

# Daniel Saldaña

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Parte 1

```
M = read.csv('países_mundo.csv')
```

```
cov(M)
```

```
##          CrecPobl      MortInf  PorcMujeres      PNB95
## CrecPobl  1.538298e+00  2.195026e+01 -6.078026e+00 -8.933379e+04
## MortInf   2.195026e+01  1.032859e+03 -9.249342e+00 -2.269332e+06
## PorcMujeres -6.078026e+00 -9.249342e+00  7.698322e+01  2.813114e+05
## PNB95      -8.933379e+04 -2.269332e+06  2.813114e+05  4.999786e+10
## ProdElec   -4.973964e+04 -1.043435e+06  2.260248e+05  2.247791e+10
## LinTelf     -1.369079e+02 -4.381366e+03  4.499750e+02  2.039550e+07
## ConsAgua    -4.827092e+01 -1.288211e+03 -1.568313e+03  1.097481e+07
## PropBosq    -3.887018e+00 -1.466316e+01  6.517895e+01  2.474311e+05
## PropDeform  3.361974e-01  1.276296e+01  2.680592e-01 -5.806203e+04
## ConsEner    -8.384169e+02 -4.442568e+04  2.855207e+02  1.415628e+08
## EmisCO2     -1.137877e+00 -9.485500e+01 -2.150132e+00  2.501673e+05
##          ProdElec      LinTelf      ConsAgua      PropBosq
## CrecPobl  -4.973964e+04 -1.369079e+02 -4.827092e+01 -3.887018
## MortInf   -1.043435e+06 -4.381366e+03 -1.288211e+03 -14.663158
## PorcMujeres 2.260248e+05 4.499750e+02 -1.568313e+03  65.178947
## PNB95      2.247791e+10 2.039550e+07  1.097481e+07 247431.122807
## ProdElec   1.821909e+10 7.583050e+06  1.399817e+07 70359.785965
## LinTelf     7.583050e+06 3.841247e+04  1.193110e+04 248.715789
## ConsAgua    1.399817e+07 1.193110e+04  3.301981e+05 -2220.757895
## PropBosq    7.035979e+04 2.487158e+02 -2.220758e+03 401.003509
## PropDeform -3.180340e+04 -9.940461e+01 -6.743793e+01 2.625263
## ConsEner    6.801296e+07 3.426262e+05  2.092242e+05 -5153.438596
## EmisCO2     1.392779e+05 6.385700e+02  4.869328e+02 -12.897193
##          PropDeform      ConsEner      EmisCO2
## CrecPobl  3.361974e-01 -8.384169e+02 -1.137877
## MortInf   1.276296e+01 -4.442568e+04 -94.855000
## PorcMujeres 2.680592e-01 2.855207e+02 -2.150132
## PNB95      -5.806203e+04 1.415628e+08 250167.323509
## ProdElec   -3.180340e+04 6.801296e+07 139277.888640
## LinTelf     -9.940461e+01 3.426262e+05  638.570000
## ConsAgua    -6.743793e+01 2.092242e+05  486.932763
## PropBosq    2.625263e+00 -5.153439e+03 -12.897193
## PropDeform  1.817253e+00 -1.051522e+03 -2.632487
## ConsEner    -1.051522e+03 5.014395e+06 10286.159781
## EmisCO2     -2.632487e+00 1.028616e+04  27.268614
```

```
cor(M)
```

```
##          CrecPobl      MortInf PorcMujeres      PNB95      ProdElec
## CrecPobl      1.00000000  0.55067948 -0.55852711 -0.32212154 -0.29711119
## MortInf       0.55067948  1.00000000 -0.03280139 -0.31579250 -0.24053689
## PorcMujeres -0.55852711 -0.03280139  1.00000000  0.14338826  0.19085114
## PNB95        -0.32212154 -0.31579250  0.14338826  1.00000000  0.74476081
## ProdElec     -0.29711119 -0.24053689  0.19085114  0.74476081  1.00000000
## LinTelf      -0.56321228 -0.69558922  0.26167018  0.46539599  0.28664508
## ConsAgua     -0.06772953 -0.06975563 -0.31106243  0.08541500  0.18047653
## PropBosq     -0.15650281 -0.02278415  0.37096694  0.05525919  0.02603078
## PropDefor    0.20107881  0.29459348  0.02266339 -0.19262327 -0.17478434
