

2019/2020



Faculté des sciences et techniques MARRAKECH

Master SDAD

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[ATELIER 1 : ACP EN R

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INTRODUCTION

L'objectif de cet atelier est d'étudier des exemples d'application de l'analyse en composantes principales avec le logiciel R ainsi pratiquer cette étude sur un exercice d'application.

EXEMPLE D'APPRENTISSAGE

Pour débiter la phase d'apprentissage on commence par charger les bibliothèques nécessaires :

```
#Installing environnement packages
install.packages("FactoMineR")
library(FactoMineR)
install.packages("ggplot2")
library(ggplot2)
install.packages("factoextra")
library(factoextra)
```

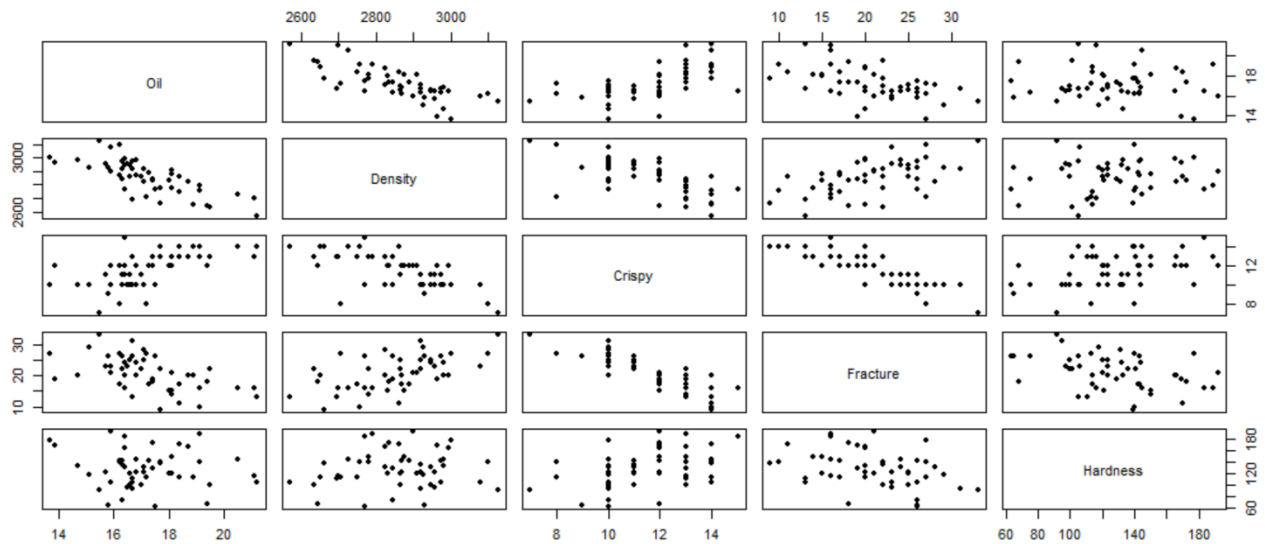
-  **FactoMineR** : Bibliothèque de la fonction ACP.
-  **ggplot2** : Bibliothèque des affichages graphiques nécessaires.
-  **factoextra** : Bibliothèque de la visualisation des nuages de points ACP.

Ensuite on va charger les données à travailler avec :

```
#Loading data
food <- read.csv("C:/Users/PC Gamer/Desktop/food.csv")
#Cheking data
head(food)
```

On affiche le nuage des points par couples de variables :

```
#Variables display
plot(food[,1:2], pch=16, cex=0.8)
```



Application de la fonction ACP sur les données :

```
#Applying ACP on data
food.pca<-PCA(food[,-1], scale.unit=TRUE, graph=F)
#Displaying ACP summary
summary(food.pca)
#Other way of displaying ACP informations
str(food.pca)
```

```
Call:
PCA(X = food[, -1], scale.unit = TRUE, graph = F)
```

Eigenvalues

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
Variance	3.031	1.296	0.310	0.242	0.121
% of var.	60.624	25.914	6.201	4.838	2.422
Cumulative % of var.	60.624	86.538	92.739	97.578	100.000

Individuals (the 10 first)

