

Analyse En Composantes Principales

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Exercice 1 : Arrestations aux Etats-Unis

Le fichier de données USarrests, accessible en R via la commande data ("USarrests"), contient des statistiques collectées en 1973 sur les taux d'arrestation pour 100000 habitants pour agression, meurtre ou viol dans chacun des $n = 50$ états des USA. Une quatrième variable indique le pourcentage de résidents dans des zones urbaines pour chaque état

Chargement des données :

```
knitr::opts_chunk$set(warning=FALSE, message=FALSE)
data = USarrests
head(data)
```

```
##           Murder Assault UrbanPop Rape
## Alabama      13.2      236        58 21.2
## Alaska       10.0      263        48 44.5
## Arizona       8.1      294        80 31.0
## Arkansas      8.8      190        50 19.5
## California    9.0      276        91 40.6
## Colorado      7.9      204        78 38.7
```

```
dim(data)
```

```
## [1] 50  4
```

Faisant une analyse descriptive des variables :

```
summary(data)
```

```
##           Murder           Assault           UrbanPop           Rape
## Min.      : 0.800   Min.      : 45.0   Min.      :32.00   Min.      : 7.30
## 1st Qu.: 4.075   1st Qu.:109.0   1st Qu.:54.50   1st Qu.:15.07
## Median : 7.250   Median :159.0   Median :66.00   Median :20.10
## Mean     : 7.788   Mean     :170.8   Mean     :65.54   Mean     :21.23
## 3rd Qu.:11.250   3rd Qu.:249.0   3rd Qu.:77.75   3rd Qu.:26.18
## Max.     :17.400   Max.      :337.0   Max.      :91.00   Max.      :46.00
```

On va effectuer en premier temps une **ACP non normée** :

```
acp = prcomp(data,scale=FALSE)
summary(acp)
```

```
## Importance of components:
##              PC1      PC2      PC3      PC4
## Standard deviation 83.7324 14.21240 6.4894 2.48279
## Proportion of Variance 0.9655 0.02782 0.0058 0.00085
## Cumulative Proportion 0.9655 0.99335 0.9991 1.00000
```

on va utiliser la bibliothèque **factoextra** pour voir le résultat de l'acp :

```
library(factoextra)
var <- get_pca_var(acp)
var
```

```
## 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"
```

Interpretation des axes

Variables :

```
var <- get_pca_var(acp)
var
```

```
## 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"
```

```
var$contrib
```

```
##           Dim.1      Dim.2      Dim.3      Dim.4
## Murder      0.1739250  0.2008981  0.6382517  98.9869251
## Assault     99.0465399  0.3452741  0.4565669  0.1516191
## UrbanPop    0.2147001  95.4250536  4.0218813  0.3383649
## Rape        0.5648349  4.0287742  94.8833000  0.5230908
```

La variable qui contribue le plus à la formation de l'axe 1 est : **Assault** 99.04%

La variable qui contribue le plus à la formation de l'axe 2 est : **UrbanPop** 95.42%

Individues :

```
ind = get_pca_ind(acp)
ind
```

```
## Principal Component Analysis Results for individuals
## =====
##   Name      Description
## 1 "$coord"   "Coordinates for the individuals"
## 2 "$cos2"    "Cos2 for the individuals"
## 3 "$contrib" "contributions of the individuals"
```

ind\$contrib

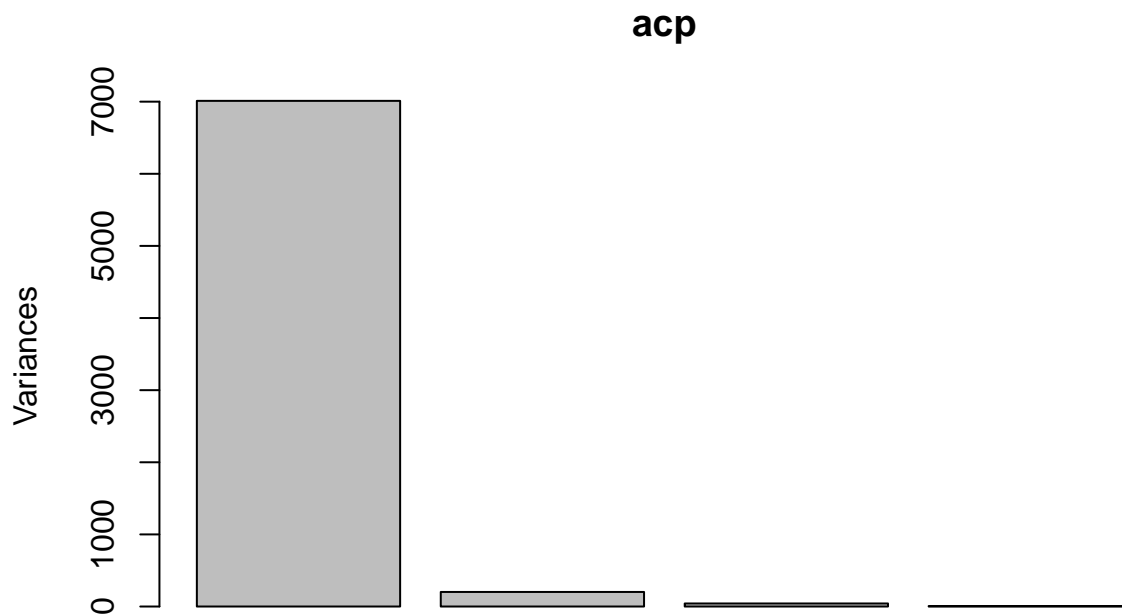
| ## | Dim.1 | Dim.2 | Dim.3 | Dim.4 |
|-------------------|-------------|--------------|--------------|-------------|
| ## Alabama | 1.197903759 | 1.2976418443 | 2.956209e-01 | 1.88116656 |
| ## Alaska | 2.458078547 | 3.2019648471 | 1.923788e+01 | 5.43820626 |
| ## Arizona | 4.391005602 | 0.7720689580 | 1.352317e-01 | 6.14984356 |
| ## Arkansas | 0.095949619 | 2.7626851500 | 2.098161e-03 | 0.08806732 |
| ## California | 3.291827649 | 5.0215119680 | 2.161194e+00 | 2.56523334 |
| ## Colorado | 0.348965784 | 1.8637039554 | 7.160924e+00 | 0.96149219 |
| ## Connecticut | 1.057538260 | 1.6560065045 | 3.367514e+00 | 0.15893694 |
| ## Delaware | 1.270277225 | 0.0181469083 | 6.043790e+00 | 4.50917869 |
| ## Florida | 7.789261053 | 0.3898338951 | 4.268363e-01 | 0.50507625 |
| ## Georgia | 0.468713058 | 0.5262336818 | 6.187548e-01 | 17.49752661 |
| ## Hawaii | 4.353421604 | 5.8424265482 | 6.587795e-01 | 3.91310880 |
| ## Idaho | 0.765336041 | 0.8878115518 | 1.097339e-01 | 3.63643961 |
| ## Illinois | 1.779959840 | 1.6469351944 | 1.643820e+00 | 0.04385280 |
| ## Indiana | 0.944817808 | 0.0802131582 | 6.636411e-01 | 0.88270967 |
| ## Iowa | 3.811178746 | 0.1105966181 | 2.031477e-02 | 0.24529302 |
| ## Kansas | 0.887873052 | 0.0986980472 | 7.016220e-03 | 0.13826086 |
| ## Kentucky | 1.110140486 | 1.1279507825 | 2.376752e-01 | 4.87490383 |
| ## Louisiana | 1.747906772 | 0.1826436974 | 6.958757e-01 | 6.52222533 |
| ## Maine | 2.272829412 | 1.3066847644 | 1.045703e+00 | 1.45299202 |
| ## Maryland | 4.771362131 | 0.2482308085 | 2.616421e-01 | 1.20645227 |
| ## Massachusetts | 0.129010803 | 3.7457798163 | 2.676501e+00 | 0.34743916 |
| ## Michigan | 2.082967849 | 0.3451990828 | 1.984568e+00 | 0.08080430 |
| ## Minnesota | 2.793294853 | 0.2687224148 | 2.052320e-06 | 0.17379944 |
| ## Mississippi | 2.152019195 | 7.4489765851 | 1.188924e+00 | 4.88381523 |
| ## Missouri | 0.018194198 | 0.2755786535 | 1.436926e+00 | 0.14976425 |
| ## Montana | 1.113718649 | 0.8955749264 | 1.605005e-01 | 0.01962536 |
| ## Nebraska | 1.361932458 | 0.0004416376 | 1.040274e-02 | 0.13986446 |
| ## Nevada | 1.994327763 | 2.2582631479 | 1.198930e+01 | 0.03623708 |
| ## New Hampshire | 3.757987534 | 0.2219494064 | 2.473979e-01 | 0.28419650 |
| ## New Jersey | 0.033369846 | 5.3005612023 | 1.891027e+00 | 0.84354897 |
| ## New Mexico | 3.763936281 | 0.0011208412 | 2.428411e-01 | 0.61900363 |
| ## New York | 2.026929368 | 2.5107154612 | 1.058603e+00 | 0.25816572 |
| ## North Carolina | 7.702876158 | 9.5746141114 | 6.496872e+00 | 1.44612484 |
| ## North Dakota | 4.636959310 | 2.5777161839 | 8.172829e-02 | 1.71778828 |
| ## Ohio | 0.715632181 | 1.4929455990 | 1.304475e-01 | 1.33586826 |
| ## Oklahoma | 0.110636533 | 0.1124575478 | 9.751884e-03 | 0.01055266 |
| ## Oregon | 0.035465927 | 0.1479905724 | 3.139037e+00 | 2.75506825 |
| ## Pennsylvania | 1.193728864 | 0.7863233982 | 4.882828e-01 | 1.14057171 |
| ## Rhode Island | 0.002678013 | 3.3427281949 | 1.449453e+01 | 1.72869888 |
| ## South Carolina | 3.283137685 | 5.4848483787 | 1.962477e-01 | 0.50837317 |
| ## South Dakota | 2.115032315 | 2.7277160686 | 8.204635e-02 | 0.50881281 |
| ## Tennessee | 0.087423839 | 0.4191794654 | 1.767239e+00 | 4.99292104 |
| ## Texas | 0.279309160 | 1.6694601113 | 7.350324e-03 | 5.83837758 |
| ## Utah | 0.710685046 | 3.0839587031 | 1.518568e-01 | 1.13179220 |
| ## Vermont | 4.436868928 | 7.3867198390 | 1.095475e+00 | 1.30428521 |
| ## Virginia | 0.062631058 | 0.0304136187 | 5.190062e-02 | 0.44706271 |
| ## Washington | 0.179371673 | 0.9838031609 | 1.085621e+00 | 2.34965359 |
| ## West Virginia | 2.390610500 | 5.2163812660 | 7.674211e-03 | 0.17617380 |
| ## Wisconsin | 3.983858420 | 0.3003423273 | 3.491236e-01 | 0.01363141 |
| ## Wyoming | 0.031059144 | 0.3475293948 | 6.837768e-01 | 0.08701362 |

