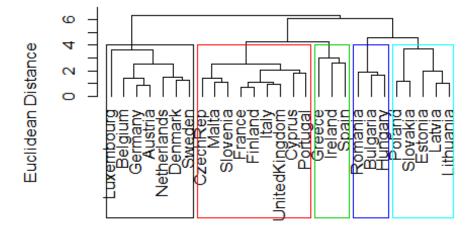
```
### Exercise 1

remove(list = ls())

euro = read.table(file = "Europe.txt", header = T, dec =".")
attach(euro)
europe = data.frame(euro)
europe = na.omit(europe)
Country = ï..Country
numbers = cbind(CPI, UNE, INP, BOP, PRC, UN)

# a)

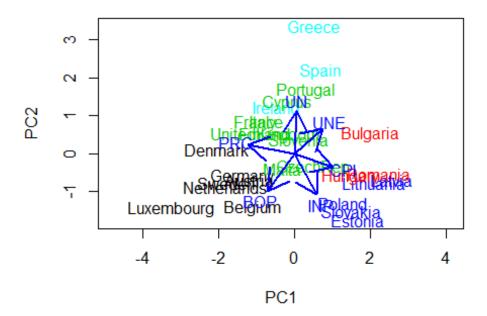
Euro = scale(numbers, center = TRUE, scale = TRUE)
EuroEuc = dist(Euro, method = "euclidean", diag = TRUE, upper = TRUE)
reshclust = hclust(EuroEuc, method = "complete")
plot(reshclust, hang = -1, labels = Country, ylab = "Euclidean Distance")
rect.hclust(reshclust, k = 5, border = c(1,2,3,4,5))
```



EuroEuc hclust (*, "complete")

```
# b)
S1 = cov(Euro)
lam1 = eigen(S1)$values[1:2]
e1 = eigen(S1)$vectors[,1:2]
Yhat = Euro%*%e1
plot(-Yhat[,1],Yhat[,2], type = "n", asp = 1, ylab = "PC2", xlab = "PC1")
text(-Yhat[,1],Yhat[,2], Country, col = cutree(reshclust, k = 5))
```

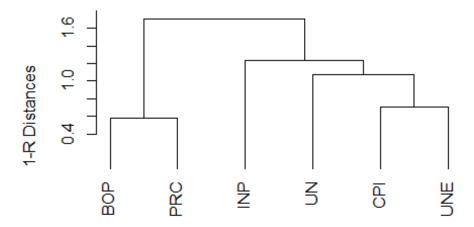
```
arrows(0,0,-2*e1[1,1],2*e1[1,2], col = "blue", lwd = 2)
text(-2.5*e1[1,1],2.5*e1[1,2], labels = variable.names(europe[2]), col = "blu
e")
arrows(0,0,-2*e1[2,1],2*e1[2,2], col = "blue", lwd = 2)
text(-2.5*e1[2,1],2.5*e1[2,2], labels = variable.names(europe[3]), col = "blu
e")
arrows(0,0,-2*e1[3,1],2*e1[3,2], col = "blue", lwd = 2)
text(-2.5*e1[3,1],2.5*e1[3,2], labels = variable.names(europe[4]), col = "blu
e")
arrows(0,0,-2*e1[4,1],2*e1[4,2], col = "blue", lwd = 2)
text(-2.5*e1[4,1],2.5*e1[4,2], labels = variable.names(europe[5]), col = "blu
e")
arrows(0,0,-2*e1[5,1],2*e1[5,2], col = "blue", lwd = 2)
text(-2.5*e1[5,1],2.5*e1[5,2], labels = variable.names(europe[6]), col = "blu
e")
arrows(0,0,-2*e1[6,1],2*e1[6,2], col = "blue", lwd = 2)
text(-2.5*e1[6,1],2.5*e1[6,2], labels = variable.names(europe[7]), col = "blu
e")
arrows(0,0,-2*e1[6,1],2*e1[6,2], labels = variable.names(europe[7]), col = "blu
e")
```



```
# c)
# d)
R1 = cor(Euro)
round(R1, digits = 2)
```

```
## CPI UNE INP BOP PRC UN
## CPI 1.00 0.29 0.21 -0.11 -0.71 -0.08
## UNE 0.29 1.00 -0.10 -0.40 -0.33 0.03
## INP 0.21 -0.10 1.00 0.04 -0.51 -0.24
## BOP -0.11 -0.40 0.04 1.00 0.42 -0.31
## PRC -0.71 -0.33 -0.51 0.42 1.00 -0.02
## UN -0.08 0.03 -0.24 -0.31 -0.02 1.00

EuroDist = as.dist(1-R1)
EuroClus = hclust(EuroDist, method = "complete")
plot(EuroClus, hang = -1, labels = variable.names(europe[2:7]), ylab = "1-R D istances")
```



