pr(>F)
## foret      2   57.26   28.632   31.87 7.81e-08 ***
## Residuals 27   24.25    0.898
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

#page 409
options(contrasts=c("contr.sum","contr.poly"))

#page 410
residus<-residuals(modele1)
shapiro.test(residus)

##
##  Shapiro-Wilk normality test
##
## data:  residus
## W = 0.95202, p-value = 0.1914

length(residus)

## [1] 30

#En plus : les résidus des deux modèles sont égaux
all(residuals(modele1)==residuals(modele1_aov))

## [1] TRUE

#page 412
bartlett.test(residus~foret,data=arbre)

##
##  Bartlett test of homogeneity of variances
##
## data:  residus by foret
## Bartlett's K-squared = 3.8798, df = 2, p-value = 0.1437

coef(modele1)

## (Intercept)      foret1      foret2
##   23.506667    1.563333   -1.796667

```

```

#En plus : les coefficients des deux modèles sont égaux
all(coef(modele1)==coef(modele1_aov))

## [1] TRUE

#page 413
-sum(coef(modele1)[2:3])

## [1] 0.2333333

dummy.coef(modele1)

## Full coefficients are
##
## (Intercept):      23.50667
## foret:              1          2          3
##                1.5633333 -1.7966667  0.2333333

#En plus : fonctionne aussi avec le modèle aov et introduction de la
#fonction model.tables
dummy.coef(modele1_aov)

## Full coefficients are
##
## (Intercept):      23.50667
## foret:              1          2          3
##                1.5633333 -1.7966667  0.2333333

model.tables(modele1_aov)

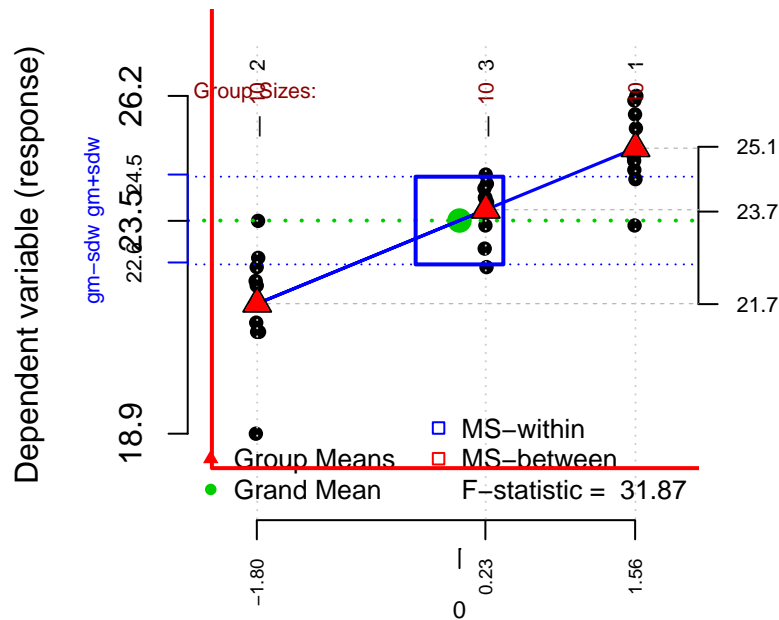
## Tables of effects
##
## foret
## foret
##      1      2      3
## 1.5633 -1.7967 0.2333

if(!("granova" %in% rownames(installed.packages()))){
  install.packages("granova")}
library(granova)

## Loading required package: car
## Loading required package: carData
granova.1w(arbre$hauteur,arbre$foret)

```

One-way ANOVA displaying 3 groups



Contrast coefficients based on group means and sizes

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat        F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
##
## $stats
##      Size Contrast Coef Wt'd Mean  Mean Trim'd Mean Var. St. Dev.
## 2    10          -1.80   21.71 21.71      21.82 1.56    1.25
## 3    10           0.23   23.74 23.74      23.87 0.40    0.63
## 1    10           1.56   25.07 25.07      25.10 0.74    0.86
```

```
pdf("chap10fig102.pdf")
print(granova.1w(arbre$hauteur, arbre$foret))
```

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat        F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
##
## $stats
##      Size Contrast Coef Wt'd Mean  Mean Trim'd Mean Var. St. Dev.
## 2    10          -1.80   21.71 21.71      21.82 1.56    1.25
## 3    10           0.23   23.74 23.74      23.87 0.40    0.63
## 1    10           1.56   25.07 25.07      25.10 0.74    0.86
```

```
dev.off()
```

```
## pdf
## 2
```

```

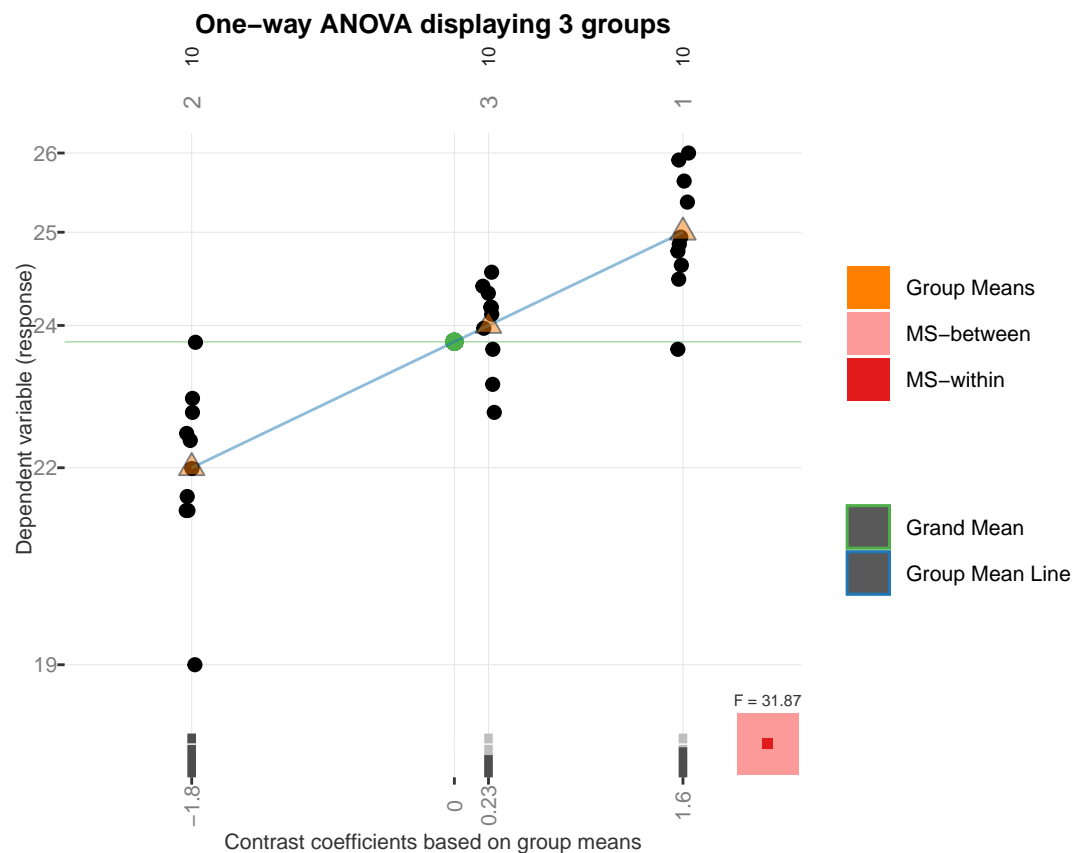
#page 415
if(!("granovaGG" %in% rownames(installed.packages()))){
  install.packages("granovaGG")}
library(granovaGG)