Les individus qui contribuent le plus à la formation de l'axe 1 sont : **North Carolina** et **Florida** 7%

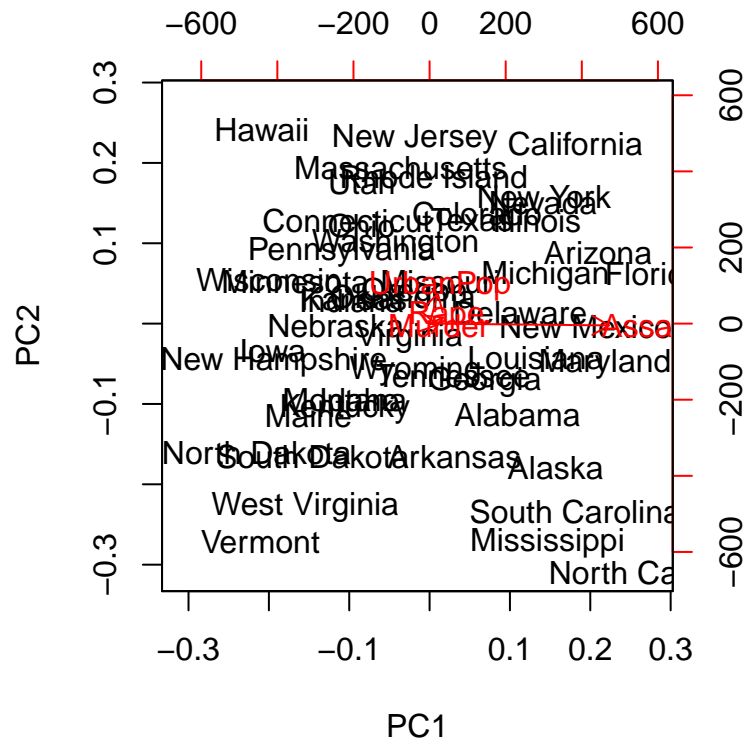
Les individus qui contribuent le plus à la formation de l'axe 2 sont : **North Carolina** 9% et **Mississippi** , **Vermont** 7%

Une représentation graphique donne le résultat suivant :

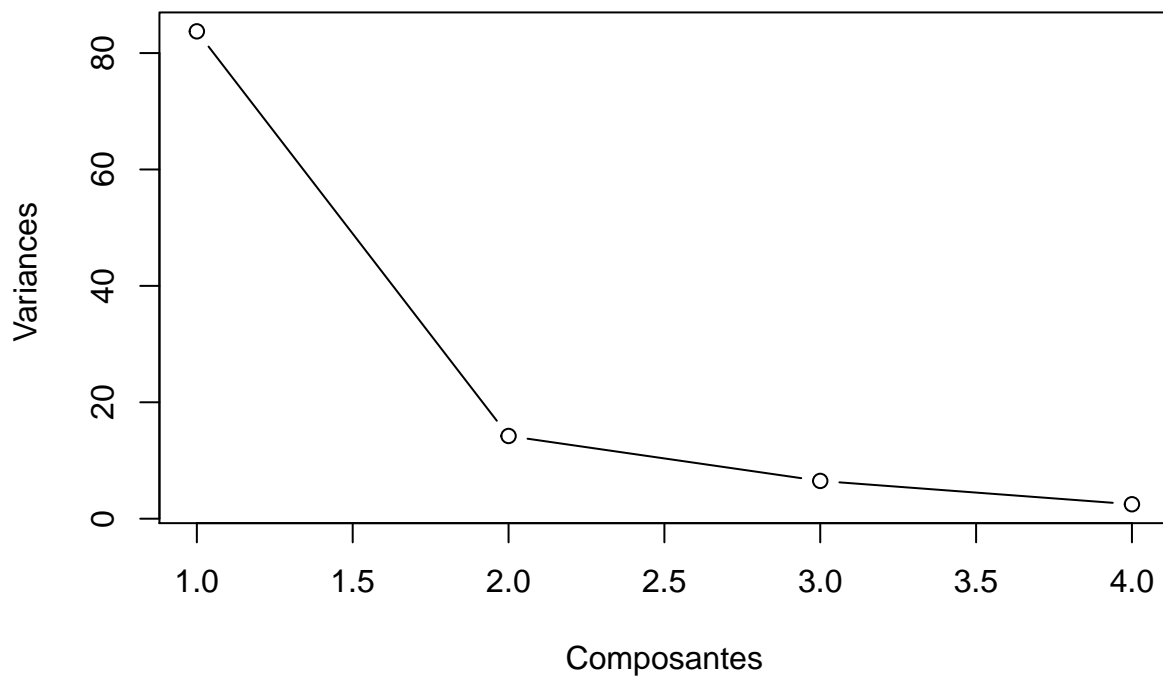
```
plot(acp)
```



```
biplot(acp , xlab = row.names(USArrests))
```



```
plot(1:4,acp$sdev,ylab="Variances ",xlab="Composantes",type = "b")
```



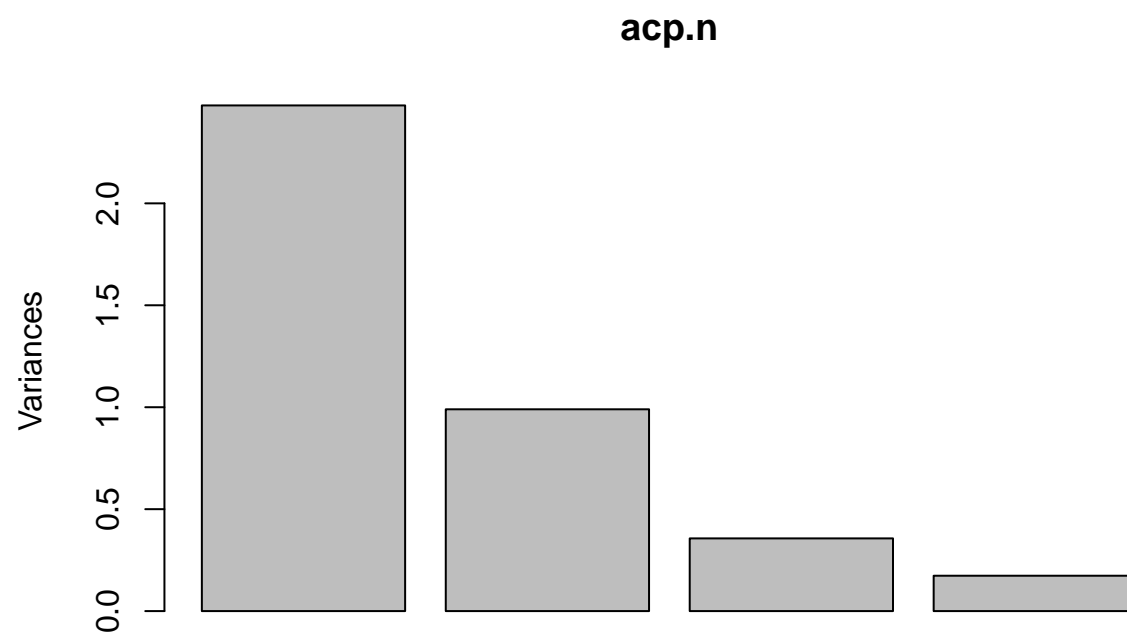
On va maintenant effectuer une **ACP normée** :

```
acp.n = prcomp(data,scale=TRUE)
summary(acp.n)
```

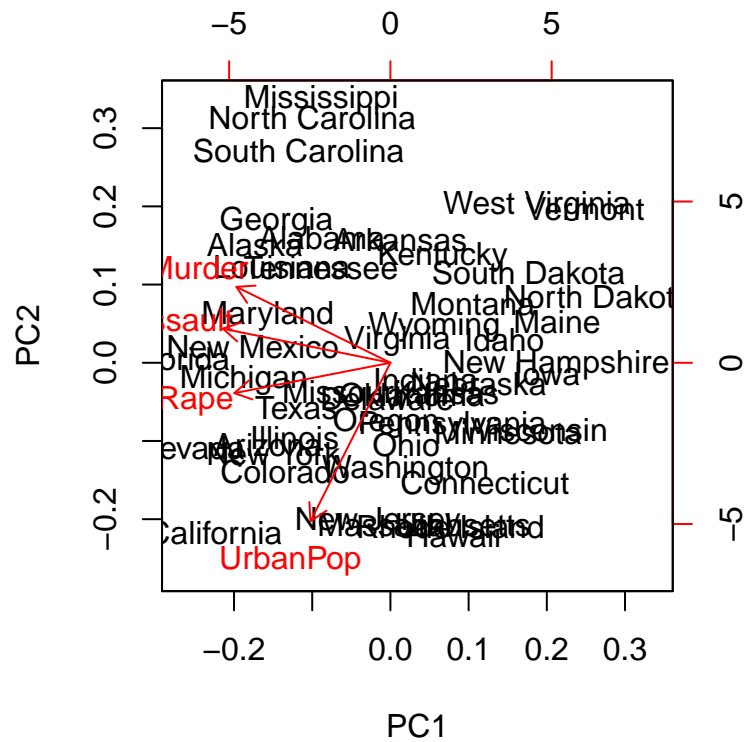
```
## Importance of components:
##              PC1      PC2      PC3      PC4
## Standard deviation    1.5749 0.9949 0.59713 0.41645
## Proportion of Variance 0.6201 0.2474 0.08914 0.04336
## Cumulative Proportion 0.6201 0.8675 0.95664 1.00000
```

La représentation graphique de l'ACP normée donne le résultat suivant:

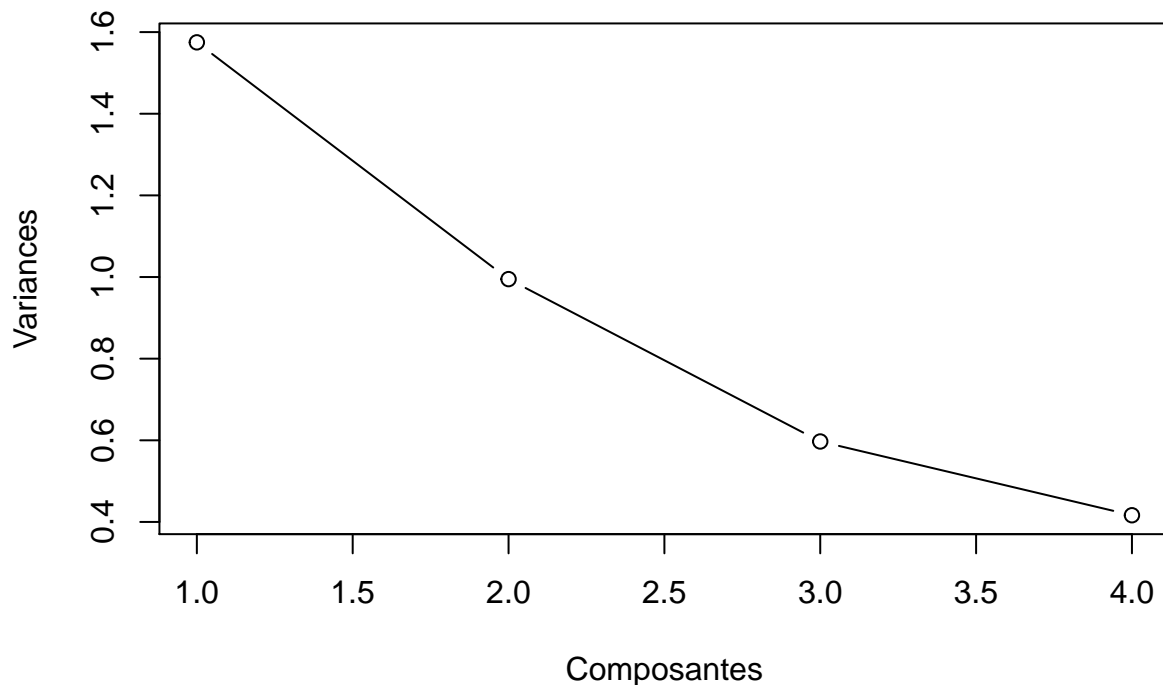
```
plot(acp.n)
```



```
biplot(acp.n , xlabs = row.names(USArrests))
```



```
plot(1:4,acp.n$sdev,ylab="Variances",xlab="Composantes",type = "b")
```

On constate qu'après la **normalisation** le **2ème** composant participe plus que dans la version non normée

Exercice 2 : Analyse textuelle d'un corpus d'emails

Dans cette partie, il est proposé d'analyser un jeu de données textuelles dans le but de tenter de déterminer les caractéristiques de spams. Une telle analyse est classiquement basée sur la fréquence d'une sélection de mots dans un ensemble d'apprentissage constitué de courriels qui appartiennent à 2 catégories possibles : spam ou non-spam. Les données analysées dans ce TP sont publiques, et elles peuvent servir de "benchmark" pour la comparaison de méthodes d'apprentissage automatique (UCI Machine Learning Repository. Irvine, CA : University of California, School of Information and Computer Science. <http://archive.ics.uci.edu/ml>): Il a été constitué un échantillon de messages électroniques dans chacun desquels a été évalué le nombre d'occurrences d'une sélection de mots et caractères. Les variables considérées sont des ratios qui correspondent au nombre d'occurrences d'un mot spécifique sur le nombre total de mots, ou nombre d'occurrences d'un caractère sur le nombre de caractères du message. Il a également considéré trois variables prenant en compte la casse (majuscule / minuscule) des caractères et une dernière variable qualitative binaire indiquant le type de chaque message : spam ou Nsp.

Chargement des données :

```
setwd("C:/Users/W 7/Desktop/Master 2/AA2/TP2_ACP")
spam <- read.csv("Data\\data_spam.csv",header=FALSE,sep=";")
nom_spam <- read.csv("Data\\names_spam.csv",header=FALSE,sep=";")
names(spam) <- sapply((1:nrow(nom_spam)),function(i) toString(nom_spam[i,1]))
spam$y = as.factor(spam$y)

head(spam)
```

```

## word_freq_make word_freq_address word_freq_all word_freq_3d
## 1      0.00      0.64      0.64      0
## 2      0.21      0.28      0.50      0
## 3      0.06      0.00      0.71      0
## 4      0.00      0.00      0.00      0
## 5      0.00      0.00      0.00      0
## 6      0.00      0.00      0.00      0
## word_freq_our word_freq_over word_freq_remove word_freq_internet
## 1      0.32      0.00      0.00      0.00
## 2      0.14      0.28      0.21      0.07
## 3      1.23      0.19      0.19      0.12
## 4      0.63      0.00      0.31      0.63
## 5      0.63      0.00      0.31      0.63
## 6      1.85      0.00      0.00      1.85
## word_freq_order word_freq_mail word_freq_receive word_freq_will
## 1      0.00      0.00      0.00      0.64
## 2      0.00      0.94      0.21      0.79
## 3      0.64      0.25      0.38      0.45
## 4      0.31      0.63      0.31      0.31
## 5      0.31      0.63      0.31      0.31
## 6      0.00      0.00      0.00      0.00
## word_freq_people word_freq_report word_freq_addresses word_freq_free
## 1      0.00      0.00      0.00      0.32
## 2      0.65      0.21      0.14      0.14
## 3      0.12      0.00      1.75      0.06
## 4      0.31      0.00      0.00      0.31
## 5      0.31      0.00      0.00      0.31
## 6      0.00      0.00      0.00      0.00
## word_freq_business word_freq_email word_freq_you word_freq_credit
## 1      0.00      1.29      1.93      0.00
## 2      0.07      0.28      3.47      0.00
## 3      0.06      1.03      1.36      0.32
## 4      0.00      0.00      3.18      0.00
## 5      0.00      0.00      3.18      0.00
## 6      0.00      0.00      0.00      0.00
## word_freq_your word_freq_font word_freq_000 word_freq_money word_freq_hp
## 1      0.96      0      0.00      0.00      0
## 2      1.59      0      0.43      0.43      0
## 3      0.51      0      1.16      0.06      0
## 4      0.31      0      0.00      0.00      0
## 5      0.31      0      0.00      0.00      0
## 6      0.00      0      0.00      0.00      0
## word_freq_hpl word_freq_george word_freq_650 word_freq_lab
## 1      0      0      0      0
## 2      0      0      0      0
## 3      0      0      0      0
## 4      0      0      0      0
## 5      0      0      0      0
## 6      0      0      0      0
## word_freq_labs word_freq_telnet word_freq_857 word_freq_data
## 1      0      0      0      0
## 2      0      0      0      0
## 3      0      0      0      0
## 4      0      0      0      0

```

```

## 5          0          0          0          0
## 6          0          0          0          0
## word_freq_415 word_freq_85 word_freq_technology word_freq_1999
## 1          0          0          0          0.00
## 2          0          0          0          0.07
## 3          0          0          0          0.00
## 4          0          0          0          0.00
## 5          0          0          0          0.00
## 6          0          0          0          0.00
## word_freq_parts word_freq_pm word_freq_direct word_freq_cs
## 1          0          0          0.00          0
## 2          0          0          0.00          0
## 3          0          0          0.06          0
## 4          0          0          0.00          0
## 5          0          0          0.00          0
## 6          0          0          0.00          0
## word_freq_meeting word_freq_original word_freq_project word_freq_re
## 1          0          0.00          0          0.00
## 2          0          0.00          0          0.00
## 3          0          0.12          0          0.06
## 4          0          0.00          0          0.00
## 5          0          0.00          0          0.00
## 6          0          0.00          0          0.00
## word_freq_edu word_freq_table word_freq_conference char_freq_;
## 1          0.00          0          0          0.00
## 2          0.00          0          0          0.00
## 3          0.06          0          0          0.01
## 4          0.00          0          0          0.00
## 5          0.00          0          0          0.00
## 6          0.00          0          0          0.00
## char_freq_( char_freq_[ char_freq_! char_freq_$ char_freq_#
## 1          0.000          0          0.778          0.000          0.000
## 2          0.132          0          0.372          0.180          0.048
## 3          0.143          0          0.276          0.184          0.010
## 4          0.137          0          0.137          0.000          0.000
## 5          0.135          0          0.135          0.000          0.000
## 6          0.223          0          0.000          0.000          0.000
## capital_run_length_average capital_run_length_longest
## 1          3.756          61
## 2          5.114          101
## 3          9.821          485
## 4          3.537          40
## 5          3.537          40
## 6          3.000          15
## capital_run_length_total y
## 1          278 1
## 2          1028 1
## 3          2259 1
## 4          191 1
## 5          191 1
## 6          54 1