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
1	1.659	-1.397	1.288	0.710	-0.626
2	3.096	2.823	5.258	0.831	0.357
3	1.109	0.240	0.038	0.047	0.870
4	2.403	-1.959	2.532	0.664	-1.153
5	1.567	-1.281	1.082	0.668	0.670
6	2.676	2.018	2.686	0.568	1.312
7	1.664	1.503	1.491	0.816	-0.639
8	2.582	-0.837	1.463	0.105	-2.407
9	1.344	-1.197	0.946	0.794	0.354
10	1.096	-1.004	0.664	0.838	0.354

	cos2
1	0.060
2	0.123
3	0.310
4	0.034
5	0.079
6	0.173
7	0.027
8	0.010
9	0.102
10	0.009

Variables

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
oil	0.797	-0.422	0.367	43.432	0.135
Density	-0.834	0.406	0.009	0.026	0.000
Crispy	0.927	0.225	-0.100	3.200	0.010
Fracture	-0.878	-0.252	0.302	29.407	0.091
Hardness	0.267	0.916	0.272	23.935	0.074

```

List of 5
 $ eig : num [1:5, 1:3] 3.031 1.296 0.31 0.242 0.121 ...
  ..- attr(*, "dimnames")=List of 2
    .. ..$ : chr [1:5] "comp 1" "comp 2" "comp 3" "comp 4" ...
    .. ..$ : chr [1:3] "eigenvalue" "percentage of variance" "cumulative percentage of variance"
 $ var :List of 4
  ..$ coord : num [1:5, 1:5] 0.797 -0.834 0.927 -0.878 0.267 ...
  .. ..- attr(*, "dimnames")=List of 2
    .. .. ..$ : chr [1:5] "oil" "Density" "Crispy" "Fracture" ...
    .. .. ..$ : chr [1:5] "Dim.1" "Dim.2" "Dim.3" "Dim.4" ...
  ..$ cor : num [1:5, 1:5] 0.797 -0.834 0.927 -0.878 0.267 ...
  .. ..- attr(*, "dimnames")=List of 2
    .. .. ..$ : chr [1:5] "oil" "Density" "Crispy" "Fracture" ...
    .. .. ..$ : chr [1:5] "Dim.1" "Dim.2" "Dim.3" "Dim.4" ...
  ..$ cos2 : num [1:5, 1:5] 0.6345 0.6947 0.8592 0.7714 0.0713 ...
  .. ..- attr(*, "dimnames")=List of 2
    .. .. ..$ : chr [1:5] "oil" "Density" "Crispy" "Fracture" ...
    .. .. ..$ : chr [1:5] "Dim.1" "Dim.2" "Dim.3" "Dim.4" ...
  ..$ contrib: num [1:5, 1:5] 20.93 22.92 28.34 25.45 2.35 ...
  .. ..- attr(*, "dimnames")=List of 2
    .. .. ..$ : chr [1:5] "oil" "Density" "Crispy" "Fracture" ...
    .. .. ..$ : chr [1:5] "Dim.1" "Dim.2" "Dim.3" "Dim.4" ...
 $ ind :List of 4

```

Génération des valeurs propres/variances :

```

#Proper values calculation
eig.val<-get_eigenvalue(food.pca)