## Loading required package: ggplot2

granovagg.lw(arbre$hauteur,arbre$foret)

##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82    -1.80      1.56              1.25
## 3      3      23.74      23.87     0.23      0.40              0.63
## 1      1      25.07      25.10     1.56      0.74              0.86
##   group.size
## 2           10
## 3           10
## 1           10
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##   Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067     0.1730 135.844 < 2e-16 ***
## group1       1.5633     0.2447   6.388 7.66e-07 ***
## group2      -1.7967     0.2447  -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF, p-value: 7.809e-08

```



```
pdf("chap10fig103.pdf")
print(granovagg.1w(arbre$hauteur, arbre$foret))
```

```
##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82    -1.80      1.56      1.25
## 3      3      23.74      23.87     0.23      0.40      0.63
## 1      1      25.07      25.10     1.56      0.74      0.86
##   group.size
## 2          10
## 3          10
## 1          10
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067     0.1730 135.844 < 2e-16 ***
```

```
## group1      1.5633      0.2447      6.388 7.66e-07 ***
## group2     -1.7967      0.2447     -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF,  p-value: 7.809e-08
dev.off()
```

```
## pdf
## 2
```

```
#page 418
modele2<-aov(hauteur~foret,data=arbre)
model.tables(modele2)
```

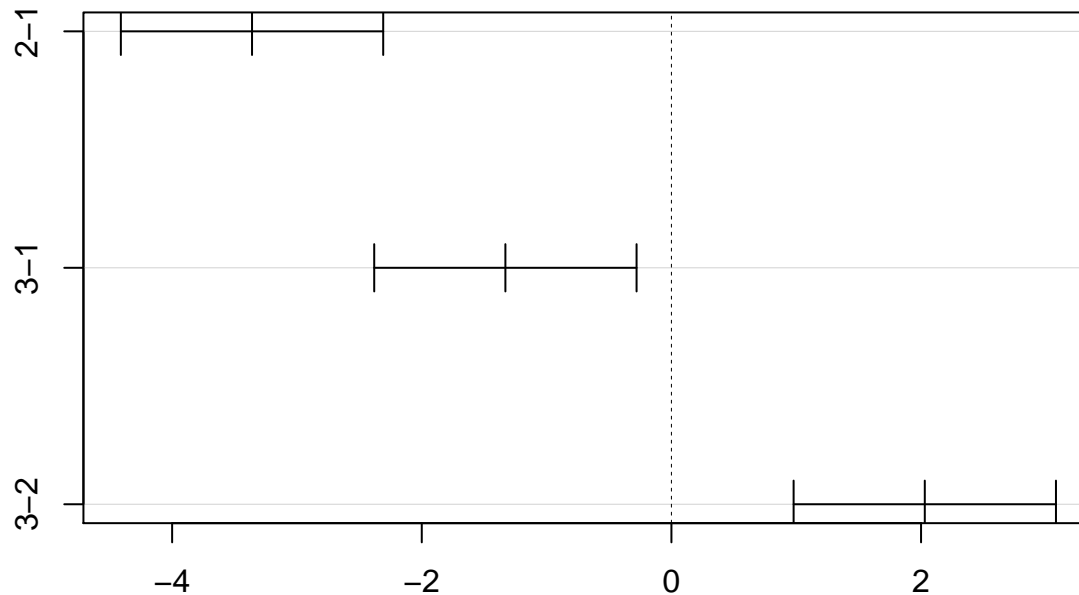
```
## Tables of effects
##
## foret
## foret
##      1      2      3
## 1.5633 -1.7967 0.2333
```

```
TukeyHSD(modele2)
```

```
## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = hauteur ~ foret, data = arbre)
##
## $foret
##      diff      lwr      upr      p adj
## 2-1 -3.36 -4.4109317 -2.3090683 0.0000000
## 3-1 -1.33 -2.3809317 -0.2790683 0.0110330
## 3-2  2.03  0.9790683  3.0809317 0.0001544
```

```
plot(TukeyHSD(modele2))
```


95% family-wise confidence level



Differences in mean levels of forest

```
pdf("chap10fig104.pdf")
plot(TukeyHSD(modele2))
dev.off()
```

```
## pdf
## 2
```

#En plus : export des graphiques en niveaux de gris et aux formats .png ou .ps

```
png("chap10fig102.png")
granova.1w(arbre$hauteur, arbre$foret)
```

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat        F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
```

```
##
## $stats
##      Size Contrast Coef Wt'd Mean Mean Trim'd Mean Var. St. Dev.
## 2      10      -1.80   21.71 21.71      21.82 1.56   1.25
## 3      10       0.23   23.74 23.74      23.87 0.40   0.63
## 1      10       1.56   25.07 25.07      25.10 0.74   0.86
```

```
dev.off()
```

```
## pdf
## 2
```

```
postscript("chap10fig102.ps")
granova.1w(arbre$hauteur, arbre$foret)
```

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat      F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
##
```

```
## $stats
##   Size Contrast Coef Wt'd Mean  Mean Trim'd Mean Var. St. Dev.
## 2   10      -1.80   21.71 21.71      21.82 1.56    1.25
## 3   10       0.23   23.74 23.74      23.87 0.40    0.63
## 1   10       1.56   25.07 25.07      25.10 0.74    0.86
```