```

```
dim(spam)
```

```
## [1] 4601 58
```

L'analyse descriptive des variables donne le résultat suivant :

```
data = spam[,1:57]
head(summary(data))
```

```
## word_freq_make      word_freq_address word_freq_all
## "Min.      :0.0000 " "Min.      : 0.000 " "Min.      :0.0000 "
## "1st Qu.:0.0000 " "1st Qu.: 0.000 " "1st Qu.:0.0000 "
## "Median :0.0000 " "Median : 0.000 " "Median :0.0000 "
## "Mean      :0.1046 " "Mean      : 0.213 " "Mean      :0.2807 "
## "3rd Qu.:0.0000 " "3rd Qu.: 0.000 " "3rd Qu.:0.4200 "
## "Max.      :4.5400 " "Max.      :14.280 " "Max.      :5.1000 "
## word_freq_3d         word_freq_our      word_freq_over
## "Min.      : 0.00000 " "Min.      : 0.0000 " "Min.      :0.0000 "
## "1st Qu.: 0.00000 " "1st Qu.: 0.0000 " "1st Qu.:0.0000 "
## "Median : 0.00000 " "Median : 0.0000 " "Median :0.0000 "
## "Mean      : 0.06542 " "Mean      : 0.3122 " "Mean      :0.0959 "
## "3rd Qu.: 0.00000 " "3rd Qu.: 0.3800 " "3rd Qu.:0.0000 "
## "Max.      :42.81000 " "Max.      :10.0000 " "Max.      :5.8800 "
## word_freq_remove    word_freq_internet word_freq_order
## "Min.      :0.0000 " "Min.      : 0.0000 " "Min.      :0.00000 "
## "1st Qu.:0.0000 " "1st Qu.: 0.0000 " "1st Qu.:0.00000 "
## "Median :0.0000 " "Median : 0.0000 " "Median :0.00000 "
## "Mean      :0.1142 " "Mean      : 0.1053 " "Mean      :0.09007 "
## "3rd Qu.:0.0000 " "3rd Qu.: 0.0000 " "3rd Qu.:0.00000 "
## "Max.      :7.2700 " "Max.      :11.1100 " "Max.      :5.26000 "
## word_freq_mail      word_freq_receive  word_freq_will
## "Min.      : 0.0000 " "Min.      :0.00000 " "Min.      :0.0000 "
## "1st Qu.: 0.0000 " "1st Qu.:0.00000 " "1st Qu.:0.0000 "
## "Median : 0.0000 " "Median :0.00000 " "Median :0.1000 "
## "Mean      : 0.2394 " "Mean      :0.05982 " "Mean      :0.5417 "
## "3rd Qu.: 0.1600 " "3rd Qu.:0.00000 " "3rd Qu.:0.8000 "
## "Max.      :18.1800 " "Max.      :2.61000 " "Max.      :9.6700 "
## word_freq_people    word_freq_report   word_freq_addresses
## "Min.      :0.00000 " "Min.      : 0.00000 " "Min.      :0.0000 "
## "1st Qu.:0.00000 " "1st Qu.: 0.00000 " "1st Qu.:0.0000 "
## "Median :0.00000 " "Median : 0.00000 " "Median :0.0000 "
## "Mean      :0.09393 " "Mean      : 0.05863 " "Mean      :0.0492 "
## "3rd Qu.:0.00000 " "3rd Qu.: 0.00000 " "3rd Qu.:0.0000 "
## "Max.      :5.55000 " "Max.      :10.00000 " "Max.      :4.4100 "
## word_freq_free      word_freq_business word_freq_email
## "Min.      : 0.0000 " "Min.      :0.0000 " "Min.      :0.0000 "
## "1st Qu.: 0.0000 " "1st Qu.:0.0000 " "1st Qu.:0.0000 "
## "Median : 0.0000 " "Median :0.0000 " "Median :0.0000 "
## "Mean      : 0.2488 " "Mean      :0.1426 " "Mean      :0.1847 "
## "3rd Qu.: 0.1000 " "3rd Qu.:0.0000 " "3rd Qu.:0.0000 "
## "Max.      :20.0000 " "Max.      :7.1400 " "Max.      :9.0900 "
## word_freq_you       word_freq_credit   word_freq_your
## "Min.      : 0.000 " "Min.      : 0.00000 " "Min.      : 0.0000 "
## "1st Qu.: 0.000 " "1st Qu.: 0.00000 " "1st Qu.: 0.0000 "
## "Median : 1.310 " "Median : 0.00000 " "Median : 0.2200 "
```

```

## "Mean : 1.662 " "Mean : 0.08558 " "Mean : 0.8098 "
## "3rd Qu.: 2.640 " "3rd Qu.: 0.00000 " "3rd Qu.: 1.2700 "
## "Max. :18.750 " "Max. :18.18000 " "Max. :11.1100 "
## word_freq_font word_freq_000 word_freq_money
## "Min. : 0.0000 " "Min. :0.0000 " "Min. : 0.00000 "
## "1st Qu.: 0.0000 " "1st Qu.:0.0000 " "1st Qu.: 0.00000 "
## "Median : 0.0000 " "Median :0.0000 " "Median : 0.00000 "
## "Mean : 0.1212 " "Mean :0.1016 " "Mean : 0.09427 "
## "3rd Qu.: 0.0000 " "3rd Qu.:0.0000 " "3rd Qu.: 0.00000 "
## "Max. :17.1000 " "Max. :5.4500 " "Max. :12.50000 "
## word_freq_hp word_freq_hpl word_freq_george
## "Min. : 0.0000 " "Min. : 0.0000 " "Min. : 0.0000 "
## "1st Qu.: 0.0000 " "1st Qu.: 0.0000 " "1st Qu.: 0.0000 "
## "Median : 0.0000 " "Median : 0.0000 " "Median : 0.0000 "
## "Mean : 0.5495 " "Mean : 0.2654 " "Mean : 0.7673 "
## "3rd Qu.: 0.0000 " "3rd Qu.: 0.0000 " "3rd Qu.: 0.0000 "
## "Max. :20.8300 " "Max. :16.6600 " "Max. :33.3300 "
## word_freq_650 word_freq_lab word_freq_labs
## "Min. :0.0000 " "Min. : 0.00000 " "Min. :0.0000 "
## "1st Qu.:0.0000 " "1st Qu.: 0.00000 " "1st Qu.:0.0000 "
## "Median :0.0000 " "Median : 0.00000 " "Median :0.0000 "
## "Mean :0.1248 " "Mean : 0.09892 " "Mean :0.1029 "
## "3rd Qu.:0.0000 " "3rd Qu.: 0.00000 " "3rd Qu.:0.0000 "
## "Max. :9.0900 " "Max. :14.28000 " "Max. :5.8800 "
## word_freq_telnet word_freq_857 word_freq_data
## "Min. : 0.00000 " "Min. :0.00000 " "Min. : 0.00000 "
## "1st Qu.: 0.00000 " "1st Qu.:0.00000 " "1st Qu.: 0.00000 "
## "Median : 0.00000 " "Median :0.00000 " "Median : 0.00000 "
## "Mean : 0.06475 " "Mean :0.04705 " "Mean : 0.09723 "
## "3rd Qu.: 0.00000 " "3rd Qu.:0.00000 " "3rd Qu.: 0.00000 "
## "Max. :12.50000 " "Max. :4.76000 " "Max. :18.18000 "
## word_freq_415 word_freq_85 word_freq_technology
## "Min. :0.00000 " "Min. : 0.0000 " "Min. :0.00000 "
## "1st Qu.:0.00000 " "1st Qu.: 0.0000 " "1st Qu.:0.00000 "
## "Median :0.00000 " "Median : 0.0000 " "Median :0.00000 "
## "Mean :0.04784 " "Mean : 0.1054 " "Mean :0.09748 "
## "3rd Qu.:0.00000 " "3rd Qu.: 0.0000 " "3rd Qu.:0.00000 "
## "Max. :4.76000 " "Max. :20.0000 " "Max. :7.69000 "
## word_freq_1999 word_freq_parts word_freq_pm
## "Min. :0.000 " "Min. :0.0000 " "Min. : 0.00000 "
## "1st Qu.:0.000 " "1st Qu.:0.0000 " "1st Qu.: 0.00000 "
## "Median :0.000 " "Median :0.0000 " "Median : 0.00000 "
## "Mean :0.137 " "Mean :0.0132 " "Mean : 0.07863 "
## "3rd Qu.:0.000 " "3rd Qu.:0.0000 " "3rd Qu.: 0.00000 "
## "Max. :6.890 " "Max. :8.3300 " "Max. :11.11000 "
## word_freq_direct word_freq_cs word_freq_meeting
## "Min. :0.00000 " "Min. :0.00000 " "Min. : 0.0000 "
## "1st Qu.:0.00000 " "1st Qu.:0.00000 " "1st Qu.: 0.0000 "
## "Median :0.00000 " "Median :0.00000 " "Median : 0.0000 "
## "Mean :0.06483 " "Mean :0.04367 " "Mean : 0.1323 "
## "3rd Qu.:0.00000 " "3rd Qu.:0.00000 " "3rd Qu.: 0.0000 "
## "Max. :4.76000 " "Max. :7.14000 " "Max. :14.2800 "
## word_freq_original word_freq_project word_freq_re
## "Min. :0.0000 " "Min. : 0.0000 " "Min. : 0.0000 "

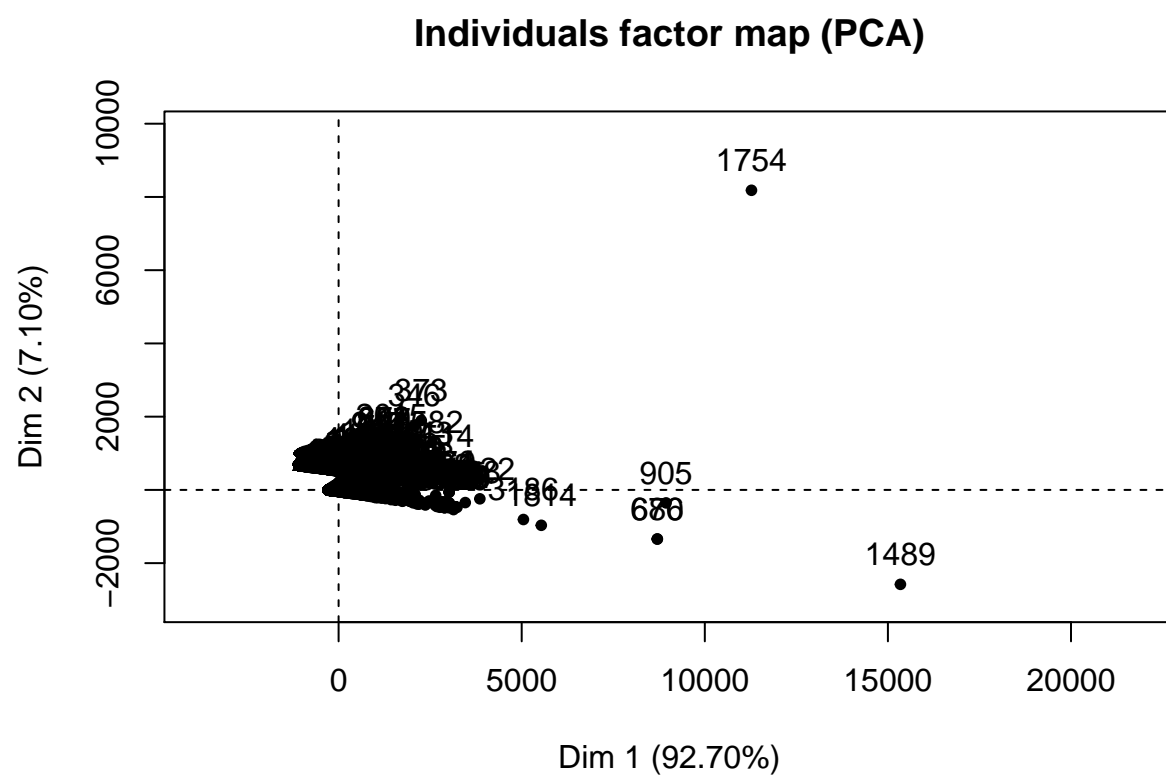
```

```
## "1st Qu.:0.0000 " "1st Qu.: 0.0000 " "1st Qu.: 0.0000 "
## "Median :0.0000 " "Median : 0.0000 " "Median : 0.0000 "
## "Mean :0.0461 " "Mean : 0.0792 " "Mean : 0.3012 "
## "3rd Qu.:0.0000 " "3rd Qu.: 0.0000 " "3rd Qu.: 0.1100 "
## "Max. :3.5700 " "Max. :20.0000 " "Max. :21.4200 "
## word_freq_edu word_freq_table word_freq_conference
## "Min. : 0.0000 " "Min. :0.000000 " "Min. : 0.00000 "
## "1st Qu.: 0.0000 " "1st Qu.:0.000000 " "1st Qu.: 0.00000 "
## "Median : 0.0000 " "Median :0.000000 " "Median : 0.00000 "
## "Mean : 0.1798 " "Mean :0.005444 " "Mean : 0.03187 "
## "3rd Qu.: 0.0000 " "3rd Qu.:0.000000 " "3rd Qu.: 0.00000 "
## "Max. :22.0500 " "Max. :2.170000 " "Max. :10.00000 "
## char_freq_; char_freq_( char_freq_[
## "Min. :0.00000 " "Min. :0.000 " "Min. :0.00000 "
## "1st Qu.:0.00000 " "1st Qu.:0.000 " "1st Qu.:0.00000 "
## "Median :0.00000 " "Median :0.065 " "Median :0.00000 "
## "Mean :0.03857 " "Mean :0.139 " "Mean :0.01698 "
## "3rd Qu.:0.00000 " "3rd Qu.:0.188 " "3rd Qu.:0.00000 "
## "Max. :4.38500 " "Max. :9.752 " "Max. :4.08100 "
## char_freq_! char_freq_$ char_freq_#
## "Min. : 0.0000 " "Min. :0.00000 " "Min. : 0.00000 "
## "1st Qu.: 0.0000 " "1st Qu.:0.00000 " "1st Qu.: 0.00000 "
## "Median : 0.0000 " "Median :0.00000 " "Median : 0.00000 "
## "Mean : 0.2691 " "Mean :0.07581 " "Mean : 0.04424 "
## "3rd Qu.: 0.3150 " "3rd Qu.:0.05200 " "3rd Qu.: 0.00000 "
## "Max. :32.4780 " "Max. :6.00300 " "Max. :19.82900 "
## capital_run_length_average capital_run_length_longest
## "Min. : 1.000 " "Min. : 1.00 "
## "1st Qu.: 1.588 " "1st Qu.: 6.00 "
## "Median : 2.276 " "Median : 15.00 "
## "Mean : 5.191 " "Mean : 52.17 "
## "3rd Qu.: 3.706 " "3rd Qu.: 43.00 "
## "Max. :1102.500 " "Max. :9989.00 "
## capital_run_length_total
## "Min. : 1.0 "
## "1st Qu.: 35.0 "
## "Median : 95.0 "
## "Mean : 283.3 "
## "3rd Qu.: 266.0 "
## "Max. :15841.0 "
```