```

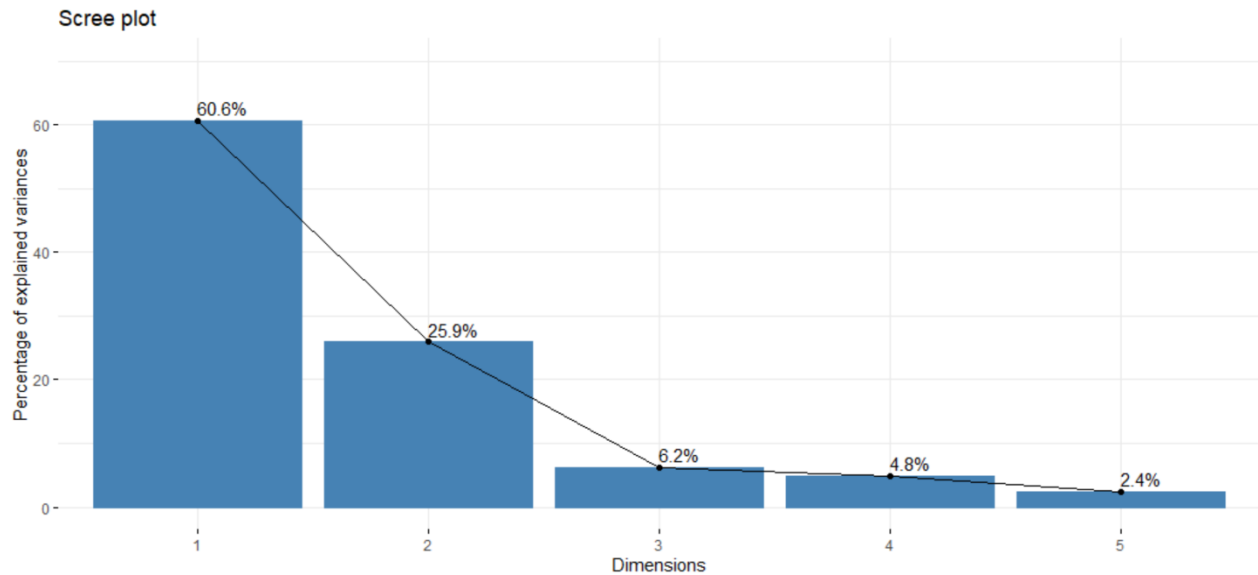
	eigenvalue	variance.percent	cumulative.variance.percent
Dim.1	3.0312132	60.624263	60.62426
Dim.2	1.2957058	25.914115	86.53838
Dim.3	0.3100493	6.200987	92.73937
Dim.4	0.2419201	4.838402	97.57777
Dim.5	0.1211116	2.422233	100.00000

Affichage du graphe des pourcentages des variances :

```

#Displaying the percentage of explained variances
fviz_eig(food.pca, addlabels = TRUE, ylim = c(0, 70))

```



Génération des résultats pour les variables de sortie ACP :

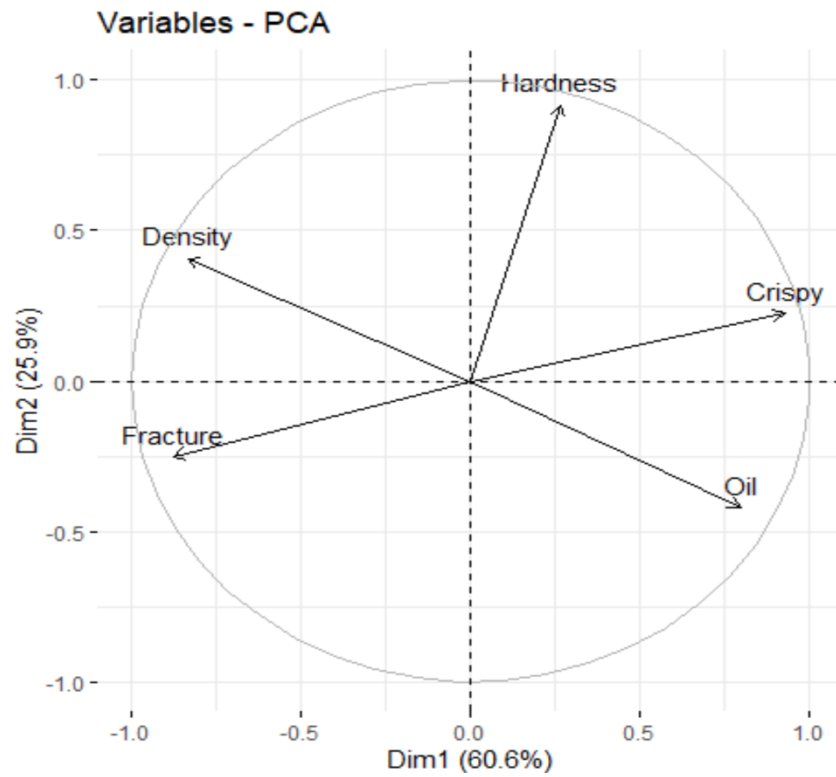
```
#Active variables results
food.var<-get_pca_var(food.pca)
```

