

QUIZ1 ADEI_1920Q2: Solutions to questions

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List of Questions

25 personality self-report items taken from the International Personality Item Pool (ipip.ori.org) were included as part of the Synthetic Aperture Personality Assessment (SAPA) web based personality assessment project (SAPA <https://sapa-project.org>). The data from 2800 subjects are included here. Three additional demographic variables (sex, education, and age) are also included. The first 25 items are organized by five putative factors: Agreeableness, Conscientiousness, Extraversion, Neuroticism, and Openness. The item data were collected using a 6 point response scale: 1 Very Inaccurate 2 Moderately Inaccurate 3 Slightly Inaccurate 4 Slightly Accurate 5 Moderately Accurate 6 Very Accurate. The items given were sampled from the International Personality Item Pool of Lewis Goldberg using the sampling technique of SAPA. This is a sample data set taken from the much larger SAPA data bank. Available variables:

- A1 I Am indifferent to the feelings of others. (q_146)
- A2 Inquire about others' well-being. (q_1162)
- A3 Know how to comfort others. (q_1206)
- A4 Love children. (q_1364)
- A5 Make people feel at ease. (q_1419)
- C1 Am exacting in my work. (q_124)
- C2 Continue until everything is perfect. (q_530)
- C3 Do things according to a plan. (q_619)
- C4 Do things in a half-way manner. (q_626)
- C5 Waste my time. (q_1949)
- E1 Don't talk a lot. (q_712)
- E2 Find it difficult to approach others. (q_901)
- E3 Know how to captivate people. (q_1205)
- E4 Make friends easily. (q_1410)
- E5 Take charge. (q_1768)
- N1 Get angry easily. (q_952)
- N2 Get irritated easily. (q_974)
- N3 Have frequent mood swings. (q_1099)
- N4 Often feel blue. (q_1479)
- N5 Panic easily. (q_1505)
- O1 Am full of ideas. (q_128)
- O2 Avoid difficult reading material. (q_316)
- O3 Carry the conversation to a higher level. (q_492)
- O4 Spend time reflecting on things. (q_1738)
- O5 Will not probe deeply into a subject. (q_1964)
- gender gender Males = 1, Females = 2
- education 1 = HS, 2 = finished HS, 3 = some college, 4 = college graduate 5 = graduate degree
- age age in years

Source: The items are from the ipip (Goldberg, 1999). The data are from the SAPA project (Revelle, Wilt and Rosenthal, 2010) , collected Spring, 2010 (<https://sapa-project.org>).

References: 1. Goldberg, L.R. (1999) A broad-bandwidth, public domain, personality inventory measuring the lower-level facets of several five-factor models. In Mervielde, I. and Deary, I. and De Fruyt, F. and Ostendorf, F. (eds) Personality psychology in Europe. 7. Tilburg University Press. Tilburg, The Netherlands. 2. Revelle, W., Wilt, J., and Rosenthal, A. (2010) Individual Differences in Cognition: New Methods

for examining the Personality-Cognition Link In Gruszka, A. and Matthews, G. and Szymura, B. (Eds.) Handbook of Individual Differences in Cognition: Attention, Memory and Executive Control, Springer. 3. Revelle, W, Condon, D.M., Wilt, J., French, J.A., Brown, A., and Elleman, L.G. (2016) Web and phone based data collection using planned missing designs. In Fielding, N.G., Lee, R.M. and Blank, G. (Eds). SAGE Handbook of Online Research Methods (2nd Ed), Sage Publications.

Firstly, load dataset and check available variables.

```
rm(list=ls())
setwd("C:/Users/lidia/Dropbox/DOCENCIA/FIB-ADEI/EXAMENS/1920Q2")
load("C:/Users/lidia/Dropbox/DOCENCIA/FIB-ADEI/EXAMENS/1920Q2/bfi_Raw.RData")
summary(df)
```

```
##           A1           A2           A3           A4
## Min.      :1.000   Min.      :1.000   Min.      :1.000   Min.      :1.0
## 1st Qu.:1.000   1st Qu.:4.000   1st Qu.:4.000   1st Qu.:4.0
## Median :2.000   Median :5.000   Median :5.000   Median :5.0
## Mean     :2.413   Mean     :4.802   Mean     :4.604   Mean     :4.7
## 3rd Qu.:3.000   3rd Qu.:6.000   3rd Qu.:6.000   3rd Qu.:6.0
## Max.     :6.000   Max.     :6.000   Max.     :6.000   Max.     :6.0
## NA's     :16     NA's     :27     NA's     :26     NA's     :19
##           A5           C1           C2           C3
## Min.      :1.00   Min.      :1.000   Min.      :1.00   Min.      :1.000
## 1st Qu.:4.00   1st Qu.:4.000   1st Qu.:4.00   1st Qu.:4.000
## Median :5.00   Median :5.000   Median :5.00   Median :5.000
## Mean     :4.56   Mean     :4.502   Mean     :4.37   Mean     :4.304
## 3rd Qu.:5.00   3rd Qu.:5.000   3rd Qu.:5.00   3rd Qu.:5.000
## Max.     :6.00   Max.     :6.000   Max.     :6.00   Max.     :6.000
## NA's     :16     NA's     :21     NA's     :24     NA's     :20
##           C4           C5           E1           E2
## Min.      :1.000   Min.      :1.000   Min.      :1.000   Min.      :1.000
## 1st Qu.:1.000   1st Qu.:2.000   1st Qu.:2.000   1st Qu.:2.000
## Median :2.000   Median :3.000   Median :3.000   Median :3.000
## Mean     :2.553   Mean     :3.297   Mean     :2.974   Mean     :3.142
## 3rd Qu.:4.000   3rd Qu.:5.000   3rd Qu.:4.000   3rd Qu.:4.000
## Max.     :6.000   Max.     :6.000   Max.     :6.000   Max.     :6.000
## NA's     :26     NA's     :16     NA's     :23     NA's     :16
##           E3           E4           E5           N1
## Min.      :1.000   Min.      :1.000   Min.      :1.000   Min.      :1.000
## 1st Qu.:3.000   1st Qu.:4.000   1st Qu.:4.000   1st Qu.:2.000
## Median :4.000   Median :5.000   Median :5.000   Median :3.000
## Mean     :4.001   Mean     :4.422   Mean     :4.416   Mean     :2.929
## 3rd Qu.:5.000   3rd Qu.:6.000   3rd Qu.:5.000   3rd Qu.:4.000
## Max.     :6.000   Max.     :6.000   Max.     :6.000   Max.     :6.000
## NA's     :25     NA's     :9     NA's     :21     NA's     :22
##           N2           N3           N4           N5
## Min.      :1.000   Min.      :1.000   Min.      :1.000   Min.      :1.00
## 1st Qu.:2.000   1st Qu.:2.000   1st Qu.:2.000   1st Qu.:2.00
## Median :4.000   Median :3.000   Median :3.000   Median :3.00
## Mean     :3.508   Mean     :3.217   Mean     :3.186   Mean     :2.97
## 3rd Qu.:5.000   3rd Qu.:4.000   3rd Qu.:4.000   3rd Qu.:4.00
## Max.     :6.000   Max.     :6.000   Max.     :6.000   Max.     :6.00
## NA's     :21     NA's     :11     NA's     :36     NA's     :29
##           O1           O2           O3           O4
## Min.      :1.000   Min.      :1.000   Min.      :1.000   Min.      :1.000
```

```
## 1st Qu.:4.000 1st Qu.:1.000 1st Qu.:4.000 1st Qu.:4.000
## Median :5.000 Median :2.000 Median :5.000 Median :5.000
## Mean :4.816 Mean :2.713 Mean :4.438 Mean :4.892
## 3rd Qu.:6.000 3rd Qu.:4.000 3rd Qu.:5.000 3rd Qu.:6.000
## Max. :6.000 Max. :6.000 Max. :6.000 Max. :6.000
## NA's :22 NA's :28 NA's :14
## 05 gender education age
## Min. :1.00 Min. :1.000 Min. :1.00 Min. : 3.00
## 1st Qu.:1.00 1st Qu.:1.000 1st Qu.:3.00 1st Qu.:20.00
## Median :2.00 Median :2.000 Median :3.00 Median :26.00
## Mean :2.49 Mean :1.672 Mean :3.19 Mean :28.78
## 3rd Qu.:3.00 3rd Qu.:2.000 3rd Qu.:4.00 3rd Qu.:35.00
## Max. :6.00 Max. :2.000 Max. :5.00 Max. :86.00
## NA's :20 NA's :223
```

1. Define a binary factor for gender `f.gender` and a polytomic factor for education `f.educ`. Justify with R commands for the procedure and your answer. Calculate thresholds to identify severe outliers for the age variable (`age`).

```
df$f.gender<-factor(df$gender,labels=c("sex.male","sex.female"))
summary(df$f.gender)
```

```
## sex.male sex.female
## 919 1881
```

```
ll<-which(is.na(df$education))
```

```
df$education[ll]<-6
```

```
df$f.educ<-factor(df$education,labels=c("HS", "finished HS", "some college", "college graduate", "graduate"),
levels(df$educ))
```

```
## NULL
```

```
summary(df$educ)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 1.000 3.000 3.000 3.414 4.000 6.000
```

```
sumres<-summary(df$age)
```

```
iqr<-as.numeric(sumres[5]-sumres[2]);iqr
```

```
## [1] 15
```

```
mildlow<-as.numeric(sumres[2]-1.5*iqr)
```

```
mildup<-as.numeric(sumres[5]+1.5*iqr)
```

```
sevlw<-as.numeric(sumres[2]-3*iqr)
```

```
sevup<-as.numeric(sumres[5]+3*iqr)
```

```
mildlow;mildup
```

```
## [1] -2.5
```

```
## [1] 57.5
```

```
sevlw;sevup
```

```
## [1] -25
```

```
## [1] 80
```

```
ll<-which(df$age>sevup);length(ll);ll
```

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

```
## [1] 1158
```

Education can be seen to have 223 missing values. Imputation is not a reasonable solution and an specific level unknown has to be defined. Gender and education are defined as factors. Age is a numeric variable without missing data. Computation of severe outliers thresholds determines that those observations greater than 80 are severe outliers: only 1 person satisfy this condition (obs. 1158). Lower severe threshold does not make sense (since is -25). Follow R commands to figure out the calculus of these thresholds, based on 1.5/3 times Inter Quartilar Range from Q1/Q3. Or check theory slide notes.

