

Assignment 3: Principal Components

Mathematics for Big Data

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The goal of these exercises is to build a function to compute principal components analysis, this will be done step by step. At the end, `PCA()` function in `FactoMineR` library is explored, and the results compared.

```
library(FactoMineR)
library(fBasics)
```

1 Preliminaries

1. Use the data `decathlon` in package `FactoMineR`:

```
data(decathlon, package="FactoMineR")
data<-decathlon
```

2. Transform the variables involving race times, so that: $t \rightarrow \max - t$:

```
time.cols<- data[c(1,5,6,10)]
max.cols<-apply(time.cols,2,max)
transf<- sweep(time.cols,2,max.cols,FUN="-")
transf<- -transf
data[c(1,5,6,10)]<-transf
```

3. Separate basic data from supplementary data: variables 11, 12 and 13 are considered supplementary, numerical or qualitative, and cases 39 to 41 are considered as supplementary too:

```
dat<-data[-(39:41),-(11:13)] # basic data subset
ind.supp<- data[39:41,-(11:13)] # supplementary individuals
var.supp<- data[-(39:41),11:13] # supplementary numerical variables
qual.supp<- as.factor(data[-(39:41),13]) # supplementary categorical variable
```

We have our data set, now we can begin performing PCA.

2 Steps in the function

We want to create a new function `pc()` to get principal components and all the relevant matrices involved in the interpretation.

2.1 Normalizing

4. Normalize (scale) the data, i.e., subtract the mean and divide by the standard deviation by columns.

```
n<-nrow(dat)
p<-ncol(dat)

centr<-apply(dat,2,mean) # global center: mean vector of the basic subset
s<-apply(dat,2,sd)      # global st.dev: st. dev of the basic subset
```

```

dat.cent<-sweep(dat,2,centr,FUN="-")
dat.std <-sweep(dat.cent,2,s,FUN="/")

# ind.sup (if not null) must be scaled with respect to global center and st.dev
if (!is.null(ind.sup)){
ind.sup.cent<-sweep(ind.sup,2,centr,FUN="-")
ind.sup.std <-sweep(ind.sup.cent,2,s,FUN="/")}

# var.sup subset (if not null) must be scaled with respect its own center
if (!is.null(var.sup)){
var.sup.cent<-sweep(var.sup,2,apply(var.sup,2,mean),FUN="-")
var.sup.std <-sweep(var.sup.cent,2,apply(var.sup,2,sd),FUN="/")}

# Warning in treating the quali.sup variable !!
# 1st: get the column means for the factor groups
if (!is.null(qual.sup)){
dat.qual.sup<-lapply(split(data.frame(dat),qual.sup),colMeans) # its a list

# 2nd: convert to a data frame
dat.qual.sup<-data.frame(dat.qual.sup)

# 3rd: transpose to trate groups as individual cases
dat.qual.sup<-t(dat.qual.sup)

# 4th: center and scale using descriptives of basic data
qual.sup.cent<-sweep(dat.qual.sup,2,centr,FUN="-")
qual.sup.std<-sweep(qual.sup.cent,2,s,FUN="/")}

```

5. Getting the data subsets to whom principal components apply (factor $1/\sqrt{n-1}$ applies to obtain the covariance).

```

corr<-TRUE # standardized data
if (corr==T){ X<-dat.std*(1/sqrt(n-1))
if (!is.null(ind.sup)){X.ind.sup<-ind.sup.std*(1/sqrt(n-1))}
if (!is.null(var.sup)){X.var.sup<-var.sup.std*(1/sqrt(n-1))}
if (!is.null(qual.sup)){X.qual.sup<-qual.sup.std*(1/sqrt(n-1))}
Sigma<-t(as.matrix(X))%*%as.matrix(X)
Sinv<-diag(rep(1,p))} # "Sinv" is the identity if corr=TRUE

#corr<-FALSE
if (corr==F) {X<-dat.cent*(1/sqrt(n-1))
if (!is.null(ind.sup)){X.ind.sup<-ind.sup.cent*(1/sqrt(n-1))}
if (!is.null(var.sup)){X.var.sup<-var.sup.cent*(1/sqrt(n-1))}
if (!is.null(qual.sup)){X.qual.sup<-qual.sup.cent*(1/sqrt(n-1))}
Sigma<-t(as.matrix(X))%*%as.matrix(X)
Sinv<-diag((diag(Sigma))^(-(1/2)))}

