

TD4

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Exercice 1

```
X0<-read.table(text="
math scie fran lati d-m
jean 6.0 6.0 5.0 5.5 8.0
aline 8.0 8.0 8.0 8.0 9.0
annie 6.0 7.0 11.0 9.5 11.0
monique 14.5 14.5 15.5 15.0 8.0
didier 14.0 14.0 12.0 12.5 10.0
andr e 11.0 10.0 5.5 7.0 13.0
pierre 5.5 7.0 14.0 11.5 10.0
brigitte 13.0 12.5 8.5 9.5 12.0
evelyne 9.0 9.5 12.5 12.0 18.0
")
```

```
X<- scale(X0,center=TRUE,scale = FALSE)
X
```

```
##           math      scie      fran      lati d.m
## jean      -3.666667 -3.833333 -5.222222 -4.555556 -3
## aline     -1.666667 -1.833333 -2.222222 -2.055556 -2
## annie      -3.666667 -2.833333  0.777778 -0.555556  0
## monique    4.833333  4.666667  5.277778  4.944444 -3
## didier     4.333333  4.166667  1.777778  2.444444 -1
## andr e     1.333333  0.166667 -4.722222 -3.055556  2
## pierre    -4.166667 -2.833333  3.777778  1.444444 -1
## brigitte   3.333333  2.666667 -1.722222 -0.555556  1
## evelyne   -0.666667 -0.333333  2.277778  1.944444  7
## attr(,"scaled:center")
##      math      scie      fran      lati      d.m
##  9.666667  9.833333 10.222222 10.055556 11.000000
```

```
n<-nrow(X)
p<-ncol(X)
S<-var(X)*(n-1)/n

U<-eigen(S)$vectors ; Lambda<-eigen(S)$values ; C = X%*%U

Lambda
```

```
## [1] 28.253249801 12.074723274  8.615733579  0.021732182  0.009869805
```

Significativit  des diff rentes valeurs propres :

```
cumsum(Lambda/sum(Lambda))
```

```
## [1] 0.5768876 0.8234348 0.9993547 0.9997985 1.0000000
```

On remarque qu'on a clairement 3 valeurs pr pond rantes qui correspondent   99,9% de la somme totale.

Matrice de rotation

```
rot_map <- sqrt(Lambda) * U
rot_map
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] -2.734950189  3.0135494 -0.272801593 -1.534807732  3.043314229
## [2,] -1.761719352  1.2926370 -0.050222155  1.921799481 -1.898505282
## [3,] -1.445188052 -1.9089548  0.317200445  1.155714453  1.202815235
## [4,] -0.071443132 -0.0476513  0.003323298 -0.099388800 -0.066844869
## [5,] -0.003042775 -0.0112156 -0.098597506  0.003421169  0.001258566
```

Tout ce qui est corrélation

On calcule maintenant la corrélation. C'est-à-dire la contribution relative des axes aux individus :

```
COR<- C^2 / rowSums(X^2)
COR
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## jean      0.88894548 0.0340427538 0.076587880 2.622775e-04 1.616070e-04
## aline     0.80365559 0.0260088320 0.169811511 4.260371e-04 9.802799e-05
## annie     0.46012197 0.5344933508 0.004038159 1.329907e-03 1.660880e-05
## monique   0.89414954 0.0004372797 0.105035734 2.827126e-04 9.472899e-05
## didier    0.87726184 0.1020725256 0.019794006 1.079034e-04 7.637245e-04
## andr e    0.23623016 0.5777346580 0.185435015 1.439185e-05 5.857764e-04
## pierre    0.02583884 0.9071461473 0.066661161 3.109728e-04 4.287757e-05
## brigitte  0.17435766 0.7357964730 0.087211720 2.629342e-03 4.805383e-06
## evelyne   0.05330060 0.1977673369 0.748407627 4.645728e-04 5.986834e-05
```

Et maintenant, la contribution relative des individus aux axes :

```
CTR<- 1/n* C^2 / matrix(eigen(S)$values,n,p,byrow = TRUE)
CTR
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## jean      0.297726930 0.0266783459 0.084116112 0.114200796 0.154939841
## aline     0.061005911 0.0046197026 0.042271263 0.042045012 0.021301584
## annie     0.040507433 0.1101020178 0.001165793 0.152211729 0.004185620
## monique   0.374291227 0.0004283019 0.144182553 0.153854355 0.113512117
## didier    0.159647400 0.0434643006 0.011812517 0.025528947 0.397858978
## andr e    0.034783719 0.1990491457 0.089538242 0.002755005 0.246906420
## pierre    0.004343771 0.3568307951 0.036748775 0.067964365 0.020634020
## brigitte  0.015425990 0.1523214081 0.025302476 0.302429457 0.001217026
## evelyne   0.012267619 0.1065059822 0.564862269 0.139010335 0.039444393
```