2. Conduct a suitable data imputation procedure to remove missing data included in dataset for numeric variables. Check imputation consistency for numeric variables.

```
library(missMDA)
```

```
## Warning: package 'missMDA' was built under R version 3.6.2
```

```
names(df)
```

```
## [1] "A1"      "A2"      "A3"      "A4"      "A5"
## [6] "C1"      "C2"      "C3"      "C4"      "C5"
## [11] "E1"      "E2"      "E3"      "E4"      "E5"
## [16] "N1"      "N2"      "N3"      "N4"      "N5"
## [21] "O1"      "O2"      "O3"      "O4"      "O5"
## [26] "gender"  "education" "age"      "f.gender" "f.educ"
```

```
#summary(df[,c(1:25)])
```

```
res.impu<-imputePCA(df[,c(1:25,28)])
```

```
dfimpu<-as.data.frame(res.impu$completeObs)
```

```
#library(psych)
```

```
#describe(df[,1:25])
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 3.6.2
```

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

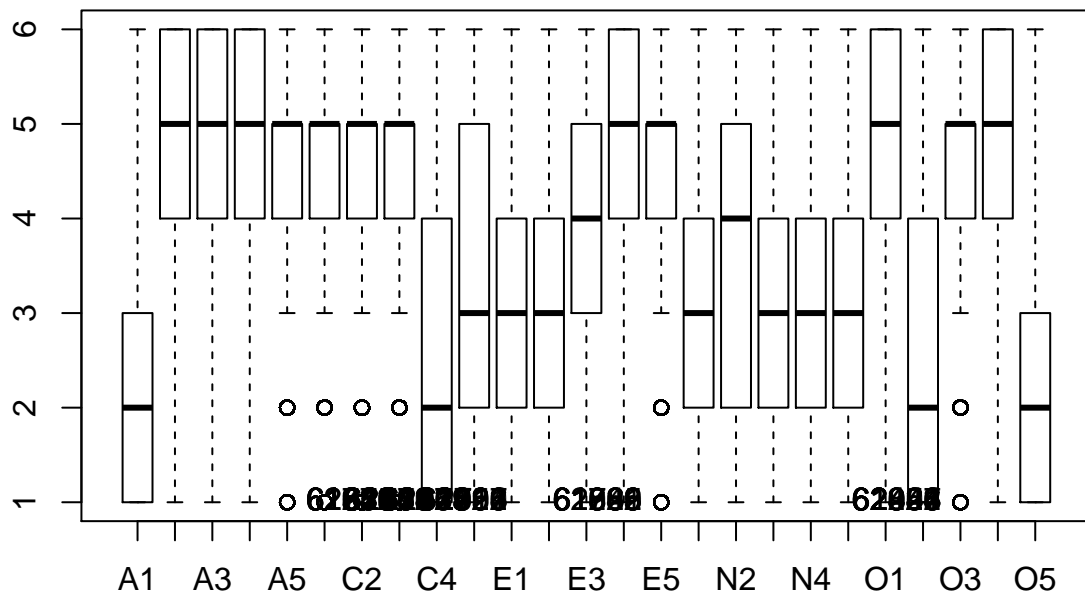
```
## Registered S3 methods overwritten by 'car':
```

```
## method                from
## influence.merMod        lme4
## cooks.distance.influence.merMod lme4
## dfbeta.influence.merMod lme4
## dfbetas.influence.merMod lme4
```

```
Boxplot(df[,1:25])
```

```
## [1] "61629" "61640" "61788" "61840" "61873" "61926" "61932" "62282"
## [9] "62551" "62552" "61654" "61682" "61761" "61921" "61979" "62038"
## [17] "62060" "62102" "62111" "62498" "61654" "61825" "61839" "61865"
## [25] "61918" "61921" "61969" "61979" "62029" "62079" "61654" "61701"
## [33] "61716" "62022" "62029" "62092" "62526" "62716" "62787" "62795"
## [41] "61629" "61682" "61761" "61788" "61825" "61840" "61865" "61989"
## [49] "62092" "62266" "61856" "61926" "62022" "62054" "62064" "62246"
## [57] "62327" "62328" "62443" "62491"
```

```
Boxplot(dfimpu[,1:25])
```



```
## [1] "61629" "61640" "61788" "61840" "61873" "61926" "61932" "62282"
## [9] "62551" "62552" "61654" "61682" "61761" "61921" "61979" "62038"
## [17] "62060" "62102" "62111" "62498" "61654" "61825" "61839" "61865"
## [25] "61918" "61921" "61969" "61979" "62029" "62079" "61654" "61701"
## [33] "61716" "62022" "62029" "62092" "62526" "62716" "62787" "62795"
## [41] "61629" "61682" "61761" "61788" "61825" "61840" "61865" "61989"
## [49] "62092" "62266" "61856" "61926" "62022" "62054" "62064" "62246"
## [57] "62327" "62328" "62443" "62491"
```

```
lapply(df[,1:25],quantile, probs=seq(0,1,0.1),na.rm=T)
```

```
## $A1
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 1 2 2 2 3 4 5 6
##
## $A2
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 5 5 5 6 6 6 6
##
## $A3
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 5 5 5 5 6 6 6
##
## $A4
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 4 4 5 5 6 6 6 6 6
##
```

```

## $A5
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 4 5 5 5 6 6 6
##
## $C1
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 4 5 5 5 6 6 6
##
## $C2
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 3 4 4 5 5 5 5 6 6
##
## $C3
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 3 4 4 5 5 5 5 6 6
##
## $C4
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 2 3 3 4 5 6
##
## $C5
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 2 2 3 3 4 4 5 6 6
##
## $E1
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 3 3 4 5 5 6
##
## $E2
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 2 2 2 3 4 4 5 5 6
##
## $E3
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 3 3 4 4 4 5 5 6 6
##
## $E4
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 3 4 4 5 5 5 6 6 6
##
## $E5
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 2 3 4 4 5 5 5 6 6 6
##
## $N1
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 3 3 4 4 5 6
##
## $N2
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 2 2 3 4 4 4 5 6 6
##
## $N3
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%

```

```
##      1      1      2      2      2      3      4      4      5      5      6
##
## $N4
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      1      2      2      2      3      4      4      5      5      6
##
## $N5
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      1      1      2      2      3      3      4      5      5      6
##
## $O1
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      3      4      4      5      5      5      6      6      6      6
##
## $O2
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      1      1      2      2      2      3      4      4      5      6
##
## $O3
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      3      4      4      4      5      5      5      5      6      6
##
## $O4
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      3      4      5      5      5      5      6      6      6      6
##
## $O5
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      1      1      2      2      2      3      3      4      4      6
```

```
lapply(dfimpu[,1:25],quantile, probs=seq(0,1,0.1),na.rm=T)
```

```
## $A1
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      1      1      1      2      2      2      3      4      5      6
##
## $A2
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      3      4      4      5      5      5      6      6      6      6
##
## $A3
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      3      4      4      5      5      5      5      6      6      6
##
## $A4
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##      1      2      4      4      5      5      6      6      6      6      6
##
## $A5
##           0%           10%           20%           30%           40%           50%           60%           70%
## 1.000000 3.000000 4.000000 4.000000 4.318546 5.000000 5.000000 5.000000
##           80%           90%          100%
## 6.000000 6.000000 6.000000
##
## $C1
```

##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	3	4	4	4	5	5	5	6	6	6
##											
##	\$C2										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	2	3	4	4	5	5	5	5	6	6
##											
##	\$C3										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	2	3	4	4	5	5	5	5	6	6
##											
##	\$C4										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	1	2	2	2	3	3	4	5	6
##											
##	\$C5										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	2	2	3	3	4	4	5	6	6
##											
##	\$E1										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	1	2	2	3	3	4	5	5	6
##											
##	\$E2										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	2	2	2	3	4	4	5	5	6
##											
##	\$E3										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	2	3	3	4	4	4	5	5	6	6
##											
##	\$E4										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	2	3	4	4	5	5	5	6	6	6
##											
##	\$E5										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	2	3	4	4	5	5	5	6	6	6
##											
##	\$N1										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	1	2	2	3	3	4	4	5	6
##											
##	\$N2										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	2	2	3	4	4	4	5	6	6
##											
##	\$N3										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	2	2	2	3	4	4	5	5	6
##											
##	\$N4										
##	0%	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%
##	1	1	2	2	2	3	4	4	5	5	6


```
##
## $N5
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 3 3 4 5 5 6
##
## $O1
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 5 5 5 6 6 6 6
##
## $O2
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 2 3 4 4 5 6
##
## $O3
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 4 4 5 5 5 5 6 6
##
## $O4
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 3 4 5 5 5 5 6 6 6 6
##
## $O5
## 0% 10% 20% 30% 40% 50% 60% 70% 80% 90% 100%
## 1 1 1 2 2 2 3 3 4 4 6
```

```
df[,1:25]<-res.impu$completeObs[,1:25]
#summary(df)
```

All 25 first variables have missing values, between 9 and 36, except variable O2. Method `imputePCA()` from `missMDA` package has to be used for imputation of numeric variables. Check for reasonable imputation values has to be done using either graphics or quantiles. No problems seems to be present.

3. Conduct a suitable data imputation procedure for factors. Summarize imputation results for f.education factor.

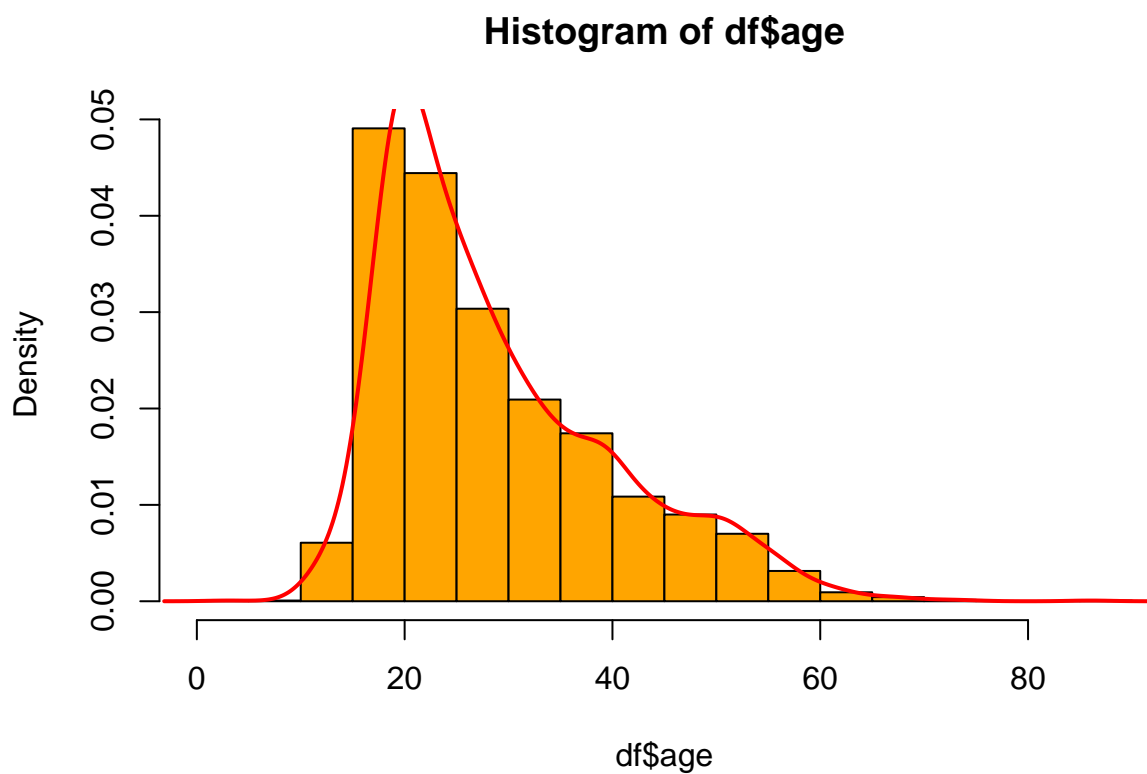
```
summary(df$f.educ)
```

```
##          HS          finished HS          some college college graduate
##          224             292             1249             394
## graduate degree          Unknown
##          418             223
```

Imputation for factors would have to use `f.educ` and `f.gender` data, so it is not likely that they contain enough information for a suitable imputation. If a set of factors had been included in dataset, then `imputeMCA()` in `missMDA` package would have to be used for imputation purposes. Actually, missing values of variable education have to be selected to define a new level in factor `f.educ` labelled as “Unknown”. There are 223 observations with unknown education level.

