

456F

```
w = read.table("wine.txt",sep = ",")
head(w)
```

```
##   V1    V2    V3    V4    V5    V6    V7    V8    V9   V10   V11   V12   V13   V14
## 1  1 14.23 1.71 2.43 15.6 127 2.80 3.06 0.28 2.29 5.64 1.04 3.92 1065
## 2  1 13.20 1.78 2.14 11.2 100 2.65 2.76 0.26 1.28 4.38 1.05 3.40 1050
## 3  1 13.16 2.36 2.67 18.6 101 2.80 3.24 0.30 2.81 5.68 1.03 3.17 1185
## 4  1 14.37 1.95 2.50 16.8 113 3.85 3.49 0.24 2.18 7.80 0.86 3.45 1480
## 5  1 13.24 2.59 2.87 21.0 118 2.80 2.69 0.39 1.82 4.32 1.04 2.93 735
## 6  1 14.20 1.76 2.45 15.2 112 3.27 3.39 0.34 1.97 6.75 1.05 2.85 1450
```

#a Covariance matrix

```
sigma = cov(w)
round(sigma,2)
```

```
##      V1    V2    V3    V4    V5    V6    V7    V8    V9   V10
## V1    0.60 -0.21  0.38 -0.01  1.34 -2.32 -0.35 -0.66  0.05 -0.22
## V2   -0.21  0.66  0.09  0.05 -0.84  3.14  0.15  0.19 -0.02  0.06
## V3    0.38  0.09  1.25  0.05  1.08 -0.87 -0.23 -0.46  0.04 -0.14
## V4   -0.01  0.05  0.05  0.08  0.41  1.12  0.02  0.03  0.01  0.00
## V5    1.34 -0.84  1.08  0.41 11.15 -3.97 -0.67 -1.17  0.15 -0.38
## V6   -2.32  3.14 -0.87  1.12 -3.97 203.99  1.92  2.79 -0.46  1.93
## V7   -0.35  0.15 -0.23  0.02 -0.67  1.92  0.39  0.54 -0.04  0.22
## V8   -0.66  0.19 -0.46  0.03 -1.17  2.79  0.54  1.00 -0.07  0.37
## V9    0.05 -0.02  0.04  0.01  0.15 -0.46 -0.04 -0.07  0.02 -0.03
## V10   -0.22  0.06 -0.14  0.00 -0.38  1.93  0.22  0.37 -0.03  0.33
## V11    0.48  1.03  0.64  0.16  0.15  6.62 -0.08 -0.40  0.04 -0.03
## V12   -0.11 -0.01 -0.14  0.00 -0.21  0.18  0.06  0.12 -0.01  0.04
## V13   -0.43  0.04 -0.29  0.00 -0.66  0.67  0.31  0.56 -0.04  0.21
## V14 -154.67 164.57 -67.55 19.32 -463.36 1769.16 98.17 155.45 -12.20 59.55
##      V11   V12   V13    V14
## V1    0.48 -0.11 -0.43 -154.67
## V2    1.03 -0.01  0.04  164.57
## V3    0.64 -0.14 -0.29 -67.55
## V4    0.16  0.00  0.00  19.32
## V5    0.15 -0.21 -0.66 -463.36
## V6    6.62  0.18  0.67 1769.16
## V7   -0.08  0.06  0.31  98.17
## V8   -0.40  0.12  0.56 155.45
## V9    0.04 -0.01 -0.04 -12.20
## V10   -0.03  0.04  0.21  59.55
## V11    5.37 -0.28 -0.71 230.77
## V12   -0.28  0.05  0.09  17.00
## V13   -0.71  0.09  0.50  69.93
## V14 230.77 17.00 69.93 99166.72
```