Correction du prof

```

mon_ACP <- function(X){
  n = nrow(X)
  X <- scale(X, scale = FALSE) # On peut mettre TRUE aussi paraît-il
  S <- cov(X) # On peut aussi faire t(X) %*% X / n
  eigen_res <- eigen(S)
  U <- eigen_res$vectors
  Lambda <- eigen_res$values
  C <- X %*% U

  rot_map <- sqrt(Lambda) * U
}

```

Exercice 2

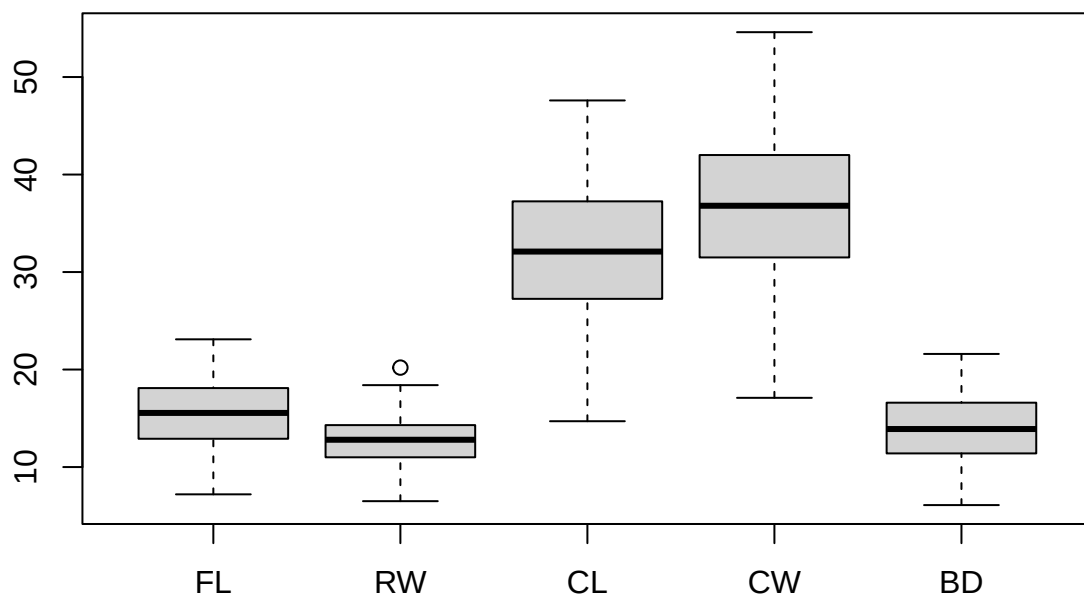
```
library(MASS)
```

```
##  
## Attachement du package : 'MASS'  
  
## L'objet suivant est masqué depuis 'package:dplyr':  
##  
##      select
```

```
data(crabs)  
crabsquant<-crabs[,4:8]
```

```
# ACP0 <- prcomp(crabs)  
# summary(ACP)  
# plot(ACP0$sdev)
```