4. Can the average of age can be argued to be the same for all education levels (f.educ) and gender (f.gender)? Which are the groups that show significant greater values than the others? Use graphic, numeric and inferential tools.

```
hist(df$age,15,freq=F,col="orange")
lines(density(df$age),col="red",lwd=2)
```



```
shapiro.test(df$age)
```

```
##  
##  Shapiro-Wilk normality test  
##  
## data:  df$age  
## W = 0.91212, p-value < 2.2e-16
```

```
Boxplot(df$age~df$f.gender,main="Age by Gender")
```

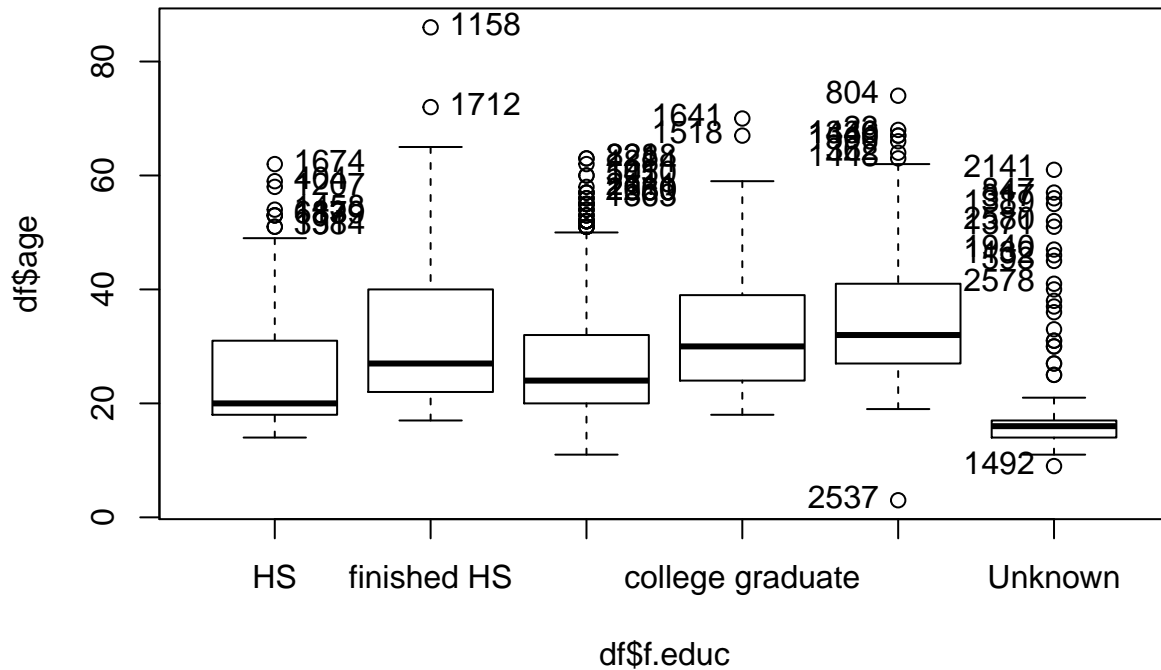
Age by Gender



```
## [1] "804" "1641" "23" "1349" "1436" "1518" "2538" "1448" "1100" "1242"
## [11] "1158" "1712" "1500" "362" "821" "2288" "1674" "1884" "2450" "2141"
```

```
Boxplot(df$age~df$f.educ,main="Age by Education Level")
```

Age by Education Level



```
## [1] "183" "338" "404" "613" "1207" "1458" "1674" "1879" "1914" "1158"
## [11] "1712" "821" "2288" "1884" "545" "1010" "2111" "1969" "2280" "2630"
## [21] "1363" "1518" "1641" "2537" "23" "362" "804" "1349" "1436" "1448"
## [31] "1500" "1492" "2141" "847" "317" "1389" "2580" "1371" "1940" "1132"
## [41] "593" "2578"
```

```
tapply(df$age,df$f.gender,summary)
```

```
## $sex.male
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   3.00  20.00   25.00  28.02  34.00   74.00
##
## $sex.female
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   9.00  20.00   26.00  29.15  36.00   86.00
```

```
tapply(df$age,df$f.educ,summary)
```

```
## $HS
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  14.00  18.00   20.00  25.13  31.00   62.00
##
## $`finished HS`
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  17.00  22.00   27.00  31.51  40.00   86.00
##
## $`some college`
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
##    11.00    20.00    24.00    27.23    32.00    63.00
##
## $`college graduate`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      18.00    24.00    30.00    32.98    39.00    70.00
##
## $`graduate degree`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##       3.0     27.0     32.0     35.3     41.0     74.0
##
## $Unknown
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##       9.00    14.00    16.00    17.95    17.00    61.00

kruskal.test(df$age,df$f.gender)

##
##  Kruskal-Wallis rank sum test
##
## data:  df$age and df$f.gender
## Kruskal-Wallis chi-squared = 7.8848, df = 1, p-value = 0.004985

kruskal.test(df$age,df$f.educ)

##
##  Kruskal-Wallis rank sum test
##
## data:  df$age and df$f.educ
## Kruskal-Wallis chi-squared = 733.88, df = 5, p-value < 2.2e-16

pairwise.wilcox.test(df$age,df$f.educ,alternative="greater")

##
##  Pairwise comparisons using Wilcoxon rank sum test
##
## data:  df$age and df$f.educ
##
##              HS      finished HS some college college graduate
## finished HS    2.4e-14 -              -              -
## some college   3.0e-09 1.0000        -              -
## college graduate < 2e-16 0.0036      < 2e-16        -
## graduate degree < 2e-16 2.6e-08      < 2e-16        0.0019
## Unknown       1.0000 1.0000        1.0000        1.0000
##
##              graduate degree
## finished HS      -
## some college     -
## college graduate -
## graduate degree  -
## Unknown         1.0000
##
## P value adjustment method: holm

library(FactoMineR)

## Warning: package 'FactoMineR' was built under R version 3.6.2
```

```
res.condes<-condes(df[,c(1:25,28:30)],num.var=26,proba=0.01)
res.condes$quanti
```

```
##      correlation      p.value
## A4  0.14447666 1.567367e-14
## A5  0.12880606 7.850631e-12
## E5  0.11507222 1.017596e-09
## A2  0.11454173 1.214740e-09
## C1  0.07945618 2.562716e-05
## A3  0.06912295 2.518337e-04
## C3  0.06772844 3.353129e-04
## C5 -0.08699401 4.027149e-06
## N1 -0.08780808 3.266923e-06
## O5 -0.09968771 1.251611e-07
## N2 -0.10334852 4.230972e-08
## N5 -0.10391457 3.565737e-08
## E2 -0.10596202 1.906176e-08
## N3 -0.11174301 3.051191e-09
## C4 -0.14765217 4.074830e-15
## A1 -0.16115729 9.489506e-18
```

```
res.condes$quali
```

```
##      R2      p.value
## f.educ 0.1703955 1.27964e-110
```

```
res.condes$category
```

```
##      Estimate      p.value
## f.educ=graduate degree 6.950567 1.064868e-39
## f.educ=college graduate 4.628827 4.548861e-16
## f.educ=finished HS     3.162830 9.046486e-06
## f.educ=HS              -3.216940 2.953164e-07
## f.educ=some college    -1.125088 2.627052e-11
## f.educ=Unknown         -10.400196 5.288444e-54
```

Numeric, graphic and inferential tools have to use to answer this question. First of all a rough assessment of age normality is performed: clearly shape is not symmetric and Shapiro-Wilk test rejects the null hypothesis of normality. Without normality, non-parametric methods have to be used. Summary statistics for groups of age defined according to f.gender/f.educ are not conclusive for f.gender, but differences appear for f.educ levels. Graphics: boxplot of age for f.gender is difficult to assess, but boxplot for each f.educ level show a clear different profile for age depending on the levels. Inferential tools: Null hypothesis for equal means age according to f.gender/f.educ are both rejected (pvalue=5e-03 for gender and pvalue=0 for education factor). Pairwise mean tests for f.educ can be computed and null hypothesis can be rejected (some of them). Null hypothesis can be defined as mean in group i greater (less) than mean in j .

A `condes()` method can be used for a fast answer: `$quali` shows global significance of f.educ and f.gender. It also shows graduate degree, college graduate, finished HS and female mean ages are over the mean and males, HS, some college and unknown are significantly under the global mean of age.

5. Let us assume that education (f.educ) is the target variable. Use a suitable feature selection and profiling tool to discuss **global association** between target and numerical variables/factors in dataset.

```
names(df)
```

```
## [1] "A1"      "A2"      "A3"      "A4"      "A5"
```

```
## [6] "C1"      "C2"      "C3"      "C4"      "C5"
## [11] "E1"      "E2"      "E3"      "E4"      "E5"
## [16] "N1"      "N2"      "N3"      "N4"      "N5"
## [21] "O1"      "O2"      "O3"      "O4"      "O5"
## [26] "gender"  "education" "age"      "f.gender" "f.educ"
```

```
res.catdes<-catdes(df[,c(1:25,28:30)],num.var=28,proba=0.01)
res.catdes$test.chi2
```

```
##                p.value df
## f.gender 0.0003503056  5
```

```
res.catdes$quanti.var
```

```
##          Eta2      P-value
## age 0.170395545 1.279640e-110
## A4 0.038149120 7.589005e-22
## A1 0.027952813 1.201449e-15
## C4 0.022730429 1.583479e-12
## C5 0.019288964 1.694486e-10
## O3 0.018711574 3.690319e-10
## E4 0.016014108 1.365382e-08
## A2 0.015117854 4.485212e-08
## C1 0.014955221 5.562131e-08
## O2 0.013667891 3.032756e-07
## A5 0.012544809 1.316374e-06
## A3 0.011749053 3.697040e-06
## C2 0.010532117 1.768863e-05
## E5 0.010161184 2.839841e-05
## N1 0.009664280 5.338181e-05
## O5 0.009540961 6.239791e-05
## N4 0.009281960 8.653265e-05
## O4 0.008702755 1.790611e-04
## N2 0.008668381 1.869235e-04
## C3 0.008279506 3.034758e-04
## E2 0.007757499 5.788560e-04
## E3 0.007101257 1.292346e-03
## N3 0.006607187 2.348837e-03
## O1 0.006073490 4.442871e-03
```

Globally associated to *f.educ* is *f.gender* factor. *f.educ* is globally associated to numeric variables *age* and 24 items more, being the most significance *A4*, *A1*, *C4*, *C5* and *O3* (all of them showing *p*values less than $1e-09$).