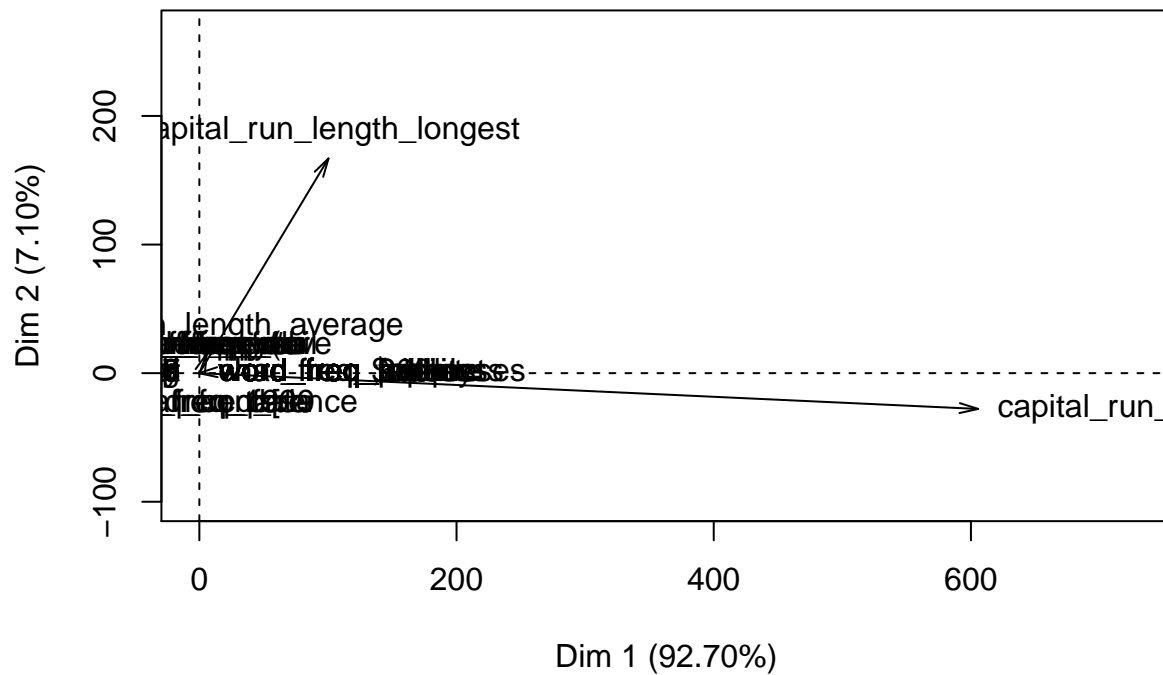
On va utiliser la bibliothèque **FactoMineR** pour effectuer l'ACP :

Une ACP non normée :

```
library(FactoMineR)
pca = PCA(data , scale.unit = F)
```



Variables factor map (PCA)



```
head(summary(pca))
```

```
##
## Call:
## PCA(X = data, scale.unit = F)
##
##
## Eigenvalues
##          Dim.1      Dim.2      Dim.3      Dim.4
## Variance    376923.652  28885.649   749.647   11.408
## % of var.      92.703    7.104    0.184    0.003
## Cumulative % of var.  92.703    99.807    99.991    99.994
##          Dim.5      Dim.6      Dim.7      Dim.8
## Variance      4.134    2.626    1.946    1.637
## % of var.      0.001    0.001    0.000    0.000
## Cumulative % of var.  99.995    99.996    99.996    99.997
##          Dim.9      Dim.10     Dim.11     Dim.12
## Variance      1.334    1.064    1.002    0.856
## % of var.      0.000    0.000    0.000    0.000
## Cumulative % of var.  99.997    99.997    99.998    99.998
##          Dim.13     Dim.14     Dim.15     Dim.16
## Variance      0.829    0.760    0.680    0.613
## % of var.      0.000    0.000    0.000    0.000
## Cumulative % of var.  99.998    99.998    99.998    99.998
##          Dim.17     Dim.18     Dim.19     Dim.20
## Variance      0.829    0.760    0.680    0.613
## % of var.      0.000    0.000    0.000    0.000
## Cumulative % of var.  99.998    99.998    99.998    99.998
```


| | | | | | | |
|-------------------------------|---------|----------|----------|---------|-------|--|
| ## Variance | 0.568 | 0.474 | 0.422 | 0.402 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 99.999 | 99.999 | 99.999 | 99.999 | | |
| ## | Dim.21 | Dim.22 | Dim.23 | Dim.24 | | |
| ## Variance | 0.377 | 0.306 | 0.289 | 0.269 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 99.999 | 99.999 | 99.999 | 99.999 | | |
| ## | Dim.25 | Dim.26 | Dim.27 | Dim.28 | | |
| ## Variance | 0.227 | 0.220 | 0.194 | 0.188 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 99.999 | 99.999 | 99.999 | 99.999 | | |
| ## | Dim.29 | Dim.30 | Dim.31 | Dim.32 | | |
| ## Variance | 0.179 | 0.175 | 0.146 | 0.139 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 99.999 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.33 | Dim.34 | Dim.35 | Dim.36 | | |
| ## Variance | 0.132 | 0.128 | 0.125 | 0.117 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.37 | Dim.38 | Dim.39 | Dim.40 | | |
| ## Variance | 0.111 | 0.108 | 0.085 | 0.082 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.41 | Dim.42 | Dim.43 | Dim.44 | | |
| ## Variance | 0.081 | 0.078 | 0.070 | 0.064 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.45 | Dim.46 | Dim.47 | Dim.48 | | |
| ## Variance | 0.055 | 0.052 | 0.050 | 0.046 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.49 | Dim.50 | Dim.51 | Dim.52 | | |
| ## Variance | 0.045 | 0.042 | 0.040 | 0.038 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.53 | Dim.54 | Dim.55 | Dim.56 | | |
| ## Variance | 0.033 | 0.022 | 0.011 | 0.006 | | |
| ## % of var. | 0.000 | 0.000 | 0.000 | 0.000 | | |
| ## Cumulative % of var. | 100.000 | 100.000 | 100.000 | 100.000 | | |
| ## | Dim.57 | | | | | |
| ## Variance | 0.000 | | | | | |
| ## % of var. | 0.000 | | | | | |
| ## Cumulative % of var. | 100.000 | | | | | |
| ## | | | | | | |
| ## Individuals (the 10 first) | | | | | | |
| ## | | Dist | Dim.1 | ctr | cos2 | |
| ## 1 | | 10.550 | -3.787 | 0.000 | 0.129 | |
| ## 2 | | 746.314 | 742.632 | 0.032 | 0.990 | |
| ## 3 | | 2022.573 | 2019.854 | 0.235 | 0.997 | |
| ## 4 | | 93.129 | -93.048 | 0.000 | 0.998 | |
| ## 5 | | 93.129 | -93.048 | 0.000 | 0.998 | |
| ## 6 | | 232.317 | -232.291 | 0.003 | 1.000 | |
| ## 7 | | 178.000 | -176.887 | 0.002 | 0.988 | |
| ## 8 | | 237.918 | -237.883 | 0.003 | 1.000 | |

```

## 9 | 1049.982 | 1024.854 0.061 0.953 |
## 10 | 465.816 | 457.883 0.012 0.966 |
## Dim.2 ctr cos2 Dim.3 ctr
## 1 9.412 0.000 0.796 | -2.247 0.000
## 2 -74.001 0.004 0.010 | -0.568 0.000
## 3 102.077 0.008 0.003 | -23.777 0.016
## 4 3.009 0.000 0.001 | -1.044 0.000
## 5 3.009 0.000 0.001 | -1.044 0.000
## 6 0.836 0.000 0.000 | -0.049 0.000
## 7 -19.587 0.000 0.012 | -0.099 0.000
## 8 -2.324 0.000 0.000 | -0.266 0.000
## 9 226.877 0.039 0.047 | -25.491 0.019
## 10 -85.593 0.006 0.034 | -0.204 0.000
## cos2
## 1 0.045 |
## 2 0.000 |
## 3 0.000 |
## 4 0.000 |
## 5 0.000 |
## 6 0.000 |
## 7 0.000 |
## 8 0.000 |
## 9 0.001 |
## 10 0.000 |
##
## Variables (the 10 first)
## Dim.1 ctr cos2 Dim.2 ctr cos2
## word_freq_make | 0.028 0.000 0.008 | 0.005 0.000 0.000 |
## word_freq_address | -0.028 0.000 0.000 | 0.018 0.000 0.000 |
## word_freq_all | 0.037 0.000 0.005 | 0.041 0.000 0.007 |
## word_freq_3d | 0.031 0.000 0.000 | 0.017 0.000 0.000 |
## word_freq_our | 0.003 0.000 0.000 | 0.039 0.000 0.003 |
## word_freq_over | 0.023 0.000 0.007 | 0.015 0.000 0.003 |
## word_freq_remove | -0.002 0.000 0.000 | 0.028 0.000 0.005 |
## word_freq_internet | 0.017 0.000 0.002 | 0.008 0.000 0.000 |
## word_freq_order | 0.070 0.000 0.064 | 0.019 0.000 0.005 |
## word_freq_mail | 0.058 0.000 0.008 | 0.043 0.000 0.004 |
## Dim.3 ctr cos2
## word_freq_make 0.007 0.000 0.001 |
## word_freq_address 0.000 0.000 0.000 |
## word_freq_all 0.026 0.000 0.003 |
## word_freq_3d -0.008 0.000 0.000 |
## word_freq_our 0.019 0.000 0.001 |
## word_freq_over -0.016 0.000 0.004 |
## word_freq_remove 0.004 0.000 0.000 |
## word_freq_internet -0.003 0.000 0.000 |
## word_freq_order 0.010 0.000 0.001 |
## word_freq_mail 0.019 0.000 0.001 |