```
dev.off()
```

```
## pdf
## 2
```

```
pdf("chap10fig102bw.pdf",colormodel="gray")
granova.1w(arbre$hauteur,arbre$foret)
```

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat      F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
##
```

```
## $stats
##   Size Contrast Coef Wt'd Mean  Mean Trim'd Mean Var. St. Dev.
## 2   10      -1.80   21.71 21.71      21.82 1.56    1.25
## 3   10       0.23   23.74 23.74      23.87 0.40    0.63
## 1   10       1.56   25.07 25.07      25.10 0.74    0.86
```

```
dev.off()
```

```
## pdf
## 2
```

```
postscript("chap10fig102bw.ps",colormodel="gray")
granova.1w(arbre$hauteur,arbre$foret)
```

```
## $grandsum
##      Grandmean      df.bet      df.with      MS.bet      MS.with
##      23.51         2.00        27.00        28.63         0.90
##      F.stat      F.prob SS.bet/SS.tot
##      31.87         0.00         0.70
##
```

```
## $stats
##   Size Contrast Coef Wt'd Mean  Mean Trim'd Mean Var. St. Dev.
## 2   10      -1.80   21.71 21.71      21.82 1.56    1.25
## 3   10       0.23   23.74 23.74      23.87 0.40    0.63
## 1   10       1.56   25.07 25.07      25.10 0.74    0.86
```

```
dev.off()
```

```
## pdf
## 2
```

```

png("chap10fig103.png")
granovagg.lw(arbre$hauteur,arbre$foret)

##
## By-group summary statistics for your input data (ordered by group means)
##
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82    -1.80      1.56          1.25
## 3      3      23.74      23.87     0.23      0.40          0.63
## 1      1      25.07      25.10     1.56      0.74          0.86
##   group.size
## 2           10
## 3           10
## 1           10
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067     0.1730 135.844 < 2e-16 ***
## group1       1.5633     0.2447   6.388 7.66e-07 ***
## group2      -1.7967     0.2447  -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF,  p-value: 7.809e-08
dev.off()

## pdf
## 2

postscript("chap10fig103.ps")
granovagg.lw(arbre$hauteur,arbre$foret)

##
## By-group summary statistics for your input data (ordered by group means)
##
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82    -1.80      1.56          1.25
## 3      3      23.74      23.87     0.23      0.40          0.63
## 1      1      25.07      25.10     1.56      0.74          0.86
##   group.size
## 2           10
## 3           10
## 1           10