6. Profile HS education group according to available data in your dataset.

```
library(FactoMineR)
res.catdes$category
```

```
## $HS
##          Cla/Mod  Mod/Cla  Global  p.value  v.test
## f.gender=sex.male 10.119695 41.51786 32.82143 0.004525328 2.839013
## f.gender=sex.female 6.964381 58.48214 67.17857 0.004525328 -2.839013
##
## $`finished HS`
## NULL
##
## $`some college`
```

```

##          Cla/Mod Mod/Cla    Global      p.value      v.test
## f.gender=sex.female 47.47475 71.4972 67.17857 1.205871e-05 4.376524
## f.gender=sex.male  38.73776 28.5028 32.82143 1.205871e-05 -4.376524
##
## `$college graduate`
## NULL
##
## `$graduate degree`
## NULL
##
## $Unknown
## NULL

```

```
res.catdes$quanti
```

```

## $HS
##          v.test Mean in category Overall mean sd in category Overall sd
## C4      2.693481          2.78952      2.553145      1.377601 1.369119
## E5     -3.099918          4.15247      4.417017      1.299913 1.331387
## age    -5.115766         25.13393     28.782143     10.375964 11.125568
##          p.value
## C4      7.071015e-03
## E5      1.935742e-03
## age     3.124698e-07
##
## `$finished HS`
##          v.test Mean in category Overall mean sd in category Overall sd
## age 4.432176          31.513699     28.782143     12.227712 11.125568
## A1 4.026571          2.726282      2.413185      1.498345 1.403699
##          p.value
## age 9.328678e-06
## A1 5.659618e-05
##
## `$some college`
##          v.test Mean in category Overall mean sd in category Overall sd
## A4 8.334781          4.959001      4.700084      1.359491 1.474833
## E4 6.135781          4.609132      4.421009      1.349902 1.455627
## O2 4.887124          2.874299      2.713214      1.596449 1.564872
## A2 4.465375          4.914508      4.804686      1.126793 1.167638
## C2 4.055532          4.482939      4.370795      1.269912 1.312823
## A3 4.006821          4.714207      4.604769      1.234469 1.296718
## O5 3.361655          2.583046      2.489364      1.312532 1.323070
## A5 2.704708          4.632075      4.560564      1.234426 1.255248
## E2 -3.467854          3.024703      3.141687      1.564404 1.601558
## N4 -4.715235          3.029375      3.184460      1.539037 1.561502
## C4 -5.751950          2.387270      2.553145      1.297967 1.369119
## C5 -6.200094          3.083966      3.296112      1.582564 1.624476
## age -6.641486         27.225781     28.782143     9.445233 11.125568
##          p.value
## A4 7.764173e-17
## E4 8.474173e-10
## O2 1.023195e-06
## A2 7.992877e-06
## C2 5.002026e-05
## A3 6.154139e-05

```



```

## 05 7.747691e-04
## A5 6.836441e-03
## E2 5.246320e-04
## N4 2.414322e-06
## C4 8.822014e-09
## C5 5.642931e-10
## age 3.105357e-11
##
## `$college graduate`
##      v.test Mean in category Overall mean sd in category Overall sd
## age 8.077477      32.979695      28.782143      10.319741      11.125568
## 02 -2.743399      2.512690      2.713214      1.486348      1.564872
## A4 -3.563921      4.454574      4.700084      1.517742      1.474833
## A1 -3.728781      2.168707      2.413185      1.241342      1.403699
##      p.value
## age 6.612060e-16
## 02 6.080667e-03
## A4 3.653562e-04
## A1 1.924086e-04
##
## `$graduate degree`
##      v.test Mean in category Overall mean sd in category Overall sd
## age 12.986646      35.301435      28.782143      10.963622      11.125568
## 03 5.498533      4.739414      4.437910      1.119506      1.215245
## 01 3.327781      4.985034      4.816070      1.071881      1.125274
## 04 3.281653      5.072769      4.892402      1.101463      1.218096
## E4 -3.207547      4.210338      4.421009      1.458926      1.455627
## 05 -3.771794      2.264193      2.489364      1.320190      1.323070
## 02 -4.544317      2.392344      2.713214      1.412229      1.564872
## A1 -5.742488      2.049475      2.413185      1.240550      1.403699
##      p.value
## age 1.456720e-38
## 03 3.829639e-08
## 01 8.754079e-04
## 04 1.032006e-03
## E4 1.338723e-03
## 05 1.620777e-04
## 02 5.511363e-06
## A1 9.329557e-09
##
## $Unknown
##      v.test Mean in category Overall mean sd in category Overall sd
## C4 6.222562      3.100556      2.553145      1.390042      1.369119
## C5 5.325793      3.852018      3.296112      1.556293      1.624476
## A1 4.575801      2.825894      2.413185      1.469612      1.403699
## N1 4.337670      3.367629      2.931033      1.613464      1.566461
## N2 4.162202      3.914401      3.507424      1.456272      1.521751
## E2 3.503690      3.502242      3.141687      1.658987      1.601558
## N3 3.344357      3.559741      3.215912      1.563547      1.600022
## N4 2.945597      3.480002      3.184460      1.613780      1.561502
## 04 -3.199153      4.642011      4.892402      1.384490      1.218096
## E5 -3.508482      4.116874      4.417017      1.435163      1.331387
## C3 -3.538053      4.012189      4.304074      1.302358      1.283936
## E3 -3.667843      3.682383      4.000051      1.413264      1.347906

```

```
## C2 -4.077042 4.026877 4.370795 1.364893 1.312823
## A3 -4.152948 4.258746 4.604769 1.470946 1.296718
## O3 -4.407056 4.093786 4.437910 1.292218 1.215245
## A2 -5.500296 4.392021 4.804686 1.330553 1.167638
## A5 -5.726959 4.098655 4.560564 1.322163 1.255248
## C1 -6.098328 4.017937 4.502660 1.349187 1.237027
## A4 -7.510122 3.988390 4.700084 1.686754 1.474833
## age -15.151719 17.950673 28.782143 8.501769 11.125568
## p.value
## C4 4.891010e-10
## C5 1.005137e-07
## A1 4.744012e-06
## N1 1.440008e-05
## N2 3.151928e-05
## E2 4.588591e-04
## N3 8.247357e-04
## N4 3.223324e-03
## O4 1.378321e-03
## E5 4.506720e-04
## C3 4.030887e-04
## E3 2.446052e-04
## C2 4.561232e-05
## A3 3.282193e-05
## O3 1.047851e-05
## A2 3.791547e-08
## A5 1.022466e-08
## C1 1.071836e-09
## A4 5.907243e-14
## age 7.381048e-52
```

Men are overrepresented in HS level (41.5% of HS group vs 32.82% globally, more than 10% of men included in the sample belong to HS group), while they are underrepresented in 'some college' groups. Specifically, numeric variables whose means are significantly different to overall mean for each f.educ level are:

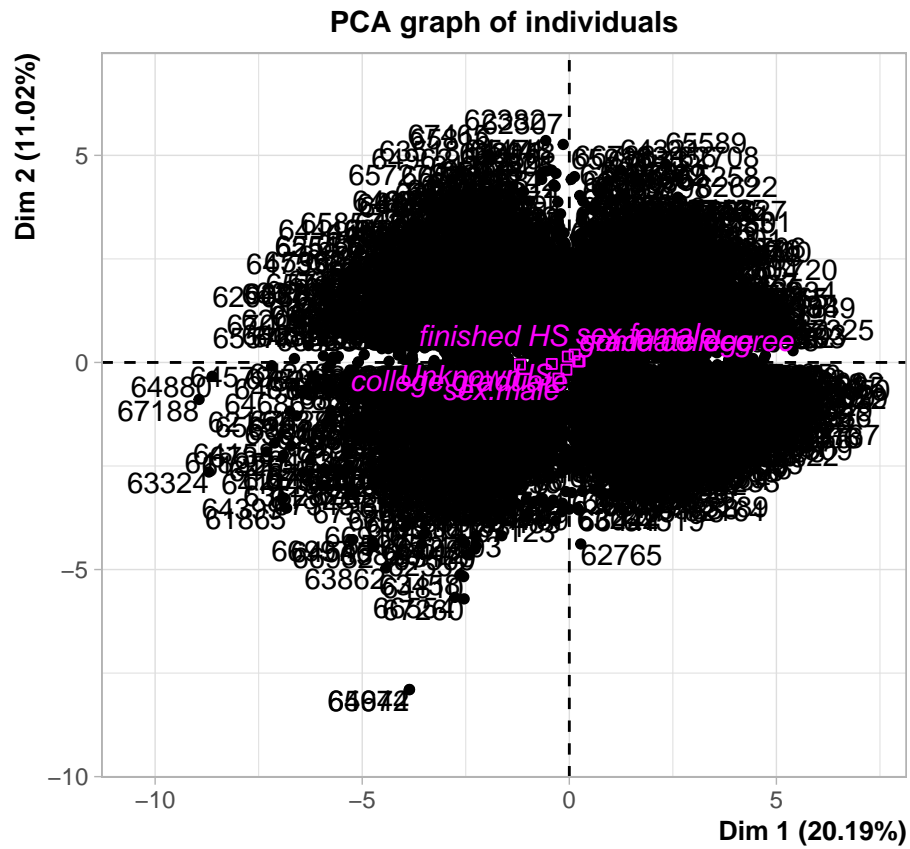
-For 'HS' level C4 and A1 are over the global mean, while age E5 and C3 are under the global mean. This is the direct answer to the question. -For 'finished HS' level age and A1 are over the global mean. -For 'some college' level A4, E4 and O2 are over the global mean, while age, C5 and C4 are under the global mean. -For 'college graduate' level age, O4 and C2 are over the global mean, while A1, A4 and O2 are under the global mean.

