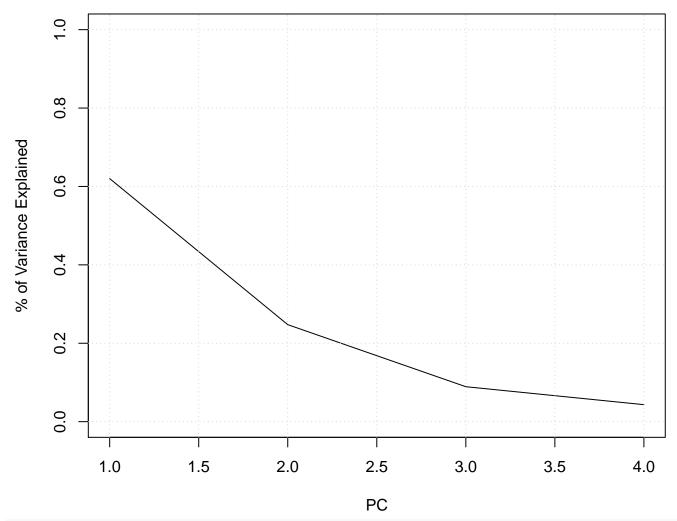
```
# pca.r
d1=USArrests
dim(d1)
                  # [1] 50 4
## [1] 50 4
head(d1)
##
             Murder Assault UrbanPop Rape
## Alabama
              13.2
                       236
                              58 21.2
## Alaska
              10.0
                       263
                                48 44.5
## Arizona
              8.1
                       294
                              80 31.0
## Arkansas
               8.8
                      190
                               50 19.5
## California
               9.0
                       276
                                91 40.6
               7.9
                       204
                               78 38.7
## Colorado
#
# a) Original data -no scaling-
# covariance matrix
var(d1)
##
              Murder
                      Assault UrbanPop
                                             Rape
## Murder
          18.970465 291.0624
                                4.386204 22.99141
## Assault 291.062367 6945.1657 312.275102 519.26906
## UrbanPop 4.386204 312.2751 209.518776 55.76808
## Rape
            22.991412 519.2691 55.768082 87.72916
apply(d1,2,var)
##
      Murder
               Assault
                        UrbanPop
                                       Rape
    18.97047 6945.16571 209.51878
##
                                   87.72916
#
# column variances appear on main diagonal of cov matrix
#
# eigenvalues
eigen(var(d1))
## eigen() decomposition
## $values
## [1] 7011.114851 201.992366 42.112651
                                          6.164246
##
## $vectors
                         [,2]
##
              [,1]
                                    [,3]
## [1,] -0.04170432 0.04482166 0.07989066 0.99492173
## [3,] -0.04633575 -0.97685748 -0.20054629 0.05816914
## [4,] -0.07515550 -0.20071807 0.97408059 -0.07232502
sum(eigen(var(d1))$values)
## [1] 7261.384
sum(apply(d1,2,var))
## [1] 7261.384
# sum of eigenvalues = sum variances
```

```
# b) SCALED DATA
m1=prcomp(d1, scale=T)
names(m1)
## [1] "sdev" "rotation" "center" "scale"
                                                  "x"
# "sdev": square-root of eigenvalues of columns of transformed data (of PCs)
# "rotation": matrix of eigenvectors
# "center" "scale": # mean and sd of original data -unscaled-
# "x": transformed data set
# means -unscaled data
#
m1$center
##
    Murder Assault UrbanPop
   7.788 170.760 65.540
##
                               21.232
# standard deviations -unscaled data
#
m1$scale
     Murder
             Assault UrbanPop
## 4.355510 83.337661 14.474763 9.366385
apply(d1,2,sd)
     Murder
             Assault UrbanPop
                                     Rape
## 4.355510 83.337661 14.474763 9.366385
#
# rotation matrix has the eigenvectors
#
m1$rotation
##
                  PC1
                             PC2
