

TD8

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```
library(car)
```

```
## Loading required package: carData
```

```
library(emmeans)
```

Question 1

```
df = read.csv("./MiniPONS.csv", sep=";")
head(df)
```

```
##      Group  Type Age Right_answers Audio_prosody Combined_channel Face_video
## 1 Bipolar  BD I  47           40           9           11           9
## 2 Bipolar  BD I  49           49          13           13          11
## 3 Bipolar  BD I  45           43           9           11          13
## 4 Bipolar  BD I  53           44          10           10          12
## 5 Bipolar BD II  50           50          14           13          11
## 6 Bipolar  BD I  31           54          13           14          14
##  Body_video Positive_valence Negative_valence Dominant Submissive
## 1           11              18              22         23         17
## 2           12              24              25         24         25
## 3           10              21              22         24         19
## 4           12              25              19         24         20
## 5           12              23              27         23         27
## 6           13              28              26         26         28
```

Question 2

```
is.factor(df$Group)
```

```
## [1] FALSE
```

```
is.factor(df$Type)
```

```
## [1] FALSE
```

```
df$Group = as.factor(df$Group)
```

```
df$Type = as.factor(df$Type)
```

```
levels(df$Type)
```

```
## [1] "BD I"    "BD II"   "Control" "UD"
```

```
levels(df$Group)
```

```
## [1] "Bipolar" "Control" "UD"
```

Question 3

```
by(df$Right_answers, df$Group, summary )
```

```
## df$Group: Bipolar
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      28.00  42.50   46.00   45.35   49.00   56.00
## -----
## df$Group: Control
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      39.00  48.00   50.00   50.24   53.00   58.00
## -----
## df$Group: UD
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      32.00  39.00   43.00   42.72   47.00   53.00
```

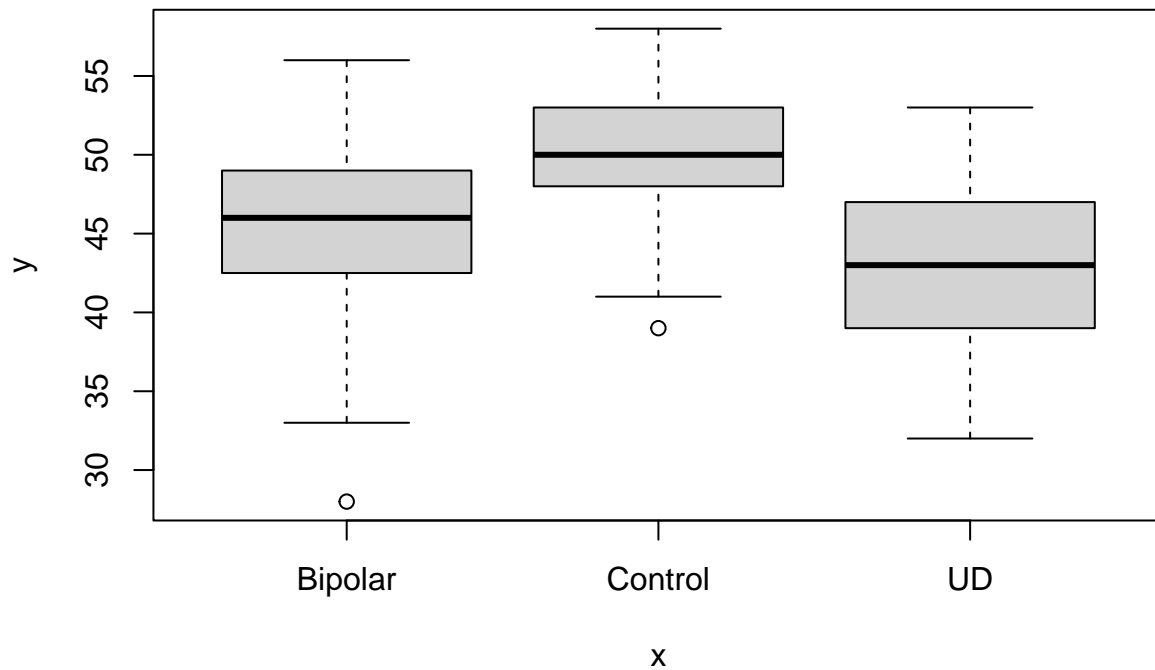
```
by(df$Right_answers, df$Group, sd )
```