```

```
## NULL
```

Interpretation :

Variables :

```
df = as.data.frame(pca$var$contrib)
head(df[order(df$Dim.1 , decreasing = T),])
```

```
##                Dim.1        Dim.2        Dim.3
## capital_run_length_total  9.731493e+01 2.682320e+00 2.709562e-03
## capital_run_length_longest 2.675963e+00 9.652560e+01 7.982684e-01
## capital_run_length_average 9.059934e-03 7.919517e-01 9.919725e+01
## word_freq_george         2.823689e-05 1.102488e-06 5.507539e-05
## word_freq_font           2.944941e-06 3.081760e-06 5.152856e-05
## word_freq_re             2.485624e-06 3.412451e-08 1.152040e-05
##                Dim.4        Dim.5
## capital_run_length_total  2.563382e-05 2.712709e-08
## capital_run_length_longest 5.075434e-06 4.469053e-05
## capital_run_length_average 1.954353e-05 3.360502e-04
## word_freq_george         9.784176e+01 1.132735e+00
## word_freq_font           2.302156e-03 4.632671e-04
## word_freq_re             4.360961e-03 1.840655e-02
```

La variable qui contribue le plus à la formation de l'axe 1 est : **capital_run_length_total** 97.3% , **capital_run_length_longest** 2.67%

```
head(df[order(df$Dim.2 , decreasing = T),])
```

```
##                Dim.1        Dim.2        Dim.3
## capital_run_length_longest 2.675963e+00 9.652560e+01 7.982684e-01
## capital_run_length_total  9.731493e+01 2.682320e+00 2.709562e-03
## capital_run_length_average 9.059934e-03 7.919517e-01 9.919725e+01
## char_freq_(               3.205798e-07 3.150584e-05 3.275013e-04
## word_freq_your            1.153215e-06 2.209680e-05 1.677025e-07
## word_freq_hp             1.488300e-06 1.033634e-05 2.198292e-05
##                Dim.4        Dim.5
## capital_run_length_longest 5.075434e-06 4.469053e-05
## capital_run_length_total  2.563382e-05 2.712709e-08
## capital_run_length_average 1.954353e-05 3.360502e-04
## char_freq_(               1.834514e-07 1.062297e-01
## word_freq_your            3.358894e-01 5.927688e+00
## word_freq_hp             2.810181e-03 3.948854e+01
```

La variable qui contribue le plus à la formation de l'axe 2 est : **capital_run_length_longest** 96.5% , **capital_run_length_total** 2.68%

Individus :

```
df = as.data.frame(pca$ind$contrib)
head(df[order(df$Dim.1 , decreasing = T),])
```

| ## | Dim.1 | Dim.2 | Dim.3 | Dim.4 | Dim.5 |
|---------|-----------|-------------|-------------|------------|--------------|
| ## 1489 | 13.572976 | 5.00414575 | 0.19009575 | 0.09942892 | 0.0005325704 |
| ## 1754 | 7.329447 | 50.37152944 | 10.55915067 | 0.08762297 | 0.2962729524 |
| ## 905 | 4.611469 | 0.09452977 | 0.06301744 | 0.02977187 | 0.0012335745 |
| ## 680 | 4.368939 | 1.35366149 | 0.06707356 | 0.03031312 | 0.0055186449 |
| ## 676 | 4.366959 | 1.35299019 | 0.06720291 | 0.03029929 | 0.0055203258 |
| ## 1814 | 1.766585 | 0.70318042 | 0.02925296 | 0.01066288 | 0.0053578463 |

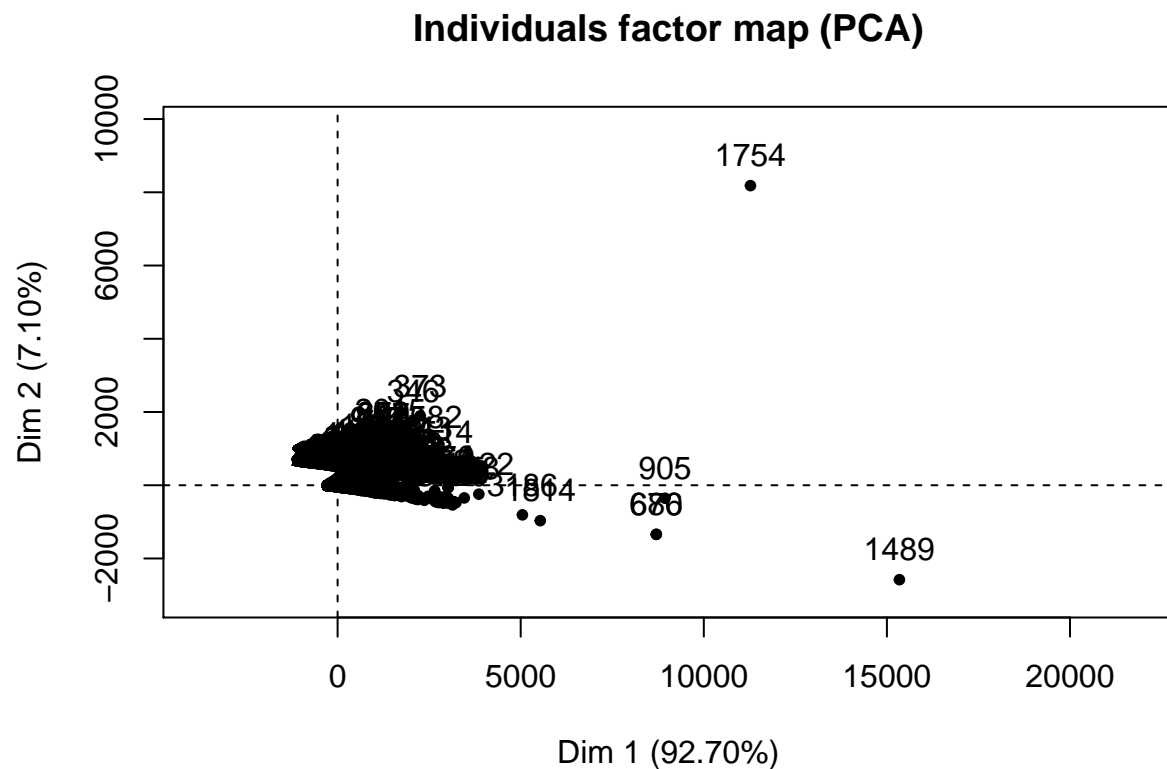
Les individus qui contribuent le plus à la formation de l'axe 1 sont : **1489** 13% et **1754** 7.32% et **905** , **680** , **676** 4%

```
df = as.data.frame(pca$ind$contrib)
head(df[order(df$Dim.2 , decreasing = T),])
```

| ## | Dim.1 | Dim.2 | Dim.3 | Dim.4 | Dim.5 |
|---------|------------|-----------|-------------|-------------|--------------|
| ## 1754 | 7.3294474 | 50.371529 | 10.55915067 | 0.087622968 | 2.962730e-01 |
| ## 1489 | 13.5729764 | 5.004146 | 0.19009575 | 0.099428921 | 5.325704e-04 |
| ## 373 | 0.2940446 | 2.707792 | 24.04310664 | 0.003644853 | 5.519886e-05 |
| ## 346 | 0.2473419 | 2.323303 | 20.63327151 | 0.002862750 | 3.148129e-04 |
| ## 680 | 4.3689388 | 1.353661 | 0.06707356 | 0.030313124 | 5.518645e-03 |
| ## 676 | 4.3669590 | 1.352990 | 0.06720291 | 0.030299291 | 5.520326e-03 |