```

6. Get U , V and D components in **SVD** of X , you can use the function `svd`.

```

sing <- svd(X)
U <- sing$u
V <- sing$v
D <- diag(sing$d)
Lambda <- D^(2)

```

7. Compute relevant matrices for the variables (column analysis).

```
# by columns: contribution of the variables into the components
var.contr<- 100*(V^2)
colSums(var.contr)
```

```
## [1] 100 100 100 100 100 100 100 100 100 100
```

```
rownames(var.contr)<-colnames(dat)
colnames(var.contr)<-paste("PC",1:ncol(var.contr),sep="")
var.contr
```

##	PC1	PC2	PC3	PC4	PC5
## 100m	17.3330712	2.327978543	3.657786	4.6373102	2.5356924
## Long.jump	14.6035731	5.779536395	4.659357	3.9880665	15.1137201
## Shot.put	13.8795326	2.110718700	17.772208	0.7399725	8.9088142
## High.jump	9.0986575	7.947955252	6.673909	7.5017364	37.8197139
## 400m	11.9861005	0.008841597	22.983984	1.0275684	8.7196020
## 110m.hurdle	15.3539305	1.630494985	4.694060	7.0498291	0.8119038
## Discus	14.5512385	0.697690566	7.452333	10.4141557	21.2259895
## Pole.vault	0.1212022	37.834947785	5.018283	12.7029744	2.1740865
## Javeline	2.6725347	8.056501774	12.305120	43.3314060	0.3200159
## 1500m	0.4001593	33.605334402	14.782960	8.6069807	2.3704617
##	PC6	PC7	PC8	PC9	PC10
## 100m	9.7878495	13.3851967	30.30745229	0.69018396	15.33747911
## Long.jump	3.3626976	27.2061645	0.81134529	20.14745141	4.32808790
## Shot.put	3.8387178	11.8963502	3.25089122	8.41964555	29.18314919
## High.jump	14.8280692	4.1101241	1.98876490	6.54803169	3.48303840
## 400m	12.0506530	0.7632945	0.96701357	38.16487809	3.32806387
## 110m.hurdle	18.9882180	8.3381721	42.03210924	0.06961239	1.03166970
## Discus	0.1102243	19.2631398	0.07420697	1.16043349	25.05058851
## Pole.vault	16.5223122	6.0402679	9.83802134	0.04840533	9.69949916
## Javeline	13.6036131	5.1786736	4.86298766	9.64042837	0.02871931
## 1500m	6.9076453	3.8186165	5.86720753	15.11092972	8.52970485

```
# correlations between variables and components
C <- Sinv%*%V%*%D
rownames(C)<-colnames(dat)
colnames(C)<-paste("PC",1:ncol(var.contr),sep="")
```

```
# coordinates of the initial variables
var.coord <- V%*%D
rownames(var.coord)<-colnames(dat)
colnames(var.coord)<-paste("PC",1:ncol(var.contr),sep="")
```

```
# squared cosines
var.cos2 <- C^2
rownames(var.cos2)<-colnames(dat)
colnames(var.cos2)<-paste("PC",1:ncol(var.contr),sep="")
```

8. Compute the relevant matrices for the individuals, as in 7.

```
#Scores
Y <- as.matrix(X)%*%V*sqrt(n)
colnames(Y)<-paste("PC",1:ncol(var.contr),sep="")
```

```
#Contribution of the individuals into the components
```

```

sqrt.scores <- Y^2
sum <- colSums(sqrt.scores)
ind.contr <- sweep(sqrt.scores, 2, sum,FUN="/")
ind.contr <- 100*ind.contr
colSums(ind.contr)

```

```

## PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9 PC10
## 100 100 100 100 100 100 100 100 100 100

```

```

#Individuals inertias
ind.iner <- rowSums(sqrt.scores)
ind.dist <- sqrt(ind.iner)

```

```

#Individuals squared cosines
ind.cos2 <- sweep(sqrt.scores, 2, ind.iner,FUN="/")

```

```

## Warning in sweep(sqrt.scores, 2, ind.iner, FUN = "/"): STATS is longer than
## the extent of 'dim(x)[MARGIN]'