```

```
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067     0.1730 135.844 < 2e-16 ***
## group1       1.5633     0.2447   6.388 7.66e-07 ***
## group2      -1.7967     0.2447  -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF,  p-value: 7.809e-08
##
## Warning in grid.Call.graphics(C_segments, x$x0, x$y0, x$x1, x$y1, x$arrow):
## semi-transparency is not supported on this device: reported only once per
## page
dev.off()

## pdf
##      2
pdf("chap10fig103bw.pdf",colormodel="gray")
granovagg.lw(arbre$hauteur,arbre$foret)

##
## By-group summary statistics for your input data (ordered by group means)
##      group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82      -1.80      1.56      1.25
## 3      3      23.74      23.87       0.23      0.40      0.63
## 1      1      25.07      25.10       1.56      0.74      0.86
##      group.size
## 2      10
## 3      10
## 1      10
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
```

```
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067    0.1730 135.844 < 2e-16 ***
## group1       1.5633    0.2447   6.388 7.66e-07 ***
## group2      -1.7967    0.2447  -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF,  p-value: 7.809e-08

dev.off()

## pdf
## 2

postscript("chap10fig103bw.ps",colormodel="gray")
granovagg.lw(arbre$hauteur,arbre$foret)

##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      21.71      21.82    -1.80      1.56      1.25
## 3      3      23.74      23.87     0.23      0.40      0.63
## 1      1      25.07      25.10     1.56      0.74      0.86
##   group.size
## 2           10
## 3           10
## 1           10
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.8100 -0.4550  0.0750  0.5425  1.7900
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  23.5067    0.1730 135.844 < 2e-16 ***
## group1       1.5633    0.2447   6.388 7.66e-07 ***
## group2      -1.7967    0.2447  -7.342 6.75e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9478 on 27 degrees of freedom
## Multiple R-squared:  0.7025, Adjusted R-squared:  0.6804
## F-statistic: 31.87 on 2 and 27 DF,  p-value: 7.809e-08
##
## Warning in grid.Call.graphics(C_segments, x$x0, x$y0, x$x1, x$y1, x$arrow):
## semi-transparency is not supported on this device: reported only once per
```

```
## page
dev.off()

## pdf
## 2

#page 425
#Exercice 10.1
#1)
options(contrasts=c(unordered="contr.sum", ordered="contr.poly"))

#page 426
#2)
variete<-rep(1:6,c(5,5,5,5,5,5))
vitamine<-c(93.6,95.3,96,93.7,96.2,95.3,96.9,95.8,97.3,97.7,94.5,97,97.8,97,
98.3,98.8,98.2,97.8,97.2,97.9,94.6,97.8,98,95,98.9,93.2,94.4,93.8,95.6,94.8)
variete<-factor(variete)
exo1<-data.frame(variete,vitamine)
modele1<-aov(vitamine~variete,data=exo1)
residus1<-residuals(modele1)
shapiro.test(residus1)

##
## Shapiro-Wilk normality test
##
## data: residus1
## W = 0.9563, p-value = 0.2485
length(residus1)

## [1] 30
bartlett.test(residus1~variete,data=exo1)

##
## Bartlett test of homogeneity of variances
##
## data: residus1 by variete
## Bartlett's K-squared = 5.6023, df = 5, p-value = 0.3469

#page 427
#3)
modele1

## Call:
## aov(formula = vitamine ~ variete, data = exo1)
##
## Terms:
##             variete Residuals
## Sum of Squares  45.836    38.512
## Deg. of Freedom      5        24
##
## Residual standard error: 1.266754
## Estimated effects may be unbalanced
summary(modele1)

##              Df Sum Sq Mean Sq F value Pr(>F)
```

```
## variete      5  45.84   9.167   5.713 0.00131 **
## Residuals   24  38.51   1.605
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

#page 428
#4)
granovagg.lw(vitamine,group=variete)

##
## By-group summary statistics for your input data (ordered by group means)

##   group group.mean trimmed.mean contrast variance standard.deviation
## 6      6      94.36      94.33    -1.92      0.85              0.92
## 1      1      94.96      95.00    -1.32      1.54              1.24
## 2      2      96.60      96.67     0.32      1.03              1.01
## 5      5      96.86      96.93     0.58      3.73              1.93
## 3      3      96.92      97.27     0.64      2.14              1.46
## 4      4      97.98      97.97     1.70      0.34              0.58
##   group.size
## 6           5
## 1           5
## 2           5
## 5           5
## 3           5
## 4           5

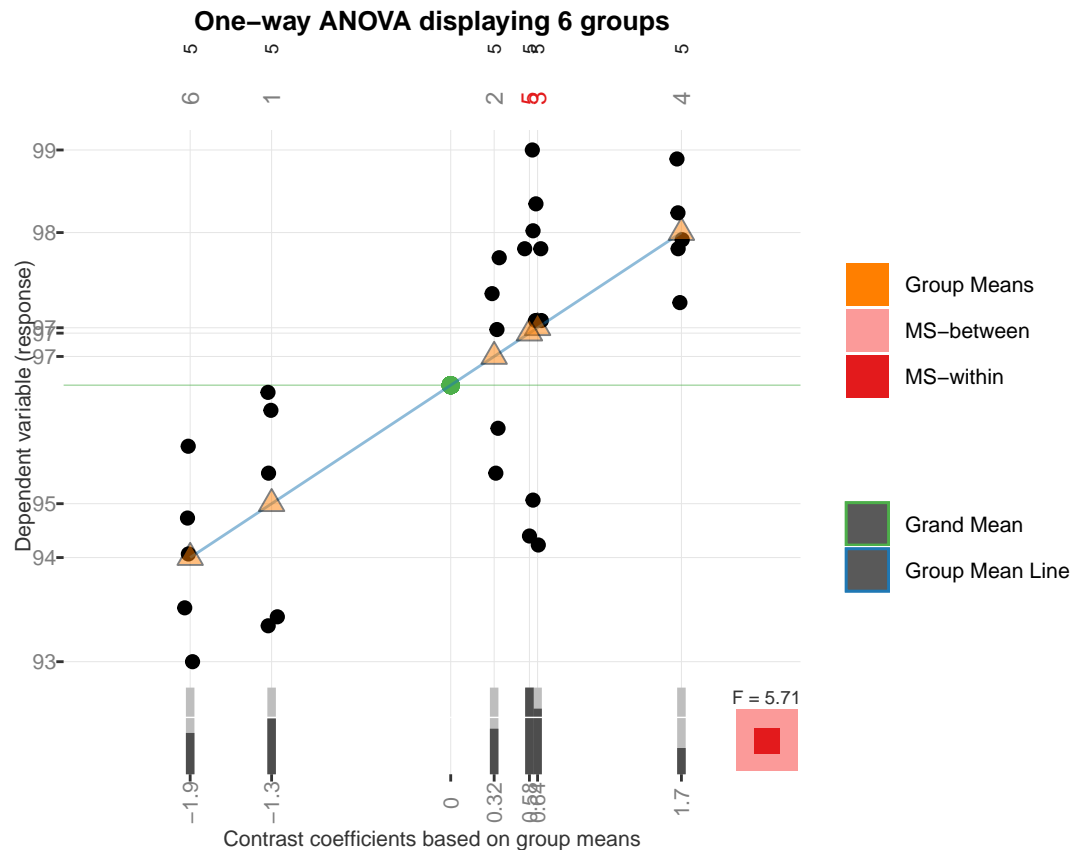
##
## The following groups are likely to be overplotted

##   group group.mean contrast
## 5      5      96.86      0.58
## 3      3      96.92      0.64

##
## Below is a linear model summary of your input data

##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.420 -0.795  0.150  0.925  2.040
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  96.2800    0.2313  416.298 < 2e-16 ***
## group1       -1.3200    0.5172  -2.552  0.01748 *
## group2        0.3200    0.5172   0.619  0.54189
## group3        0.6400    0.5172   1.238  0.22785
## group4        1.7000    0.5172   3.287  0.00311 **
## group5        0.5800    0.5172   1.122  0.27316
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.267 on 24 degrees of freedom
```

```
## Multiple R-squared:  0.5434, Adjusted R-squared:  0.4483
## F-statistic: 5.713 on 5 and 24 DF,  p-value: 0.001311
```



```
pdf("chap10fig105.pdf")
granovagg.1w(vitamine,group=variete)
```

```
##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 6      6      94.36      94.33    -1.92      0.85      0.92
## 1      1      94.96      95.00    -1.32      1.54      1.24
## 2      2      96.60      96.67     0.32      1.03      1.01
## 5      5      96.86      96.93     0.58      3.73      1.93
## 3      3      96.92      97.27     0.64      2.14      1.46
## 4      4      97.98      97.97     1.70      0.34      0.58
##   group.size
## 6           5
## 1           5
## 2           5
## 5           5
## 3           5
## 4           5
##
## The following groups are likely to be overplotted
##   group group.mean contrast
## 5      5      96.86      0.58
```



```
## 3      3      96.92      0.64

##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.420 -0.795  0.150  0.925  2.040
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  96.2800     0.2313  416.298 < 2e-16 ***
## group1       -1.3200     0.5172   -2.552  0.01748 *
## group2         0.3200     0.5172    0.619  0.54189
## group3         0.6400     0.5172    1.238  0.22785
## group4         1.7000     0.5172    3.287  0.00311 **
## group5         0.5800     0.5172    1.122  0.27316
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.267 on 24 degrees of freedom
## Multiple R-squared:  0.5434, Adjusted R-squared:  0.4483
## F-statistic: 5.713 on 5 and 24 DF,  p-value: 0.001311
```

```
dev.off()
```

```
## pdf
## 2
```

```
#page 430
#6)
```

```
Tukey1 <- TukeyHSD(modele1, conf.level = 0.95)
Tukey1
```

```
## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = vitamine ~ variete, data = exo1)
##
## $variete
##      diff      lwr      upr      p adj
## 2-1  1.64 -0.8371495  4.1171497  0.3468489
## 3-1  1.96 -0.5171495  4.4371497  0.1803668
## 4-1  3.02  0.5428505  5.4971497  0.0107386
## 5-1  1.90 -0.5771495  4.3771497  0.2058535
## 6-1 -0.60 -3.0771495  1.8771497  0.9733815
## 3-2  0.32 -2.1571495  2.7971497  0.9985151
## 4-2  1.38 -1.0971495  3.8571497  0.5310572
## 5-2  0.26 -2.2171495  2.7371497  0.9994551
## 6-2 -2.24 -4.7171495  0.2371497  0.0926651
## 4-3  1.06 -1.4171495  3.5371497  0.7697394
## 5-3 -0.06 -2.5371495  2.4171497  0.9999996
## 6-3 -2.56 -5.0371495 -0.08285053 0.0399329
```

```
## 5-4 -1.12 -3.5971495 1.35714947 0.7278111
## 6-4 -3.62 -6.0971495 -1.14285053 0.0017510
## 6-5 -2.50 -4.9771495 -0.02285053 0.0470143

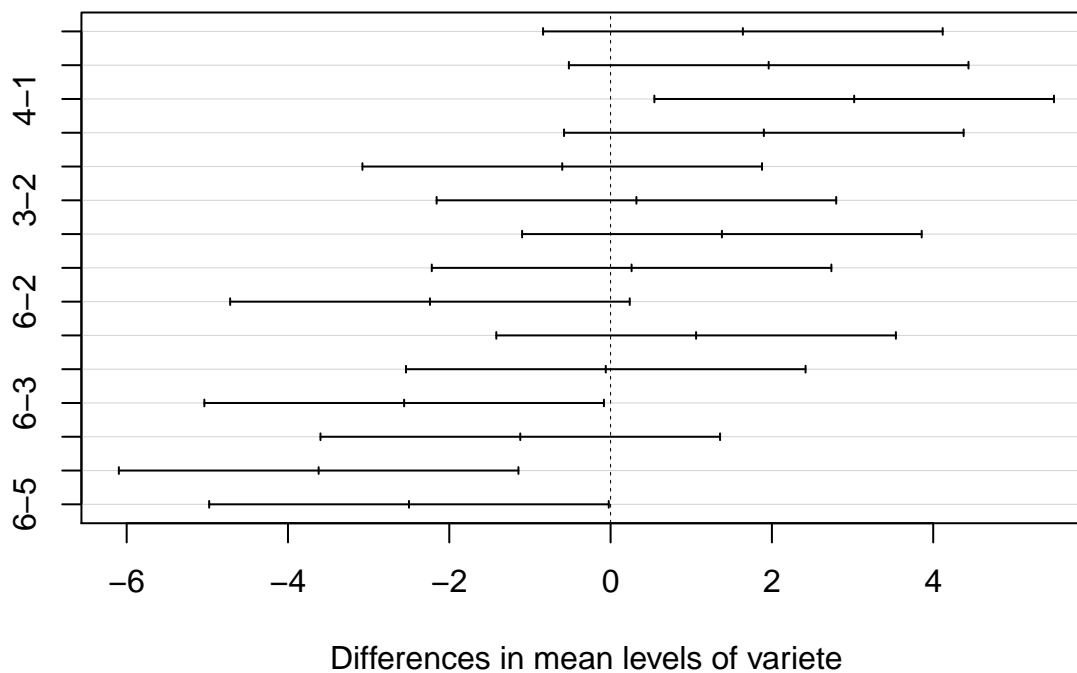
#page 431
#4)
if(!("multcomp" %in% rownames(installed.packages()))){
  install.packages("multcomp")}
library(multcomp)

## Loading required package: mvtnorm
## Loading required package: survival
## Loading required package: TH.data
## Loading required package: MASS
##
## Attaching package: 'TH.data'
## The following object is masked from 'package:MASS':
##
##      geyser
wht = glht(modele1, linfct = mcp(variete = "Tukey"))
cld(wht)

##      1      2      3      4      5      6
## "ab" "ac" "bc" "c" "bc" "a"

plot(Tukey1)
```