#b

```
sigma.eval = eigen(sigma)$values
sigma.evec = eigen(sigma)$vectors
round(sigma.eval,2)
```

```
## [1] 99202.03    172.54      9.53      5.10      1.29      0.87      0.29
## [8]      0.16      0.11      0.09      0.05      0.03      0.02      0.01
```

```
round(sigma.evec ,2)
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11]
## [1,] 0.00 0.00 -0.10 0.15 -0.22 0.17 -0.18 -0.19 -0.17 0.51 -0.62
## [2,] 0.00 0.00 -0.02 0.14 0.05 -0.20 0.89 -0.35 -0.10 0.06 -0.09
## [3,] 0.00 0.00 -0.12 0.16 -0.53 -0.80 -0.15 0.08 -0.01 -0.03 0.01
## [4,] 0.00 0.00 -0.05 -0.01 0.03 -0.04 0.05 0.15 -0.07 -0.11 0.35
## [5,] 0.00 -0.03 -0.93 -0.36 0.07 0.02 0.04 -0.01 0.00 -0.01 0.00
## [6,] -0.02 -1.00 0.03 0.00 -0.01 0.00 0.00 0.00 0.00 0.00 0.00
## [7,] 0.00 0.00 0.04 -0.08 0.32 -0.23 -0.07 0.09 -0.36 0.72 0.39
## [8,] 0.00 0.00 0.09 -0.17 0.54 -0.36 -0.08 0.20 -0.38 -0.32 -0.49
## [9,] 0.00 0.00 -0.01 0.01 -0.03 0.02 0.00 0.00 -0.03 -0.02 0.18
## [10,] 0.00 -0.01 0.03 -0.05 0.25 -0.20 -0.35 -0.85 0.12 -0.10 0.13
## [11,] 0.00 -0.02 -0.30 0.86 0.37 0.01 -0.09 0.11 0.11 -0.04 0.04
## [12,] 0.00 0.00 0.03 -0.06 0.05 0.03 0.03 0.00 -0.03 -0.05 0.08
## [13,] 0.00 0.00 0.07 -0.18 0.27 -0.25 0.08 0.18 0.80 0.30 -0.18
## [14,] -1.00 0.02 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
##      [,12] [,13] [,14]
## [1,] -0.37 -0.10 0.05
## [2,] -0.03 0.00 0.01
## [3,] 0.04 -0.06 -0.01
## [4,] -0.89 0.09 -0.16
## [5,] 0.06 0.00 0.00
## [6,] 0.00 0.00 0.00
## [7,] 0.14 0.02 -0.03
## [8,] -0.08 0.00 0.09
## [9,] -0.12 -0.13 0.97
## [10,] -0.09 0.02 -0.02
## [11,] 0.04 -0.04 -0.01
## [12,] -0.02 -0.98 -0.15
## [13,] -0.13 -0.04 0.06
## [14,] 0.00 0.00 0.00
```

```
##(c) Perform principle component analysis on raw data??? scale choose
##true or false?
```

```
w.pca = prcomp(w,center=T,scale=F)
print(w.pca)
```

```
## Standard deviations (1, ..., p=14):
## [1] 314.96353873 13.13531864 3.08726351 2.25849728 1.13392604
## [6] 0.93175457 0.53573032 0.39405452 0.33724371 0.29390207
## [11] 0.21494794 0.18688848 0.14409259 0.08995921
##
## Rotation (n x k) = (14 x 14):
##      PC1      PC2      PC3      PC4      PC5
## V1  0.0015593428 -2.778618e-03 -0.10091996 -0.147645845 0.219197410
## V2 -0.0016592619 -1.204267e-03 -0.01830354 -0.136988601 -0.048932608
## V3  0.0006810190 -2.159531e-03 -0.12455946 -0.156739376 0.528930958
## V4 -0.0001949052 -4.593996e-03 -0.05129289 0.012047122 -0.025845659
## V5  0.0046713046 -2.646070e-02 -0.92809986 0.356990103 -0.069611267
## V6 -0.0178679787 -9.993399e-01 0.02998454 0.004754117 0.006452123
## V7 -0.0009898316 -8.747309e-04 0.04276110 0.076452064 -0.320080904
```

```
## V8 -0.0015672929 5.865826e-05 0.09026712 0.172191025 -0.535695633
## V9 0.0001230870 1.353994e-03 -0.01372151 -0.010593546 0.029288874
## V10 -0.0006006091 -5.002227e-03 0.02623713 0.051606406 -0.253787072
## V11 -0.0023271272 -1.511422e-02 -0.30320298 -0.856516199 -0.367299181
## V12 -0.0001713811 7.640425e-04 0.02699227 0.059055310 -0.045942828
## V13 -0.0007049359 3.500684e-03 0.07436620 0.178526137 -0.269296632
## V14 -0.9998217211 1.776936e-02 -0.00462656 0.002950572 0.002713180
## PC6 PC7 PC8 PC9 PC10
## V1 0.168685378 -1.788007e-01 -1.909952e-01 -1.658280e-01 -0.5109932771
## V2 -0.202080449 8.901163e-01 -3.490280e-01 -1.042972e-01 -0.0575902207
## V3 -0.803839930 -1.468581e-01 7.552074e-02 -1.022552e-02 0.0336100827
## V4 -0.041693904 5.034930e-02 1.505281e-01 -6.888147e-02 0.1071613371
## V5 0.023859516 3.593720e-02 -1.139207e-02 2.441283e-03 0.0099803583
## V6 0.001349008 2.065027e-03 3.568638e-03 1.632307e-03 -0.0008647093
## V7 -0.228319660 -6.567984e-02 8.990297e-02 -3.642319e-01 -0.7196957875
## V8 -0.357645070 -7.945528e-02 2.049005e-01 -3.849304e-01 0.3184589500
## V9 0.016951591 -5.403887e-05 -9.899527e-04 -2.887276e-02 0.0177935045
## V10 -0.197181268 -3.517404e-01 -8.477276e-01 1.212945e-01 0.1040244264
## V11 0.005507929 -9.328967e-02 1.083719e-01 1.108008e-01 0.0445052328
## V12 0.030413173 2.958250e-02 2.816679e-03 -2.708614e-02 0.0519432038
## V13 -0.249150932 7.998405e-02 1.809240e-01 8.045947e-01 -0.2960758515
## V14 0.001211259 -1.178372e-03 -9.491742e-05 -2.980272e-05 -0.0003880366
## PC11 PC12 PC13 PC14
## V1 -0.6235291164 3.663543e-01 9.598765e-02 -4.715808e-02
## V2 -0.0890749178 3.477671e-02 -4.240326e-03 -1.342199e-02
## V3 0.0077436199 -3.573628e-02 6.323570e-02 1.205487e-02
## V4 0.3532557832 8.914869e-01 -8.648809e-02 1.623446e-01
## V5 -0.0026586178 -5.931059e-02 2.694858e-04 6.483270e-05
## V6 -0.0016384061 -2.620277e-03 5.009661e-04 -2.271269e-03
## V7 0.3933791381 -1.356768e-01 -1.908158e-02 3.464768e-02
## V8 -0.4852223379 8.051266e-02 -1.971202e-04 -8.667150e-02
## V9 0.1750692832 1.164653e-01 1.329541e-01 -9.672182e-01
## V10 0.1313895856 8.704446e-02 -1.502505e-02 1.929259e-02
## V11 0.0357823455 -3.763496e-02 4.256684e-02 7.836890e-03
## V12 0.0827103412 2.479200e-02 9.786547e-01 1.523550e-01
## V13 -0.1794317102 1.302390e-01 3.801870e-02 -5.675071e-02
## V14 -0.0005285291 4.693355e-05 -4.006096e-05 -6.726096e-05
```