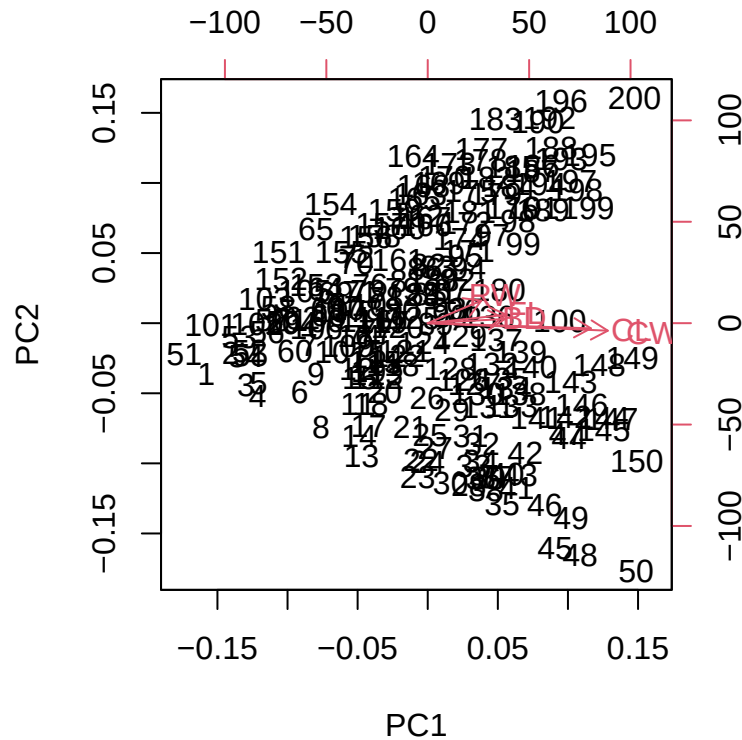
```
boxplot(crabsquant)
```



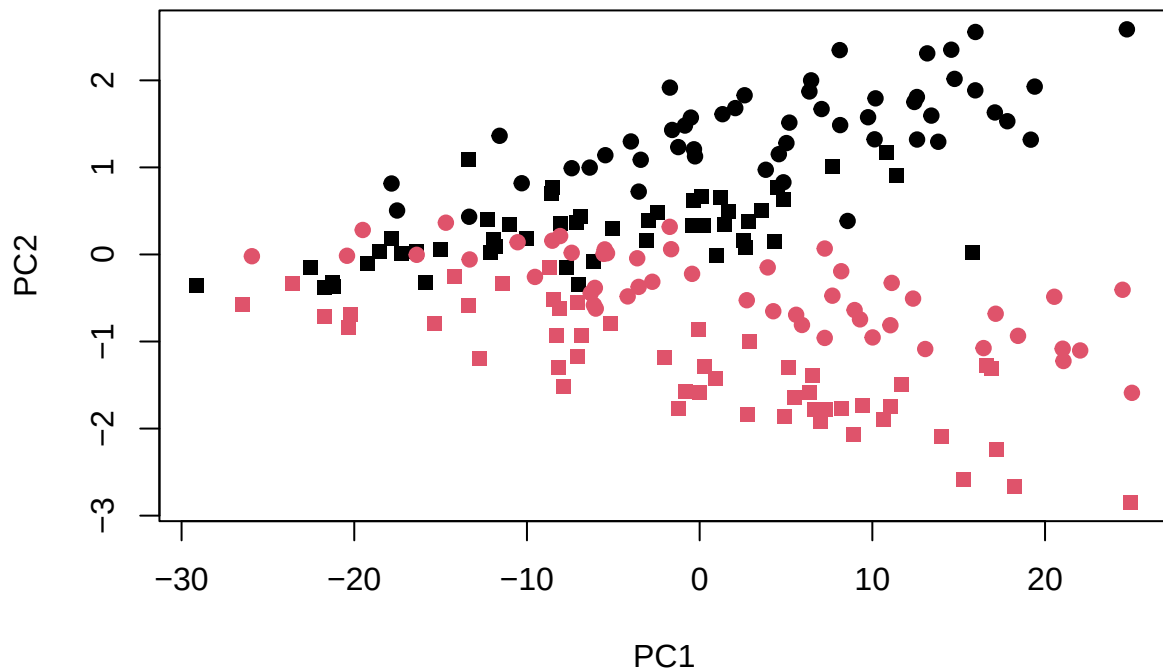
```
res <- prcomp(crabsquant)
res
```

```
## Standard deviations (1, ..., p=5):
## [1] 11.8619441  1.1387874  1.0001346  0.3678306  0.2791312
##
## Rotation (n x k) = (5 x 5):
##      PC1      PC2      PC3      PC4      PC5
## FL 0.2889810  0.3232500 -0.5071698  0.7342907  0.1248816
## RW 0.1972824  0.8647159  0.4141356 -0.1483092 -0.1408623
## CL 0.5993986 -0.1982263 -0.1753299 -0.1435941 -0.7416656
## CW 0.6616550 -0.2879790  0.4913755  0.1256282  0.4712202
## BD 0.2837317  0.1598447 -0.5468821 -0.6343657  0.4386868
```

```
biplot(res)
```



```
plot(res$x, col=crabs$sex, pch = c(15,19)[crabs$sp])
```



On remarque qu'il y a qu'une seule variable vraiment significative.

On va reconstruire notre X en enlevant la principale composante 1 (trop significative) :

```
u1 <- prcomp(crabsquant)$rotation[,1, drop=FALSE]
newX <- as.matrix(crabsquant) %*% u1 %*% t(u1)
head(newX)
```

```
##          FL          RW          CL          CW          BD
## 1  8.054021  5.498343  16.70549  18.44060  7.907722
## 2  8.892886  6.071021  18.44545  20.36128  8.731350
## 3  9.418411  6.429789  19.53548  21.56453  9.247329
## 4  9.822893  6.705922  20.37445  22.49064  9.644463
## 5  9.860819  6.731813  20.45312  22.57747  9.681700
## 6 11.269428  7.693446  23.37483  25.80265 11.064723
```