## ConsEner     -0.30187731 -0.61731132  0.01453216  0.28272492  0.22501894
## EmisCO2      -0.17568860 -0.56520778 -0.04692837  0.21425123  0.19760017
##          LinTelf      ConsAgua      PropBosq      PropDefor      ConsEner
## CrecPobl     -0.56321228 -0.06772953 -0.15650281  0.20107881 -0.30187731
## MortInf     -0.69558922 -0.06975563 -0.02278415  0.29459348 -0.61731132
## PorcMujeres  0.26167018 -0.31106243  0.37096694  0.02266339  0.01453216
## PNB95       0.46539599  0.08541500  0.05525919 -0.19262327  0.28272492
## ProdElec    0.28664508  0.18047653  0.02603078 -0.17478434  0.22501894
## LinTelf     1.00000000  0.10593934  0.06337138 -0.37623801  0.78068385
## ConsAgua    0.10593934  1.00000000 -0.19299225 -0.08705811  0.16259804
## PropBosq    0.06337138 -0.19299225  1.00000000  0.09725032 -0.11492480
## PropDefor   -0.37623801 -0.08705811  0.09725032  1.00000000 -0.34833836
## ConsEner    0.78068385  0.16259804 -0.11492480 -0.34833836  1.00000000
## EmisCO2     0.62393719  0.16227447 -0.12333592 -0.37396154  0.87965517
##          EmisCO2
## CrecPobl    -0.17568860
## MortInf     -0.56520778
## PorcMujeres -0.04692837
## PNB95       0.21425123
## ProdElec    0.19760017
## LinTelf     0.62393719
## ConsAgua    0.16227447
## PropBosq    -0.12333592
## PropDefor   -0.37396154
## ConsEner    0.87965517
## EmisCO2     1.00000000
```

```
eigen(cor(M))
```

```
## eigen() decomposition
## $values
## [1] 4.02987902 1.92999195 1.37041115 0.86451597 0.79414057 0.72919997
## [7] 0.57130511 0.32680096 0.16806846 0.14632819 0.06935866
##
## $vectors
##          [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,] -0.314119414  0.34835747 -0.07352541 -0.44028717 -0.32972147 -0.18392437
## [2,] -0.392395442 -0.04136238 -0.17759254 -0.13398483  0.08340489 -0.08656390
## [3,]  0.116546319 -0.58283641  0.16686305  0.05865031  0.18654100  0.16835650
## [4,]  0.295393771 -0.17690839 -0.53343025 -0.26248209 -0.14110658  0.04653378
## [5,]  0.258964724 -0.17356372 -0.61438847 -0.17389644 -0.07521971  0.02821905
## [6,]  0.446082934 -0.02719077  0.15177250  0.04959796 -0.05416498  0.02442175
```

```
## [7,] 0.092410503 0.32060987 -0.37024258 0.73603097 0.02671021 -0.30940890
## [8,] 0.005692925 -0.45742697 0.16480339 0.04024882 -0.41531702 -0.75356463
## [9,] -0.243652293 -0.15408201 -0.02961449 0.33650345 -0.73261463 0.50894232
## [10,] 0.415029554 0.23286257 0.20608749 -0.06730166 -0.23100421 0.05806466
## [11,] 0.374531032 0.29168698 0.20631751 -0.14843513 -0.24028756 -0.02809233
##      [,7]      [,8]      [,9]     [,10]     [,11]
## [1,] 0.1628974320 -0.09481963 -0.52181220 0.34674573 -0.10062784
## [2,] 0.6398040762 -0.32307802 0.29031618 -0.38959240 0.17487096
## [3,] 0.5310867107 0.05209889 -0.23599758 0.42854658 -0.16786800
## [4,] -0.1490207046 -0.44913216 0.36995675 0.34911534 -0.15247432
## [5,] 0.1082745817 0.50343911 -0.30681318 -0.33770404 0.12366382
## [6,] -0.0008501608 -0.56975094 -0.44733110 -0.20997673 0.44992596
## [7,] 0.2357666690 -0.05962470 -0.08358225 0.20561803 -0.07067780
## [8,] -0.0806036686 0.04275404 0.07438520 -0.08671232 -0.01493710
## [9,] 0.0112333588 -0.01607505 0.01868615 -0.03209758 0.07259619
## [10,] 0.2711228006 -0.05023582 0.04339752 -0.36147417 -0.67912543
## [11,] 0.3352822144 0.30978009 0.37666244 0.28779437 0.46737561
```