```
#above code returns 14 principle components as rotation
# Compute principal components from the covariance matrix.
# sdev: standard deviations of the component scores (
# square roots of eigenvalues of the covariance matrix)
# rotation: The coefficients needed to compute the scores
# (elements of eigenvectors)
```

```
# Print the results
```

```
round(w.pca$sdev,2)
```

```
## [1] 314.96 13.14 3.09 2.26 1.13 0.93 0.54 0.39 0.34 0.29
## [11] 0.21 0.19 0.14 0.09
```

```
round(w.pca$rotation,2)
```

```
## PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9 PC10 PC11
## V1 0.00 0.00 -0.10 -0.15 0.22 0.17 -0.18 -0.19 -0.17 -0.51 -0.62
```

```
## V2  0.00  0.00 -0.02 -0.14 -0.05 -0.20  0.89 -0.35 -0.10 -0.06 -0.09
## V3  0.00  0.00 -0.12 -0.16  0.53 -0.80 -0.15  0.08 -0.01  0.03  0.01
## V4  0.00  0.00 -0.05  0.01 -0.03 -0.04  0.05  0.15 -0.07  0.11  0.35
## V5  0.00 -0.03 -0.93  0.36 -0.07  0.02  0.04 -0.01  0.00  0.01  0.00
## V6 -0.02 -1.00  0.03  0.00  0.01  0.00  0.00  0.00  0.00  0.00  0.00
## V7  0.00  0.00  0.04  0.08 -0.32 -0.23 -0.07  0.09 -0.36 -0.72  0.39
## V8  0.00  0.00  0.09  0.17 -0.54 -0.36 -0.08  0.20 -0.38  0.32 -0.49
## V9  0.00  0.00 -0.01 -0.01  0.03  0.02  0.00  0.00 -0.03  0.02  0.18
## V10 0.00 -0.01  0.03  0.05 -0.25 -0.20 -0.35 -0.85  0.12  0.10  0.13
## V11 0.00 -0.02 -0.30 -0.86 -0.37  0.01 -0.09  0.11  0.11  0.04  0.04
## V12 0.00  0.00  0.03  0.06 -0.05  0.03  0.03  0.00 -0.03  0.05  0.08
## V13 0.00  0.00  0.07  0.18 -0.27 -0.25  0.08  0.18  0.80 -0.30 -0.18
## V14 -1.00  0.02  0.00  0.00  0.00  0.00  0.00  0.00  0.00  0.00  0.00
##      PC12  PC13  PC14
## V1  0.37  0.10 -0.05
## V2  0.03  0.00 -0.01
## V3 -0.04  0.06  0.01
## V4  0.89 -0.09  0.16
## V5 -0.06  0.00  0.00
## V6  0.00  0.00  0.00
## V7 -0.14 -0.02  0.03
## V8  0.08  0.00 -0.09
## V9  0.12  0.13 -0.97
## V10 0.09 -0.02  0.02
## V11 -0.04  0.04  0.01
## V12 0.02  0.98  0.15
## V13 0.13  0.04 -0.06
## V14 0.00  0.00  0.00
```

```
#d
s = w.pca$sdev^2
pvar = s/sum(s)
cpvar = cumsum(s) / sum(s)
cat("Cumulative proportion of the total variance explained by each component: ", cpvar, fill = T)
```

```
## Cumulative proportion of the total variance explained by each component:
## 0.9980876 0.9998235 0.9999194 0.9999708 0.9999837 0.9999924 0.9999953
## 0.9999969 0.999998 0.9999989 0.9999994 0.9999997 0.9999999 1
```

```
#second way to confirm this:
summary(w.pca)
```

```
## Importance of components%s:
##          PC1      PC2      PC3      PC4      PC5      PC6
## Standard deviation  314.9635 13.13532 3.0873 2.25850 1.13393 0.93175
## Proportion of Variance  0.9981  0.00174 0.0001 0.00005 0.00001 0.00001
## Cumulative Proportion  0.9981  0.99982 0.9999 0.99997 0.99998 0.99999
##          PC7      PC8      PC9     PC10     PC11     PC12     PC13
## Standard deviation  0.5357 0.3941 0.3372 0.2939 0.2149 0.1869 0.1441
## Proportion of Variance 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000
## Cumulative Proportion 1.0000 1.0000 1.0000 1.0000 1.0000 1.0000 1.0000
##          PC14
## Standard deviation  0.08996
## Proportion of Variance 0.00000
## Cumulative Proportion 1.00000
```

```
#use summary to compute the total variance explained by each component
```

```
#e
```

```
#It appears that 5 components needed to explain 75 percents
```

```
 #(f) For the number of components in part e, give the formula for each component  
 #and a brief interpretation.
```

```
#Provide formula for each component???
```

```
#PC1 = 0.393669533 * V1 - 0.136325011 * V2 + ...
```

```
#PC2 = ...
```

```
#g
```

```
# 7 components are needed in this case
```

```
#Give the four plots that helps explain
```

```
pcaCharts <- function(x) {
```

```
  x.var <- x$sdev ^ 2
```

```
  x.pvar <- x.var/sum(x.var)
```

```
  print("proportions of variance:")
```

```
  print(x.pvar)
```

```
  par(mfrow=c(2,2))
```

```
  plot(x.pvar,xlab="Principal component", ylab="Proportion of variance explained", ylim=c(0,1), type="n")
```

```
  plot(cumsum(x.pvar),xlab="Principal component", ylab="Cumulative Proportion of variance explained", type="n")
```

```
  screeplot(x)
```

```
  screeplot(x,type="l")
```

```
  par(mfrow=c(1,1))
```

```
}
```

```
pcaCharts(w.pca)
```