EuroDist hclust (*, "complete")

```
### Exercis 2
remove(list = ls())
pers = read.table(file = "Personality2019.txt", header = T, dec =".")
Pers = as.matrix(pers[,2:33])
pers$Name
## [1] Toni
                  Lion
                            Borui
                                      Sarah
                                                Abbas
                                                          Marc
                                                                    Dmitrii
## [8] Ivan
                  Benjamin
                                      Samuel
                                                Theresa
                                                          Rickmer
                                                                    Florentia
                            Julius
                  Frederik Carmen
## [15] Anton
                                      Louis
                                                Felipe
                                                          Stepan
                                                                    Nikolai
## [22] Tom
                  Tobias
                            Xiao
                                                Jungin
                                      Hendrik
## 26 Levels: Abbas Anton Benjamin Borui Carmen Dmitrii Felipe ... Xiao
# a)
X = scale(Pers, center = TRUE, scale = FALSE)
S = cov(X)
lam = eigen(S)$values[1:2]
e = eigen(S)$vectors[,1:2]
Yhat = X%*%e
plot(Yhat[,1],Yhat[,2], type = "n", asp = 1)
text(Yhat[,1],Yhat[,2], pers$Name)
Expl = cumsum(eigen(S)$values)/sum(eigen(S)$values)
Expl[2]
## [1] 0.3017799
# The first two principal components explain 30.18% of the total variance.
# b)
lengthOfAxes = sqrt(e[,1]^2+e[,2]^2)
o = order(lengthOfAxes, decreasing = TRUE)
arrows(0,0,3*e[o[1],1],3*e[o[1],2], col = "blue")
text(3.5*e[o[1],1],3.5*e[o[1],2], labels = variable.names(pers[o[1]+1]), col
= "blue")
arrows(0,0,3*e[o[2],1],3*e[o[2],2], col = "blue")
text(3.5*e[o[2],1],3.5*e[o[2],2], labels = variable.names(pers[o[2]+1]), col
arrows(0,0,3*e[o[3],1],3*e[o[3],2], col = "blue")
text(3.5*e[o[3],1],3.5*e[o[3],2], labels = variable.names(pers[o[3]+1]), col
= "blue")
arrows(0,0,3*e[o[4],1],3*e[o[4],2], col = "blue")
text(3.5*e[o[4],1],3.5*e[o[4],2], labels = variable.names(pers[o[4]+1]), col
= "blue")
# c)
Chol = c(0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0)
San = c(0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0)
```

Mel = c(1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0)phle = c(rep(1,32))Phle = phle-(Chol+San+Mel) Choleric = Chol-colMeans(Pers) C = Choleric%*%e text(C[1,1],C[1,2], labels = "Cholerica", col = "green") Melancholic = Mel-colMeans(Pers) M = Melancholic%*%e

```
Sanguine = San-colMeans(Pers)
Sa = Sanguine%*%e
text(Sa[1,1],Sa[1,2], labels = "Sanguine", col = "green")
Phlegmatic = Phle-colMeans(Pers)
P = Phlegmatic%*%e
```