```
eigen(cov(M))
```

```
## eigen() decomposition
## $values
## [1] 6.163576e+10 6.581612e+09 4.636256e+06 3.107232e+05 1.216015e+04
## [6] 5.137767e+02 3.627885e+02 4.542082e+01 5.800868e+00 1.438020e+00
## [11] 4.768083e-01
##
## $vectors
##      [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] -1.658168e-06 4.706785e-07 0.0001263736 -1.928408e-05 -0.0055373971
## [2,] -4.048139e-05 -1.774254e-05 0.0082253821 -2.493257e-03 -0.0944030204
## [3,] 5.739096e-06 -1.084543e-05 0.0001318149 5.538307e-03 0.0314036410
## [4,] 8.880376e-01 4.597632e-01 0.0026022071 -3.893588e-04 -0.0003327409
## [5,] 4.597636e-01 -8.880405e-01 0.0005694896 1.096305e-03 0.0002207819
## [6,] 3.504341e-04 4.016179e-04 -0.0619424889 7.641174e-03 0.9921404486
## [7,] 2.625508e-04 -1.122118e-03 -0.0401453227 -9.991411e-01 0.0057795144
## [8,] 4.089564e-06 7.790843e-06 0.0012719918 6.435797e-03 0.0419331615
## [9,] -1.073825e-06 2.350808e-07 0.0001916177 4.043796e-05 -0.0018090751
## [10,] 2.547156e-03 7.126782e-04 -0.9972315499 3.973568e-02 -0.0625729475
## [11,] 4.643724e-06 -1.315731e-06 -0.0020679047 -5.626049e-05 -0.0042367120
##      [,6]      [,7]      [,8]      [,9]     [,10]
## [1,] 1.243456e-02 5.359089e-03 -8.390810e-02 -6.778358e-02 -1.158091e-01
## [2,] 9.917515e-01 2.258019e-02 -7.891128e-02 -1.637836e-02 4.264872e-04
## [3,] 8.552991e-02 -1.136481e-01 9.856498e-01 -1.468464e-02 8.241465e-03
## [4,] -8.621005e-06 -7.566477e-06 1.217248e-05 -3.971469e-07 4.274451e-07
## [5,] 1.955408e-05 1.544658e-05 -2.558998e-05 1.059471e-06 -1.353881e-06
## [6,] 9.109622e-02 4.748682e-02 -3.416812e-02 -5.379549e-03 -3.409423e-03
## [7,] -1.087229e-03 -6.863294e-03 4.698731e-03 7.965261e-05 3.621425e-05
## [8,] 1.721948e-02 -9.920538e-01 -1.169638e-01 1.416566e-03 5.891758e-03
## [9,] 1.758667e-03 -7.455427e-03 1.811443e-02 1.283039e-01 -9.859317e-01
## [10,] 2.639673e-03 -3.764707e-03 1.267052e-03 2.262931e-03 2.672618e-04
## [11,] -1.877994e-02 -1.709137e-03 -5.204823e-03 -9.891529e-01 -1.200519e-01
##      [,11]
## [1,] 9.872887e-01
## [2,] -2.092491e-02
## [3,] 8.344324e-02
```

```
## [4,] 2.723996e-07
## [5,] -2.086857e-07
## [6,] 4.944397e-04
## [7,] 4.780416e-04
## [8,] -3.748976e-03
## [9,] -1.052934e-01
## [10,] 5.906241e-05
## [11,] -8.221371e-02
```

```
eigen(cov(M))[1]
```

```
## $values
## [1] 6.163576e+10 6.581612e+09 4.636256e+06 3.107232e+05 1.216015e+04
## [6] 5.137767e+02 3.627885e+02 4.542082e+01 5.800868e+00 1.438020e+00
## [11] 4.768083e-01
```

```
sum(diag(cov(M)))
```

```
## [1] 68222335253
```

```
my.sum = 0
for (x in eigen(cov(M))[1]) {
  y = x/sum(diag(cov(M)))
  print(y)
  my.sum = my.sum + y
}
```

```
## [1] 9.034543e-01 9.647298e-02 6.795804e-05 4.554567e-06 1.782429e-07
## [6] 7.530917e-09 5.317738e-09 6.657763e-10 8.502887e-11 2.107843e-11
## [11] 6.989035e-12
```

```
cat('sum:', sum(my.sum))
```

```
## sum: 1
```

Los componentes más importantes son CrecPobl y MortInf, estos dos componentes nos dieron los valores más altos en el análisis anterior.