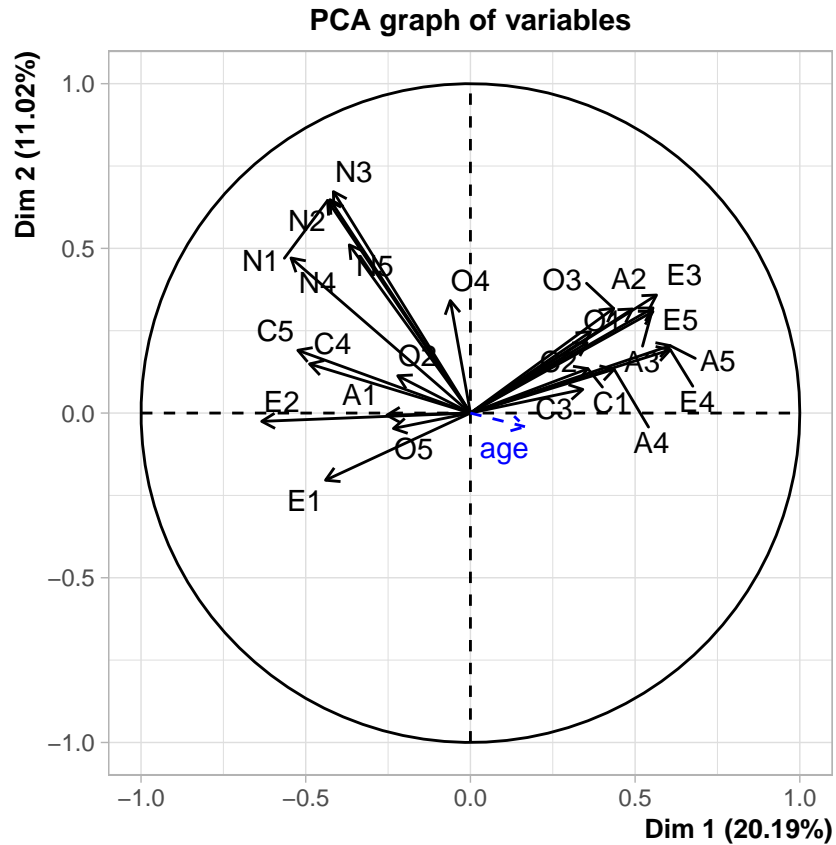
7. A Normalized Principal Component Analysis is addressed using as supplementary variables gender, education and age. How many axes do you have to retain according to Kaiser criteria? What's the inertia explained by retained Kaiser-based principal components?.

```
library(FactoMineR)
names(df)
```

```
## [1] "A1" "A2" "A3" "A4" "A5"
## [6] "C1" "C2" "C3" "C4" "C5"
## [11] "E1" "E2" "E3" "E4" "E5"
## [16] "N1" "N2" "N3" "N4" "N5"
## [21] "O1" "O2" "O3" "O4" "O5"
## [26] "gender" "education" "age" "f.gender" "f.educ"
```

```
res.pca<-PCA(df[,c(1:25,28:30)],quali.sup=27:28,quanti.sup=26)
```





```
summary(res.pca,nbind=0,nbelements = 25)
```

```
##
## Call:
## PCA(X = df[, c(1:25, 28:30)], quanti.sup = 26, quali.sup = 27:28)
##
##
## Eigenvalues
##
##          Dim.1  Dim.2  Dim.3  Dim.4  Dim.5  Dim.6
## Variance    5.048    2.755    2.098    1.823    1.527    1.111
## % of var.   20.192   11.021    8.394    7.291    6.110    4.443
## Cumulative % of var. 20.192   31.214   39.608   46.898   53.008   57.451
##
##          Dim.7  Dim.8  Dim.9  Dim.10  Dim.11  Dim.12
## Variance    0.847    0.811    0.733    0.697    0.682    0.659
## % of var.    3.387    3.244    2.932    2.788    2.728    2.636
## Cumulative % of var. 60.837   64.082   67.014   69.802   72.530   75.166
##
##          Dim.13  Dim.14  Dim.15  Dim.16  Dim.17  Dim.18
## Variance    0.629    0.596    0.563    0.541    0.524    0.499
## % of var.    2.516    2.385    2.251    2.162    2.097    1.998
## Cumulative % of var. 77.683   80.068   82.319   84.481   86.578   88.576
##
##          Dim.19  Dim.20  Dim.21  Dim.22  Dim.23  Dim.24
## Variance    0.490    0.457    0.433    0.409    0.406    0.386
## % of var.    1.960    1.828    1.733    1.635    1.622    1.542
## Cumulative % of var. 90.536   92.364   94.097   95.732   97.354   98.897
##
##          Dim.25
## Variance    0.276
```

```

## % of var.          1.103
## Cumulative % of var. 100.000
##
## Variables
##          Dim.1    ctr    cos2    Dim.2    ctr    cos2    Dim.3
## A1      | -0.254  1.276  0.064 | -0.007  0.002  0.000 |  0.141
## A2      |  0.492  4.800  0.242 |  0.316  3.623  0.100 | -0.179
## A3      |  0.554  6.079  0.307 |  0.319  3.691  0.102 | -0.266
## A4      |  0.436  3.773  0.190 |  0.133  0.642  0.018 | -0.147
## A5      |  0.607  7.289  0.368 |  0.205  1.519  0.042 | -0.291
## C1      |  0.357  2.520  0.127 |  0.135  0.659  0.018 |  0.529
## C2      |  0.354  2.488  0.126 |  0.213  1.652  0.046 |  0.515
## C3      |  0.341  2.306  0.116 |  0.072  0.187  0.005 |  0.408
## C4      | -0.488  4.726  0.239 |  0.149  0.808  0.022 | -0.470
## C5      | -0.524  5.438  0.275 |  0.191  1.318  0.036 | -0.290
## E1      | -0.440  3.834  0.194 | -0.204  1.508  0.042 |  0.340
## E2      | -0.634  7.953  0.401 | -0.025  0.023  0.001 |  0.285
## E3      |  0.565  6.324  0.319 |  0.358  4.660  0.128 | -0.155
## E4      |  0.607  7.299  0.368 |  0.192  1.341  0.037 | -0.381
## E5      |  0.554  6.084  0.307 |  0.308  3.447  0.095 |  0.084
## N1      | -0.434  3.728  0.188 |  0.645 15.117  0.417 |  0.020
## N2      | -0.426  3.602  0.182 |  0.648 15.234  0.420 |  0.072
## N3      | -0.416  3.429  0.173 |  0.672 16.408  0.452 |  0.042
## N4      | -0.545  5.877  0.297 |  0.471  8.046  0.222 |  0.111
## N5      | -0.368  2.677  0.135 |  0.511  9.462  0.261 | -0.032
## O1      |  0.365  2.634  0.133 |  0.247  2.211  0.061 |  0.257
## O2      | -0.221  0.965  0.049 |  0.113  0.466  0.013 | -0.390
## O3      |  0.435  3.742  0.189 |  0.318  3.663  0.101 |  0.190
## O4      | -0.061  0.074  0.004 |  0.342  4.234  0.117 |  0.282
## O5      | -0.234  1.083  0.055 | -0.047  0.080  0.002 | -0.353
##          ctr    cos2
## A1      0.949  0.020 |
## A2      1.535  0.032 |
## A3      3.377  0.071 |
## A4      1.029  0.022 |
## A5      4.026  0.084 |
## C1     13.349  0.280 |
## C2     12.649  0.265 |
## C3      7.923  0.166 |
## C4     10.533  0.221 |
## C5      4.005  0.084 |
## E1      5.517  0.116 |
## E2      3.875  0.081 |
## E3      1.141  0.024 |
## E4      6.930  0.145 |
## E5      0.340  0.007 |
## N1      0.019  0.000 |
## N2      0.246  0.005 |
## N3      0.084  0.002 |
## N4      0.583  0.012 |
## N5      0.050  0.001 |
## O1      3.158  0.066 |
## O2      7.239  0.152 |
## O3      1.714  0.036 |

```

```
## 04          3.787  0.079 |
## 05          5.943  0.125 |
##
## Supplementary continuous variable
##          Dim.1  cos2  Dim.2  cos2  Dim.3  cos2
## age        |  0.166  0.027 | -0.041  0.002 |  0.044  0.002 |
##
## Supplementary categories
##          Dist  Dim.1  cos2 v.test  Dim.2  cos2 v.test
## sex.male    |  0.701 | -0.263  0.140 -4.321 | -0.388  0.307 -8.650 |
## sex.female  |  0.342 |  0.128  0.140  4.321 |  0.190  0.307  8.650 |
## HS          |  0.505 | -0.423  0.702 -2.938 | -0.036  0.005 -0.340 |
## finished HS |  0.340 | -0.037  0.012 -0.301 |  0.142  0.174  1.546 |
## some college |  0.390 |  0.247  0.403  5.226 |  0.031  0.006  0.874 |
## college graduate |  0.431 | -0.070  0.026 -0.668 | -0.176  0.167 -2.269 |
## graduate degree |  0.586 |  0.216  0.136  2.131 |  0.022  0.001  0.290 |
## Unknown    |  1.338 | -1.192  0.795 -8.259 | -0.051  0.001 -0.478 |
##
##          Dim.3  cos2 v.test
## sex.male    0.143  0.042  3.647 |
## sex.female  -0.070  0.042 -3.647 |
## HS          -0.045  0.008 -0.481 |
## finished HS  -0.019  0.003 -0.236 |
## some college -0.058  0.022 -1.887 |
## college graduate 0.116  0.072  1.712 |
## graduate degree 0.224  0.147  3.434 |
## Unknown     -0.233  0.030 -2.506 |
```

Strictly following Kaiser criteria, we have to retain as many axes as eigenvalues greater than 1.0 (mean eigenvalue value). 6 axes satisfy the condition and explain 57.25% of the total inertia.

8. Try to explain the meaning of the axes in the first factorial plane. Which 3 variables have the greatest correlation with each factor in the first factorial plane?.

```
summary(res.pca,nb.dec=2,nbind=0,nbelements = 25,ncp=2)
```

```
##
## Call:
## PCA(X = df[, c(1:25, 28:30)], quanti.sup = 26, quali.sup = 27:28)
##
##
## Eigenvalues
##          Dim.1  Dim.2  Dim.3  Dim.4  Dim.5  Dim.6  Dim.7
## Variance      5.05   2.76   2.10   1.82   1.53   1.11   0.85
## % of var.     20.19  11.02   8.39   7.29   6.11   4.44   3.39
## Cumulative % of var. 20.19  31.21  39.61  46.90  53.01  57.45  60.84
##          Dim.8  Dim.9  Dim.10  Dim.11  Dim.12  Dim.13  Dim.14
## Variance      0.81   0.73   0.70   0.68   0.66   0.63   0.60
## % of var.      3.24   2.93   2.79   2.73   2.64   2.52   2.39
## Cumulative % of var. 64.08  67.01  69.80  72.53  75.17  77.68  80.07
##          Dim.15  Dim.16  Dim.17  Dim.18  Dim.19  Dim.20  Dim.21
## Variance      0.56   0.54   0.52   0.50   0.49   0.46   0.43
## % of var.      2.25   2.16   2.10   2.00   1.96   1.83   1.73
## Cumulative % of var. 82.32  84.48  86.58  88.58  90.54  92.36  94.10
##          Dim.22  Dim.23  Dim.24  Dim.25
## Variance      0.41   0.41   0.39   0.28
## % of var.      1.64   1.62   1.54   1.10
```

```

## Cumulative % of var.  95.73  97.35  98.90 100.00
##
## Variables
##          Dim.1   ctr  cos2   Dim.2   ctr  cos2
## A1          | -0.25  1.28  0.06 | -0.01  0.00  0.00 |
## A2          |  0.49  4.80  0.24 |  0.32  3.62  0.10 |
## A3          |  0.55  6.08  0.31 |  0.32  3.69  0.10 |
## A4          |  0.44  3.77  0.19 |  0.13  0.64  0.02 |
## A5          |  0.61  7.29  0.37 |  0.20  1.52  0.04 |
## C1          |  0.36  2.52  0.13 |  0.13  0.66  0.02 |
## C2          |  0.35  2.49  0.13 |  0.21  1.65  0.05 |
## C3          |  0.34  2.31  0.12 |  0.07  0.19  0.01 |
## C4          | -0.49  4.73  0.24 |  0.15  0.81  0.02 |
## C5          | -0.52  5.44  0.27 |  0.19  1.32  0.04 |
## E1          | -0.44  3.83  0.19 | -0.20  1.51  0.04 |
## E2          | -0.63  7.95  0.40 | -0.03  0.02  0.00 |
## E3          |  0.57  6.32  0.32 |  0.36  4.66  0.13 |
## E4          |  0.61  7.30  0.37 |  0.19  1.34  0.04 |
## E5          |  0.55  6.08  0.31 |  0.31  3.45  0.09 |
## N1          | -0.43  3.73  0.19 |  0.65 15.12  0.42 |
## N2          | -0.43  3.60  0.18 |  0.65 15.23  0.42 |
## N3          | -0.42  3.43  0.17 |  0.67 16.41  0.45 |
## N4          | -0.54  5.88  0.30 |  0.47  8.05  0.22 |
## N5          | -0.37  2.68  0.14 |  0.51  9.46  0.26 |
## O1          |  0.36  2.63  0.13 |  0.25  2.21  0.06 |
## O2          | -0.22  0.96  0.05 |  0.11  0.47  0.01 |
## O3          |  0.43  3.74  0.19 |  0.32  3.66  0.10 |
## O4          | -0.06  0.07  0.00 |  0.34  4.23  0.12 |
## O5          | -0.23  1.08  0.05 | -0.05  0.08  0.00 |
##
## Supplementary continuous variable
##          Dim.1   cos2   Dim.2   cos2
## age          |  0.17  0.03 | -0.04  0.00 |
##
## Supplementary categories
##          Dist   Dim.1   cos2 v.test   Dim.2   cos2 v.test
## sex.male      |  0.70 | -0.26  0.14  -4.32 | -0.39  0.31  -8.65 |
## sex.female    |  0.34 |  0.13  0.14   4.32 |  0.19  0.31   8.65 |
## HS            |  0.50 | -0.42  0.70  -2.94 | -0.04  0.01  -0.34 |
## finished HS   |  0.34 | -0.04  0.01  -0.30 |  0.14  0.17   1.55 |
## some college  |  0.39 |  0.25  0.40   5.23 |  0.03  0.01   0.87 |
## college graduate |  0.43 | -0.07  0.03  -0.67 | -0.18  0.17  -2.27 |
## graduate degree |  0.59 |  0.22  0.14   2.13 |  0.02  0.00   0.29 |
## Unknown       |  1.34 | -1.19  0.79  -8.26 | -0.05  0.00  -0.48 |

ddd<-dimdesc(res.pca,axes=1:2)
ddd$Dim.1

## $quanti
##      correlation      p.value
## E4    0.60699466 1.488184e-281
## A5    0.60660985 4.188300e-281
## E3    0.56503469 5.795883e-236
## E5    0.55420354 3.134229e-225
## A3    0.55396728 5.319307e-225