95% family-wise confidence level



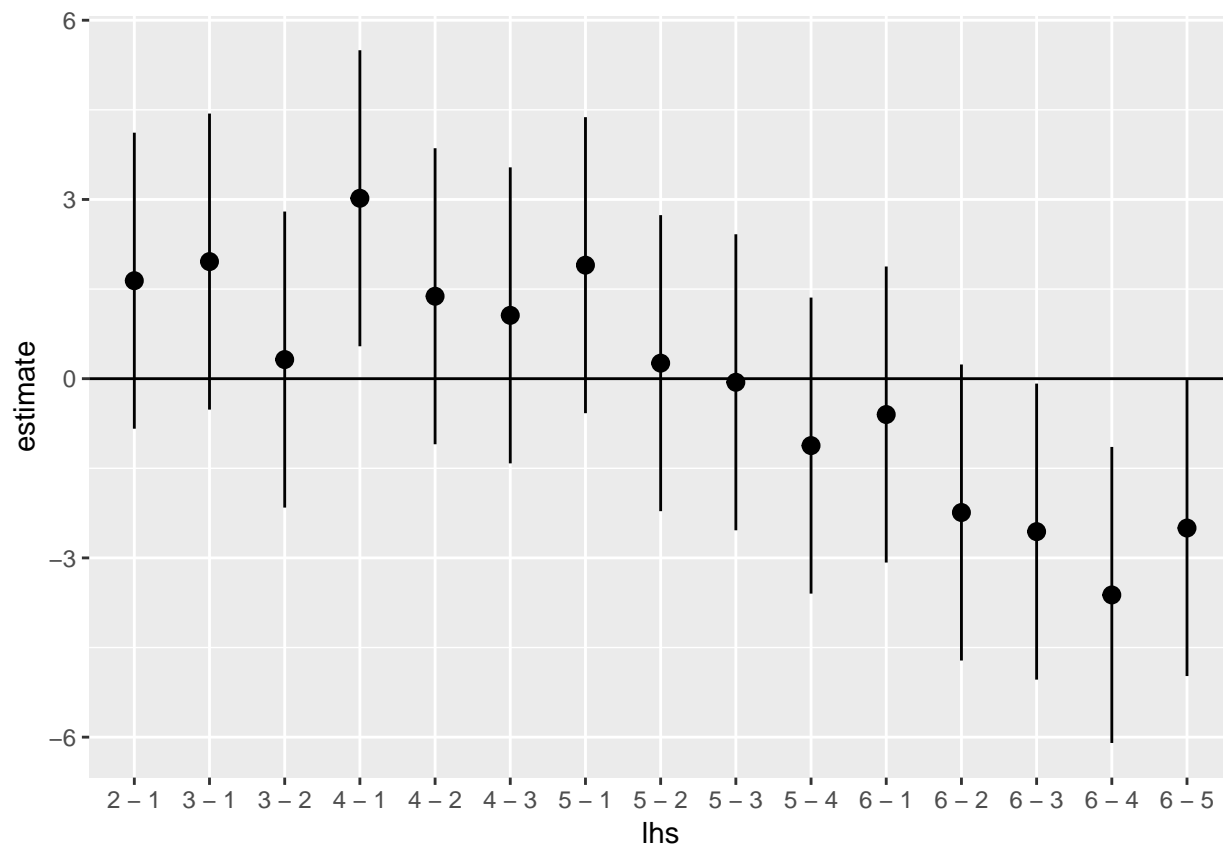
```
pdf("chap10fig106.pdf")
plot(Tukey1)
dev.off()
```

```
## pdf
## 2
```

```
#page 432
CI <- confint(wht)
fortify(CI)
```