```
## df$Group: Bipolar
## [1] 4.795104
## -----
## df$Group: Control
## [1] 3.704569
## -----
## df$Group: UD
## [1] 4.973343
```

On peut constater que les Bipolaire et les patients diagnostiqué UD ont en moyenne moins de bonne réponse que les sujets contrôles. De plus leurs écarts type est moins bon.

Question 4

```
plot(df$Group, df$Right_answers, type="b")
```



Question 5

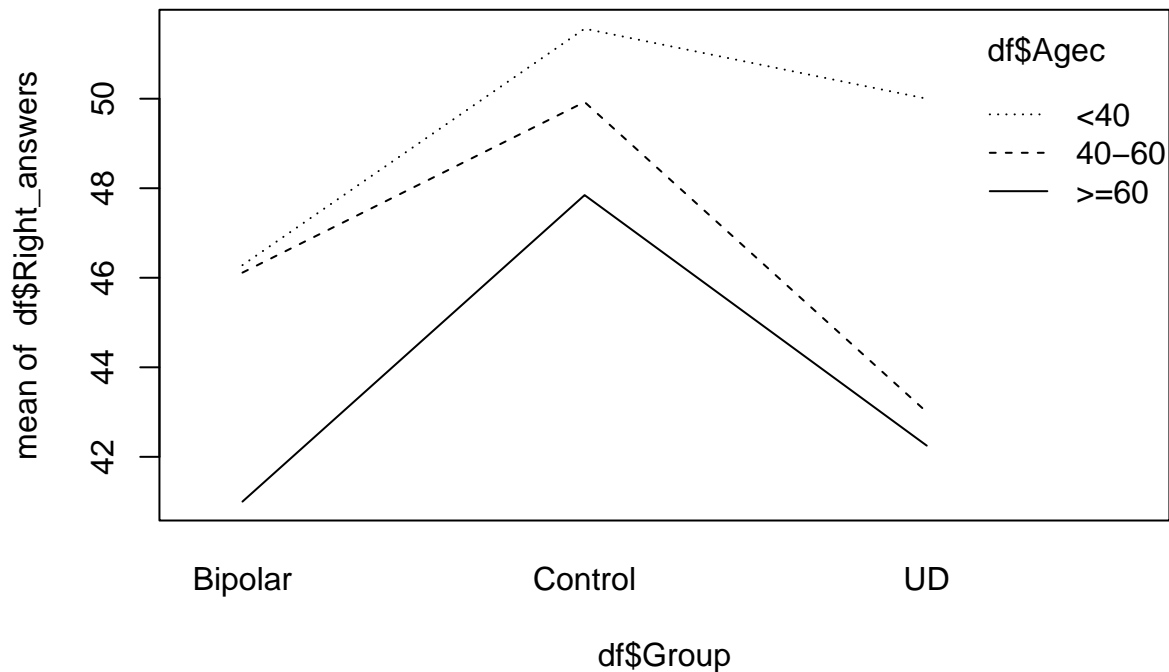
```
df$Agec=rep(2,length(df$Group))
df$Agec[which(df$Age<40)]=1
df$Agec[which(df$Age>=60)]=3
df$Agec=as.factor(df$Agec)
levels(df$Agec)=c("<40", "40-60", ">=60")
```

Question 6

```
table(df$Agec, df$Group)
```

```
##
##      Bipolar Control UD
## <40      39      39  1
## 40-60     61     67 14
## >=60     19     13 24
```

```
interaction.plot(df$Group, df$Agec, df$Right_answers)
```



Les ef-

fectifs ne sont pas vraiment équilibré dans avec le groupe UD.

Ils pourrait avoir de l'interaction. Considérons le plus grand des modèles possibles donc avec interaction. Puis on vas regarder si le les interaction valent le coup. On vas comparer le modèle avec interaction à celui sans interaction. Si le modèle sans interaction est mieux, on regardera ensuite si chaque facteur vaut le coup.

Question 7

Modèle retenu

