## T1-E04-multi COL.R

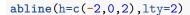
jordi

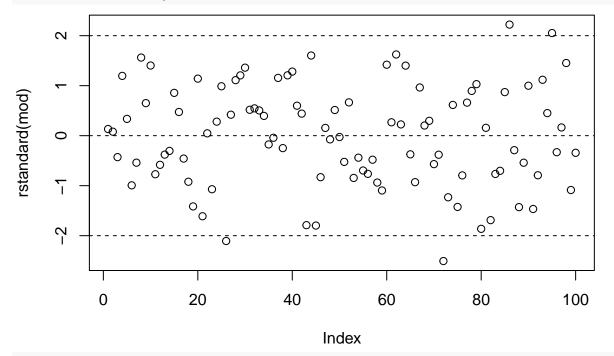
Tue Oct 23 12:21:06 2018

```
setwd("~/Documents/CURS 2018-2019/PIE2")
#setwd("F:/windows")
COL <- read.csv2("./Dades/COL.csv")</pre>
n<-dim(COL)[1]</pre>
library(car)
## Loading required package: carData
scatterplotMatrix(COL, smooth=F, diagonal=F)
                            140
                    100
                                                           100 200 300 400
          Α
180
                             Н
                                               W
                                                                             80
                                                                             9
                                        40 50 60 70 80 90
    10
               18
          14
# MODEL
mod < -lm(C \sim W + A + H, COL)
write("______
## ______
write("a), b), c)","")
## a), b), c)
# RESUM DEL MODEL
summary(mod)
```

```
##
## Call:
## lm(formula = C \sim W + A + H, data = COL)
## Residuals:
##
             1Q Median
      Min
                             3Q
                                    Max
## -74.608 -22.137 1.888 21.156 65.410
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 490.9978 35.0517 14.008 < 2e-16 ***
                        0.7365 14.090 < 2e-16 ***
             10.3773
## W
                        3.8530 -3.379 0.00105 **
## A
             -13.0195
## H
              -5.0989
                        0.7227 -7.055 2.68e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 30.11 on 96 degrees of freedom
## Multiple R-squared: 0.8101, Adjusted R-squared: 0.8041
## F-statistic: 136.5 on 3 and 96 DF, p-value: < 2.2e-16
#Càlculs opcionals: Intervals de confiança dels paràmetres
confint(mod,level=0.99)
##
                  0.5 %
                            99.5 %
## (Intercept) 398.881272 583.114304
              8.441792 12.312821
## W
## A
             -23.145228 -2.893732
              -6.998311 -3.199551
## H
write("______
write("b), d)","")
## b), d)
#Câlculs opcionals: SS1 Test dels paràmetres amb ordenació predeterminada
anova(mod)
## Analysis of Variance Table
## Response: C
##
            Df Sum Sq Mean Sq F value
## W
            1 62396 62396 68.826 6.686e-13 ***
            1 263670 263670 290.841 < 2.2e-16 ***
## H
            1 45123 45123 49.773 2.676e-10 ***
## Residuals 96 87031
                        907
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
anova(lm(C~H+W+A,COL))
## Analysis of Variance Table
## Response: C
           Df Sum Sq Mean Sq F value Pr(>F)
```