##	lhs	rhs	estimate	lwr	upr
## 1	2 - 1	0	1.64	-0.8374285	4.11742852
## 2	3 - 1	0	1.96	-0.5174285	4.43742852
## 3	4 - 1	0	3.02	0.5425715	5.49742852
## 4	5 - 1	0	1.90	-0.5774285	4.37742852
## 5	6 - 1	0	-0.60	-3.0774285	1.87742852
## 6	3 - 2	0	0.32	-2.1574285	2.79742852
## 7	4 - 2	0	1.38	-1.0974285	3.85742852
## 8	5 - 2	0	0.26	-2.2174285	2.73742852
## 9	6 - 2	0	-2.24	-4.7174285	0.23742852
## 10	4 - 3	0	1.06	-1.4174285	3.53742852
## 11	5 - 3	0	-0.06	-2.5374285	2.41742852
## 12	6 - 3	0	-2.56	-5.0374285	-0.08257148
## 13	5 - 4	0	-1.12	-3.5974285	1.35742852
## 14	6 - 4	0	-3.62	-6.0974285	-1.14257148
## 15	6 - 5	0	-2.50	-4.9774285	-0.02257148

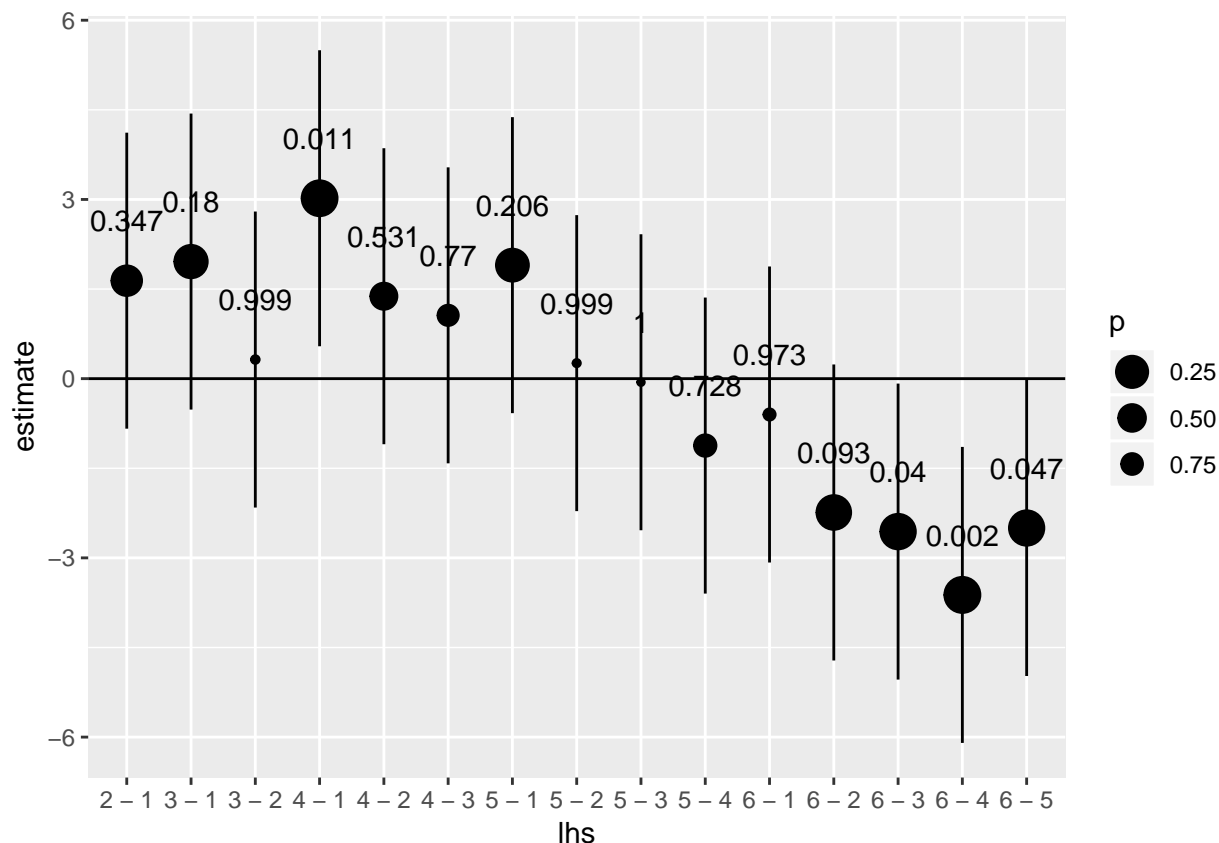
```
ggplot(CI,aes(lhs,estimate,ymin=lwr,ymax=upr))+geom_pointrange()+
  geom_hline(yintercept = 0)
```



```
pdf("chap10fig107.pdf")
print(ggplot(CI,aes(lhs,estimate,ymin=lwr,ymax=upr))+geom_pointrange()+
      geom_hline(yintercept = 0))
dev.off()
```

```
## pdf
## 2
```

```
ggplot(aes(lhs,estimate),data=fortify(summary(wht))) +
  geom_linerange(aes(ymin=lwr,ymax=upr),data=CI) +
  geom_text(aes(y=estimate+1,label=round(p,3)))+geom_hline(yintercept = 0) +
  geom_point(aes(size=p),data=summary(wht)) +scale_size(trans="reverse")
```