Principal Component Analysis Results for variables

```
=====
  Name      Description
1 "$coord"  "Coordinates for the variables"
2 "$cor"    "Correlations between variables and dimensions"
3 "$cos2"   "Cos2 for the variables"
4 "$contrib" "contributions of the variables"
```

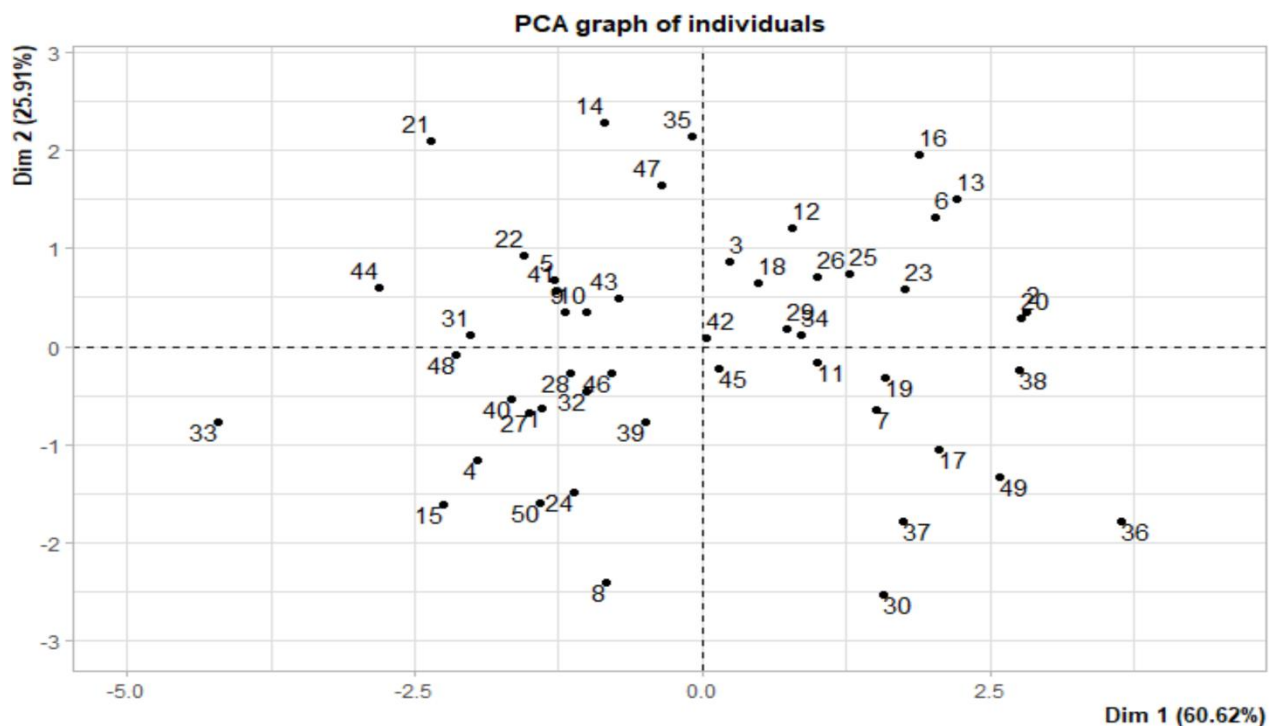
Affichage du graphe de la corrélation des variables :

```
#Displaying variables correlation
fviz_pca_var(food.pca, col.var = "black")
```