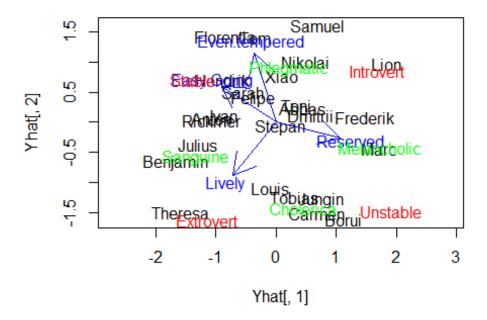
text(P[1,1],P[1,2], labels = "Phlegmatic", col = "green")

text(M[1,1],M[1,2], labels = "Melancholic", col = "green")

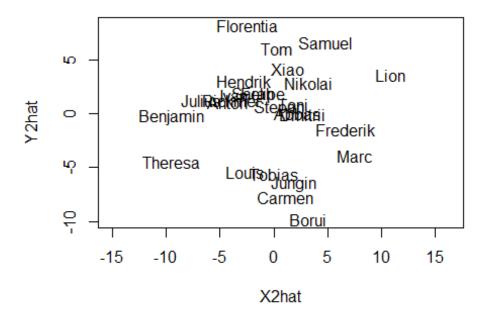
d)

```
Intro = Phle+Mel
Unst = Mel+Chol
Extro = Chol+San
Stabel = San+Phle
Introvert = Intro-colMeans(Pers)
Int = Introvert%*%e
text(Int[1,1],Int[1,2], labels = "Introvert", col = "red")
Unstable = Unst-colMeans(Pers)
Uns = Unstable%*%e
text(Uns[1,1],Uns[1,2], labels = "Unstable", col = "red")
Extrovert = Extro-colMeans(Pers)
Ex = Extrovert%*%e
text(Ex[1,1],Ex[1,2], labels = "Extrovert", col = "red")
Stabele = Stabel-colMeans(Pers)
St = Stabele%*%e
```

text(St[1,1],St[1,2], labels = "Stable", col = "red")

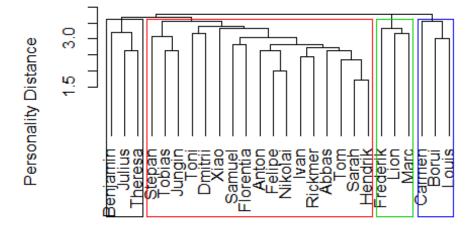


```
# e)
Delta = as.matrix(dist(Pers, method = "manhattan", upper = TRUE, diag = TRUE)
Deltastar = Delta^2
I = diag(1, nrow = 26, ncol = 26)
J = matrix(c(rep(1, 26*26)), nrow = 26, ncol = 26)
n = 1/26
H = I - n*J
B = -0.5*H%*%Deltastar%*%H
reseigen = eigen(B)
lambda = round(reseigen$values, digits = 4)
lambda
## [1] 580.1522 507.7782 333.5213 296.7573 194.7462 123.1985 103.0154
## [8] 85.5694 76.6256 52.2152 45.5587 34.9044 26.5002
                   0.0000 -7.3348 -11.6107 -14.1745 -20.3273 -32.1523
## [15]
          5.2770
## [22] -35.9884 -50.6483 -54.9865 -74.0686 -86.2971
# Since there are negative eigenvalues, B is not n.n.d.
# f)
QE = reseigen$vectors
Qlam = reseigen$values
QLam = diag(Qlam[1:2])
QYhat = QE[,1:2]%*%sqrt(QLam)
plot(-QYhat[,1],QYhat[,2], asp = 1, type = "n", ylab = "Y2hat", xlab = "X2hat
")
text(-QYhat[,1],QYhat[,2], pers$Name)
```