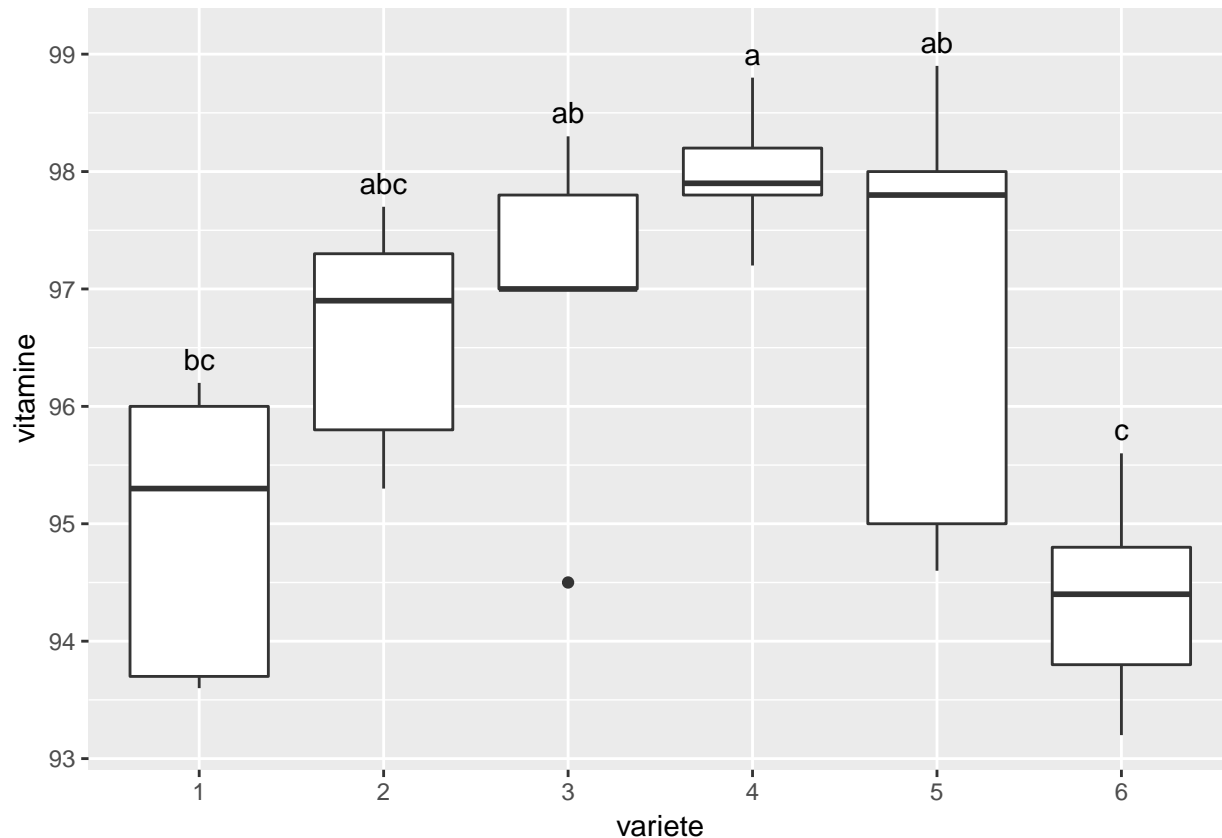
```
pdf("chap10fig108.pdf")
ggplot(aes(lhs,estimate),data=fortify(summary(wht))) +
  geom_linerange(aes(ymin=lwr,ymax=upr),data=CI) +
  geom_text(aes(y=estimate+1,label=round(p,3)))+geom_hline(yintercept = 0) +
  geom_point(aes(size=p),data=summary(wht)) +scale_size(trans="reverse")
dev.off()
```

```
## pdf
## 2
```

```
#page 433
if(!("multcompView" %in% rownames(installed.packages()))){
  install.packages("multcompView")}
library(multcompView)
if(!("plyr" %in% rownames(installed.packages()))){install.packages("plyr")}
library(plyr)
generate_label_df <- function(HSD,flev){
  Tukey.levels <- HSD[[flev]][,4]
  Tukey.labels <- multcompLetters(Tukey.levels)['Letters']
  plot.labels <- names(Tukey.labels[['Letters']])
  boxplot.df <- ddply(exo1, flev, function(x) max(fivenum(x$vitamine)) + 0.2)
  plot.levels <- data.frame(plot.labels, labels = Tukey.labels[['Letters']],
    stringsAsFactors = FALSE)
  labels.df <- merge(plot.levels, boxplot.df, by.x = 'plot.labels', by.y = flev,
    sort = FALSE)
  return(labels.df)
}
```

#page 434

```
p_base <- ggplot(exo1,aes(x=variete,y=vitamine)) + geom_boxplot() +
  geom_text(data = generate_label_df(Tukey1, 'variete'), aes(x = plot.labels,
    y = V1, label = labels))
p_base
```



```
pdf("chap10fig109.pdf")
print(p_base)
dev.off()
```

```
## pdf
## 2
```