```

9. Make a list with all the results above computed, using a reduced number of digits d , for instance $d = 3$. Add to the list result the coordinates for the supplementary objects:

```

# assume k=3
var.supp.coord <- as.matrix(t(X.var.supp))%*%U
qual.supp.coord <- as.matrix(X.qual.supp)%*%V*sqrt(n)
ind.supp.coord <- as.matrix(X.ind.supp)%*%V*sqrt(n)

result<-list(left.sing.vec = round(U, digits = 3),
             right.sing.vec = round(V, digits = 3),
             diag.sing.vec = round(D, digits = 3),
             lambda. = round(Lambda, digits = 3),
             diag.sing = round(D, digits = 3),
             var.contr = round(var.contr, digits = 3),
             corr = round(C, digits = 3),
             var.coord = round(var.coord, digits = 3),
             var.cos2 = round(var.cos2, digits = 3),
             ind.coord = round(Y, digits = 3),
             ind.contr = round(ind.contr, digits = 3),
             ind.iner= round(ind.iner, digits = 3),
             ind.dist = round(ind.dist, digits = 3),
             ind.cos2 = round(ind.cos2, digits = 3),
             qual.supp.coord = round(qual.supp.coord, digits = 3),
             ind.supp.coord = round(ind.supp.coord, digits = 3))

```

The function pc()

11. Write the complete code for the function.

```

pc<-function(dat,ind.supp=NULL,var.supp=NULL,qual.supp=NULL,corr=TRUE,k=3) #default
{
  data(decathlon, package="FactoMineR")
  data<-decathlon

  time.cols<- data[c(1,5,6,10)]

```

```

max.cols<-apply(time.cols,2,max)
transf<- sweep(time.cols,2,max.cols,FUN="-")
transf<- -transf
data[c(1,5,6,10)]<-transf

n<-nrow(dat)
p<-ncol(dat)

centr<-apply(dat,2,mean)  # global center: mean vector of the basic subset
s<-apply(dat,2,sd)        # global st.dev: st. dev of the basic subset

dat.centr<-sweep(dat,2,centr,FUN="-")
dat.std <-sweep(dat.centr,2,s,FUN="/")

# ind.supp (if not null) must be scaled with respect to global center and st.dev
if (!is.null(ind.supp)){
  ind.supp.centr<-sweep(ind.supp,2,centr,FUN="-")
  ind.supp.std <-sweep(ind.supp.centr,2,s,FUN="/")}

# var.supp subset (if not null) must be scaled with respect its own center
if (!is.null(var.supp)){
  var.supp.centr<-sweep(var.supp,2,apply(var.supp,2,mean),FUN="-")
  var.supp.std <-sweep(var.supp.centr,2,apply(var.supp,2,sd),FUN="/")}

# Warning in treating the quali.supp variable !!
# 1st: get the column means for the factor groups
if (!is.null(qual.supp)){
  dat.qual.supp<-lapply(split(data.frame(dat),qual.supp),colMeans) # its a list

# 2nd: convert to a data frame
dat.qual.supp<-data.frame(dat.qual.supp)

# 3rd: transpose to trate groups as individual cases
dat.qual.supp<-t(dat.qual.supp)

# 4th: center and scale using descriptives of basic data
qual.supp.centr<-sweep(dat.qual.supp,2,centr,FUN="-")
qual.supp.std<-sweep(qual.supp.centr,2,s,FUN="/")}

#corr<-TRUE
if (corr==T){ X<-dat.std*(1/sqrt(n-1))
  if (!is.null(ind.supp)){X.ind.supp<-ind.supp.std*(1/sqrt(n-1))}
  if (!is.null(var.supp)){X.var.supp<-var.supp.std*(1/sqrt(n-1))}
  if (!is.null(qual.supp)){X.qual.supp<-qual.supp.std*(1/sqrt(n-1))}
  Sigma<-t(as.matrix(X))%*%as.matrix(X)
  Sinv<-diag(rep(1,p))} # "Sinv" is the identity if corr=TRUE

#corr<-FALSE
if (corr==F) {X<-dat.centr*(1/sqrt(n-1))
  if (!is.null(ind.supp)){X.ind.supp<-ind.supp.centr*(1/sqrt(n-1))}
  if (!is.null(var.supp)){X.var.supp<-var.supp.centr*(1/sqrt(n-1))}
  if (!is.null(qual.supp)){X.qual.supp<-qual.supp.centr*(1/sqrt(n-1))}
  Sigma<-t(as.matrix(X))%*%as.matrix(X)