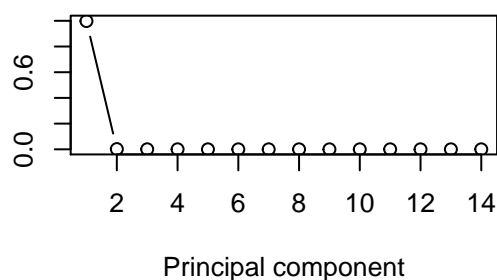
```
## [1] "proportions of variance:"
```

```
## [1] 9.980876e-01 1.735919e-03 9.589490e-05 5.132007e-05 1.293652e-05
```

```
## [6] 8.734764e-06 2.887623e-06 1.562287e-06 1.144289e-06 8.690673e-07
```

```
## [11] 4.648520e-07 3.514092e-07 2.088966e-07 8.142155e-08
```

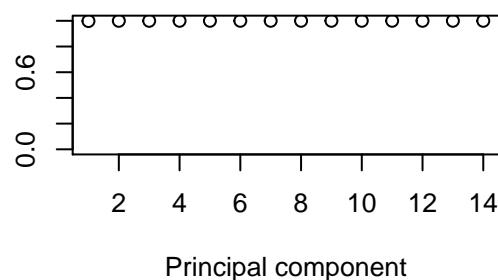
Proportion of variance explained

**x**

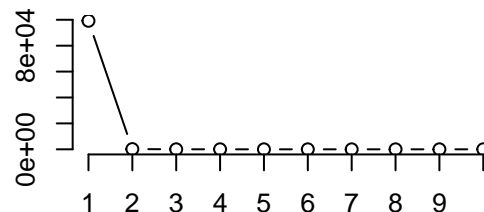
Variances



Cumulative Proportion of variance explained

**x**

Variances



#h

```
w.pca0 = princomp(w,cor=TRUE)
head(w.pca0$scores)
```

```
##          Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## [1,] -3.522934 -1.4530984  0.1647955  0.01327282  0.7373454 -0.3007162
## [2,] -2.528858  0.3300193  2.0267067  0.41714436 -0.2832137 -0.8843094
## [3,] -2.785029 -1.0369360 -0.9832377 -0.66423241 -0.3875650  0.4688416
## [4,] -3.922588 -2.7682103  0.1749682 -0.56508782 -0.3243597  0.2626164
## [5,] -1.407511 -0.8677731 -2.0258295  0.44090580  0.2279492 -0.5937621
## [6,] -3.288130 -2.1301751  0.6289923  0.60537190 -0.4096264  0.2582760
##          Comp.7    Comp.8    Comp.9    Comp.10    Comp.11
## [1,]  0.57387557  0.05563728  0.45876506  1.06556955 -0.4204939
## [2,] -0.02971648  1.01037189 -0.21880609 -0.02018203 -0.1300196
## [3,]  0.48830540 -0.26895705 -1.22276949 -0.10624913 -0.2790741
## [4,] -0.39836411  0.61884369  0.11465832  0.10767113  0.7738658
## [5,]  0.44787555  0.43493730  0.26154774  0.11243432 -0.5379312
## [6,]  0.37857594  0.36682547 -0.04491229 -0.21796620  0.4077854
##          Comp.12    Comp.13    Comp.14
## [1,] -0.552927766  0.30297818  0.20085745
## [2,] -0.394971160  0.14664531  0.12640235
## [3,] -0.001897993 -0.02127802 -0.05575152
## [4,]  0.230929232  0.50127681  0.01990285
## [5,]  0.226686179 -0.27410891 -0.51749790
## [6,]  0.376715445  0.01744264  0.23956231
```

```
#(h)
# plot component scores????????????
component_scores = w.pca$x[,c(1,2,3,5,6,7)]
head(component_scores)
```