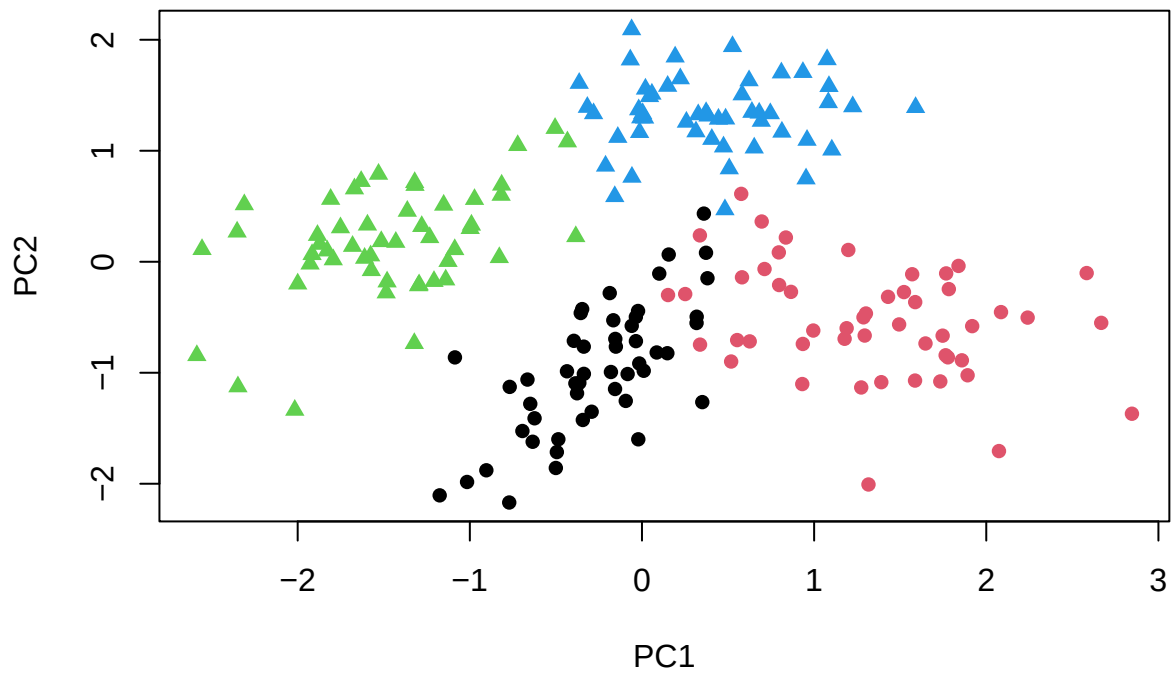
```
# Soustraction correcte pour obtenir les données ajustées
adjustedData <- crabsquant - newX

# Application de l'ACP sur les données ajustées
res2 <- prcomp(adjustedData)
res2
```

```
## Standard deviations (1, ..., p=5):
```

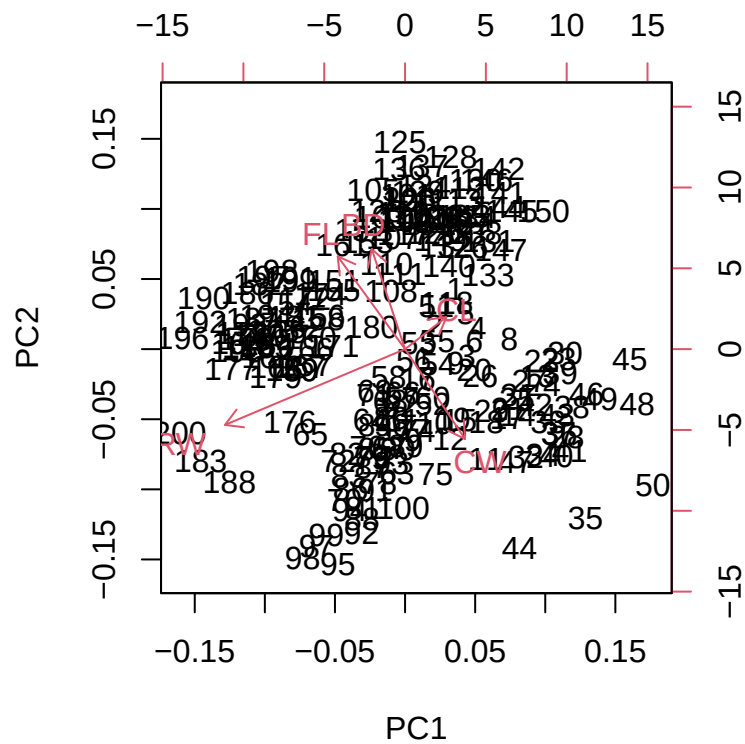
```
## [1] 1.138787e+00 1.000135e+00 3.678306e-01 2.791312e-01 4.960538e-15
##
## Rotation (n x k) = (5 x 5):
##      PC1      PC2      PC3      PC4      PC5
## FL -0.3232500  0.5071698  0.7342907 -0.1248816  0.2889810
## RW -0.8647159 -0.4141356 -0.1483092  0.1408623  0.1972824
## CL  0.1982263  0.1753299 -0.1435941  0.7416656  0.5993986
## CW  0.2879790 -0.4913755  0.1256282 -0.4712202  0.6616550
## BD -0.1598447  0.5468821 -0.6343657 -0.4386868  0.2837317
```

```
plot(res2$x, col=interaction(crabs$sex,crabs$sp), pch = c(16,17)[crabs$sp])
```



4 axes à garder j'ai l'impression

```
biplot(res2)
```

Effectivement, sans la première CP trop significative, cette nouvelle ACP est beaucoup plus intéressante !