```

```

Sinv<-diag((diag(Sigma))^(-(1/2)))}

#SVD
sing <- svd(X)
U <- sing$u
V <- sing$v
D <- diag(sing$d)
Lambda <- D^(2)

# by columns: contribution of the variables into the components
var.contr<- 100*(V^2)
colSums(var.contr)
rownames(var.contr)<-colnames(dat)
colnames(var.contr)<-paste("PC",1:ncol(var.contr),sep="")

# correlations between variables and components
C <- Sinv%*%V%*%D
rownames(C)<-colnames(dat)
colnames(C)<-paste("PC",1:ncol(var.contr),sep="")

# coordinates of the initial variables
var.coord <- V%*%D
rownames(var.coord)<-colnames(dat)
colnames(var.coord)<-paste("PC",1:ncol(var.contr),sep="")

# squared cosines
var.cos2 <- C^2
rownames(var.cos2)<-colnames(dat)
colnames(var.cos2)<-paste("PC",1:ncol(var.contr),sep="")

var.suppl.coord <- as.matrix(t(X.var.suppl))%*%U
qual.suppl.coord <- as.matrix(X.qual.suppl)%*%V*sqrt(n)
ind.suppl.coord <- as.matrix(X.ind.suppl)%*%V*sqrt(n)

result<-list(left.sing.vec = round(U, digits = k),
             right.sing.vec = round(V, digits = k),
             diag.sing.vec = round(D, digits = k),
             lambda.sing.vec = round(Lambda, digits = k),
             var.contr = round(var.contr, digits = k),
             corr = round(C, digits = k),
             var.coord = round(var.coord, digits = k),
             var.cos2 = round(var.cos2, digits = k),
             ind.coord = round(Y, digits = k),
             ind.contr = round(ind.contr, digits = k),
             ind.iner = round(ind.iner, digits = k),
             ind.dist = round(ind.dist, digits = k),
             ind.cos2 = round(ind.cos2, digits = k),
             qual.suppl.coord = round(qual.suppl.coord, digits = k),
             ind.suppl.coord = round(ind.suppl.coord, digits = k),
             var.suppl.coord = round(var.suppl.coord, digits = k))
}

```

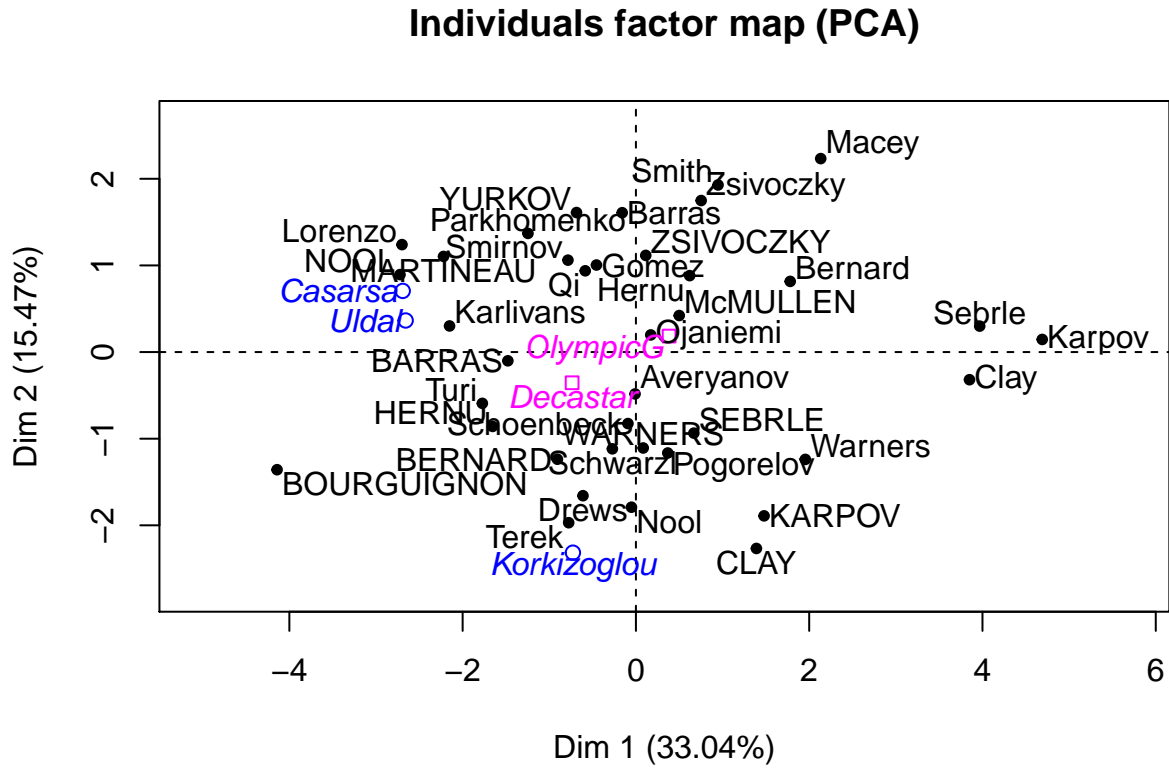
12. Apply function `pc()` in the following two cases and explore within objects `out` and `out2`.

```
out<-pc(dat, ind.supp=NULL, var.supp=NULL, qual.supp=NULL, corr=TRUE)
out2<-pc(dat, ind.supp=ind.supp, var.supp=var.supp, qual.supp=qual.supp, corr=TRUE, k=8)
```