```
res = lm(df$Right_answers~df$Group*df$Agec)
res

##
## Call:
## lm(formula = df$Right_answers ~ df$Group * df$Agec)
##
## Coefficients:
##              (Intercept)              df$GroupControl
##                46.2821                5.2821
##              df$GroupUD              df$Agec40-60
##                3.7179               -0.1673
##              df$Agec>=60 df$GroupControl:df$Agec40-60
##               -5.2821               -1.4714
## df$GroupUD:df$Agec40-60 df$GroupControl:df$Agec>=60
##               -6.8327                1.5641
##              df$GroupUD:df$Agec>=60
##               -2.4679

summary(res)

##
## Call:
## lm(formula = df$Right_answers ~ df$Group * df$Agec)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13.0000  -2.8462   0.0746   2.7179  10.0000
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      46.2821     0.6659  69.504 < 2e-16 ***
## df$GroupControl    5.2821     0.9417   5.609 5.06e-08 ***
## df$GroupUD         3.7179     4.2115   0.883  0.378
## df$Agec40-60      -0.1673     0.8526  -0.196  0.845
## df$Agec>=60       -5.2821     1.1634  -4.540 8.50e-06 ***
## df$GroupControl:df$Agec40-60 -1.4714     1.1952  -1.231  0.219
## df$GroupUD:df$Agec40-60     -6.8327     4.3881  -1.557  0.121
## df$GroupControl:df$Agec>=60  1.5641     1.7684   0.884  0.377
## df$GroupUD:df$Agec>=60     -2.4679     4.4008  -0.561  0.575
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.158 on 268 degrees of freedom
## Multiple R-squared:  0.3865, Adjusted R-squared:  0.3682
## F-statistic: 21.11 on 8 and 268 DF,  p-value: < 2.2e-16
```

Le summary n'est pas très intéressant. Car c'est des comparaisons étrange et difficile à interpréter.

Le modèle :

$$y_{i,j,k} = \mu + \alpha_i + \beta_j + \gamma_{i,j} + \epsilon_{i,j,k}$$

Avec i= Groupe, j=groupe_age, k=individue. y = intercept + effet groupe + effet age + interaction + erreur.

Est-ce que tous les paramètres sont estimable ? Non le modèle est *****. Pour contrer ça, R ajoute des contraintes :

$$\alpha_1 = 0, \alpha_{bipolar} = 0, \beta_1 = 0, \beta_{<40} = 0, \gamma_{1,1} = 0, \gamma_{bipolar,<40} = 0, \gamma_{1,2} = 0, \gamma_{bipolar,40-60} = 0, \gamma_{1,3} = 0, \gamma_{bipolar,>60} = 0, \gamma_{1,1} = 0, \gamma_{bipolar,>60} = 0$$

Eviter de s'aventurer dans l'interprétation car les contraintes dépendent du logiciel. Elle a insisté sur le pas interpréter ça.

Néanmoins, on peut voir que le modèle est utile (p value tout en bas)

```
anova(res)

## Analysis of Variance Table
##
## Response: df$Right_answers
##              Df Sum Sq Mean Sq F value    Pr(>F)
## df$Group      2 2281.8  1140.88  65.9740 < 2.2e-16 ***
## df$Agec       2  491.1   245.56  14.1999 1.375e-06 ***
## df$Group:df$Agec  4  146.9    36.72   2.1232  0.07823 .
## Residuals    268 4634.5    17.29
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Nous donne les tables des somme des effet de type 1 (car anova sans majuscule). On vas comparer les modèles dans l'ordre ou on a introduit les variables.

Ligne 4 : on test l'interaction : H_0 = sans interaction vs H_1 avec interaction. Dans notre cas avec erreur 5%, on choisit le modèle sans interaction

Ligne 3 : On test H_0 = modèle groupe VS H_1 =sans interaction mais avec les deux facteur (additif) modèle age+groupe. On choisit l'age+groupe.