Exercice 3

```
data <- read.table("neighbor_globin.txt", header=FALSE)
X <- as.matrix(data[, -1])
head(X)
```

```
##          V2      V3      V4      V5      V6      V7      V8      V9      V10     V11
## [1,] 0.0000 0.1806 0.2434 0.3964 0.5656 0.4987 1.9654 2.1040 2.1278 2.0965
## [2,] 0.1806 0.0000 0.1929 0.2997 0.4852 0.4271 1.9675 2.0689 2.2427 2.1483
## [3,] 0.2434 0.1929 0.0000 0.3432 0.5312 0.4635 1.8727 2.1478 2.1478 2.1092
## [4,] 0.3964 0.2997 0.3432 0.0000 0.3657 0.3196 1.8520 2.0577 2.0649 1.8216
## [5,] 0.5656 0.4852 0.5312 0.3657 0.0000 0.2970 1.8912 2.0551 2.0572 1.7896
## [6,] 0.4987 0.4271 0.4635 0.3196 0.2970 0.0000 1.7142 1.9036 1.9751 1.6927
##          V12     V13     V14     V15     V16     V17     V18     V19     V20     V21
## [1,] 2.2725 2.0807 1.9645 1.9928 1.9195 2.0944 1.9867 1.9486 1.8515 1.9880
## [2,] 2.2753 2.0387 2.0941 2.1273 1.9495 2.0628 2.1114 1.9951 1.9200 2.0044
## [3,] 2.2318 1.9386 2.0581 2.0567 1.9920 2.1235 2.1776 2.0310 1.9519 2.0735
## [4,] 1.9345 2.0096 1.9935 2.0463 1.8520 1.9878 2.1320 1.9407 1.8823 2.0378
## [5,] 1.9478 1.9237 1.7647 1.9622 1.9429 1.9423 2.0500 1.9352 1.9823 2.0511
## [6,] 1.8907 1.8523 1.8770 1.8414 1.7849 1.8503 1.9604 1.9075 1.8643 1.7584
##          V22
## [1,] 2.6100
## [2,] 2.5663
## [3,] 2.6225
## [4,] 2.5424
## [5,] 2.3154
## [6,] 2.4536
```

```
n <- nrow(X)

sum.of.squares<- rowSums(X^2)
D2 <- cbind(sum.of.squares) %*% matrix(1,1,n) +
matrix(1,n,1) %*% rbind(sum.of.squares) -
2 * X%*%t(X)

print(D2,digits=2)
```