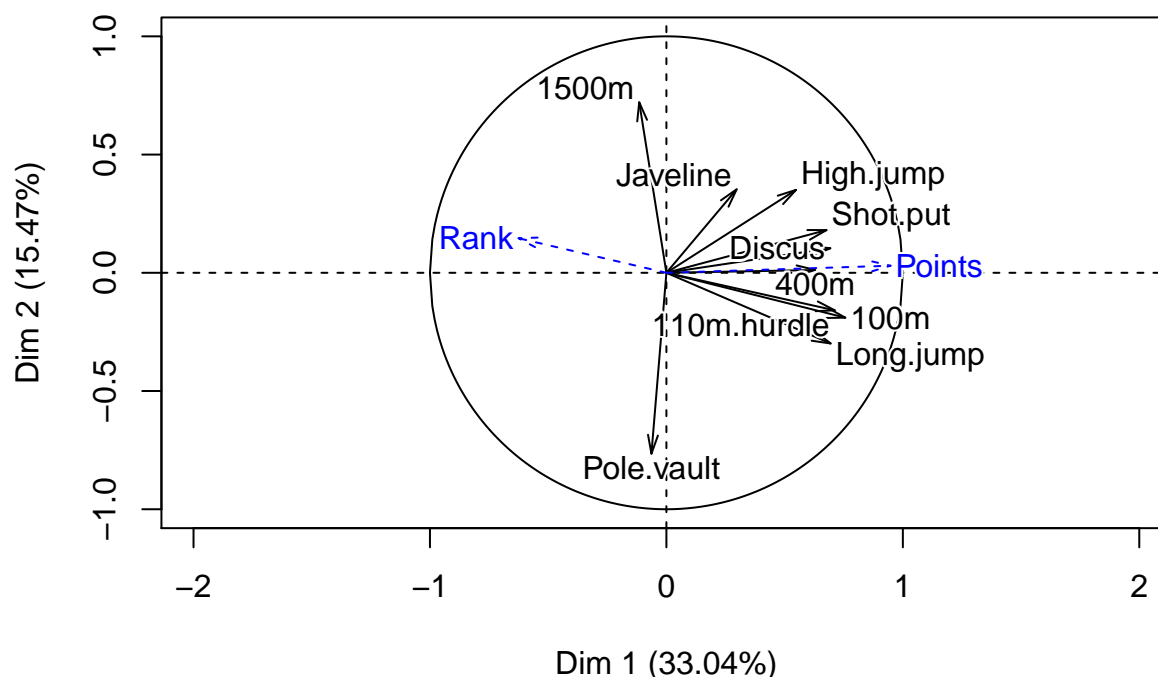
4 PCA() function in package FactoMineR

13. Compare the result in `out2` with the ones provided by the function `PCA()`.

```
res<-PCA(data, ncp = 10, ind.sup = 39:41, quanti.sup = c(11: 12), quali.sup = 13, graph=T)
```