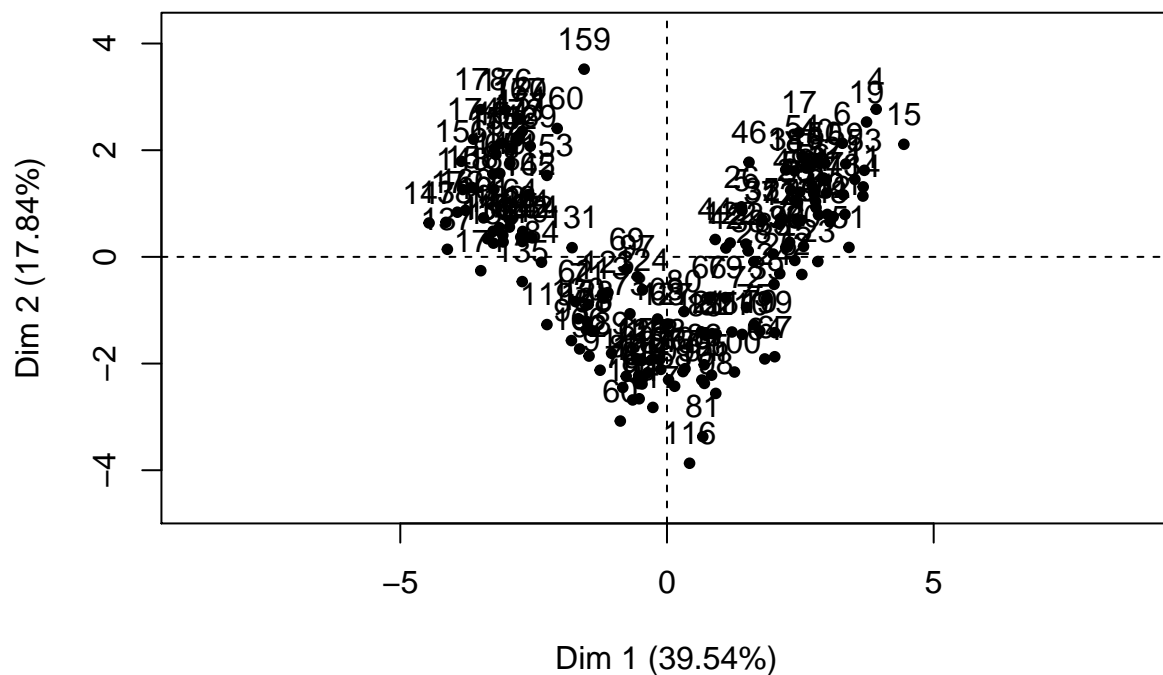
```
##          PC1          PC2          PC3          PC5          PC6          PC7
## [1,] -318.56405 -21.4907729  3.1646131 -0.7520919 -0.520108716  0.59053749
## [2,] -303.09851  5.3660827  6.8169925  0.5010454 -0.052439234 -0.08217096
## [3,] -438.06206  6.5379861 -1.0667167 -0.4532081 -0.634321618 -0.79376803
## [4,] -733.24071 -0.1933200 -0.9383123 -0.8799180 -0.480809678 -0.10023671
## [5,]  11.56995 -18.4872549 -0.4249173 -0.5067331 -0.862531264  0.40524787
## [6,] -703.23180  0.3317232  0.9229641 -0.1133199 -0.006724859 -0.07490001
```

```
#(h)
library(FactoMineR)
```

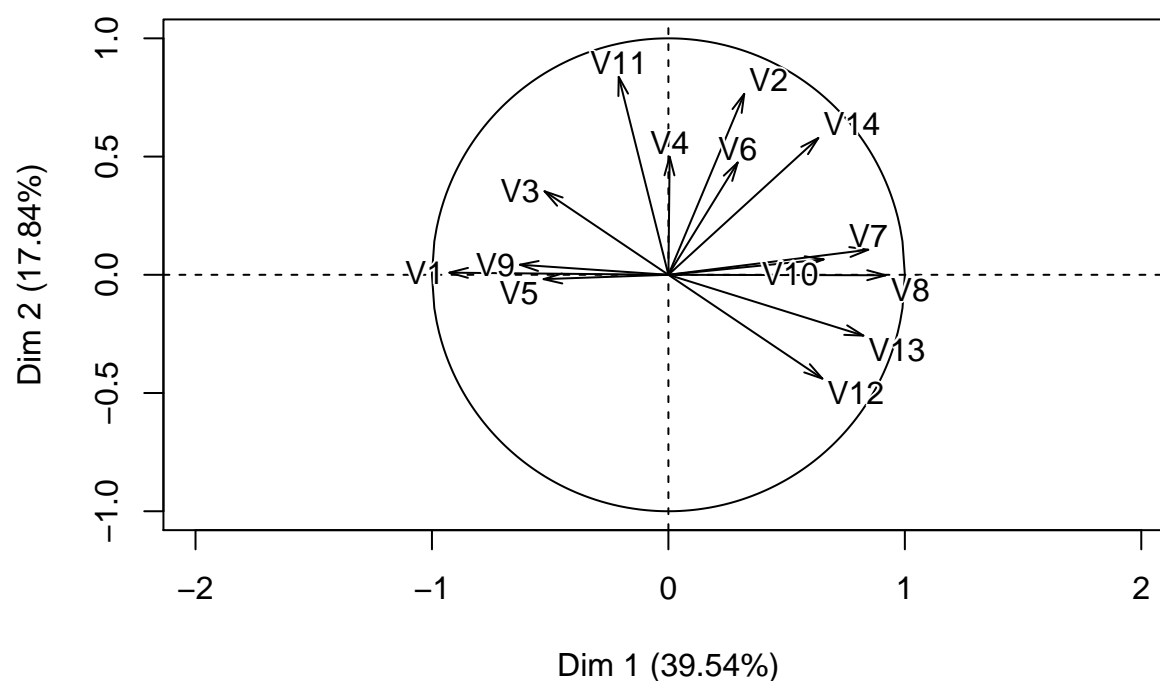
```
## Warning: package 'FactoMineR' was built under R version 3.4.2
```

```
w.pca1 = PCA(w)
```