```
my.sum = 0
for (x in eigen(cor(M))[1]) {
  y = x/sum(diag(cor(M)))
  print(y)
  my.sum = my.sum + y
}
```

```
## [1] 0.366352638 0.175453813 0.124582832 0.078592361 0.072194597 0.066290906
## [7] 0.051936828 0.029709178 0.015278951 0.013302563 0.006305332
```

```
cat('sum: ',sum(my.sum))
```

```
## sum: 1
```

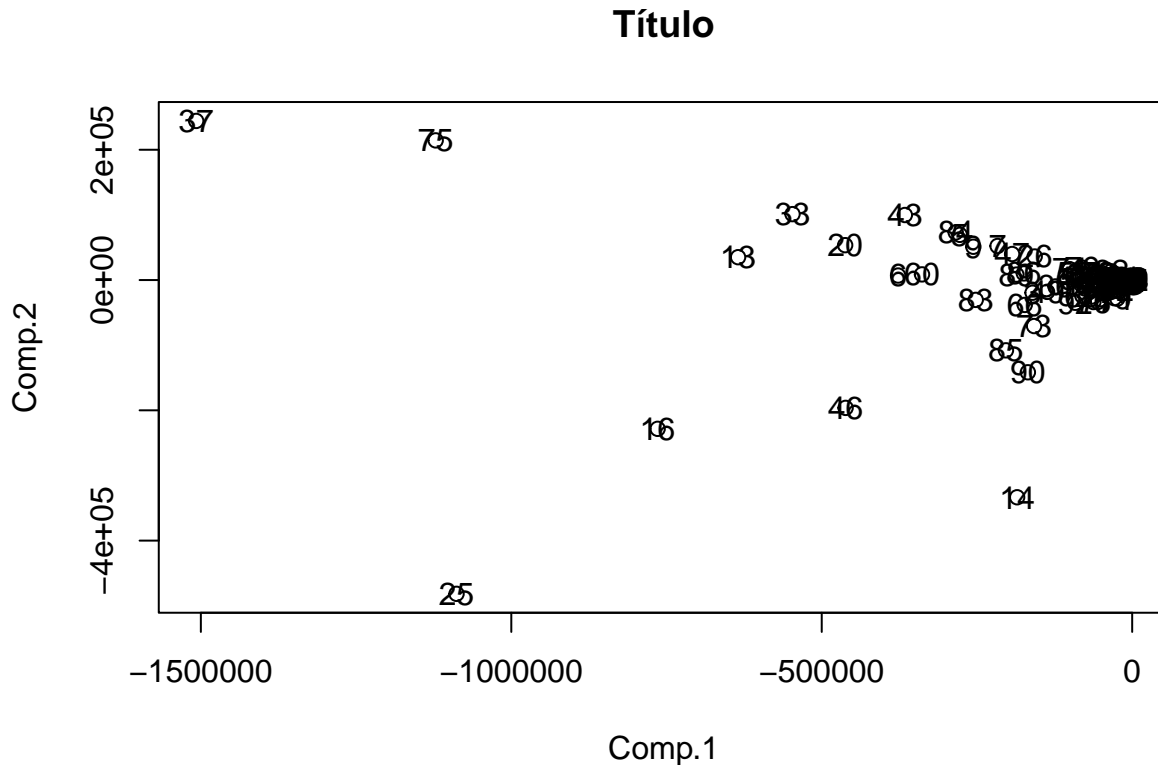
Los componentes más importantes son CrecPobl, MortInf y PorcMujeres, estos tres componentes nos dieron valores mayores al 0.1 en el análisis anterior.