Variables factor map (PCA)



We can compare some matrices of `pca()` and `PCA()` to see if they match. Some eigenvectors are multiplied by a factor (-1) . In order to compare them properly we take the absolute value of the matrices:

```
#SVD
all.equal(abs(out2$right.sing.vec),abs(res$svd$V))

## [1] TRUE

#Var. Coord. (Change the name of the columns, if not it gives an error...)
colnames(out2$var.coord)<-paste("Dim.",1:ncol(var.contr),sep="")
all.equal(abs(out2$var.coord),abs(res$var$coord))

## [1] TRUE

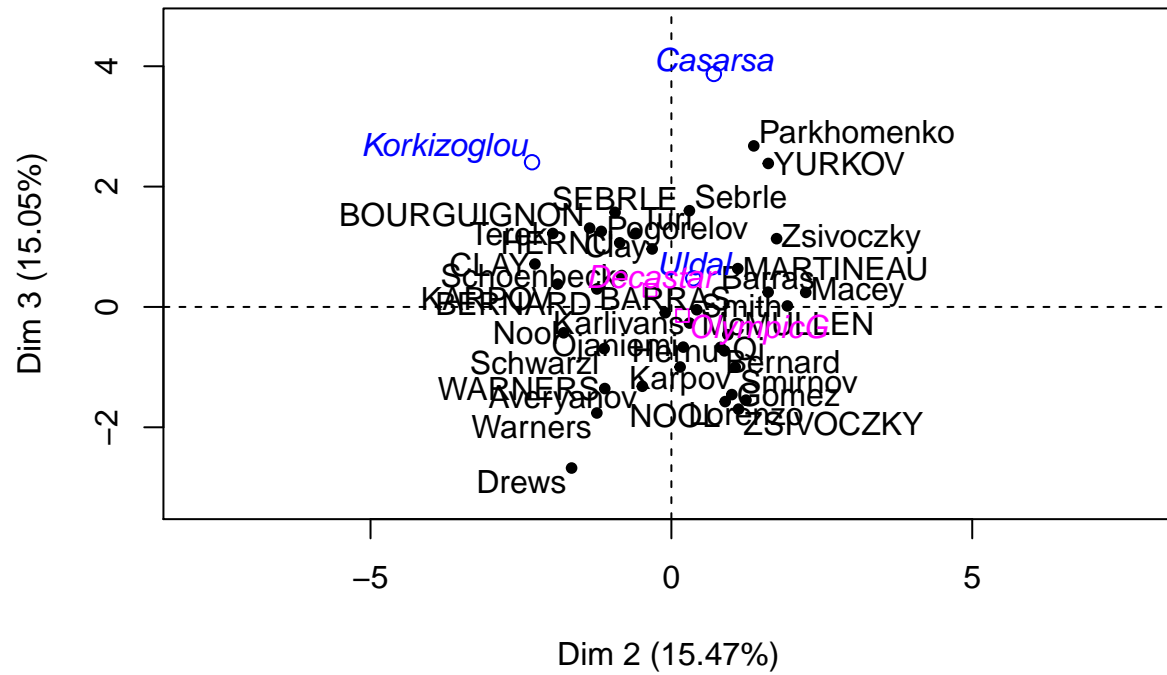
#Ind. Coord.
colnames(out2$ind.coord)<-paste("Dim.",1:ncol(var.contr),sep="")
all.equal(abs(out2$ind.coord),abs(res$ind$coord))

## [1] TRUE
```

14. Function `PCA()` has the default option `graph=TRUE` that displays 2 graphs. In particular, explore the arguments options: `axes`, `choix`, `ellipse`, `xlim`, `ylim`, `habillage`, `invisible`, `lim.cos2.var`, `select`, etc. Use these options to explore other components 3 and 4, to represent the athletes in a different color depending on Decastar or OlympicG, etc.