Ligne 2 : H_0 = Contrainte VS H_0 = constante + groupe

```
Anova(res)
```

```
## Anova Table (Type II tests)
##
## Response: df$Right_answers
##           Sum Sq Df F value    Pr(>F)
## df$Group      1585.4  2 45.8382 < 2.2e-16 ***
## df$Agec        491.1  2 14.1999 1.375e-06 ***
## df$Group:df$Agec 146.9  4  2.1232  0.07823 .
## Residuals      4634.5 268
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Si on fait une Anova avec majuscule

Ligne Groupe : H_0 Argg j'ai pas réussi

```
res2 = lm(df$Right_answers~df$Group+df$Agec)
res2
```

```
##
## Call:
## lm(formula = df$Right_answers ~ df$Group + df$Agec)
##
## Coefficients:
##      (Intercept) df$GroupControl      df$GroupUD      df$Agec40-60
##           46.6137           4.7285           -0.8968           -1.1507
##      df$Agec>=60
##           -4.2021
```

```
summary(res2)
```

```
##
## Call:
## lm(formula = df$Right_answers ~ df$Group + df$Agec)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.4117  -2.4630   0.3863   2.6578   9.5370
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    46.6137    0.5447  85.582 < 2e-16 ***
## df$GroupControl  4.7285    0.5447   8.680 3.70e-16 ***
## df$GroupUD     -0.8968    0.8410  -1.066  0.2872
## df$Agec40-60   -1.1507    0.5926  -1.942  0.0532 .
## df$Agec>=60    -4.2021    0.8057  -5.216 3.63e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.193 on 272 degrees of freedom
## Multiple R-squared:  0.3671, Adjusted R-squared:  0.3578
## F-statistic: 39.44 on 4 and 272 DF, p-value: < 2.2e-16
```

```
anova(res2)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: df$Right_answers
```

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