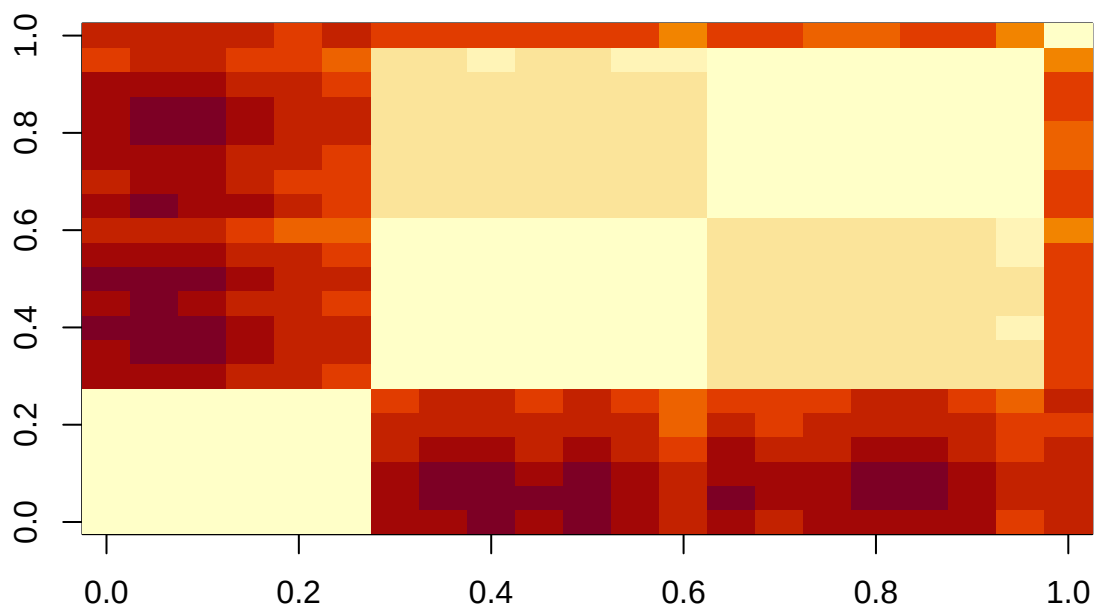
```
##          [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]      [,9]     [,10]
## [1,] 0.00 1.7e-01 2.4e-01 0.67 1.3e+00 1.36 41.47 4.3e+01 44.37 41.99
## [2,] 0.17 2.8e-14 1.5e-01 0.58 1.3e+00 1.44 43.72 4.6e+01 46.83 44.16
## [3,] 0.24 1.5e-01 -2.8e-14 0.58 1.2e+00 1.41 42.44 4.4e+01 45.47 42.89
## [4,] 0.67 5.8e-01 5.8e-01 0.00 5.1e-01 0.54 38.88 4.1e+01 42.01 38.82
## [5,] 1.27 1.3e+00 1.2e+00 0.51 -1.4e-14 0.46 36.34 3.8e+01 39.31 36.07
## [6,] 1.36 1.4e+00 1.4e+00 0.54 4.6e-01 0.00 33.86 3.6e+01 37.06 33.91
## [7,] 41.47 4.4e+01 4.2e+01 38.88 3.6e+01 33.86 0.00 4.2e-01 0.59 0.77
## [8,] 43.44 4.6e+01 4.4e+01 41.00 3.8e+01 36.07 0.42 -1.4e-14 0.38 0.85
## [9,] 44.37 4.7e+01 4.5e+01 42.01 3.9e+01 37.06 0.59 3.8e-01 0.00 0.92
## [10,] 41.99 4.4e+01 4.3e+01 38.82 3.6e+01 33.91 0.77 8.5e-01 0.92 0.00
## [11,] 45.71 4.8e+01 4.7e+01 42.55 4.0e+01 37.58 1.07 9.7e-01 0.84 0.28
## [12,] 41.66 4.4e+01 4.2e+01 39.07 3.6e+01 34.16 0.84 1.0e+00 0.94 0.49
## [13,] 36.38 3.9e+01 3.8e+01 34.53 3.1e+01 30.02 1.42 1.3e+00 1.57 1.91
## [14,] 41.69 4.4e+01 4.4e+01 40.82 3.8e+01 35.68 9.37 1.0e+01 9.18 9.99
```

```

## [15,] 38.90 4.1e+01 4.1e+01 37.82 3.6e+01 32.98 8.76 9.5e+00 8.89 9.44
## [16,] 39.96 4.2e+01 4.2e+01 38.88 3.7e+01 34.05 9.80 1.0e+01 9.56 10.01
## [17,] 43.11 4.6e+01 4.6e+01 42.45 4.0e+01 37.41 10.03 1.1e+01 9.64 10.47
## [18,] 42.17 4.5e+01 4.4e+01 41.23 3.9e+01 36.27 9.71 1.1e+01 9.66 10.32
## [19,] 40.78 4.3e+01 4.3e+01 39.87 3.8e+01 35.03 9.35 1.0e+01 9.40 9.91
## [20,] 35.84 3.8e+01 3.8e+01 35.18 3.3e+01 30.32 8.23 8.6e+00 7.72 8.83
## [21,] 36.70 3.8e+01 3.7e+01 37.05 3.4e+01 35.96 35.06 3.2e+01 32.73 33.66
##      [,11] [,12] [,13] [,14] [,15] [,16] [,17] [,18] [,19]
## [1,] 4.6e+01 41.66 3.6e+01 4.2e+01 38.90 4.0e+01 4.3e+01 4.2e+01 4.1e+01
## [2,] 4.8e+01 43.80 3.9e+01 4.4e+01 41.39 4.2e+01 4.6e+01 4.5e+01 4.3e+01