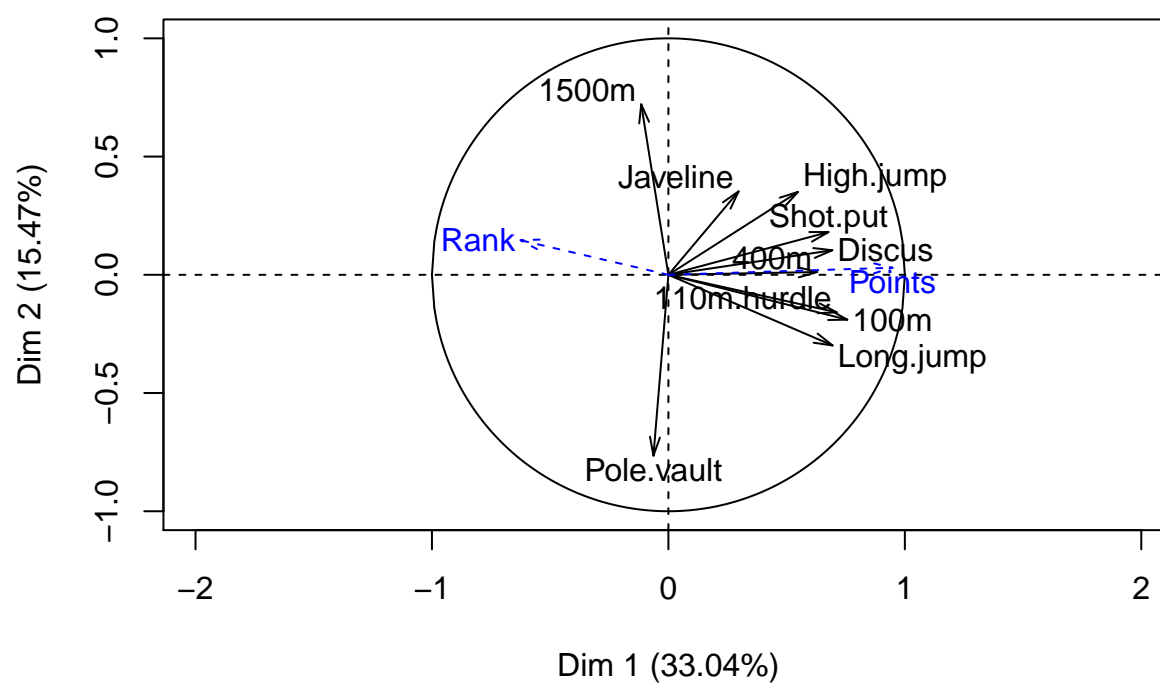
```
#Axes selects the components to plot
plot.PCA(res,axes = c(2,3))
```


Individuals factor map (PCA)



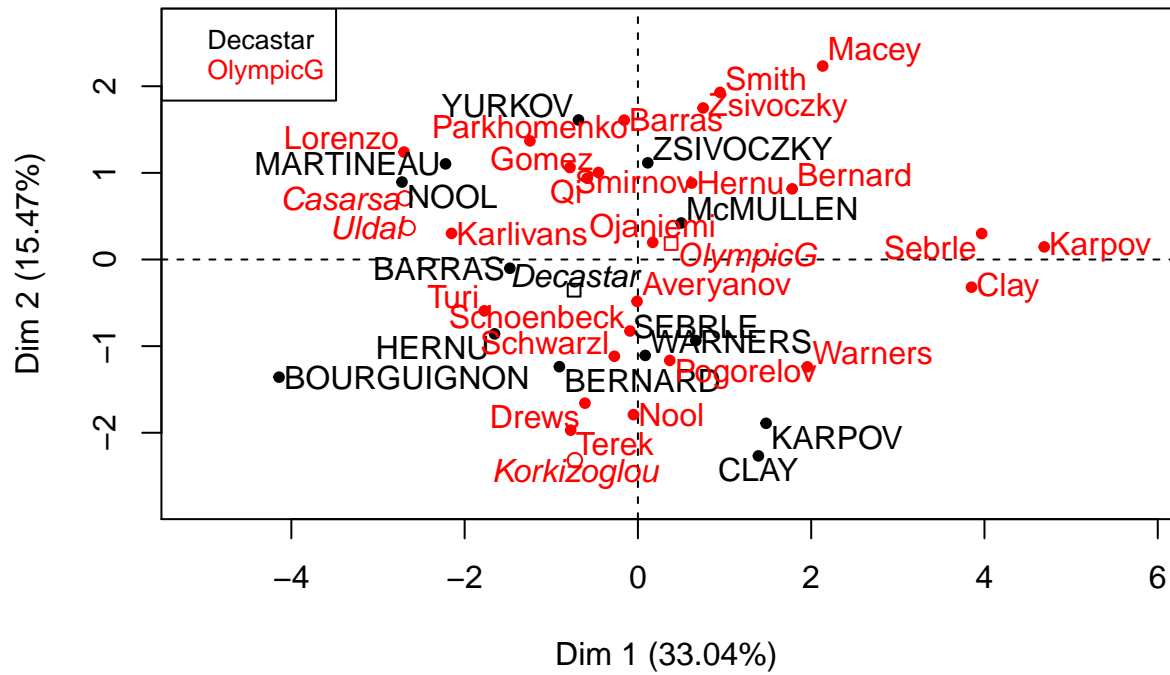
```
#plot the variables
plot.PCA(res, choix = "var")
```