```
## df$Group    2 2281.8 1140.88  64.902 < 2.2e-16 ***
```

```
## df$Agec     2  491.1  245.56  13.969 1.679e-06 ***
```

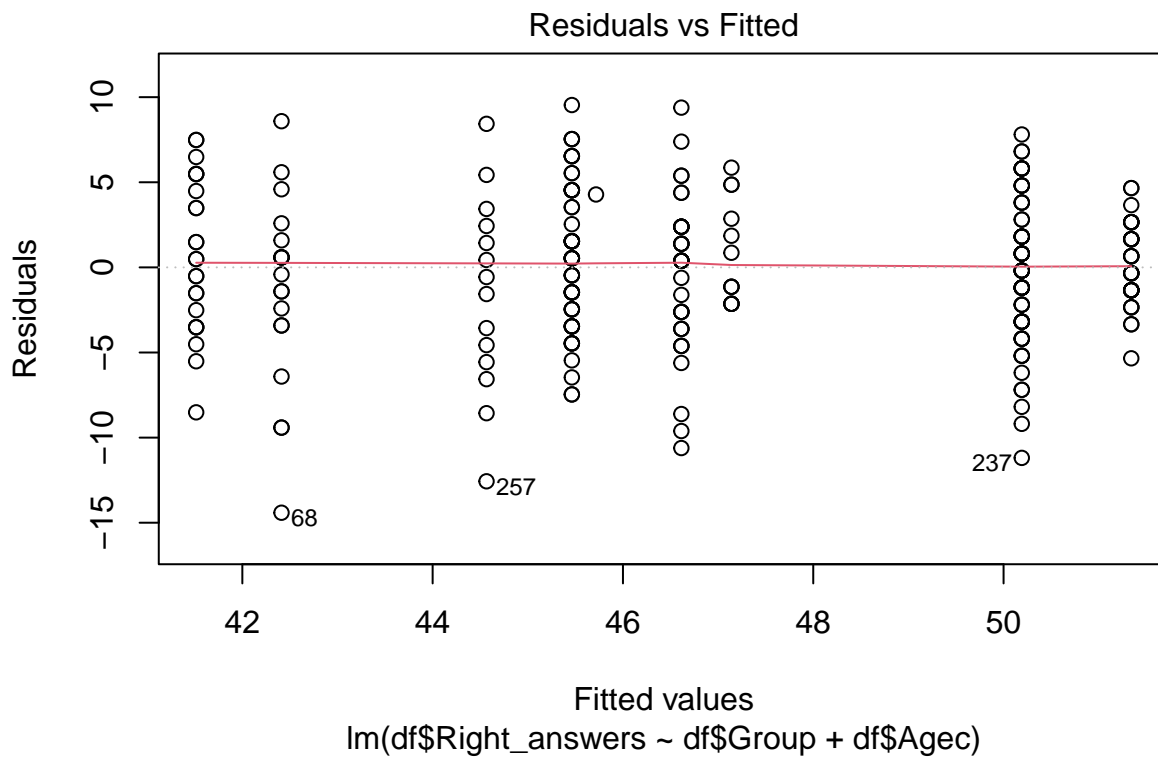
```
## Residuals 272 4781.4   17.58
```

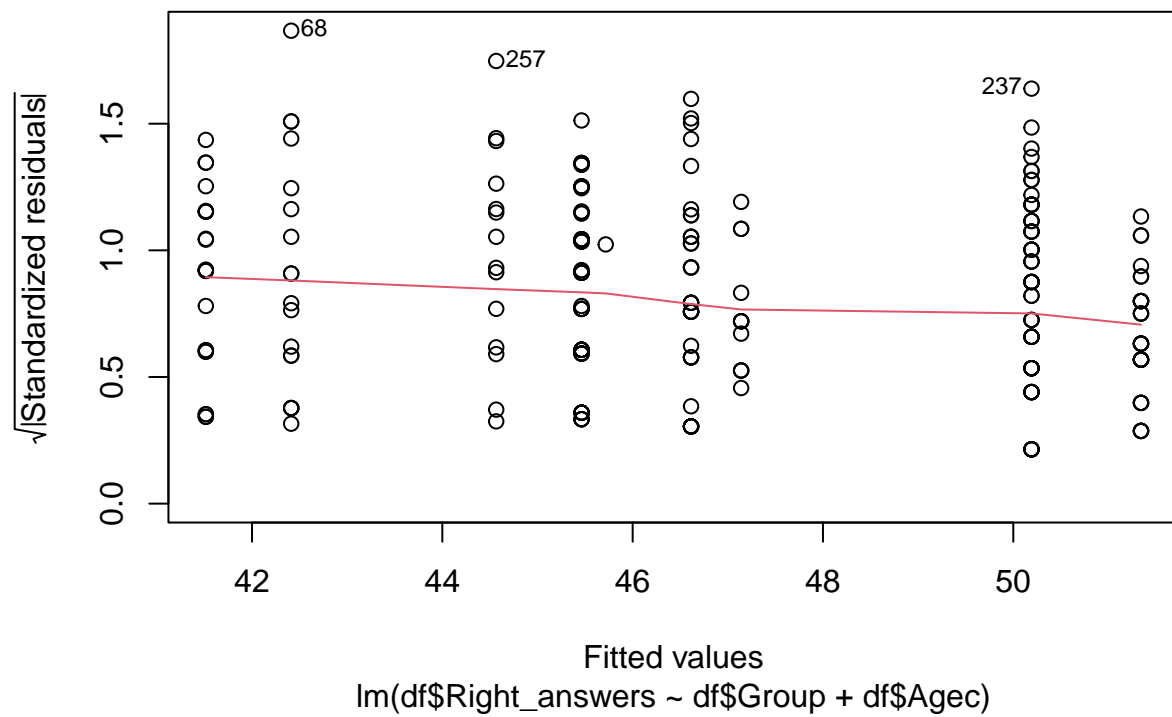