Individuals factor map (PCA)



Variables factor map (PCA)



```
w.pca1$var$coord
```

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5
V1	-0.926249119	0.008992068	0.001464623	0.11796747	-0.14761693
V2	0.320753602	0.765077776	-0.249405338	-0.07891083	0.23502535
V3	-0.523926254	0.353321542	0.106779776	0.45263625	0.17667138
V4	0.005312596	0.499119886	0.752905787	-0.24066828	0.08760833
V5	-0.527742841	-0.018355351	0.735934791	0.06934999	-0.04362215
V6	0.293237259	0.474935577	0.157512659	-0.15722170	-0.72910189
V7	0.845297832	0.106063692	0.176179709	0.18397317	0.13551568
V8	0.919289787	-0.002075539	0.181536969	0.13930706	0.10492129
V9	-0.628216328	0.042647926	0.204400357	-0.31596971	0.40521898
V10	0.656594877	0.065140471	0.180234438	0.44576752	-0.08579534
V11	-0.210153399	0.837170096	-0.165066603	0.06946486	0.04334310
V12	0.651324814	-0.439153088	0.102609980	-0.41870738	0.02797754
V13	0.824738871	-0.257221308	0.199865440	0.15096930	0.13507348
V14	0.634131533	0.578451334	-0.152344513	-0.24640337	0.07906923

```
#On the second output table, Dim1 symbolizes the first principle component :
#same for rest to interpret.
```

```
#Using 0.5 as threshold:
```

```
###there are only 5 dims but to explain 90 percent, there needs about 7??
```

```
#The first principle component: V1 has highest and V4 has the lowest
#The second principle component:...
```



```
#...
```

```
#i. Give the correlation between Alcohol and the first principal component
```

```
pca.out = prcomp(w)
round(cor(pca.out$x,w),2)
```

```
##      V1    V2    V3    V4    V5    V6    V7    V8    V9    V10   V11
## PC1  0.63 -0.64  0.19 -0.22  0.44 -0.39 -0.50 -0.49  0.31 -0.33 -0.32
## PC2 -0.05 -0.02 -0.03 -0.22 -0.10 -0.92 -0.02  0.00  0.14 -0.11 -0.09
## PC3 -0.40 -0.07 -0.34 -0.58 -0.86  0.01  0.21  0.28 -0.34  0.14 -0.40
## PC4 -0.43 -0.38 -0.32  0.10  0.24  0.00  0.28  0.39 -0.19  0.20 -0.83
## PC5  0.32 -0.07  0.54 -0.11 -0.02  0.00 -0.58 -0.61  0.27 -0.50 -0.18
## PC6  0.20 -0.23 -0.67 -0.14  0.01  0.00 -0.34 -0.33  0.13 -0.32  0.00
## PC7 -0.12  0.59 -0.07  0.10  0.01  0.00 -0.06 -0.04  0.00 -0.33 -0.02
## PC8 -0.10 -0.17  0.03  0.22  0.00  0.00  0.06  0.08  0.00 -0.58  0.02
## PC9 -0.07 -0.04  0.00 -0.08  0.00  0.00 -0.20 -0.13 -0.08  0.07  0.02
## PC10 -0.19 -0.02  0.01  0.11  0.00  0.00 -0.34  0.09  0.04  0.05  0.01
## PC11 -0.17 -0.02  0.00  0.28  0.00  0.00  0.14 -0.10  0.30  0.05  0.00
## PC12  0.09  0.01 -0.01  0.61  0.00  0.00 -0.04  0.02  0.17  0.03  0.00
## PC13  0.02  0.00  0.01 -0.05  0.00  0.00  0.00  0.00  0.15  0.00  0.00
## PC14 -0.01  0.00  0.00  0.05  0.00  0.00  0.00  0.00 -0.01 -0.70  0.00
##      V12    V13 V14
## PC1 -0.24 -0.31 -1
## PC2  0.04  0.06  0
## PC3  0.36  0.32  0
## PC4  0.58  0.57  0
## PC5 -0.23 -0.43  0
## PC6  0.12 -0.33  0
## PC7  0.07  0.06  0
## PC8  0.00  0.10  0
## PC9 -0.04  0.38  0
## PC10 0.07 -0.12  0
## PC11 0.08 -0.05  0
## PC12 0.02  0.03  0
## PC13 0.62  0.01  0
## PC14 0.06 -0.01  0
```

```
#correlation between Alcohol and the first principal component is 0.63, which
#suggests those 2 are highly correlated
```

```
#####
```

```
##### PART2
```

```
#####
```

```
# To compute principal components from a correlation matrix, I first
#standardize the data aka change the scale to true.