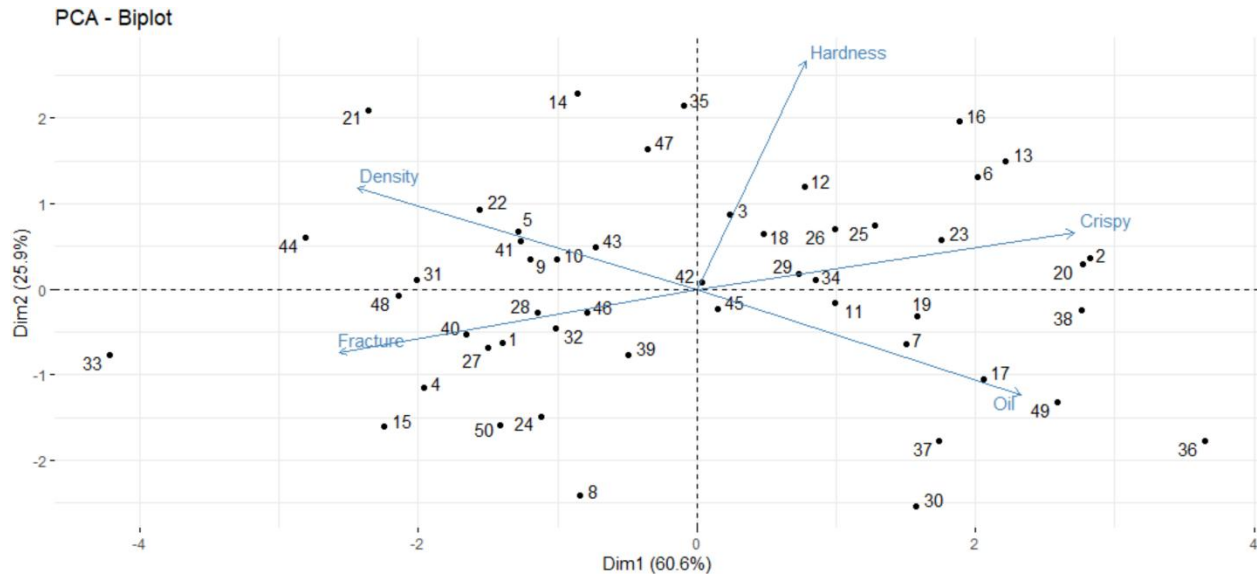
Affichage du graphe de la corrélation des individus :

```
#Displaying individuals
plot(food.pca)
```



Affichage du graphe de la corrélation des individus et des variables en même temps :

```
#Displaying individuals and variables  
fviz_pca_biplot(food.pca, repel=TRUE)
```



EXERCICE D'APPLICATION

On commence par charger les 30 premières lignes des données de « quakes » :

```
#Cheking Data
quakes<-head(quakes,30)
```

Application de la fonction ACP sur les données :

```
#Applying ACP on data
quakes.pca=PCA(quakes, scale.unit=TRUE, graph=F)
```

```
Call:
PCA(X = quakes[1:30, ], scale.unit = TRUE, graph = F)
```

Eigenvalues

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
Variance	2.642	1.357	0.730	0.215	0.055
% of var.	52.850	27.145	14.606	4.294	1.105
Cumulative % of var.	52.850	79.994	94.600	98.895	100.000

Individuals (the 10 first)