## [3,] 4.7e+01 42.48 3.8e+01 4.4e+01 41.02 4.2e+01 4.6e+01 4.4e+01 4.3e+01
## [4,] 4.3e+01 39.07 3.5e+01 4.1e+01 37.82 3.9e+01 4.2e+01 4.1e+01 4.0e+01
## [5,] 4.0e+01 36.34 3.1e+01 3.8e+01 35.72 3.7e+01 4.0e+01 3.9e+01 3.8e+01
## [6,] 3.8e+01 34.16 3.0e+01 3.6e+01 32.98 3.4e+01 3.7e+01 3.6e+01 3.5e+01
## [7,] 1.1e+00 0.84 1.4e+00 9.4e+00 8.76 9.8e+00 1.0e+01 9.7e+00 9.4e+00
## [8,] 9.7e-01 1.01 1.3e+00 1.0e+01 9.52 1.0e+01 1.1e+01 1.1e+01 1.0e+01
## [9,] 8.4e-01 0.94 1.6e+00 9.2e+00 8.89 9.6e+00 9.6e+00 9.7e+00 9.4e+00
## [10,] 2.8e-01 0.49 1.9e+00 1.0e+01 9.44 1.0e+01 1.0e+01 1.0e+01 9.9e+00
## [11,] 1.4e-14 0.41 2.5e+00 9.7e+00 9.37 9.7e+00 9.8e+00 9.7e+00 9.5e+00
## [12,] 4.1e-01 0.00 2.1e+00 8.7e+00 8.36 8.9e+00 9.0e+00 8.8e+00 8.6e+00
## [13,] 2.5e+00 2.10 1.4e-14 8.7e+00 8.28 8.8e+00 9.3e+00 9.6e+00 9.2e+00
## [14,] 9.7e+00 8.75 8.7e+00 1.4e-14 0.32 4.1e-01 7.6e-01 6.7e-01 6.7e-01
## [15,] 9.4e+00 8.36 8.3e+00 3.2e-01 0.00 4.7e-01 1.1e+00 8.4e-01 8.0e-01
## [16,] 9.7e+00 8.94 8.8e+00 4.1e-01 0.47 -1.4e-14 1.1e+00 1.0e+00 9.4e-01
## [17,] 9.8e+00 9.00 9.3e+00 7.6e-01 1.09 1.1e+00 1.4e-14 1.9e-01 3.4e-01
## [18,] 9.7e+00 8.82 9.6e+00 6.7e-01 0.84 1.0e+00 1.9e-01 -1.4e-14 1.1e-01
## [19,] 9.5e+00 8.56 9.2e+00 6.7e-01 0.80 9.4e-01 3.4e-01 1.1e-01 -1.4e-14
## [20,] 8.6e+00 7.54 7.2e+00 1.6e+00 1.66 1.7e+00 1.8e+00 1.9e+00 2.0e+00
## [21,] 3.3e+01 32.05 2.7e+01 3.3e+01 32.73 3.2e+01 3.1e+01 3.3e+01 3.4e+01
##      [,20] [,21]
## [1,] 3.6e+01 37
## [2,] 3.8e+01 38
## [3,] 3.8e+01 37
## [4,] 3.5e+01 37
## [5,] 3.3e+01 34
## [6,] 3.0e+01 36
## [7,] 8.2e+00 35
## [8,] 8.6e+00 32
## [9,] 7.7e+00 33
## [10,] 8.8e+00 34
## [11,] 8.6e+00 33
## [12,] 7.5e+00 32
## [13,] 7.2e+00 27
## [14,] 1.6e+00 33
## [15,] 1.7e+00 33
## [16,] 1.7e+00 32
## [17,] 1.8e+00 31
## [18,] 1.9e+00 33
## [19,] 2.0e+00 34
## [20,] 1.4e-14 28
## [21,] 2.8e+01 0

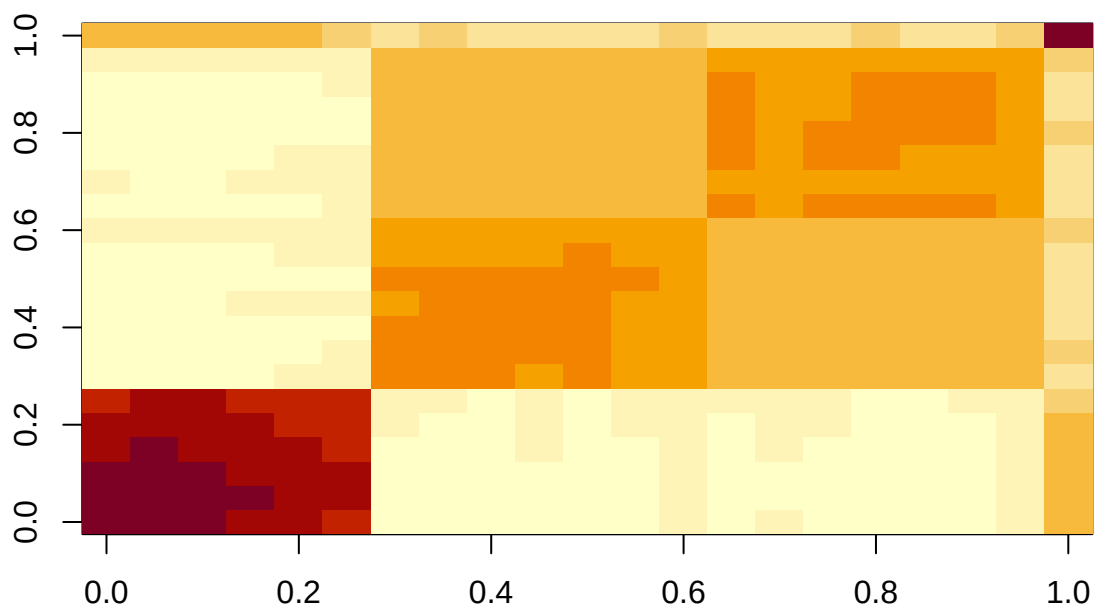
```