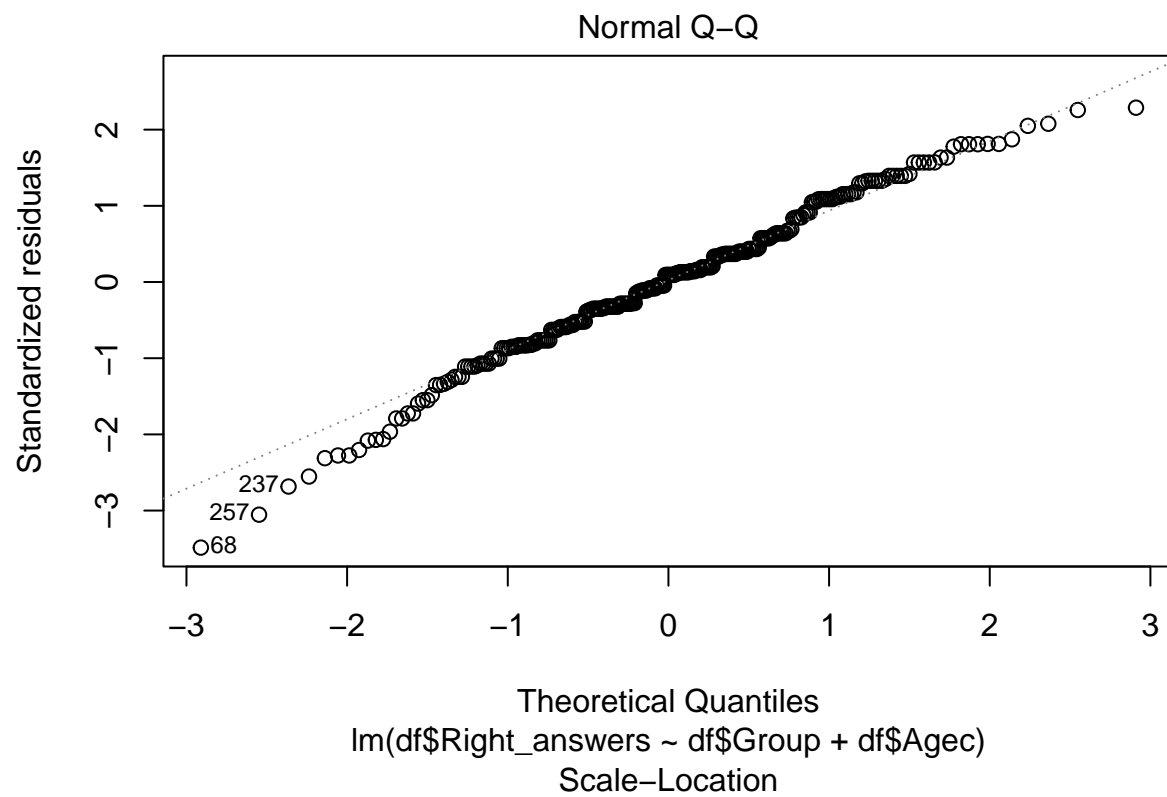
```
## ---
```

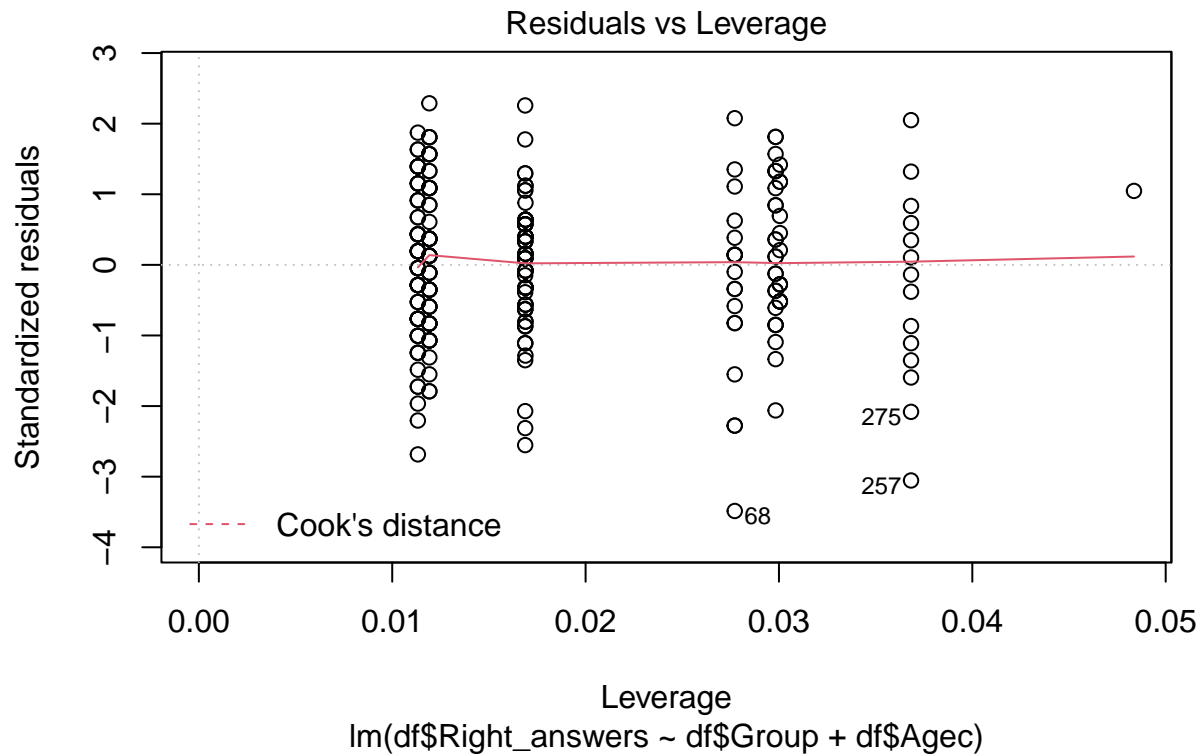
```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Ici on voit un peu la même chose, surtout que le modèle est significatif

```
plot(res2)
```