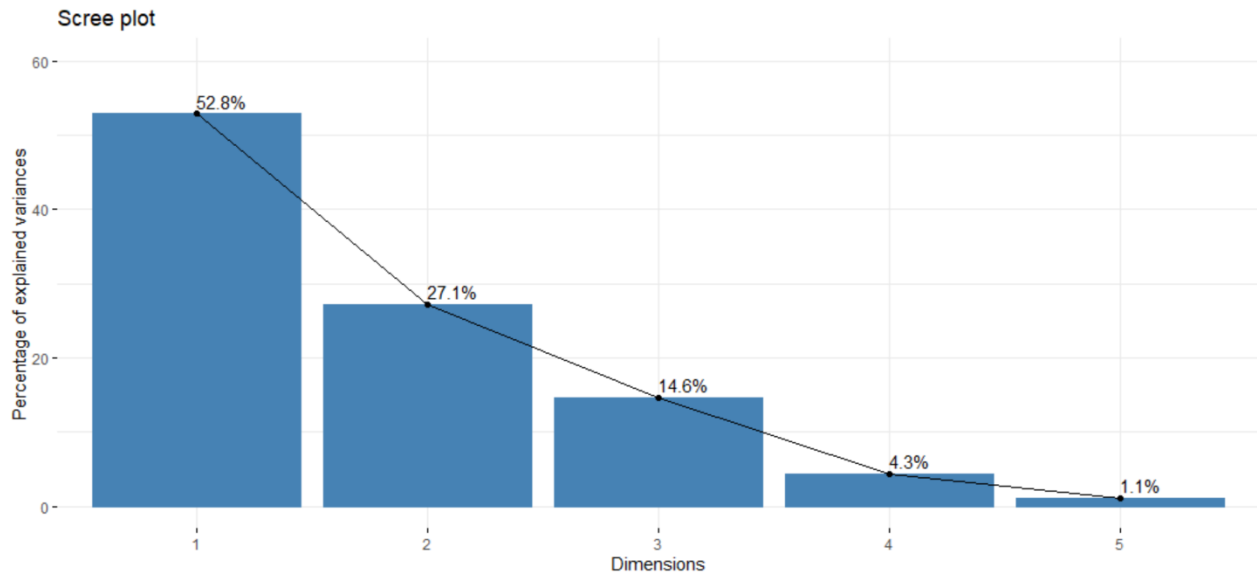
	Dist	Dim.1	ctr	cos2	Dim.2	ctr	cos2
1	1.163	-0.078	0.008	0.005	0.597	0.874	0.263
2	1.592	-1.442	2.622	0.820	-0.236	0.136	0.022
3	2.896	1.186	1.776	0.168	2.320	13.224	0.642
4	1.628	-1.320	2.196	0.657	-0.605	0.900	0.138
5	1.907	-1.800	4.085	0.891	-0.386	0.366	0.041
6	1.924	-0.986	1.227	0.263	-0.123	0.037	0.004
7	3.240	2.590	8.459	0.639	-1.819	8.123	0.315
8	2.230	-0.739	0.688	0.110	1.331	4.353	0.357
9	2.226	-0.020	0.000	0.000	1.898	8.851	0.728
10	1.373	-0.921	1.070	0.450	-0.712	1.244	0.269

Variables

	Dim.1	ctr	cos2	Dim.2	ctr	cos2
lat	0.375	5.318	0.141	-0.851	53.400	0.725
long	-0.746	21.077	0.557	0.560	23.103	0.314
depth	-0.699	18.478	0.488	-0.094	0.653	0.009
mag	0.870	28.630	0.757	0.419	12.956	0.176
stations	0.837	26.497	0.700	0.366	9.888	0.134

Affichage du graphe des pourcentages des variances :

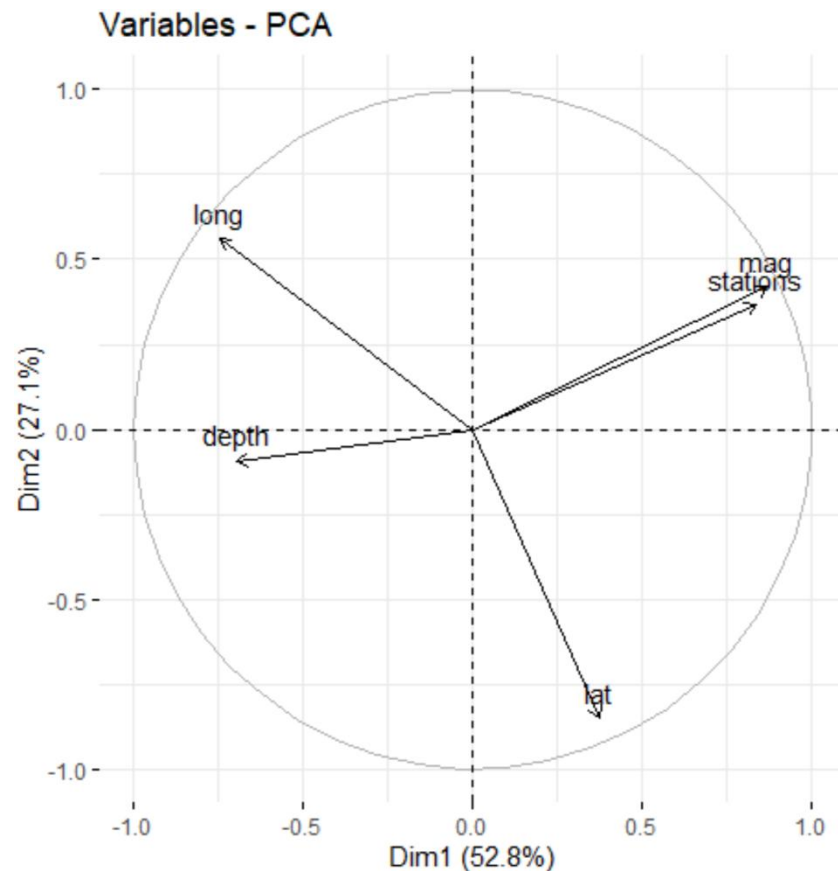
```
#Displaying the percentage of explained variances  
fviz_eig(quakes.pca, addlabels = TRUE, ylim = c(0, 60))
```