image(D2)



```
J <- diag(n) - matrix(1/n, n, n)
```

```
B <- -(1/2) * J %*% D2 %*% J  
image(B)
```



B correspond à la matrice XX^T lorsque X est centrée.

```
B_diag <- eigen(B)
U <- B_diag$vectors
Lambda <- diag(B_diag$values)
```

```
barplot(B_diag$values)
```

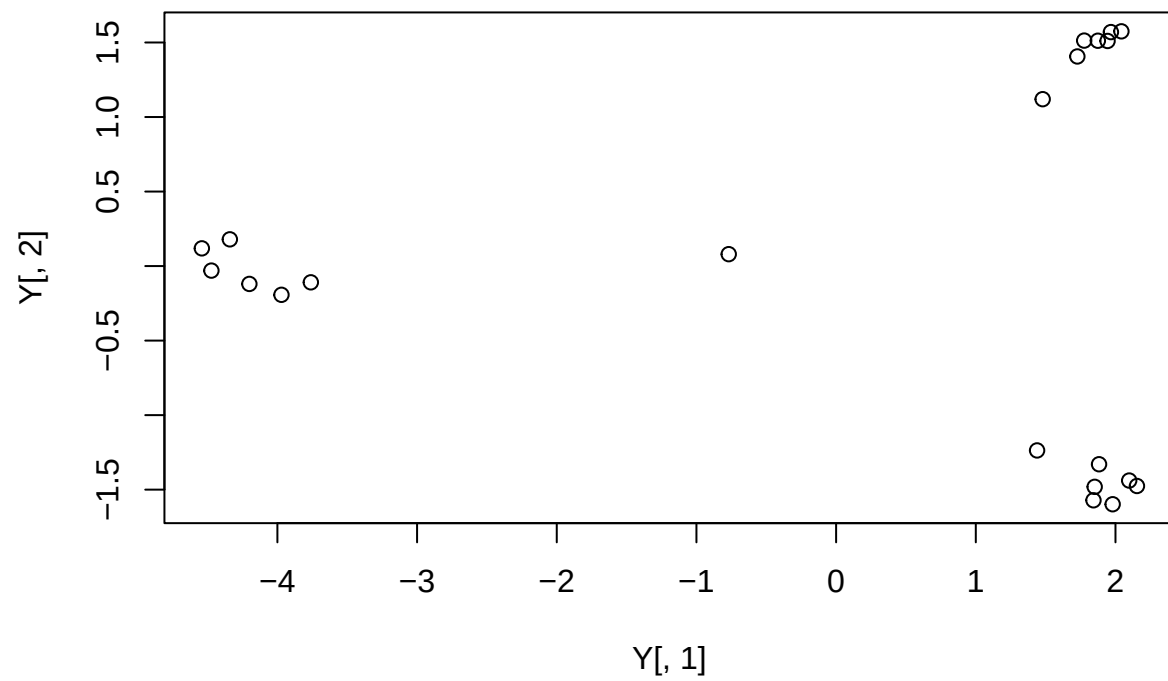


3 valeurs propres significatives.

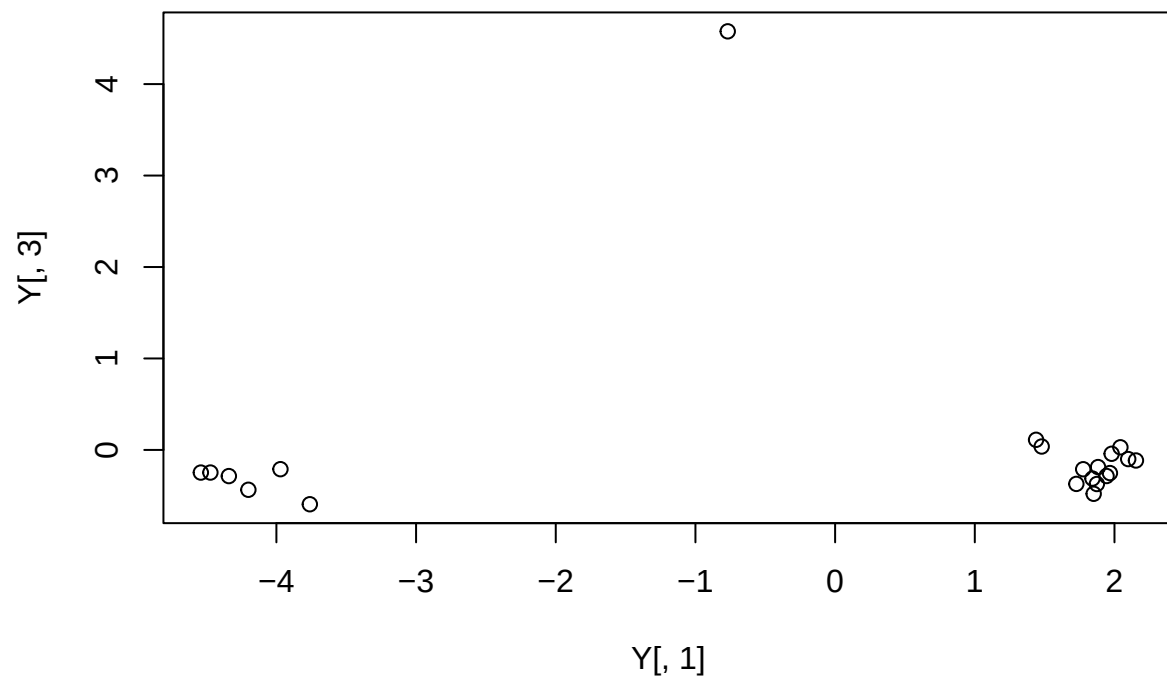
```
U_3 <- U[, 1:3] # les trois premiers vecteurs propres
Lambda_3 <- Lambda[1:3, 1:3] # les trois premières valeurs propres

Y <- U_3 %*% sqrt(Lambda_3)

plot(Y[,1], Y[,2])
```



```
plot(Y[,1], Y[,3])
```



```
plot(Y[,2], Y[,3])
```