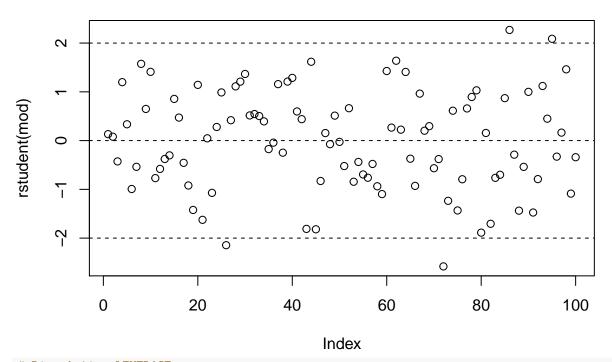
```
1 171564 171564 189.244 < 2.2e-16 ***
## W
             1 189273 189273 208.778 < 2.2e-16 ***
               10351
                       10351 11.418 0.001052 **
## A
## Residuals 96 87031
                         907
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#Nota: SS3, els tests (F) sempre coincideixen amb els del resum (t), F=t^2
Anova (mod)
## Anova Table (Type II tests)
## Response: C
            Sum Sq Df F value
## W
            ## A
             10351 1 11.418 0.001052 **
## H
             45123 1 49.773 2.676e-10 ***
## Residuals 87031 96
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
write("f)","")
## f)
# Diagnostic: TENDÈNCIES
plot(predict(mod),resid(mod))
abline(h=0,lty=2)
                                    0
     9
                           0
           0
     40
                                                                           0
     20
resid(mod)
     0
                                            0
              0
                                                         0
                                                          0
                                             00
                                                         0
                0
                                  B
                0
         100
                    150
                              200
                                         250
                                                    300
                                                               350
                                                                          400
                                      predict(mod)
# Diagnòstic: OUTLIERS (rstudent)
plot(rstandard(mod))
```



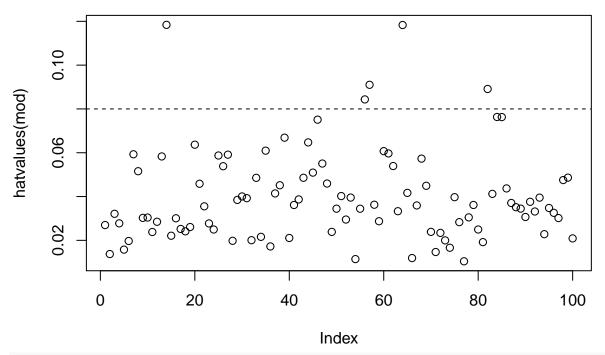


plot(rstudent(mod),main="rstudent")
abline(h=c(-2,0,2),lty=2)

## rstudent

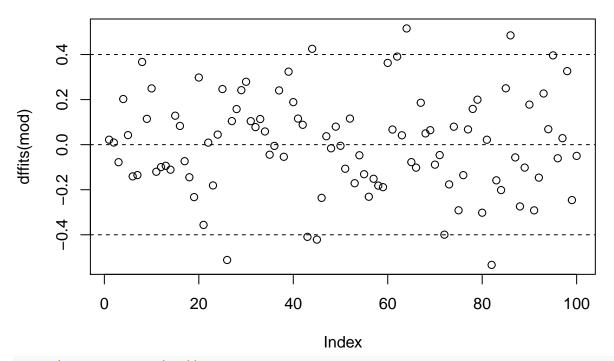


# Diagnostic: LEVERAGE
plot(hatvalues(mod))
abline(h=c(0,2\*mean(hatvalues(mod))),lty=2)



```
# Diagnostic: INFLUENCIA (dffits)
plot(dffits(mod), main="dffits")
abline(h=c(-2*sqrt(p/n),0,2*sqrt(p/n)),lty=2)
```