Variables factor map (PCA)



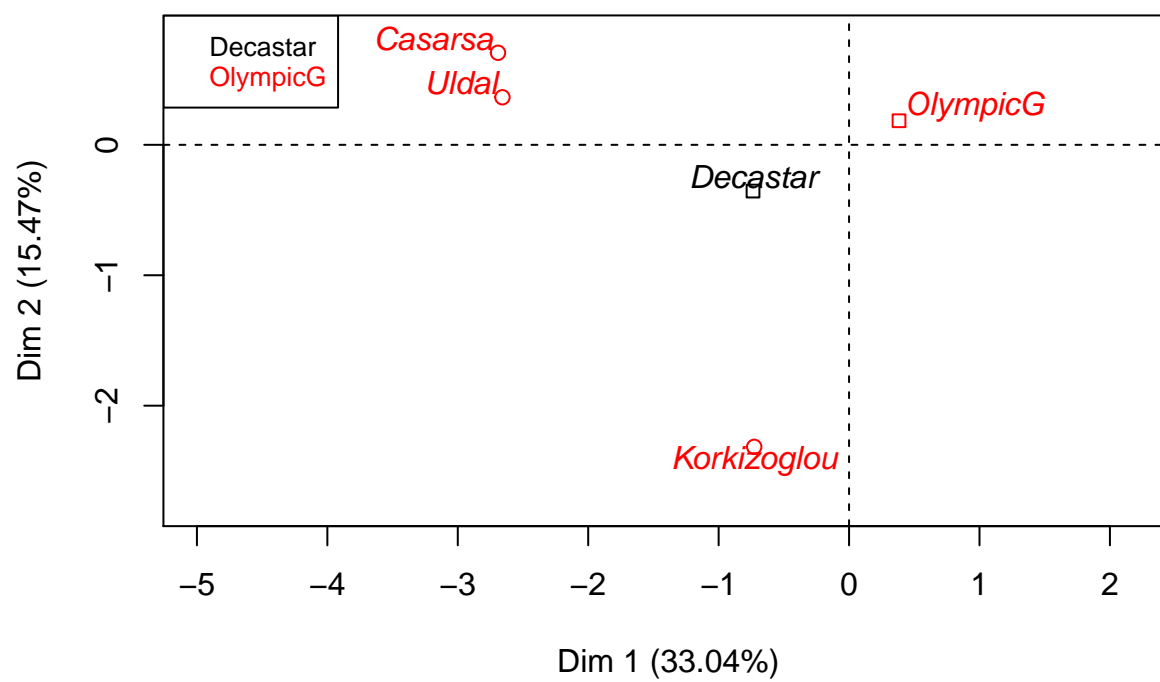
```
#plot ind. by group  
plot.PCA(res, choix="ind", habillage=13)
```

Individuals factor map (PCA)



```
#Make ind invisible
plot.PCA(res, choix="ind", habillage=13, invisible = "ind")
```

Individuals factor map (PCA)



```
#Comp 3 and 4  
plot.PCA(res, axes = c(3,4), habillage = 13)
```

Individuals factor map (PCA)