#page 435

#Exercice 10.1

#2)

```
traitement<-rep(1:5,c(7,7,7,7,7))
taux<-c(4.5,2.5,6,4.5,3,5.5,3.5,7.5,3,2.5,4,2,4,5.5,8,6.5,6,3.5,5,
  7,5,2,7.5,4,2.5,5,3.5,6.5,6.5,5.5,6,4.5,4,7,5.5)
traitement<-factor(traitement)
exo2<-data.frame(traitement,taux)
modele2<-aov(taux~traitement,data=exo2)
residus2<-residuals(modele2)
shapiro.test(residus2)
```

```
##
## Shapiro-Wilk normality test
##
```

```
## data: residus2
## W = 0.97436, p-value = 0.5734
length(residus2)

## [1] 35
bartlett.test(residus2~traitement,data=exo2)

##
## Bartlett test of homogeneity of variances
##
## data: residus2 by traitement
## Bartlett's K-squared = 3.1361, df = 4, p-value = 0.5353
#page 436
#3)
modele1<-lm(taux~traitement,data=exo2)
anova(modele1)

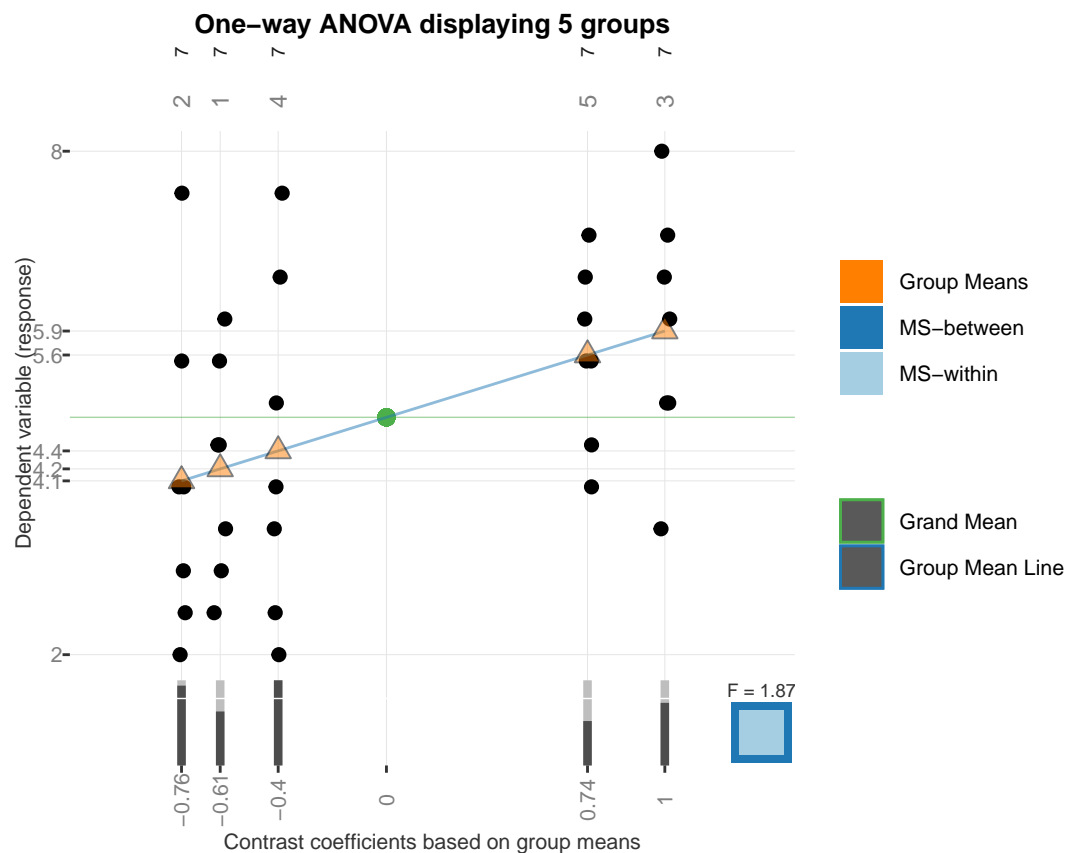
## Analysis of Variance Table
##
## Response: taux
##          Df Sum Sq Mean Sq F value Pr(>F)
## traitement  4 19.043   4.7607   1.8687 0.1419
## Residuals  30 76.429   2.5476
#4)
power.anova.test(5,7,19.043,76.42857)

##
##      Balanced one-way analysis of variance power calculation
##
##      groups = 5
##      n = 7
##      between.var = 19.043
##      within.var = 76.42857
##      sig.level = 0.05
##      power = 0.4684833
##
## NOTE: n is number in each group
#page 437
power.anova.test(groups=5,between.var=19.043,within.var=76.42857,power=.80)

##
##      Balanced one-way analysis of variance power calculation
##
##      groups = 5
##      n = 12.96035
##      between.var = 19.043
##      within.var = 76.42857
##      sig.level = 0.05
##      power = 0.8
##
## NOTE: n is number in each group
```

```
granovagg.lw(taux,group=traitement)
```

```
##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2      4.07         3.8    -0.76      3.62          1.90
## 1      1      4.21         4.2    -0.61      1.65          1.29
## 4      4      4.43         4.3    -0.40      4.12          2.03
## 5      5      5.57         5.6     0.74      1.12          1.06
## 3      3      5.86         5.9     1.03      2.23          1.49
##   group.size
## 2           7
## 1           7
## 4           7
## 5           7
## 3           7
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.4286 -1.0714 -0.0714  1.0357  3.4286
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   4.8286     0.2698  17.897  <2e-16 ***
## group1        -0.6143     0.5396  -1.138  0.2639
## group2        -0.7571     0.5396  -1.403  0.1708
## group3         1.0286     0.5396   1.906  0.0662 .
## group4        -0.4000     0.5396  -0.741  0.4643
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.596 on 30 degrees of freedom
## Multiple R-squared:  0.1995, Adjusted R-squared:  0.09272
## F-statistic: 1.869 on 4 and 30 DF,  p-value: 0.1419
```

```
pdf("chap10fig1010.pdf",colormodel="gray")
granovagg.1w(taux,group=traitement)
```

```
##
## By-group summary statistics for your input data (ordered by group means)
##   group group.mean trimmed.mean contrast variance standard.deviation
## 2      2          4.07          3.8   -0.76          3.62            1.90
## 1      1          4.21          4.2   -0.61          1.65            1.29
## 4      4          4.43          4.3   -0.40          4.12            2.03
## 5      5          5.57          5.6    0.74          1.12            1.06
## 3      3          5.86          5.9    1.03          2.23            1.49
##   group.size
## 2           7
## 1           7
## 4           7
## 5           7
## 3           7
##
## Below is a linear model summary of your input data
##
## Call:
## lm(formula = score ~ group, data = owp$data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.4286 -1.0714 -0.0714  1.0357  3.4286
```

```
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## (Intercept)   4.8286     0.2698  17.897 <2e-16 ***
## group1        -0.6143     0.5396  -1.138  0.2639
## group2        -0.7571     0.5396  -1.403  0.1708
## group3         1.0286     0.5396   1.906  0.0662 .
## group4        -0.4000     0.5396  -0.741  0.4643
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.596 on 30 degrees of freedom
## Multiple R-squared:  0.1995, Adjusted R-squared:  0.09272
## F-statistic: 1.869 on 4 and 30 DF,  p-value: 0.1419
dev.off()

## pdf
## 2
```