The 2000 World's ten largest companies (shown in the table) are available in file "10largest.txt" on Blackboard. In this example we find principal components PC1, PC2 of variables sales, profit

World's Ten Largest Companies

Company	Sales	Profits	Assets
Citigroup	108.28	17.05	1484.10
GE	152.36	16.59	750.33
American Intl. Group	95.04	10.91	766.42
Bank of America	65.45	14.14	1110.46
HSBC Group	62.97	9.52	1031.29
Exxon Mobil	263.99	25.33	195.26
Royal Dutch/Shell	265.19	18.54	193.83
BP	285.06	15.73	191.11
ING Group	92.01	8.10	1175.16
Toyota	165.68	11.13	211.15

Forbes, The Forbes Global 2000

- a) Find principal components (PC1, PC2) when the data is centered and not scaled. Find their covariance matrix. Verify that their eigenvalues do not add up to one.
- b) Find principal components when the data is centered and scaled. Verify that their eigenvalues add up to one.

```
d0=read.table("10largest.txt")
names(d0)=c("sales","profit","assets")
# only sales and profit
d1=d0[,c(1,2)]
# centering, no scaling
#-----
pr1=prcomp(d1)
names(pr1)
# [1] "sdev"
               "rotation" "center"
                                              "x"
                                    "scale"
# rotation matrix
rot = pr1$rotation
rot
#
             PC1
                         PC2
#sales 0.99917338 0.04065165
#profit 0.04065165 -0.99917338
# columns are eigenvectors (orthogonal)
d1b = scale(d1,scale=F)
        sales profit
# [1,] -47.323 2.346
# [2,] -3.243 1.886
# [3,] -60.563 -3.794
# [4,] -90.153 -0.564
# [5,] -92.633 -5.184
# [6,] 108.387 10.626
# [7,] 109.587 3.836
# [8,] 129.457 1.026
# [9,] -63.593 -6.604
#[10,] 10.077 -3.574
colMeans(d1b)
        sales
                    profit
#-4.263256e-15 -7.105427e-16
```

```
# transformed data (score vectors)
as.matrix(d1b)%*%rot
             PC1
                         PC2
# [1,] -47.188513 -4.2678188
# [2,] -3.163650 -2.0162743
# [3,] -60.667170 1.3288779
# [4,] -90.101405 -3.1013344
# [5,] -92.767166 1.4140305
# [6,] 108.729370 -6.2111060
# [7,] 109.652353 0.6220633
# [8,] 129.391697 4.2374888
# [9,] -63.808896 4.0133806
#[10,]
        9.923381 3.9806923
pr1$x
#
              PC1
                         PC2
# [1,] -47.188513 -4.2678188
# [2,] -3.163650 -2.0162743
# [3,] -60.667170 1.3288779
# [4,] -90.101405 -3.1013344
# [5,] -92.767166 1.4140305
# [6,] 108.729370 -6.2111060
# [7,] 109.652353 0.6220633
# [8,] 129.391697 4.2374888
# [9,] -63.808896 4.0133806
#[10,] 9.923381 3.9806923
 prcomp() agrees with eigen()
# eigenvalues
pr1$sdev^2
# [1] 7488.80605
                   13.83751
eigen(var(d1))
# $values
# [1] 7488.80605
                   13.83751
# $vectors
              [,1]
                          [,2]
# [1,] -0.99917338 0.04065165
# [2,] -0.04065165 -0.99917338
# eigen(var(d1b)) same results
```

```
d1x = pr1$x
var(d1x)
                       PC2
            PC1
#PC1 7.488806e+03 1.275985e-15
#PC2 1.275985e-15 1.383751e+01
# PCs uncorrelated
\mbox{\tt\#} but sum of eigenvals do not add up to p=2
# centering and scaling
pr3=prcomp(d1,scale=T)
d3x = pr3$x
var(d3x)
            PC1
                       PC2
#PC1 1.686136e+00 1.387252e-16
#PC2 1.387252e-16 3.138640e-01
sum(diag(var(d3x)))
                  # [1] 2
```