```

```

## A2 0.49222293 8.238256e-171
## A4 0.43639863 1.498031e-130
## O3 0.43464926 2.088312e-129
## O1 0.36462129 8.715772e-89
## C1 0.35665465 9.185889e-85
## C2 0.35438275 1.228486e-83
## C3 0.34119245 2.802821e-77
## age 0.16564985 1.119648e-18
## O4 -0.06102942 1.233824e-03
## O2 -0.22069147 3.125168e-32
## O5 -0.23377252 4.585558e-36
## A1 -0.25382053 2.045460e-42
## N5 -0.36763735 2.438712e-90
## N3 -0.41603183 1.225228e-117
## N2 -0.42638872 4.294669e-124
## N1 -0.43383833 7.046188e-129
## E1 -0.43994580 6.831921e-133
## C4 -0.48846284 7.392605e-168
## C5 -0.52394638 2.978100e-197
## N4 -0.54467322 4.154837e-216
## E2 -0.63361021 3.371755e-314
##
## $quali
## R2 p.value
## f.educ 0.032215754 3.187109e-18
## f.gender 0.006669853 1.512737e-05
##
## $category
## Estimate p.value
## f.educ=some college 0.4572505 1.627991e-07
## f.gender=sex.female 0.1953874 1.512737e-05
## f.educ=graduate degree 0.4259802 3.305420e-02
## f.educ=HS -0.2132007 3.286679e-03
## f.gender=sex.male -0.1953874 1.512737e-05
## f.educ=Unknown -0.9823412 9.780837e-17
##
## attr("class")
## [1] "condes" "list "

```

```
ddd$Dim.2
```

```

## $quanti
## correlation p.value
## N3 0.67238557 0.000000e+00
## N2 0.64788637 0.000000e+00
## N1 0.64539007 0.000000e+00
## N5 0.51058660 8.937682e-186
## N4 0.47084892 1.662478e-154
## E3 0.35831722 1.358530e-85
## O4 0.34157340 1.854528e-77
## A3 0.31888557 3.297121e-67
## O3 0.31769525 1.076535e-66
## A2 0.31595203 6.031240e-66
## E5 0.30817227 1.147153e-62
## O1 0.24679642 4.000931e-40

```



```
## C2  0.21336473  3.444210e-30
## A5  0.20458220  7.711571e-28
## E4  0.19221975  1.038573e-24
## C5  0.19057214  2.617404e-24
## C4  0.14916588  2.121687e-15
## C1  0.13470439  8.225381e-13
## A4  0.13300524  1.591790e-12
## O2  0.11336165  1.796037e-09
## C3  0.07183851  1.420592e-04
## age -0.04133694  2.872039e-02
## O5  -0.04681709  1.322768e-02
## E1  -0.20384340  1.202263e-27
##
## $quali
##                R2      p.value
## f.gender 0.02672902 3.151797e-18
##
## $category
##                Estimate      p.value
## f.gender=sex.female  0.2889712 3.151797e-18
## f.educ=college graduate -0.1644619 2.326792e-02
## f.gender=sex.male    -0.2889712 3.151797e-18
##
## attr(,"class")
## [1] "condes" "list "
```

It is difficult to summarize, but positive correlation to axis 1 are E4, A5 and negative correlated to E2 (Find it difficult to approach others. (q_901)) and N4 (Often feel blue. (q_1479)). It seems an axis of sociability. For axis 2, positive correlation appears for N3-Have frequent mood swings. (q_1099), N2 (Get irritated easily. (q_974)), N1 (Get angry easily. (q_952)) and inversely associated to E1 (not so intense): it seems an axis of psychological stability.

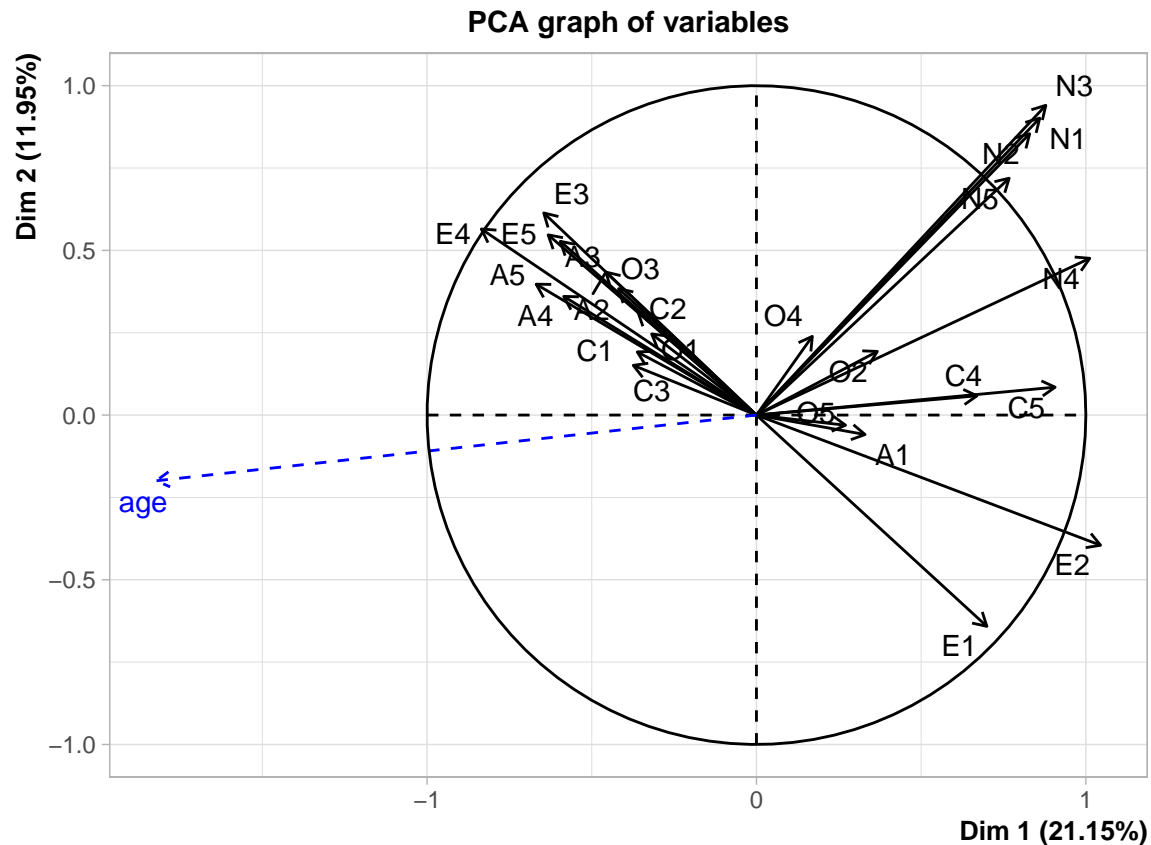
9. A Non-normalized Principal Component Analysis is addressed using as supplementary variables gender, education and age. How many axes do you have to retain according to Kaiser criteria? What's the inertia explained by retained Kaiser-based principal components?

Strictly following Kaiser criteria, we have to retain as many axes as eigenvalues greater than 1.995566 (mean eigenvalue value). 6 axes satisfy the condition and explain 58.50% of the total inertia.

```
names(df)

## [1] "A1"      "A2"      "A3"      "A4"      "A5"
## [6] "C1"      "C2"      "C3"      "C4"      "C5"
## [11] "E1"      "E2"      "E3"      "E4"      "E5"
## [16] "N1"      "N2"      "N3"      "N4"      "N5"
## [21] "O1"      "O2"      "O3"      "O4"      "O5"
## [26] "gender"  "education" "age"      "f.gender" "f.educ"

res.pcan<-PCA(df[,c(1:25,28:30)],quali.sup=27:28,quanti.sup=26,scale.unit = FALSE )
```

```
summary(res.pcan, nb.dec=2, nbind=0, ncp=2, nbelements = 25)
```

```
##
## Call:
## PCA(X = df[, c(1:25, 28:30)], scale.unit = FALSE, quanti.sup = 26,
##     quali.sup = 27:28)
##
##
## Eigenvalues
##
```

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6	Dim.7
## Variance	10.55	5.96	4.04	3.49	2.99	2.16	1.80
## % of var.	21.15	11.95	8.09	6.99	6.00	4.33	3.61
## Cumulative % of var.	21.15	33.10	41.19	48.17	54.17	58.50	62.11

```
##
```

	Dim.8	Dim.9	Dim.10	Dim.11	Dim.12	Dim.13	Dim.14
## Variance	1.75	1.50	1.43	1.34	1.22	1.17	1.14
## % of var.	3.51	3.01	2.87	2.69	2.44	2.35	2.29
## Cumulative % of var.	65.61	68.63	71.50	74.18	76.62	78.97	81.27

```
##
```

	Dim.15	Dim.16	Dim.17	Dim.18	Dim.19	Dim.20	Dim.21
## Variance	1.09	0.99	0.98	0.93	0.89	0.84	0.80
## % of var.	2.19	1.98	1.97	1.86	1.79	1.68	1.61
## Cumulative % of var.	83.46	85.44	87.41	89.26	91.06	92.74	94.35

```
##
```

	Dim.22	Dim.23	Dim.24	Dim.25
## Variance	0.76	0.74	0.69	0.63
## % of var.	1.53	1.48	1.38	1.27
## Cumulative % of var.	95.88	97.35	98.73	100.00

```
##
```