```

```
ws = scale(w,center=T,scale = T)
ws.cor = var(ws)
cat("covariance matrix:", fill=T)
```

```
## covariance matrix:
```

```
round(ws.cor,2)
```

```
##      V1    V2    V3    V4    V5    V6    V7    V8    V9    V10   V11
## V1  1.00 -0.33  0.44 -0.05  0.52 -0.21 -0.72 -0.85  0.49 -0.50  0.27
## V2 -0.33  1.00  0.09  0.21 -0.31  0.27  0.29  0.24 -0.16  0.14  0.55
```

```
## V3  0.44  0.09  1.00  0.16  0.29 -0.05 -0.34 -0.41  0.29 -0.22  0.25
## V4 -0.05  0.21  0.16  1.00  0.44  0.29  0.13  0.12  0.19  0.01  0.26
## V5  0.52 -0.31  0.29  0.44  1.00 -0.08 -0.32 -0.35  0.36 -0.20  0.02
## V6 -0.21  0.27 -0.05  0.29 -0.08  1.00  0.21  0.20 -0.26  0.24  0.20
## V7 -0.72  0.29 -0.34  0.13 -0.32  0.21  1.00  0.86 -0.45  0.61 -0.06
## V8 -0.85  0.24 -0.41  0.12 -0.35  0.20  0.86  1.00 -0.54  0.65 -0.17
## V9  0.49 -0.16  0.29  0.19  0.36 -0.26 -0.45 -0.54  1.00 -0.37  0.14
## V10 -0.50  0.14 -0.22  0.01 -0.20  0.24  0.61  0.65 -0.37  1.00 -0.03
## V11  0.27  0.55  0.25  0.26  0.02  0.20 -0.06 -0.17  0.14 -0.03  1.00
## V12 -0.62 -0.07 -0.56 -0.07 -0.27  0.06  0.43  0.54 -0.26  0.30 -0.52
## V13 -0.79  0.07 -0.37  0.00 -0.28  0.07  0.70  0.79 -0.50  0.52 -0.43
## V14 -0.63  0.64 -0.19  0.22 -0.44  0.39  0.50  0.49 -0.31  0.33  0.32
##      V12  V13  V14
## V1 -0.62 -0.79 -0.63
## V2 -0.07  0.07  0.64
## V3 -0.56 -0.37 -0.19
## V4 -0.07  0.00  0.22
## V5 -0.27 -0.28 -0.44
## V6  0.06  0.07  0.39
## V7  0.43  0.70  0.50
## V8  0.54  0.79  0.49
## V9 -0.26 -0.50 -0.31
## V10 0.30  0.52  0.33
## V11 -0.52 -0.43  0.32
## V12 1.00  0.57  0.24
## V13 0.57  1.00  0.31
## V14 0.24  0.31  1.00
```

```
#Start repeating from the first step
#a Covariance matrix
sigma = var(ws)
round(sigma,2)
```

```
##      V1  V2  V3  V4  V5  V6  V7  V8  V9  V10  V11
## V1  1.00 -0.33  0.44 -0.05  0.52 -0.21 -0.72 -0.85  0.49 -0.50  0.27
## V2 -0.33  1.00  0.09  0.21 -0.31  0.27  0.29  0.24 -0.16  0.14  0.55
## V3  0.44  0.09  1.00  0.16  0.29 -0.05 -0.34 -0.41  0.29 -0.22  0.25
## V4 -0.05  0.21  0.16  1.00  0.44  0.29  0.13  0.12  0.19  0.01  0.26
## V5  0.52 -0.31  0.29  0.44  1.00 -0.08 -0.32 -0.35  0.36 -0.20  0.02
## V6 -0.21  0.27 -0.05  0.29 -0.08  1.00  0.21  0.20 -0.26  0.24  0.20
## V7 -0.72  0.29 -0.34  0.13 -0.32  0.21  1.00  0.86 -0.45  0.61 -0.06
## V8 -0.85  0.24 -0.41  0.12 -0.35  0.20  0.86  1.00 -0.54  0.65 -0.17
## V9  0.49 -0.16  0.29  0.19  0.36 -0.26 -0.45 -0.54  1.00 -0.37  0.14
## V10 -0.50  0.14 -0.22  0.01 -0.20  0.24  0.61  0.65 -0.37  1.00 -0.03
## V11  0.27  0.55  0.25  0.26  0.02  0.20 -0.06 -0.17  0.14 -0.03  1.00
## V12 -0.62 -0.07 -0.56 -0.07 -0.27  0.06  0.43  0.54 -0.26  0.30 -0.52
## V13 -0.79  0.07 -0.37  0.00 -0.28  0.07  0.70  0.79 -0.50  0.52 -0.43
## V14 -0.63  0.64 -0.19  0.22 -0.44  0.39  0.50  0.49 -0.31  0.33  0.32
##      V12  V13  V14
## V1 -0.62 -0.79 -0.63
## V2 -0.07  0.07  0.64
## V3 -0.56 -0.37 -0.19
## V4 -0.07  0.00  0.22
## V5 -0.27 -0.28 -0.44
## V6  0.06  0.07  0.39
```

```
## V7  0.43  0.70  0.50
## V8  0.54  0.79  0.49
## V9 -0.26 -0.50 -0.31
## V10 0.30  0.52  0.33
## V11 -0.52 -0.43  0.32
## V12 1.00  0.57  0.24
## V13 0.57  1.00  0.31
## V14 0.24  0.31  1.00
```