Continuer l'analyse avec deux composantes principales est suffisant car on a la plupart d'inertie totale sur le premier et le deuxième axe.

Affichage du graphe de la corrélation des variables :

```
#Displaying variables correlation  
fviz_pca_var(quakes.pca, col.var = "black")
```



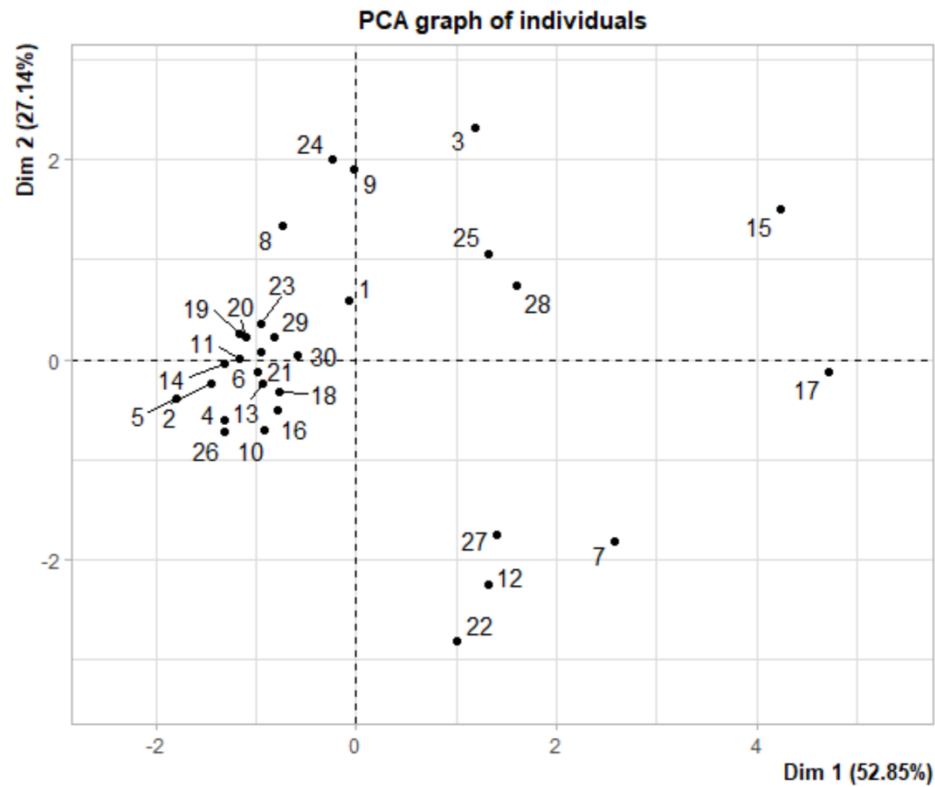
```
> quakes.var$contrib
```

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
lat	5.318355	53.4001866	9.717830	29.0296906	2.5339379
long	21.077181	23.1032349	2.439826	51.5265728	1.8531859
depth	18.477540	0.6530227	63.971971	16.5221774	0.3752884
mag	28.629742	12.9559030	5.062486	0.7282807	52.6235875
stations	26.497182	9.8876528	18.807887	2.1932785	42.6140003

Les variables importantes pour la première composante sont : Long, Mag, Stations (x > 20) et pour la deuxième composante : Lat, Long (x > 20)

Affichage du graphe de la corrélation des individus :

```
#Displaying individuals correlation
plot(quakes.pca)
```



Affichage du graphe de la corrélation des individus et des variables :

```
#Displaying individuals and variables
fviz_pca_biplot(quakes.pca, repel=TRUE)
```