                                        PC3
## Murder -0.5358995 0.4181809 -0.3412327 0.64922780
## Assault -0.5831836 0.1879856 -0.2681484 -0.74340748
## UrbanPop -0.2781909 -0.8728062 -0.3780158 0.13387773
          -0.5434321 -0.1673186 0.8177779 0.08902432
## Rape
#
eigen(var(scale(d1)))
## eigen() decomposition
## $values
## [1] 2.4802416 0.9897652 0.3565632 0.1734301
##
## $vectors
              [,1]
                        [,2]
                                   [,3]
##
## [1,] -0.5358995  0.4181809 -0.3412327  0.64922780
## [2,] -0.5831836  0.1879856 -0.2681484 -0.74340748
## [3,] -0.2781909 -0.8728062 -0.3780158 0.13387773
## [4,] -0.5434321 -0.1673186  0.8177779  0.08902432
# squared-root of eigenvalues of scaled data
```

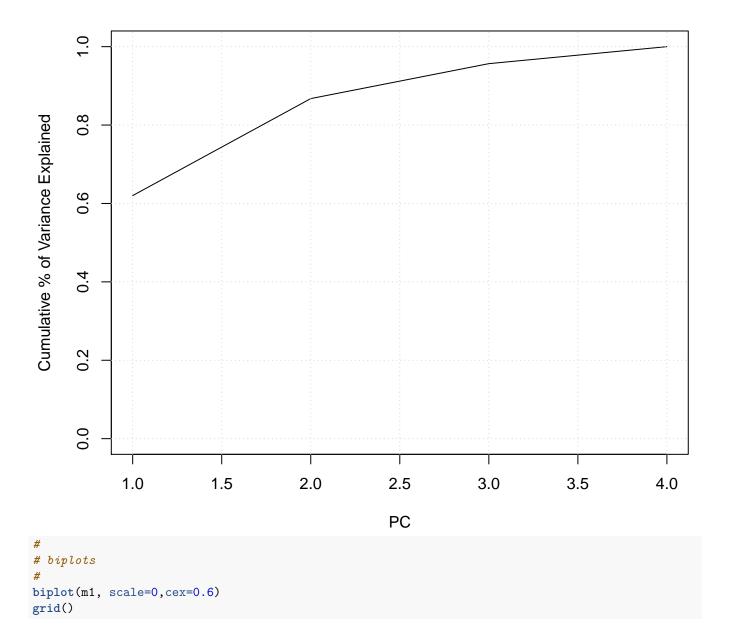
```
m1$sdev
## [1] 1.5748783 0.9948694 0.5971291 0.4164494
# transformed data on PC axes
d2 = m1$x
dim(d2)
## [1] 50 4
head(d2)
                               PC2
                                                        PC4
##
                    PC1
                                           PC3
## Alabama
             -0.9756604 1.1220012 -0.43980366 0.154696581
            -1.9305379 1.0624269 2.01950027 -0.434175454
## Alaska
## Arizona -1.7454429 -0.7384595 0.05423025 -0.826264240
## Arkansas 0.1399989 1.1085423 0.11342217 -0.180973554
## California -2.4986128 -1.5274267 0.59254100 -0.338559240
## Colorado -1.4993407 -0.9776297 1.08400162 0.001450164
apply(d2,2,mean)
                                        PC3
##
            PC1
                          PC2
                                                      PC4
## 3.695720e-17 3.619582e-16 2.375205e-16 -1.916184e-16
#
# PCs are centered (their means are zero)
#
# eigenvalues of scaled data
#
m1$sdev^2
## [1] 2.4802416 0.9897652 0.3565632 0.1734301
apply(d2,2,var)
##
        PC1
                  PC2
                            PC3
## 2.4802416 0.9897652 0.3565632 0.1734301
# eigenvalues of cov matrix (of scaled data) = variances of Principal components
#
# eigen() function
cova=var(scale(d1))
m2 = eigen(cova)
# covariance matrix of transformed data
var(d2)
                PC1
                              PC2
                                            PC3
## PC1 2.480242e+00 -3.812359e-16 1.126674e-16 1.778258e-17
## PC2 -3.812359e-16 9.897652e-01 -2.024956e-16 9.907132e-17
## PC3 1.126674e-16 -2.024956e-16 3.565632e-01 -1.564569e-16
## PC4 1.778258e-17 9.907132e-17 -1.564569e-16 1.734301e-01
round(var(d2),5)
          PC1
                  PC2
                          PC3
                                  PC4
## PC1 2.48024 0.00000 0.00000 0.00000
```

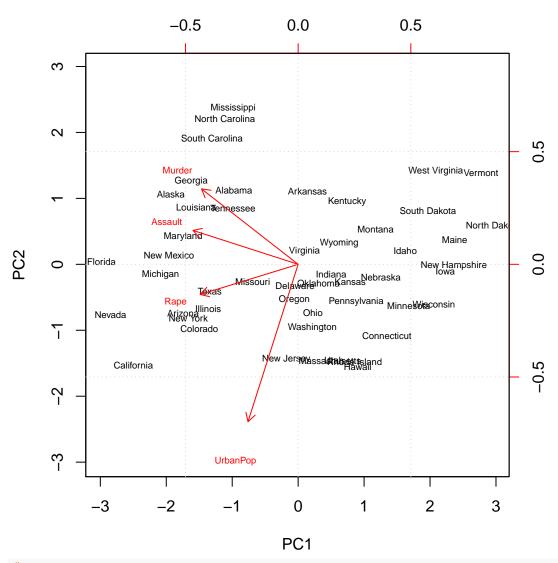
```
## PC2 0.00000 0.98977 0.00000 0.00000
## PC3 0.00000 0.00000 0.35656 0.00000
## PC4 0.00000 0.00000 0.00000 0.17343
# This is Big lambda diagonal matrix (eigenvalues on main diagonal)
sum(diag(var(d2)))