#why same results??

#should I put true or false?

```
ws.pca = prcomp(ws,center=T,scale=F)
print(ws.pca)
```

```
## Standard deviations (1, ..., p=14):
## [1] 2.3528595 1.5802140 1.2025283 0.9632849 0.9367510 0.8202307 0.7441767
## [8] 0.5916369 0.5427172 0.5121583 0.4752351 0.4108488 0.3599502 0.2404419
##
## Rotation (n x k) = (14 x 14):
##          PC1          PC2          PC3          PC4          PC5
## V1  0.393669533 -0.005690412  0.001217953 -0.12246373  0.15758395
## V2 -0.136325011 -0.484160868 -0.207400812  0.08191848 -0.25089415
## V3  0.222676383 -0.223590947  0.088796064 -0.46988824 -0.18860015
## V4 -0.002257932 -0.315855884  0.626102363  0.24984122 -0.09352360
## V5  0.224298489  0.011615737  0.611989600 -0.07199322  0.04656750
## V6 -0.124630159 -0.300551432  0.130984580  0.16321412  0.77833048
## V7 -0.359264042 -0.067119829  0.146507749 -0.19098521 -0.14466563
## V8 -0.390711715  0.001313454  0.150962746 -0.14461667 -0.11200553
## V9  0.267001203 -0.026988703  0.169975512  0.32801272 -0.43257916
## V10 -0.279062504 -0.041222563  0.149879586 -0.46275771  0.09158820
## V11  0.089318293 -0.529782740 -0.137266298 -0.07211248 -0.04626960
## V12 -0.276822650  0.277907354  0.085328539  0.43466618 -0.02986657
## V13 -0.350526181  0.162776250  0.166204360 -0.15672341 -0.14419358
## V14 -0.269515252 -0.366058862 -0.126686846  0.25579490 -0.08440794
##          PC6          PC7          PC8          PC9          PC10
## V1 -0.20033864  0.05938234 -0.07179553 -0.162368819  0.19899373
## V2  0.13517139  0.09269887 -0.42154435 -0.450190708 -0.31127983
## V3  0.59841948 -0.37436980 -0.08757556 -0.006025687  0.32592413
## V4  0.10799983  0.16708856  0.17208034  0.262494455  0.12452347
## V5 -0.08811224  0.26872469 -0.41324857 -0.118633417 -0.15716811
## V6  0.14483831 -0.32957951  0.14881189 -0.252536278 -0.12773363
## V7 -0.14809748  0.03789829  0.36343884 -0.406373544  0.30772263
## V8 -0.06247252  0.06773223  0.17540500 -0.090919334  0.14044000
## V9 -0.25868639 -0.61111195  0.23075135 -0.159122818 -0.24054263
## V10 -0.46627764 -0.42292282 -0.34373920  0.265786794 -0.10869629
## V11 -0.42525454  0.18613617  0.04069617 -0.075264592  0.21704255
## V12  0.01565089 -0.19204101 -0.48362564 -0.212416815  0.50966073
## V13  0.21770365  0.07850980  0.06865116 -0.084264837 -0.45570504
## V14  0.06656550 -0.05420370 -0.11146671  0.544905394  0.04620802
##          PC11          PC12          PC13          PC14
## V1 -0.01444169  0.01575769 -0.49224318 -0.669045280
## V2  0.22154641 -0.26411262 -0.05610645 -0.090626055
## V3 -0.06839251  0.11921210  0.06675544  0.025225306
## V4  0.49452428 -0.04502305 -0.19201787  0.001635816
```

```
## V5 -0.47461722 -0.06131271 0.20007784 0.095361066
## V6 -0.07119731 0.06116074 0.05829909 -0.022300745
## V7 -0.29740957 -0.30087591 -0.35952714 0.253037788
## V8 0.03219187 -0.05001396 0.59834288 -0.601909165
## V9 -0.12200984 0.04266558 0.06403952 -0.082230935
## V10 0.23292405 -0.09334264 -0.11013538 0.058641979
## V11 -0.01972448 0.59795428 0.15917751 0.178821145
## V12 0.06140493 0.25774292 -0.04923091 0.022582562
## V13 -0.06646166 0.61109218 -0.32941979 -0.135092159
## V14 -0.55130818 -0.07268036 -0.17322892 -0.216043617
```