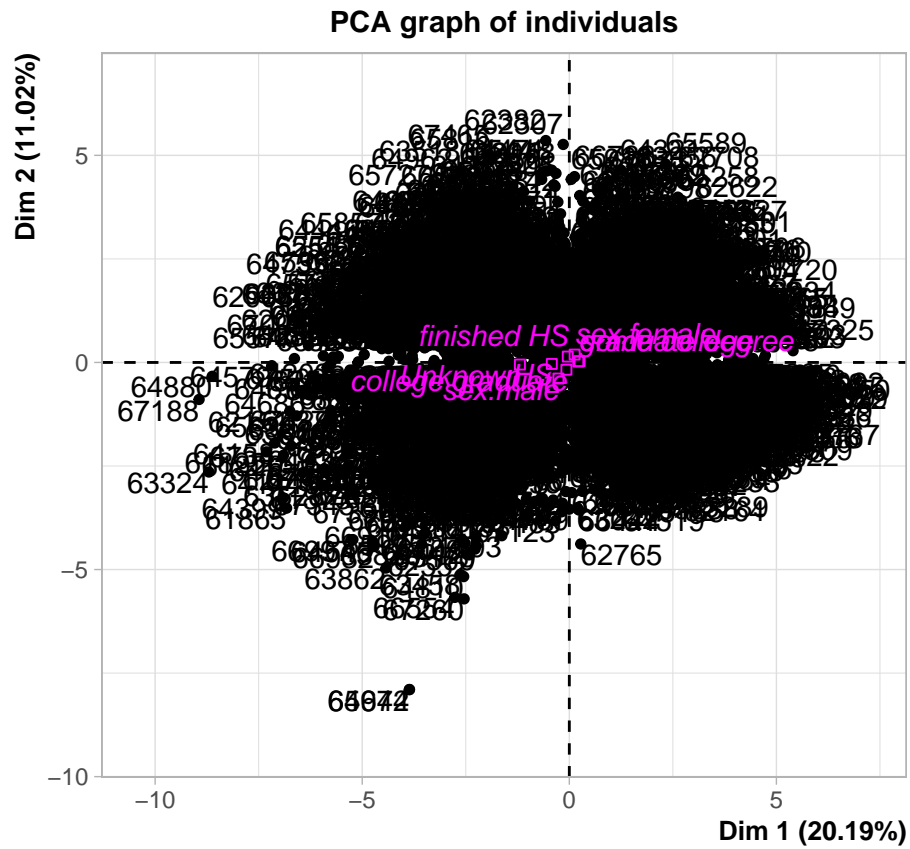
```
## Variables
##           Dim.1   ctr  cos2   Dim.2   ctr  cos2
## A1          |  0.33  1.03  0.06 | -0.06  0.06  0.00 |
## A2          | -0.46  1.98  0.15 |  0.43  3.16  0.14 |
## A3          | -0.60  3.36  0.21 |  0.53  4.67  0.17 |
## A4          | -0.59  3.24  0.16 |  0.36  2.18  0.06 |
## A5          | -0.67  4.24  0.28 |  0.40  2.65  0.10 |
## C1          | -0.36  1.24  0.09 |  0.19  0.62  0.02 |
## C2          | -0.36  1.23  0.08 |  0.31  1.64  0.06 |
## C3          | -0.37  1.33  0.08 |  0.15  0.38  0.01 |
## C4          |  0.67  4.25  0.24 |  0.06  0.06  0.00 |
## C5          |  0.91  7.79  0.31 |  0.08  0.12  0.00 |
## E1          |  0.70  4.64  0.19 | -0.64  6.90  0.16 |
## E2          |  1.04 10.34  0.43 | -0.40  2.62  0.06 |
## E3          | -0.65  3.95  0.23 |  0.61  6.32  0.21 |
## E4          | -0.84  6.61  0.33 |  0.56  5.35  0.15 |
## E5          | -0.63  3.79  0.23 |  0.55  5.02  0.17 |
## N1          |  0.86  7.01  0.30 |  0.90 13.63  0.33 |
## N2          |  0.83  6.52  0.30 |  0.85 12.24  0.31 |
## N3          |  0.88  7.31  0.30 |  0.94 14.83  0.35 |
## N4          |  1.01  9.70  0.42 |  0.48  3.80  0.09 |
## N5          |  0.77  5.58  0.23 |  0.72  8.67  0.20 |
## O1          | -0.32  0.96  0.08 |  0.25  1.02  0.05 |
## O2          |  0.37  1.28  0.06 |  0.19  0.63  0.02 |
## O3          | -0.42  1.66  0.12 |  0.38  2.47  0.10 |
## O4          |  0.17  0.27  0.02 |  0.24  0.96  0.04 |
## O5          |  0.27  0.69  0.04 | -0.03  0.02  0.00 |
##
## Supplementary continuous variable
##           Dim.1   cos2   Dim.2   cos2
## age          | -1.82  0.03 | -0.20  0.00 |
##
## Supplementary categories
##           Dist   Dim.1   cos2 v.test   Dim.2   cos2 v.test
## sex.male      |  1.01 |  0.22  0.05  2.48 | -0.72  0.51 -10.87 |
## sex.female    |  0.49 | -0.11  0.05 -2.48 |  0.35  0.51  10.87 |
## HS            |  0.70 |  0.57  0.67  2.76 | -0.17  0.06 -1.06 |
## finished HS   |  0.48 |  0.10  0.05  0.57 |  0.20  0.17  1.45 |
## some college  |  0.56 | -0.34  0.38 -5.03 |  0.16  0.08  3.08 |
## college graduate | 0.62 |  0.05  0.01  0.34 | -0.35  0.31 -3.04 |
## graduate degree | 0.81 | -0.25  0.10 -1.72 | -0.07  0.01 -0.67 |
## Unknown       |  1.86 |  1.60  0.74  7.65 | -0.23  0.01 -1.45 |

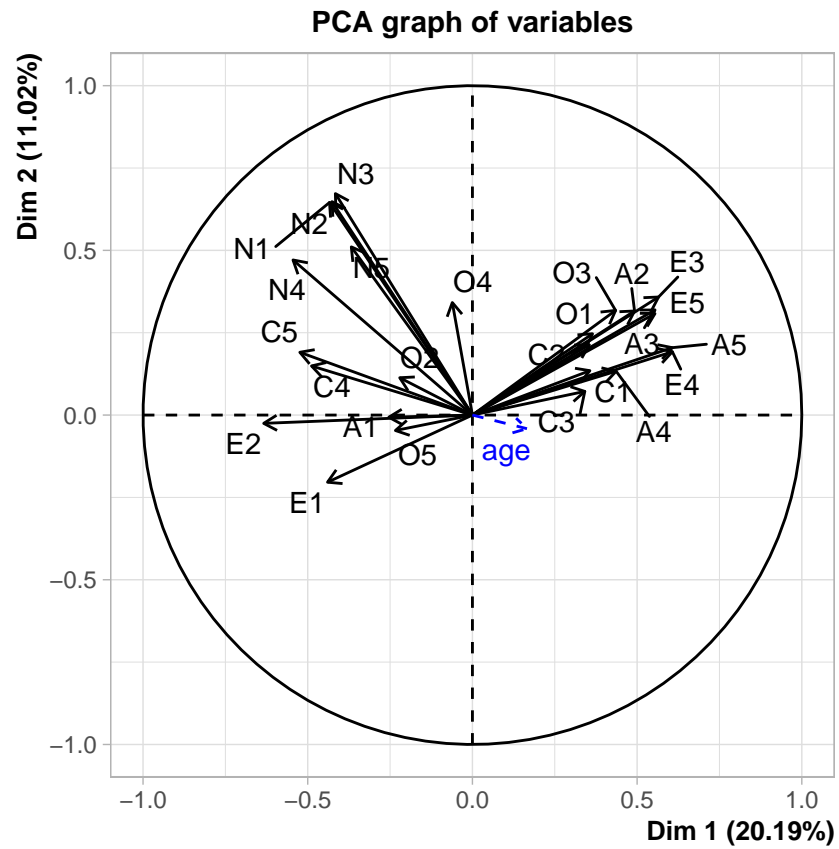
mean(res.pca$eig[,1])

## [1] 1.995566
```

10. A Hierarchical Clustering is addressed. A non-default criteria for selecting the number of clusters to 3 has to be set. Explain the characteristics of cluster number 1.

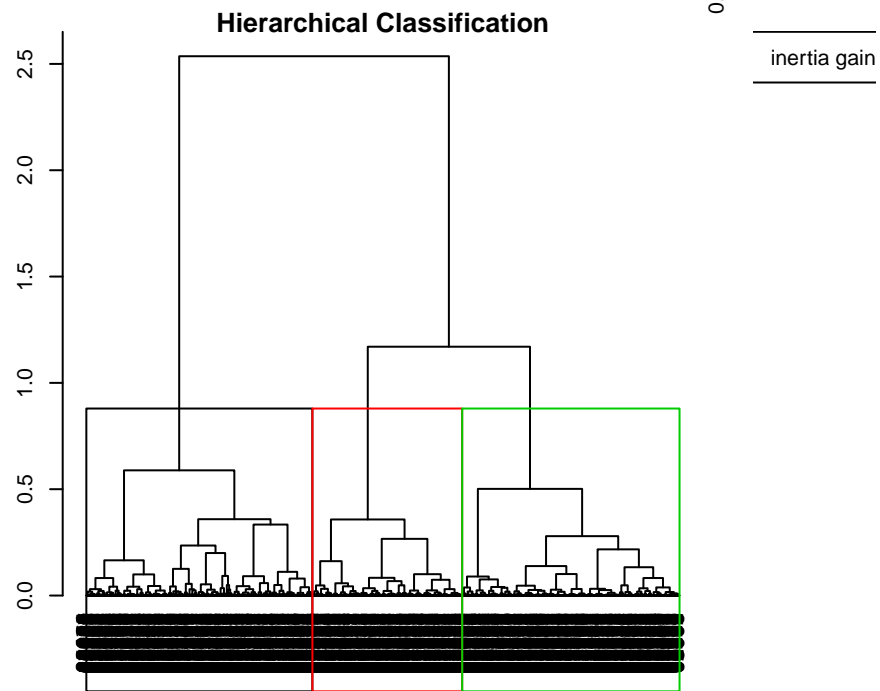
```
# 6 dimensions have to be selected according to Kaiser's criteria
res.pca<-PCA(df[,c(1:25,28:30)],quali.sup=27:28,quanti.sup=26,ncp=6)
```



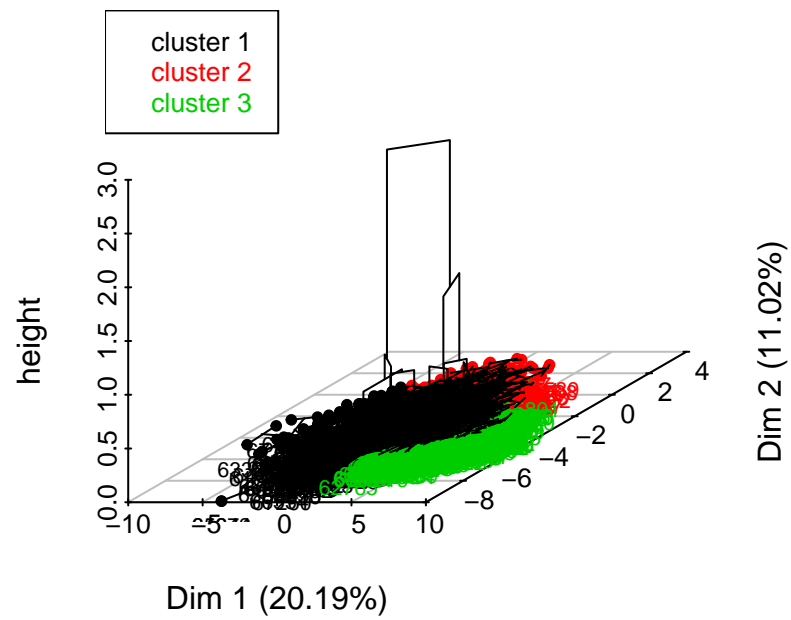


```
res.hcpc<-HCPC(res.pca,nb.clust=1,graph=T)
```