## [1] 4
# covariances (off diagonal) all equal to 0 (PCs uncorrelated)
# PC1 with largest variance across states
# Use eigenvectors to define the PC variables.
m1$rotation
##
                 PC1
                          PC2
                                    PC3
                                               PC4
          ## Murder
## Assault -0.5831836 0.1879856 -0.2681484 -0.74340748
## UrbanPop -0.2781909 -0.8728062 -0.3780158 0.13387773
          -0.5434321 -0.1673186  0.8177779  0.08902432
## Rape
# Score vectors are PC1, PC2, defined as follows
# PC1 = 0.536 Murder + 0.58Assault + 0.28 UrbanPop + 0.543 Rape
# A weighted average of crime rates (almost exclude UrbanPop)
# PC2 = 0.4 Murder - 0.87 UrbanPop
# Weighted average of Urban Pop and Murder
# transformed variables in the principal components space.
#-----
# eigenvectors span a new p-dimensional space
# score vectors are the transformed data in this new space
d2 = m1\$x
head(d2)
                            PC2
##
                  PC1
                                       PC3
                                                   PC4
## Alabama
            -0.9756604 1.1220012 -0.43980366 0.154696581
## Alaska -1.9305379 1.0624269 2.01950027 -0.434175454
## Arizona
          -1.7454429 -0.7384595 0.05423025 -0.826264240
            0.1399989 1.1085423 0.11342217 -0.180973554
## Arkansas
## California -2.4986128 -1.5274267 0.59254100 -0.338559240
           -1.4993407 -0.9776297 1.08400162 0.001450164
## Colorado
tail(m1$x)
##
                    PC1
                              PC2
                                         PC3
                                                   PC4
## Vermont
               2.7732561 1.3881944 0.83280797 -0.1434337
## Virginia
               0.0953667  0.1977278  0.01159482  0.2092464
## Washington
               ## West Virginia 2.0873931 1.4105263 0.10372163 0.1305831
               2.0588120 -0.6051251 -0.13746933 0.1822534
## Wisconsin
               ## Wyoming
# Variance of the PCs are the eigenvalues
```

```
#-----
apply(d2,2,var)
              PC2
                        PC3
       PC1
                                PC4
## 2.4802416 0.9897652 0.3565632 0.1734301
m2$values
## [1] 2.4802416 0.9897652 0.3565632 0.1734301
\# proportion of variance explained (PVE) by each PC
# variance of PCs
aux=m1$sdev^2
sum(aux) # 4
## [1] 4
pve=aux/sum(aux)
## [1] 0.62006039 0.24744129 0.08914080 0.04335752
m2$values/4
## [1] 0.62006039 0.24744129 0.08914080 0.04335752
# each eigenvalue divided by 4
cumsum(pve)
## [1] 0.6200604 0.8675017 0.9566425 1.0000000
\# 87% variability in the dataset explained by PC1 and PC2
# plots
plot(pve, xlab="PC", ylab="% of Variance Explained", ylim=c(0,1),type='1')
grid()
```



plot(cumsum(pve), xlab="PC", ylab="Cumulative % of Variance Explained", ylim=c(0,1),type='l')
grid()





```
head(d2)
                     PC1
                                PC2