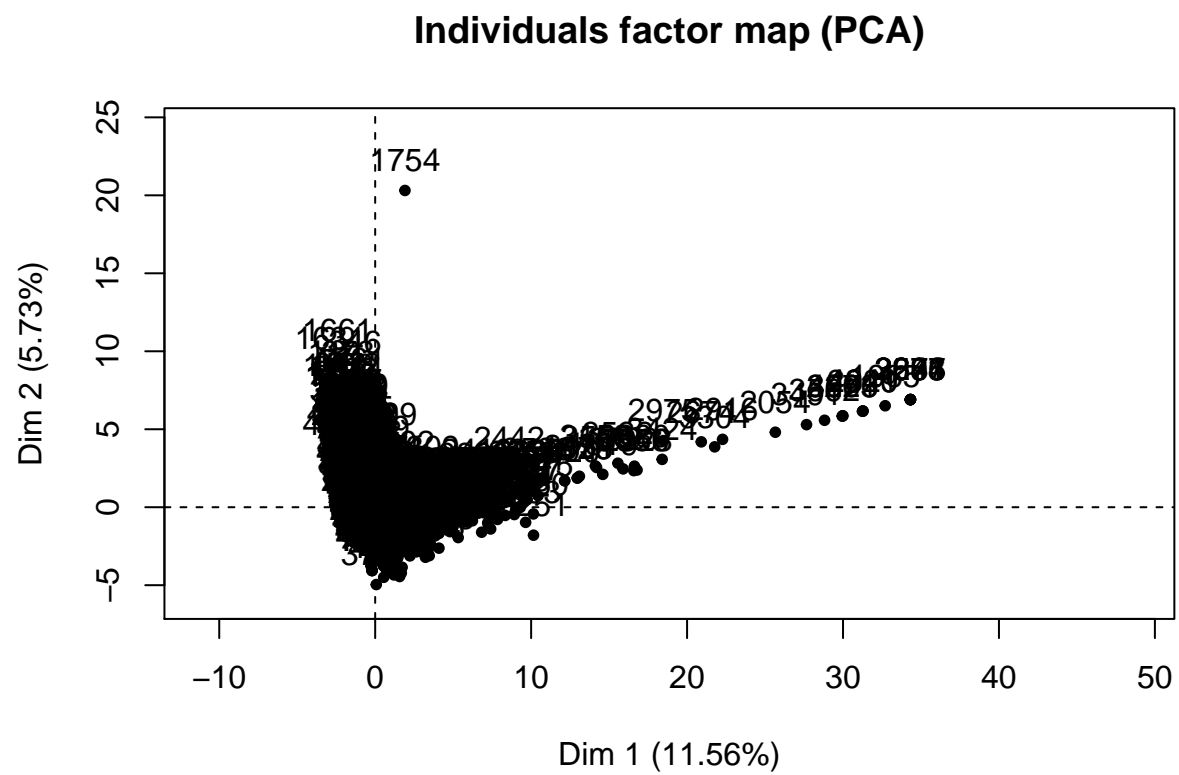
Les individus qui contribuent le plus à la formation de l'axe 2 sont : **1754** 50.37% et **1489** 5%

Une représentation graphique de l'ACP non normée :

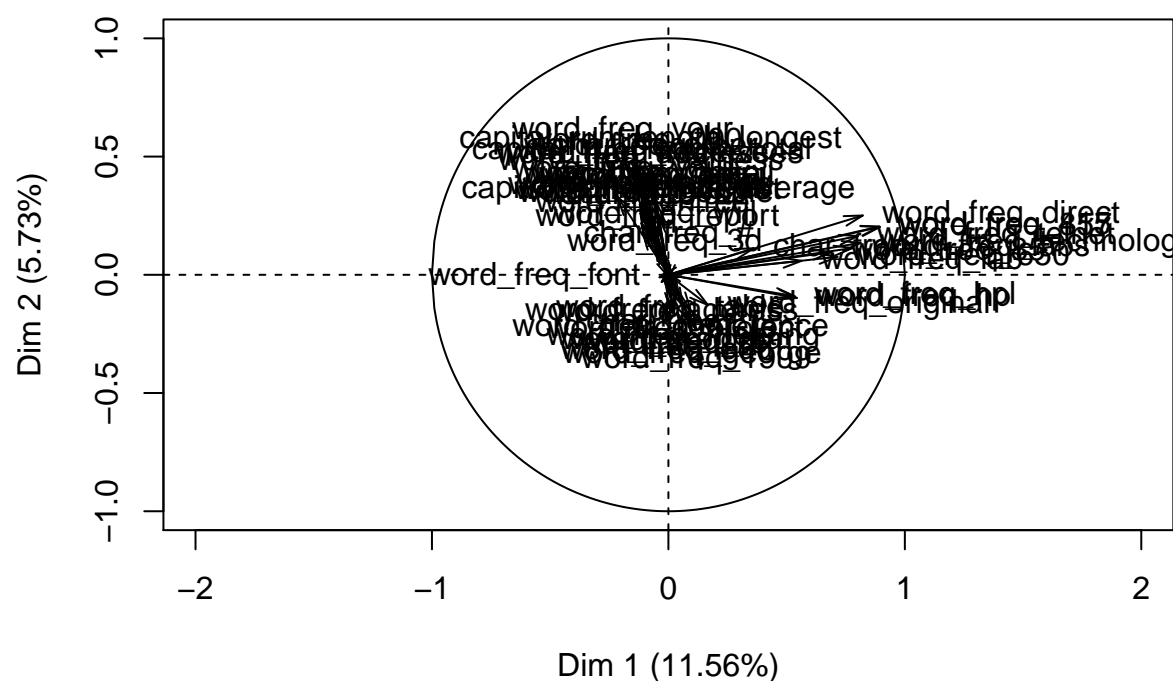


Une ACP normée :

```
acp.n = PCA(data,scale.unit =TRUE)
```



Variables factor map (PCA)



```
summary(acp.n)
```

```
##
## Call:
## PCA(X = data, scale.unit = TRUE)
##
##
## Eigenvalues
##
```

| | Dim.1 | Dim.2 | Dim.3 | Dim.4 | Dim.5 | Dim.6 |
|-------------------------|--------|--------|--------|--------|--------|--------|
| ## Variance | 6.592 | 3.267 | 2.003 | 1.613 | 1.546 | 1.463 |
| ## % of var. | 11.565 | 5.732 | 3.514 | 2.830 | 2.713 | 2.566 |
| ## Cumulative % of var. | 11.565 | 17.297 | 20.811 | 23.642 | 26.354 | 28.920 |

```
##
```

| | Dim.7 | Dim.8 | Dim.9 | Dim.10 | Dim.11 | Dim.12 |
|-------------------------|--------|--------|--------|--------|--------|--------|
| ## Variance | 1.414 | 1.375 | 1.295 | 1.277 | 1.217 | 1.130 |
| ## % of var. | 2.481 | 2.412 | 2.272 | 2.240 | 2.135 | 1.983 |
| ## Cumulative % of var. | 31.401 | 33.813 | 36.085 | 38.326 | 40.460 | 42.443 |

```
##
```

| | Dim.13 | Dim.14 | Dim.15 | Dim.16 | Dim.17 | Dim.18 |
|-------------------------|--------|--------|--------|--------|--------|--------|
| ## Variance | 1.112 | 1.095 | 1.087 | 1.063 | 1.049 | 1.023 |
| ## % of var. | 1.950 | 1.921 | 1.907 | 1.866 | 1.840 | 1.795 |
| ## Cumulative % of var. | 44.394 | 46.315 | 48.222 | 50.088 | 51.927 | 53.723 |

```
##
```

| | Dim.19 | Dim.20 | Dim.21 | Dim.22 | Dim.23 | Dim.24 |
|-------------------------|--------|--------|--------|--------|--------|--------|
| ## Variance | 1.013 | 1.003 | 0.996 | 0.978 | 0.965 | 0.941 |
| ## % of var. | 1.777 | 1.759 | 1.747 | 1.716 | 1.692 | 1.651 |
| ## Cumulative % of var. | 55.499 | 57.259 | 59.006 | 60.722 | 62.414 | 64.066 |

```
##
```

| | Dim.25 | Dim.26 | Dim.27 | Dim.28 | Dim.29 | Dim.30 |
|-------------------------|--------|--------|--------|--------|--------|--------|
| ## Variance | 1.013 | 1.003 | 0.996 | 0.978 | 0.965 | 0.941 |
| ## % of var. | 1.777 | 1.759 | 1.747 | 1.716 | 1.692 | 1.651 |
| ## Cumulative % of var. | 55.499 | 57.259 | 59.006 | 60.722 | 62.414 | 64.066 |

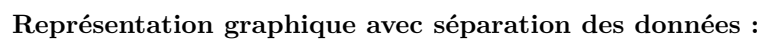
```

## Variance          0.937   0.924   0.915   0.905   0.873   0.866
## % of var.        1.643   1.622   1.606   1.587   1.532   1.519
## Cumulative % of var. 65.709 67.331 68.936 70.524 72.056 73.575
## Dim.31 Dim.32 Dim.33 Dim.34 Dim.35 Dim.36
## Variance          0.836   0.827   0.798   0.782   0.777   0.756
## % of var.        1.468   1.450   1.399   1.371   1.362   1.325
## Cumulative % of var. 75.043 76.493 77.893 79.264 80.626 81.952
## Dim.37 Dim.38 Dim.39 Dim.40 Dim.41 Dim.42
## Variance          0.734   0.723   0.705   0.690   0.675   0.666
## % of var.        1.288   1.269   1.236   1.210   1.184   1.168
## Cumulative % of var. 83.240 84.508 85.744 86.955 88.138 89.307
## Dim.43 Dim.44 Dim.45 Dim.46 Dim.47 Dim.48
## Variance          0.619   0.608   0.582   0.577   0.524   0.489
## % of var.        1.086   1.067   1.021   1.012   0.920   0.857
## Cumulative % of var. 90.393 91.460 92.481 93.494 94.414 95.271
## Dim.49 Dim.50 Dim.51 Dim.52 Dim.53 Dim.54
## Variance          0.450   0.409   0.376   0.366   0.335   0.305
## % of var.        0.790   0.718   0.659   0.642   0.587   0.536
## Cumulative % of var. 96.061 96.779 97.438 98.079 98.667 99.203
## Dim.55 Dim.56 Dim.57
## Variance          0.260   0.190   0.004
## % of var.        0.457   0.334   0.007
## Cumulative % of var. 99.659 99.993 100.000
##
## Individuals (the 10 first)
##
##          Dist   Dim.1   ctr   cos2   Dim.2   ctr
## 1 | 2.826 | -0.732 0.002 0.067 | -0.043 0.000
## 2 | 3.512 | -1.185 0.005 0.114 | 2.068 0.028
## 3 | 9.018 | -1.468 0.007 0.026 | 5.024 0.168
## 4 | 2.910 | -0.805 0.002 0.077 | 0.428 0.001
## 5 | 2.910 | -0.806 0.002 0.077 | 0.427 0.001
## 6 | 5.392 | -0.493 0.001 0.008 | -0.490 0.002
## 7 | 5.635 | -1.026 0.003 0.033 | 1.021 0.007
## 8 | 5.470 | -0.504 0.001 0.009 | -0.488 0.002
## 9 | 8.901 | -1.266 0.005 0.020 | 3.641 0.088
## 10 | 2.227 | -0.841 0.002 0.143 | 0.401 0.001
##          cos2   Dim.3   ctr   cos2
## 1 0.000 | -0.581 0.004 0.042 |
## 2 0.347 | 0.036 0.000 0.000 |
## 3 0.310 | 3.278 0.117 0.132 |
## 4 0.022 | -0.583 0.004 0.040 |
## 5 0.022 | -0.585 0.004 0.040 |
## 6 0.008 | -0.373 0.002 0.005 |
## 7 0.033 | -1.738 0.033 0.095 |
## 8 0.008 | -0.419 0.002 0.006 |
## 9 0.167 | 1.492 0.024 0.028 |
## 10 0.032 | 0.030 0.000 0.000 |
##
## Variables (the 10 first)
##
##          Dim.1   ctr   cos2   Dim.2   ctr   cos2
## word_freq_make | -0.112 0.191 0.013 | 0.307 2.877 0.094 |
## word_freq_address | -0.029 0.012 0.001 | -0.030 0.028 0.001 |
## word_freq_all | -0.121 0.222 0.015 | 0.299 2.729 0.089 |
## word_freq_3d | -0.016 0.004 0.000 | 0.020 0.012 0.000 |