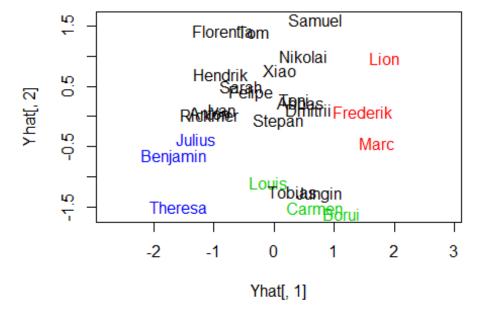
```
# The biplot differs from the MDS-plot.

# g)
Personality = dist(X, method = "euclidean", upper = TRUE, diag = TRUE)
PersALM = hclust(Personality, method = "average")
plot(PersALM, hang = -1, labels = pers$Name, ylab = "Personality Distance")
rect.hclust(PersALM, k = 4, border = c(1,2,3,4))
```



Personality hclust (*, "average")

```
plot(Yhat[,1],Yhat[,2], type = "n", asp = 1)
text(Yhat[,1],Yhat[,2], pers$Name, col = cutree(PersALM, k = 4))
```



The cluster does correspond more or less to the biplot except the observations Tobias and Jungin.

```
### Exercise 3

remove(list = ls())

pers3 = read.table(file = "Personality.txt", header = T, dec =".")

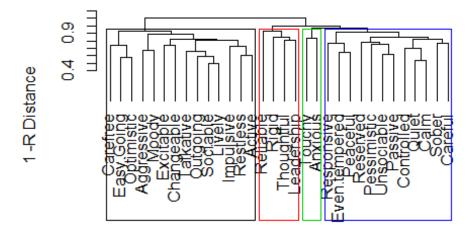
# a)

Pers3 = scale(pers3, scale = FALSE, center = TRUE)

R = cor(pers3)

PersDelt = as.dist(1-R)

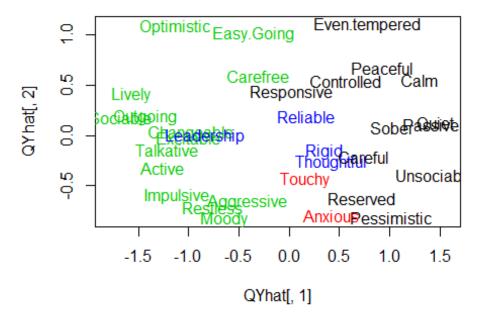
Persreshc = hclust(PersDelt, method = "average")
plot(Persreshc, hang = -1, labels = pers3$Name, ylab = "1 -R Distance")
rect.hclust(Persreshc, k = 4, border = c(1,2,3,4))
```



PersDelt hclust (*, "average")

```
Group = sort(cutree(Persreshc, k = 4))
Group[1:12]
                                   Unsociable Even.tempered
##
           Quiet
                          Calm
                                                                  Reserved
##
               1
                             1
                                            1
                                                          1
                                                                         1
##
     Pessimistic
                    Responsive
                                   Controlled
                                                      Sober
                                                                  Peaceful
##
                                            1
                                                          1
                                                                         1
         Careful
##
                       Passive
##
                             1
# Cluster 1
Group[13:14]
```

```
Touchy Anxious
##
         2
# Cluster 2
Group[15:28]
##
     Sociable
                Restless
                            Outgoing Aggressive Talkative Excitable
##
## Changeable Easy.Going
                                         Lively Optimistic
                           Impulsive
                                                              Carefree
##
                                              3
                                                          3
##
        Moody
                  Active
##
                        3
# Cluster 3
Group[29:32]
                   Rigid Thoughtful Leadership
##
     Reliable
##
            4
# Cluster 4
# b)
X3 = R
Q = X3\%*\%t(X3)
QE = eigen(Q)$vectors
Qlam = eigen(Q)$values
QLam = diag(Qlam[1:2])
QYhat = QE[,1:2]%*%sqrt(QLam)
plot(QYhat[,1],QYhat[,2], asp = 1, type = "n")
text(QYhat[,1],QYhat[,2], variable.names(pers3), col = cutree(Persreshc, k =
4))
```



```
# c)
Chol = c(0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0)
San = c(0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,0,1,0,0,1,0,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,1,0,
```