                                            PC3
##
## Alabama
              -0.9756604 1.1220012 -0.43980366 0.154696581
## Alaska
              -1.9305379 1.0624269
                                    2.01950027 -0.434175454
## Arizona
              -1.7454429 -0.7384595
                                     0.05423025 -0.826264240
## Arkansas
               0.1399989 1.1085423
                                     0.11342217 -0.180973554
                                    0.59254100 -0.338559240
## California -2.4986128 -1.5274267
## Colorado
              -1.4993407 -0.9776297 1.08400162 0.001450164
# rowname is State name, located at (PC1,PC2) coordinates
# Rotate original axes (red colored)
#
# mirror image
m1$rotation=-m1$rotation
m1$x=-m1$x
biplot(m1, scale=0,cex=0.6)
grid()
#
```

```
rot=m1$rotation

#

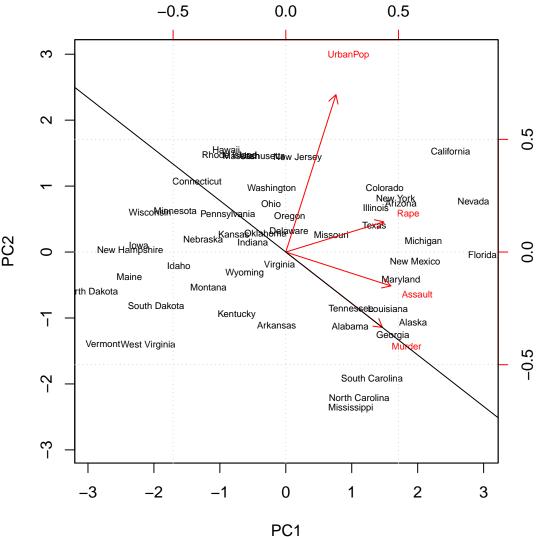
# Murder axis identified

#

slope1=rot[1,2]/rot[1,1]
slope1 # -0.7803345

## [1] -0.7803345

abline(0,slope1)
```



```
## Murder 0.5358995 -0.4181809 0.3412327 -0.64922780

## Assault 0.5831836 -0.1879856 0.2681484 0.74340748

## UrbanPop 0.2781909 0.8728062 0.3780158 -0.13387773

## Rape 0.5434321 0.1673186 -0.8177779 -0.08902432

# states with large values in PC1 have high crime rates

# (PC1 weights -col1- in rotation are 0.5359, 0.5831, 0.5434)

# California, Nevada, Florida vs North Dakota, Vermont
```

```
# states with large values in PC2 have large urban areas
# (PC2 largest weight -col2- in rotation is 0.8728)
# California vs Mississippi
# original vs transformed values
d3=data.frame(d1,d2)
head(d3)
##
               Murder Assault UrbanPop Rape PC1 PC2
                                                                                  PC3
## Alabama
                13.2 236 58 21.2 -0.9756604 1.1220012 -0.43980366
## Alaska 10.0 263 48 44.5 -1.9305379 1.0624269 2.01950027 ## Arizona 8.1 294 80 31.0 -1.7454429 -0.7384595 0.05423025 ## Arkansas 8.8 190 50 19.5 0.1399989 1.1085423 0.11342217 ## California 9.0 276 91 40.6 -2.4986128 -1.5274267 0.59254100 ## Colorado 7.9 204 78 38.7 -1.4993407 -0.9776297 1.08400162
##
                         PC4
## Alabama 0.154696581
## Alaska -0.434175454
## Arizona -0.826264240
## Arkansas -0.180973554
## California -0.338559240
## Colorado 0.001450164
tail(d3)
```

##		Murder	Assault	UrbanPop	Rape	PC1	PC2	PC3
##	Vermont	2.2	48	32	11.2	2.7732561	1.3881944	0.83280797
##	Virginia	8.5	156	63	20.7	0.0953667	0.1977278	0.01159482
##	Washington	4.0	145	73	26.2	0.2147234	-0.9603739	0.61859067
##	West Virginia	5.7	81	39	9.3	2.0873931	1.4105263	0.10372163
##	Wisconsin	2.6	53	66	10.8	2.0588120	-0.6051251	-0.13746933
##	Wyoming	6.8	161	60	15.6	0.6231006	0.3177866	-0.23824049
##			PC4					
##	Vermont	-0.1434	337					
##	Virginia	0.2092	464					
##	Washington	-0.2186	282					
##	West Virginia	0.1305	831					
##	Wisconsin	0.1822	534					
##	Wyoming	-0.1649	769					