```

Une représentation graphique de l'ACP normée :

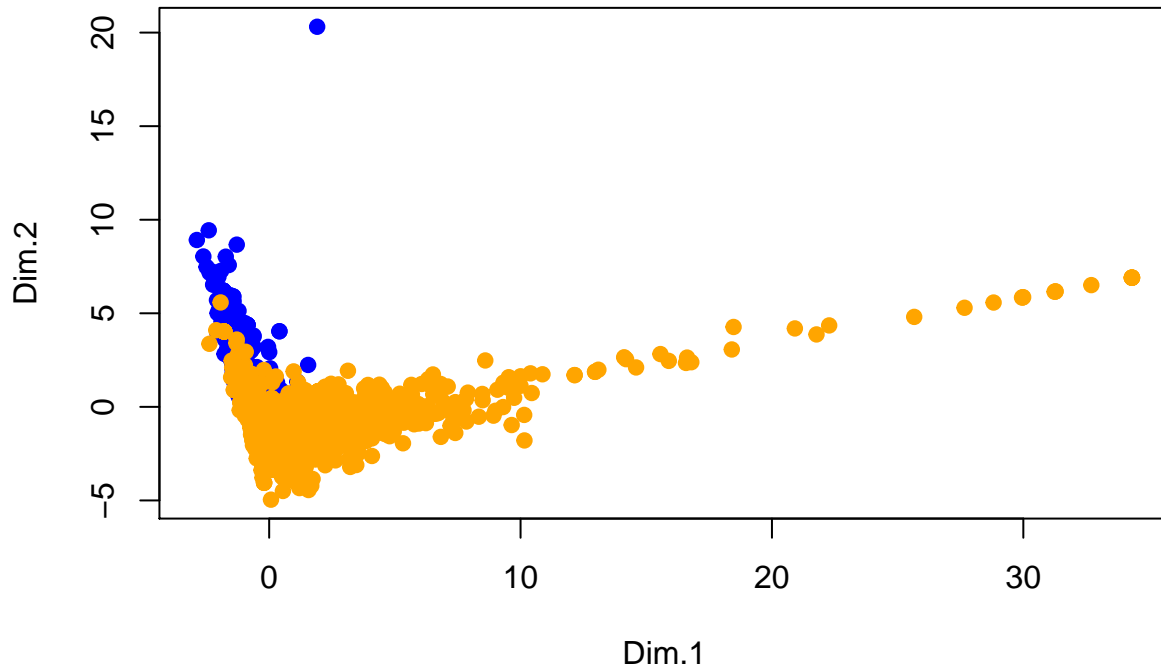
Individuals factor map (PCA)




```

coord <- acp.n$ind$coord
col = rep("blue",nrow(data))
col[spam$y == 0] = "orange"
plot(coord[,1:2],col=col , pch = 19)

```



recodage en forme de facteurs :

```

make=factor(spam[, "word_freq_make"] > 0, c(TRUE, FALSE), labels=c("make", "Nmk"))
table(make)

```

```

## make
## make  Nmk
## 1053 3548

```

```

CapLMq=cut(spam[, "capital_run_length_total"], breaks=quantile(spam[, "capital_run_length_total"], probs =
labels = c("Mm1", "Mm2", "Mm3"), include.lowest = TRUE)
table(CapLMq)

```

```

## CapLMq
## Mm1  Mm2  Mm3
## 1537 1530 1534

```

```
table(make, CapLMq)
```

```
##           CapLMq
## make      Mm1  Mm2  Mm3
##   make      75  264  714
##   Nmk  1462 1266  820
```

```
data = cbind(as.numeric(make), as.numeric(CapLMq))
names(data) = c("make", "CapLMq")
```

Analayse descriptive :

```
summary(data)
```

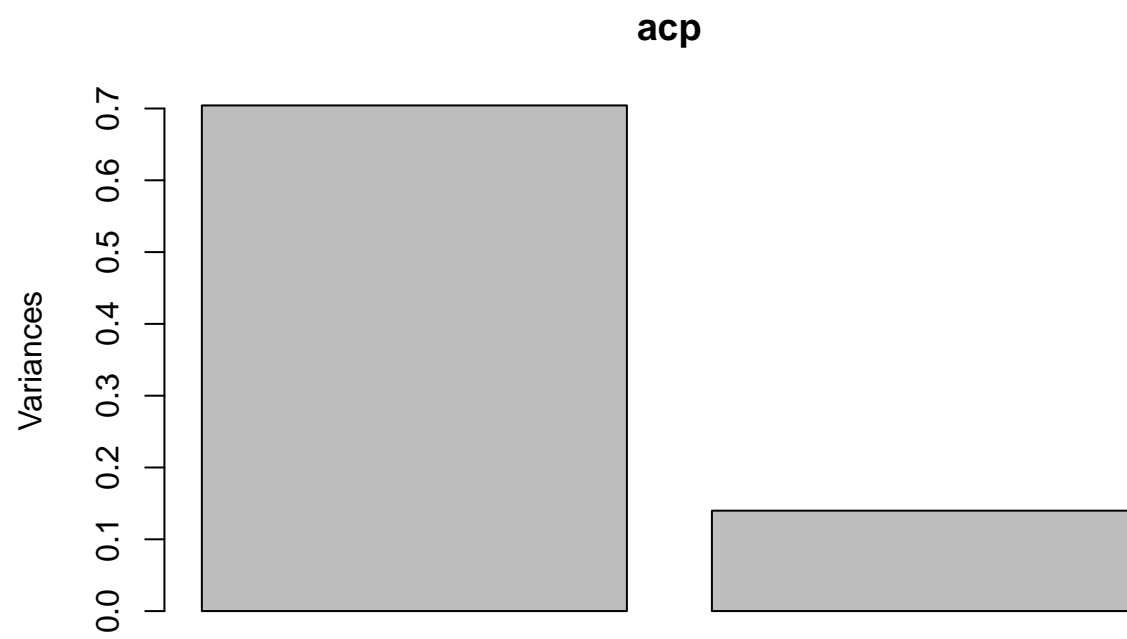
```
##           V1           V2
## Min.      :1.000   Min.      :1.000
## 1st Qu.:2.000   1st Qu.:1.000
## Median :2.000   Median :2.000
## Mean    :1.771   Mean    :1.999
## 3rd Qu.:2.000   3rd Qu.:3.000
## Max.    :2.000   Max.    :3.000
```

```
acp = prcomp(data)
```

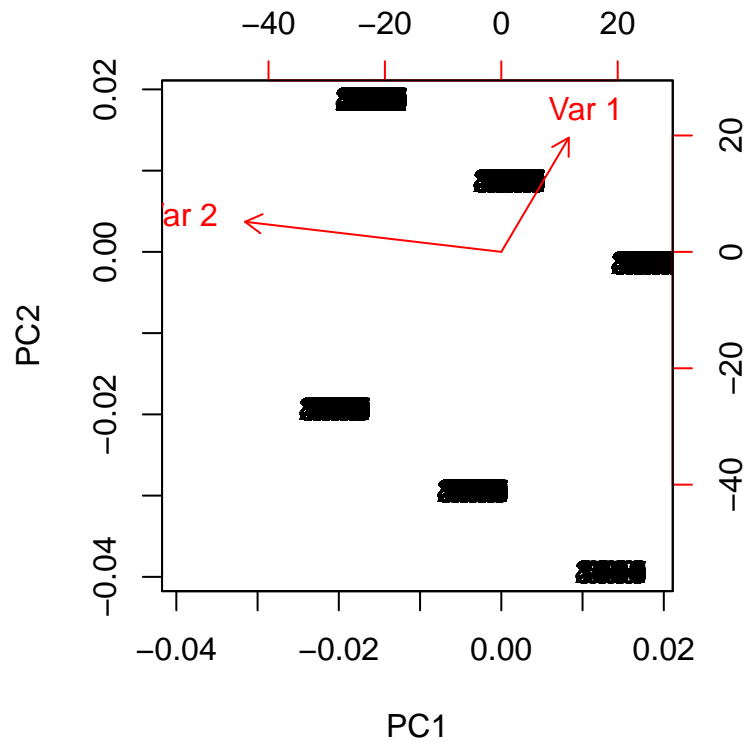
```
summary(acp)
```

```
## Importance of components:
##               PC1    PC2
## Standard deviation    0.8392 0.3740
## Proportion of Variance 0.8343 0.1657
## Cumulative Proportion 0.8343 1.0000
```

```
plot(acp)
```



```
biplot(acp)
```



Générateur Aléatoire de visages

```
setwd("C:\\Users\\W 7\\Desktop\\Master 2\\AA2\\TP2_ACP\\Images\\")
n = 10;
img = list()
names = c('img1.dat', 'img2.dat', 'img3.dat', 'img4.dat', 'img5.dat', 'img6.dat', 'img7.dat', 'img8.dat', 'img9')

for (i in 1:n) {
  aux = t(as.matrix(read.table(names[i], sep=",", as.is=TRUE)))
  img[[i]] = aux[,112:1]
}

N1 = dim(img[[1]])[1]
N2 = dim(img[[1]])[2]

for (i in 1:n) {
  image(img[[i]], col=gray(0:255/256), xlab="", ylab="", axes=FALSE)
}
```





















Image moyenne :

```
# Image moyenne
moy = matrix(0,N1,N2)

for (i in 1:n) {
  moy = moy + img[[i]]
}
moy = moy/n

image(moy, col=gray(0:256/256) , xlab="", ylab="",axes=FALSE)
```



```
# Matrice de donnees
X = matrix(0,n,N1*N2)
for (i in 1:n) {
  X[i,] = as.vector(img[[i]]-moy)
}
```

Génération Aléatoire :

On multiplie la matrice des composantes principales par la transposé de la matrice de corrélation, on obtient une matrice qui représente les images à l'aide des 2 premières composantes

```
acp = prcomp(X)
mat = acp$x[,1:10]%*%t(acp$rotation[,1:10])
mat = scale(mat, center = -1*moy, scale=FALSE)
for (i in 1:5){
  img <- matrix((mat[i,]), 92, 112)
  image(1:92, 1:112,img,col=gray((0:256)/256) , xlab = "" , ylab = "" , axes =F)
}
```