## dffits



```
#Diagnôstics de R
oldpar <- par( mfrow=c(2,2))</pre>
plot(mod,ask=F)
                                                 Standardized residuals
                                                                    Normal Q-Q
                Residuals vs Fitted
Residuals
     50
                                                      ^{\circ}
                                                      0
     -50
                                                      7
                                                                                       2
        100 150 200 250 300
                                   350
                                        400
                                                              -2
                    Fitted values
                                                                 Theoretical Quantiles
Standardized residuals
                                                 Standardized residuals
                  Scale-Location
                                                               Residuals vs Leverage
                                                                                         640
                                                                     ok'§ istance 820
             150 200 250 300 350 400
                                                          0.00
                                                                    0.04
                                                                               0.08
                                                                                         0.12
                    Fitted values
                                                                       Leverage
par(oldpar)
write("______
write("g)","")
## g)
#Diagnostics: Col·linealitat
vif(mod)
##
## 9.489406 20.904776 31.695499
write("e)","")
\#C\`{a}lculs opcionals: Per alguns casos predeterminats, IC de E(Y)
(CO < -data.frame(cbind(W = c(65,75,65),A = c(15,15,12),H = c(150,150,150)), row.names = 1:3))
      W A
## 1 65 15 150
```

```
## 2 75 15 150
## 3 65 12 150
predict(mod, CO, interval="confidence", level=.95, se.fit=T)
## $fit
##
          fit
                   lwr
## 1 205.3908 199.1668 211.6148
## 2 309.1639 294.6188 323.7089
## 3 244.4492 219.8210 269.0774
##
## $se.fit
## 3.135539 7.327533 12.407261
##
## $df
## [1] 96
##
## $residual.scale
## [1] 30.1094
#Càlculs opcionals: Per alguns casos predeterminats, IPredicció de Y
predict(mod, CO, interval="prediction", level=.95, se.fit=F)
##
          fit
                   lwr
## 1 205.3908 145.3009 265.4807
## 2 309.1639 247.6528 370.6749
## 3 244.4492 179.8071 309.0914
write("h), i)","")
## h), i)
#Canvis lineals en les variables independents:
#canvi en alguna variable, per exemple, exces de pes,
# pes patró 0.5*H-10, WE=W-(0.5*H-10)
#COL$WE<-COL$W-0.5*COL$H+10
summary(mod2 < -lm(C \sim I(W-0.5 * H+10) + A+H, COL))
##
## Call:
## lm(formula = C \sim I(W - 0.5 * H + 10) + A + H, data = COL)
##
## Residuals:
##
       Min
                1Q Median
                                30
## -74.608 -22.137
                   1.888 21.156 65.410
##
## Coefficients:
                        Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                     387.22473 33.69605 11.492 < 2e-16 ***
## I(W - 0.5 * H + 10) 10.37731 0.73649 14.090 < 2e-16 ***
## A
                       -13.01948
                                  3.85300 -3.379 0.00105 **
```

```
## H
                        0.08972
                                   0.58736 0.153 0.87891
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 30.11 on 96 degrees of freedom
## Multiple R-squared: 0.8101, Adjusted R-squared: 0.8041
## F-statistic: 136.5 on 3 and 96 DF, p-value: < 2.2e-16
vif(mod2)
## I(W - 0.5 * H + 10)
                                                            Η
                                        Α
             1.009937
                                20.904776
                                                    20.933520
#Nota: Només canvia algun paràmetre i la col·linealitat
#Canvis lineals en les variables independents:
#eliminar alguna variable independent no significativa i/o amb molta col·linealitat
#per exemple H, si ja s'utilitza l'exces de pes
summary(mod3<-lm(C~I(W-0.5*H+10)+A,COL))
##
## Call:
## lm(formula = C \sim I(W - 0.5 * H + 10) + A, data = COL)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
## -74.286 -22.638
                   1.755 20.935 66.244
##
## Coefficients:
                      Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                      391.9885
                                 12.6975 30.87
                                                   <2e-16 ***
## I(W - 0.5 * H + 10) 10.3882
                                          14.24
                                   0.7294
                                                   <2e-16 ***
## A
                      -12.4452
                                   0.8387 -14.84
                                                  <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 29.96 on 97 degrees of freedom
## Multiple R-squared: 0.81, Adjusted R-squared: 0.8061
## F-statistic: 206.8 on 2 and 97 DF, p-value: < 2.2e-16
vif(mod3)
## I(W - 0.5 * H + 10)
                                        Α
             1.000527
                                 1.000527
COL$WE<-COL$W-0.5*COL$H+10
scatterplotMatrix(COL, smooth=F, diagonal=F)
```