Distance de cook parfaite sauf pour un point qui dépasse $(p+1)/n = 2(i+j)/n = 0.043$. Ce point et le sujet qui était le seul dans son groupe, car il est seul il attire fortement les poids.

Q-plot ok décroche un peu sur les bord, variance ok

Question 8

```
?predict.lm
predict(res2, data.frame(Group="Control", Agec="<40"), interval="confidence") # IC à 95%
```

```
## Warning: 'newdata' had 1 row but variables found have 277 rows
```

```
##      fit      lwr      upr
## 1  45.46300 44.56182 46.36417
## 2  45.46300 44.56182 46.36417
## 3  45.46300 44.56182 46.36417
## 4  45.46300 44.56182 46.36417
## 5  45.46300 44.56182 46.36417
## 6  46.61374 45.54143 47.68604
## 7  45.46300 44.56182 46.36417
## 8  45.46300 44.56182 46.36417
## 9  45.46300 44.56182 46.36417
## 10 45.46300 44.56182 46.36417
## 11 45.46300 44.56182 46.36417
## 12 42.41166 41.03753 43.78579
## 13 45.46300 44.56182 46.36417
## 14 45.46300 44.56182 46.36417
## 15 45.46300 44.56182 46.36417
## 16 42.41166 41.03753 43.78579
## 17 45.46300 44.56182 46.36417
## 18 46.61374 45.54143 47.68604
```

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29 45.46300 44.56182 46.36417
30 45.46300 44.56182 46.36417
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117 45.46300 44.56182 46.36417
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119 46.61374 45.54143 47.68604
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124 50.19150 49.31302 51.06999
125 51.34224 50.26945 52.41503
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187 47.14016 45.70937 48.57096
188 51.34224 50.26945 52.41503
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200 47.14016 45.70937 48.57096
201 51.34224 50.26945 52.41503
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204 50.19150 49.31302 51.06999
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218 47.14016 45.70937 48.57096
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221 47.14016 45.70937 48.57096
222 51.34224 50.26945 52.41503
223 51.34224 50.26945 52.41503
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225 50.19150 49.31302 51.06999
226 51.34224 50.26945 52.41503
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```

## 235 50.19150 49.31302 51.06999
## 236 50.19150 49.31302 51.06999
## 237 50.19150 49.31302 51.06999
## 238 51.34224 50.26945 52.41503
## 239 41.51485 40.08931 42.94039
## 240 41.51485 40.08931 42.94039
## 241 41.51485 40.08931 42.94039
## 242 41.51485 40.08931 42.94039
## 243 44.56619 42.98209 46.15029
## 244 41.51485 40.08931 42.94039
## 245 44.56619 42.98209 46.15029
## 246 41.51485 40.08931 42.94039
## 247 41.51485 40.08931 42.94039
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## 272 44.56619 42.98209 46.15029
## 273 44.56619 42.98209 46.15029
## 274 45.71693 43.90174 47.53212
## 275 44.56619 42.98209 46.15029
## 276 44.56619 42.98209 46.15029
## 277 41.51485 40.08931 42.94039

predict(res2, data.frame(Group="Control", Agec="<40"), interval="prediction") # IC à 95% + la variance