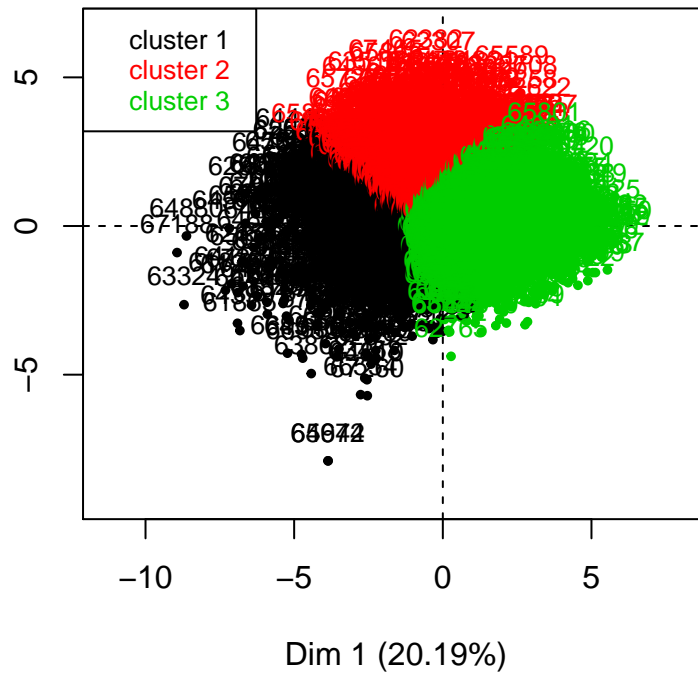
Hierarchical Clustering



Hierarchical clustering on the factor map



Factor map



```
res.hcpc$desc.var
```

```
##
## Link between the cluster variable and the categorical variables (chi-square test)
## =====
##           p.value df
## f.gender 2.303901e-13 2
## f.educ   8.067390e-12 10
##
## Description of each cluster by the categories
## =====
## $`1`
##           Cla/Mod  Mod/Cla  Global  p.value  v.test
## f.gender=sex.male  38.95539 42.61905 32.821429 9.031198e-13 7.144518
## f.educ=Unknown    47.08520 12.50000 7.964286 2.270547e-08 5.590011
## f.educ=college graduate 35.27919 16.54762 14.071429 1.481146e-02 2.436957
## f.educ=some college 25.70056 38.21429 44.607143 7.925025e-06 -4.467199
## f.gender=sex.female 25.62467 57.38095 67.178571 9.031198e-13 -7.144518
##
## $`2`
##           Cla/Mod  Mod/Cla  Global  p.value  v.test
## f.gender=sex.female 33.70548 74.32591 67.17857 6.973169e-08 5.391861
## f.gender=sex.male  23.83025 25.67409 32.82143 6.973169e-08 -5.391861
##
## $`3`
##           Cla/Mod  Mod/Cla  Global  p.value  v.test
```

```

## f.educ=some college 43.31465 48.870822 44.607143 2.469734e-04 3.665378
## f.educ=Unknown 18.38565 3.703704 7.964286 1.896052e-12 -7.041923
##
##
## Link between the cluster variable and the quantitative variables
## =====
##          Eta2          P-value
## N3  0.34876858 3.227595e-261
## N2  0.32558352 5.702340e-240
## N1  0.32520323 1.254388e-239
## N4  0.30655043 4.563696e-223
## E4  0.29104254 1.236849e-209
## E3  0.28985213 1.292062e-208
## A5  0.26969800 1.283022e-191
## A3  0.26508472 8.567006e-188
## E2  0.25215023 3.383701e-177
## E5  0.24919838 8.355890e-175
## N5  0.21034708 3.669376e-144
## A2  0.20972241 1.108828e-143
## C5  0.18862254 1.117575e-127
## E1  0.18189876 1.149895e-122
## O3  0.15326770 8.973411e-102
## C4  0.14420090 2.642392e-95
## A4  0.14261871 3.498180e-94
## O1  0.09751128 4.845401e-63
## C2  0.07599473 9.901835e-49
## C1  0.06582140 4.428313e-42
## C3  0.05940283 6.383109e-38
## A1  0.04203449 8.275538e-27
## O2  0.03405374 9.051483e-22
## O4  0.03282860 5.328050e-21
## O5  0.02832719 3.522014e-18
## age 0.01728051 2.586750e-11
##
## Description of each cluster by quantitative variables
## =====
## $`1`
##          v.test Mean in category Overall mean sd in category Overall sd
## E2  23.454477      4.226251      3.141687      1.369127      1.601558
## E1  22.454916      4.028386      2.974798      1.493125      1.625071
## C5  14.513925      3.976859      3.296112      1.472112      1.624476
## C4  13.338115      3.080402      2.553145      1.344438      1.369119
## N4  11.882418      3.720176      3.184460      1.450143      1.561502
## A1   8.424656      2.754623      2.413185      1.337596      1.403699
## O5   7.352829      2.770246      2.489364      1.329124      1.323070
## O2   3.817889      2.885714      2.713214      1.572467      1.564872
## N2   3.497848      3.661109      3.507424      1.369696      1.521751
## N5   2.890577      3.103213      2.968657      1.523331      1.612245
## N1   2.871027      3.060884      2.931033      1.477797      1.566461
## N3   2.230828      3.318970      3.215912      1.505534      1.600022
## O4  -2.004696      4.821898      4.892402      1.255727      1.218096
## age -4.476912     27.344048     28.782143     11.040055     11.125568
## C3 -11.992404      3.859507      4.304074      1.319944      1.283936
## C1 -13.265351      4.028871      4.502660      1.311911      1.237027

```

```

## C2 -14.572952      3.818411      4.370795      1.387198      1.312823
## O1 -16.366670      4.284322      4.816070      1.268400      1.125274
## A4 -19.274909      3.879313      4.700084      1.582601      1.474833
## O3 -20.699502      3.711619      4.437910      1.326620      1.215245
## A2 -24.190360      3.989161      4.804686      1.221139      1.167638
## E5 -26.345503      3.404277      4.417017      1.344742      1.331387
## A5 -26.419397      3.603063      4.560564      1.213978      1.255248
## A3 -27.161159      3.587863      4.604769      1.311775      1.296718
## E4 -27.885006      3.249062      4.421009      1.414524      1.455627
## E3 -28.348570      2.896792      4.000051      1.217736      1.347906
##           p.value
## E2 1.189928e-121
## E1 1.145743e-111
## C5 9.889614e-48
## C4 1.389284e-40
## N4 1.460803e-32
## A1 3.618111e-17
## O5 1.940555e-13
## O2 1.345987e-04
## N2 4.690281e-04
## N5 3.845359e-03
## N1 4.091402e-03
## N3 2.569250e-02
## O4 4.499553e-02
## age 7.573054e-06
## C3 3.894368e-33
## C1 3.676956e-40
## C2 4.174562e-48
## O1 3.308245e-60
## A4 8.725377e-83
## O3 3.499698e-95
## A2 2.810242e-129
## E5 5.778956e-153
## A5 8.203233e-154
## A3 1.869332e-162
## E4 4.055636e-171
## E3 8.716657e-177
##
## $`2`
##           v.test Mean in category Overall mean sd in category Overall sd
## N3 27.120140      4.455065      3.215912      1.2166416      1.600022
## N1 25.823267      4.086181      2.931033      1.3490254      1.566461
## N2 25.507693      4.615888      3.507424      1.1050891      1.521751
## N5 20.459869      3.910635      2.968657      1.5470715      1.612245
## N4 18.980980      4.030846      3.184460      1.3136976      1.561502
## A3 9.912763      4.971837      4.604769      1.0059997      1.296718
## E3 9.790369      4.376899      4.000051      1.1085967      1.347906
## C5 9.764682      3.749092      3.296112      1.5662436      1.624476
## E5 9.749020      4.787674      4.417017      1.0688579      1.331387
## O4 9.316749      5.216483      4.892402      0.9983749      1.218096
## A2 9.259691      5.113439      4.804686      0.9109201      1.167638
## O3 8.315184      4.726475      4.437910      1.0212739      1.215245
## C4 7.760448      2.856558      2.553145      1.4081918      1.369119
## E4 6.596893      4.695227      4.421009      1.2477534      1.455627

```

```

## O2 6.443819 3.001172 2.713214 1.6488295 1.564872
## C2 5.790455 4.587878 4.370795 1.2424899 1.312823
## O1 5.063380 4.978777 4.816070 1.0451340 1.125274
## A5 4.649632 4.727233 4.560564 1.1168769 1.255248
## A4 3.611554 4.852189 4.700084 1.4083383 1.474833
## C1 3.157163 4.614188 4.502660 1.1402270 1.237027
## A1 2.506996 2.513677 2.413185 1.4883777 1.403699
## age -2.856649 27.874560 28.782143 10.2486998 11.125568
## E1 -7.732733 2.615949 2.974798 1.5132280 1.625071
## p.value
## N3 5.699381e-162
## N1 4.859224e-147
## N2 1.619557e-143
## N5 4.907036e-93
## N4 2.449831e-80
## A3 3.663709e-23
## E3 1.238428e-22
## C5 1.596113e-22
## E5 1.862582e-22
## O4 1.199595e-20
## A2 2.050248e-20
## O3 9.160984e-17
## C4 8.463015e-15
## E4 4.198643e-11
## O2 1.165044e-10
## C2 7.019587e-09
## O1 4.118871e-07
## A5 3.325275e-06
## A4 3.043672e-04
## C1 1.593121e-03
## A1 1.217621e-02
## age 4.281387e-03
## E1 1.052622e-14
##
## $`3`
## v.test Mean in category Overall mean sd in category Overall sd
## A5 20.385174 5.158696 4.560564 0.9116163 1.255248
## E4 19.925777 5.098991 4.421009 1.0399977 1.455627
## E3 17.354055 4.546833 4.000051 1.0914402 1.347906
## A3 16.125912 5.093560 4.604769 1.0198486 1.296718
## E5 15.515562 4.899881 4.417017 1.0553382 1.331387
## A4 14.666053 5.205686 4.700084 1.1349854 1.474833
## A2 13.956242 5.185602 4.804686 0.9769061 1.167638
## O3 11.573480 4.766672 4.437910 1.0118126 1.215245
## O1 10.573555 5.094190 4.816070 0.9099437 1.125274
## C3 10.368259 4.615247 4.304074 1.1303353 1.283936
## C1 9.461211 4.776237 4.502660 1.1436575 1.237027
## C2 8.207915 4.622674 4.370795 1.1740210 1.312823
## age 6.885214 30.572719 28.782143 11.5865580 11.125568
## O4 -6.891459 4.696181 4.892402 1.2904182 1.218096
## O5 -8.153247 2.237209 2.489364 1.2040214 1.323070
## O2 -9.644343 2.360434 2.713214 1.4184190 1.564872
## A1 -10.256173 2.076663 2.413185 1.3075449 1.403699
## E1 -13.767075 2.451840 2.974798 1.4172126 1.625071

```

```
## C4 -19.806800      1.919262      2.553145      1.0761019      1.369119
## N5 -21.969333      2.140713      2.968657      1.2595977      1.612245
## C5 -22.795557      2.430513      3.296112      1.3715291      1.624476
## E2 -23.000762      2.280618      3.141687      1.2882043      1.601558
## N1 -26.999893      1.942401      2.931033      1.0624035      1.566461
## N2 -27.290326      2.536678      3.507424      1.2592351      1.521751
## N3 -27.620675      2.182881      3.215912      1.1646548      1.600022
## N4 -29.004964      2.125771      3.184460      1.1624081      1.561502
##      p.value
## A5  2.264107e-92
## E4  2.432423e-88
## E3  1.837998e-67
## A3  1.677651e-58
## E5  2.722392e-54
## A4  1.063539e-48
## A2  2.882363e-44
## O3  5.615780e-31
## O1  3.952219e-26
## C3  3.457680e-25
## C1  3.043967e-21
## C2  2.250616e-16
## age 5.770091e-12
## O4  5.522286e-12
## O5  3.542807e-16
## O2  5.194253e-22
## A1  1.110203e-24
## E1  4.021685e-43
## C4  2.600975e-87
## N5  5.659355e-107
## C5  5.074829e-115
## E2  4.580121e-117
## N1  1.482190e-160
## N2  5.525391e-164
## N3  6.282566e-168
## N4  5.696362e-185
```

```
(res.hcpc$call$t$within[1]-res.hcpc$call$t$within[3])/res.hcpc$call$t$within[1]
```

```
## [1] 0.2580477
```

Three clusters are selected which it is enough to represent the complexity of this dataset (it explains less than 26% of total inertia in data). Cluster 1 contains 39% of the male observations in the sample. On average 32.82% of the data units belong to male gender, but in Cluster 1 males are overrepresented (42.6%). 'Unknown' is also overrepresented being 12.5% in Cluster 1 and 8% globally and 'some graduate' educated people represents 44.61% of the sample, but only 38.21% of them are included in Cluster 1.

Cluster 1 shows mean values of E2, E1, C5 and C4 remarkably over the global mean, while the global mean in the sample, while E3, E4, A3 are clearly under the global mean in this cluster. It indicates difficult approach and not very social and communicative behavior.

Do not forget to Knit to .pdf before posting your answers in ATENEA.