```
library(stats)
library(factoextra)
```

```
## Loading required package: ggplot2
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(ggplot2)
datos= M
cpS=princomp(datos,cor=FALSE)
cpaS=as.matrix(datos)%*%cpS$loadings
plot(cpaS[,1:2],type="p", main = "Título")
text(cpaS[,1],cpaS[,2],1:nrow(cpaS))
```



```
biplot(cpS)
```

```
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

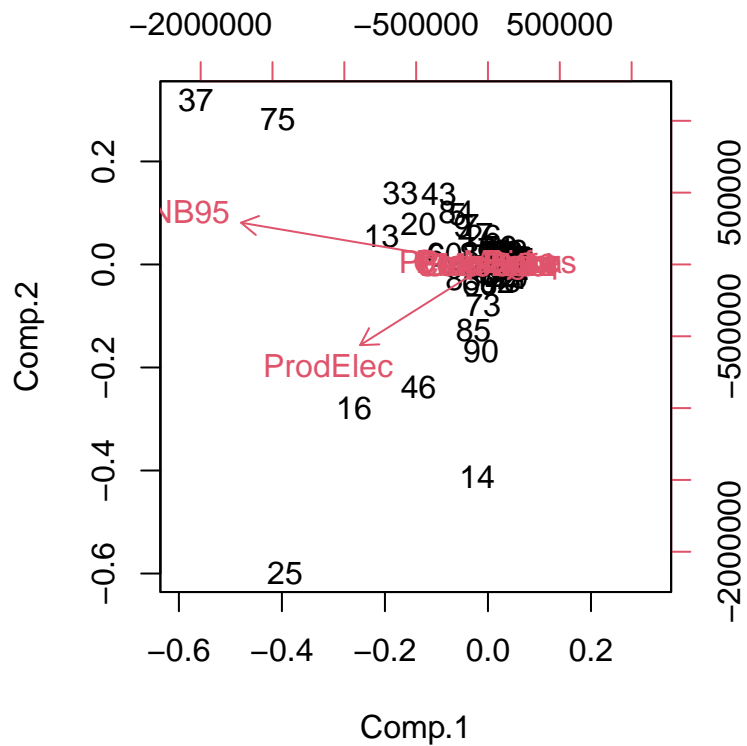
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

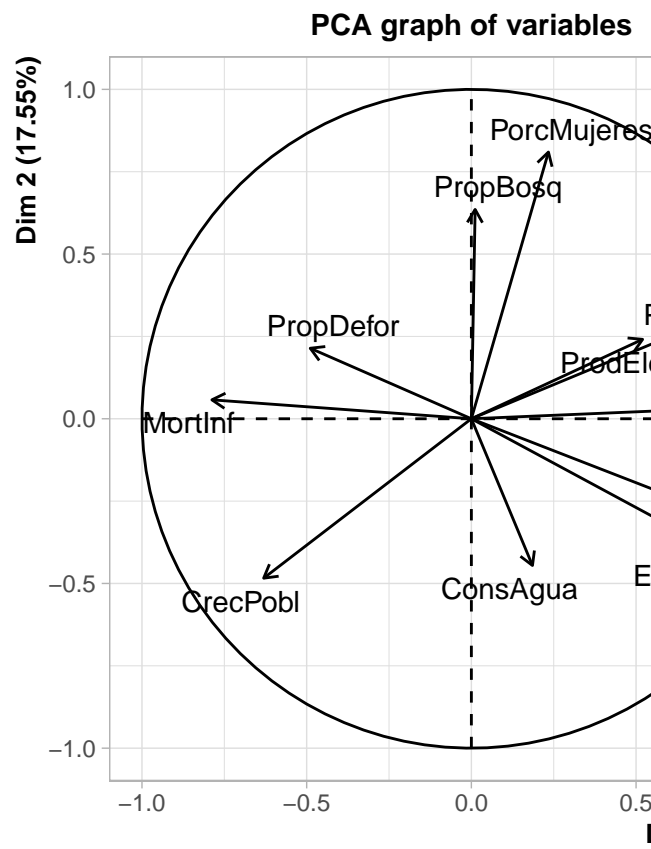
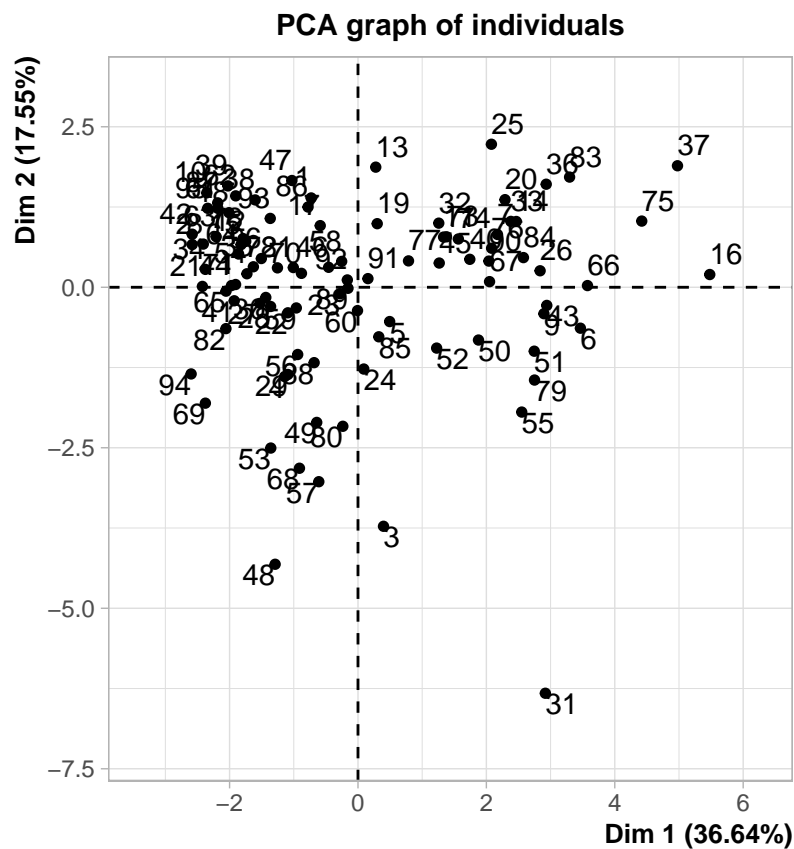
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped

## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped
```

```
## Warning in arrows(0, 0, y[, 1L] * 0.8, y[, 2L] * 0.8, col = col[2L], length =