## Warning: 'newdata' had 1 row but variables found have 277 rows

##      fit      lwr      upr
## 1  45.46300 37.15972 53.76627
## 2  45.46300 37.15972 53.76627
## 3  45.46300 37.15972 53.76627
## 4  45.46300 37.15972 53.76627
## 5  45.46300 37.15972 53.76627
## 6  46.61374 38.29015 54.93732

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12 42.41166 34.04383 50.77948
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18 46.61374 38.29015 54.93732
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266 44.56619 36.16133 52.97105
267 44.56619 36.16133 52.97105
268 44.56619 36.16133 52.97105
269 41.51485 33.13843 49.89127
270 41.51485 33.13843 49.89127
271 41.51485 33.13843 49.89127
272 44.56619 36.16133 52.97105
273 44.56619 36.16133 52.97105
274 45.71693 37.26547 54.16839
275 44.56619 36.16133 52.97105
276 44.56619 36.16133 52.97105

```
## 277 41.51485 33.13843 49.89127
```

Question 9

```
library(emmeans)
emmeans(res2, ~Group)
```

```
## Group emmean SE df lower.CL upper.CL
## Bipolar 44.8 0.404 272 44.0 45.6
## Control 49.6 0.417 272 48.7 50.4
## UD 43.9 0.711 272 42.5 45.3
##
## Results are averaged over the levels of: Agec
## Confidence level used: 0.95
```

Dans ce cas, ça utilise une moyenne marginale, on vas faire comme si on avait une répartition de sujet parfaite dans chaque classe. Ca se construit en pondérant les moyenne en m'étant le même coef pour chaque classe d'age.

La moyenne de UD change beaucoup car peu de jeune dans ce groupe, il faut compenser l'impact des score moins bon des vieux.

Ici on veut garder l'information de l'age dans la prediction. Car sinon on pourrait juste faire une anova à 1 facteur.

Question 11

On cherche à savoir si $\alpha_{BD} = \alpha_{UD}$. On vas tester $H_0 = \{\alpha_{BD} = \alpha_{UD}\}, H_1 = \{\alpha_{BD} \neq \alpha_{UD}\}$. C'est un test de contraste. $1 * \alpha_{BD} + 0 * \alpha_{ctrl} - 1 * \alpha_{UD} = 0$

```
c1 = c(1,0,-1)
contrast(emmeans(res2, ~Group), list(Group=c1)) # Pas d'écart significatif
```

```
## contrast estimate SE df t.ratio p.value
## Group 0.897 0.841 272 1.066 0.2872
##
## Results are averaged over the levels of: Agec
```

Question 11

```
contrast(emmeans(res2, ~Group), "pairwise", adjust="Tukey")
```

```
## contrast estimate SE df t.ratio p.value
## Bipolar - Control -4.729 0.545 272 -8.680 <.0001
## Bipolar - UD 0.897 0.841 272 1.066 0.5358
## Control - UD 5.625 0.854 272 6.590 <.0001
##
## Results are averaged over the levels of: Agec
## P value adjustment: tukey method for comparing a family of 3 estimates
```

```
cat("\n\n\n")
```

```
contrast(emmeans(res2, ~Group), "pairwise", adjust="Bonferroni")
```

```
## contrast estimate SE df t.ratio p.value
```

```
## Bipolar - Control    -4.729 0.545 272  -8.680 <.0001
## Bipolar - UD         0.897 0.841 272   1.066 0.8616
## Control - UD        5.625 0.854 272   6.590 <.0001
##
## Results are averaged over the levels of: Agec
## P value adjustment: bonferroni method for 3 tests
```

Ici ça fait plusieurs (3) test into une p-value je crois, ce qui explique la différence de p-value avec le test question d'avant.