## arrow.len): zero-length arrow is of indeterminate angle and so skipped
```

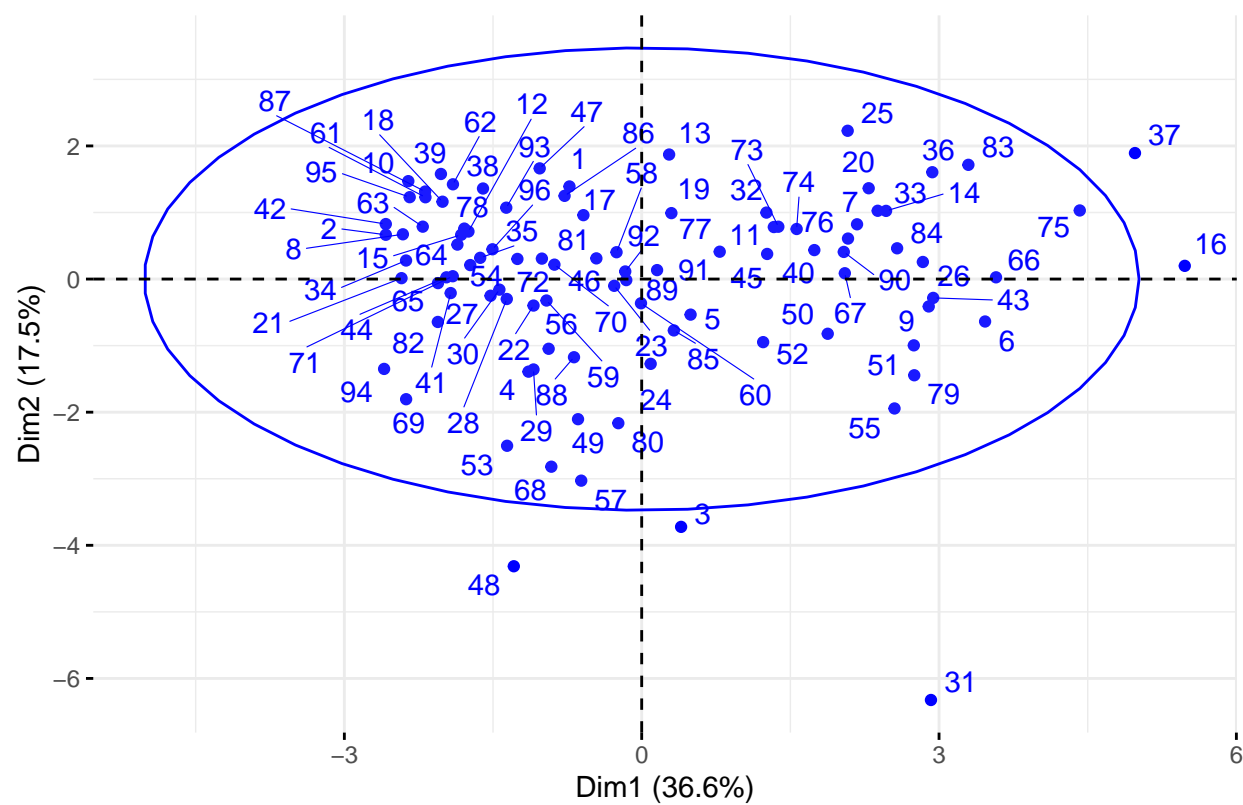


```
library(FactoMineR)
cp3 = PCA(datos)
```



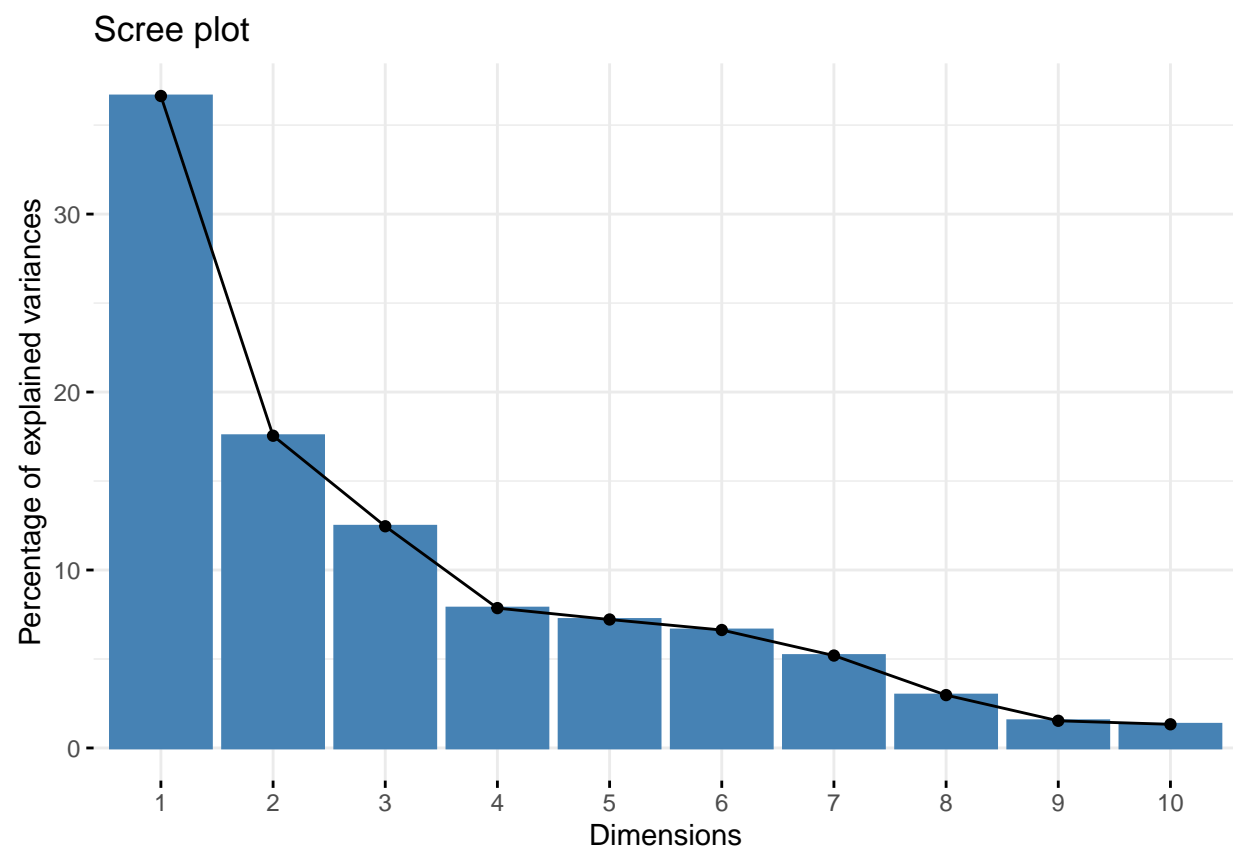
```
fviz_pca_ind(cp3, col.ind = "blue", addEllipses = TRUE, repel = TRUE)
```

## Individuals – PCA



```
fviz_screepplot(cp3)
```





```
fviz_contrib(cp3, choice = c("var"))
```

