Use PCA to find the principal components and associated eigenvalues. (8 pts)

What is the minimum number of eigenvalues to keep (> 1.0 rule)? (1 pt)

What is the minimum number of eigenvalues to keep (scree plot)? (3 pts)

Interpret the first two principal components. (3 pts)

- > antho<-read.csv('anthro.csv', header=T)
- > rownames(antho)<-colnames(antho)
- > antho

HeadL HeadB FaceB LeftFL LeftFAL LeftFTL Height

HeadL 1.000 0.402 0.396 0.301 0.305 0.339 0.340

HeadB 0.402 1.000 0.618 0.150 0.135 0.206 0.183

FaceB 0.396 0.618 1.000 0.321 0.289 0.363 0.345

LeftFL 0.301 0.150 0.321 1.000 0.846 0.759 0.661

LeftFAL 0.305 0.135 0.289 0.846 1.000 0.797 0.800

LeftFTL 0.339 0.206 0.363 0.759 0.797 1.000 0.736

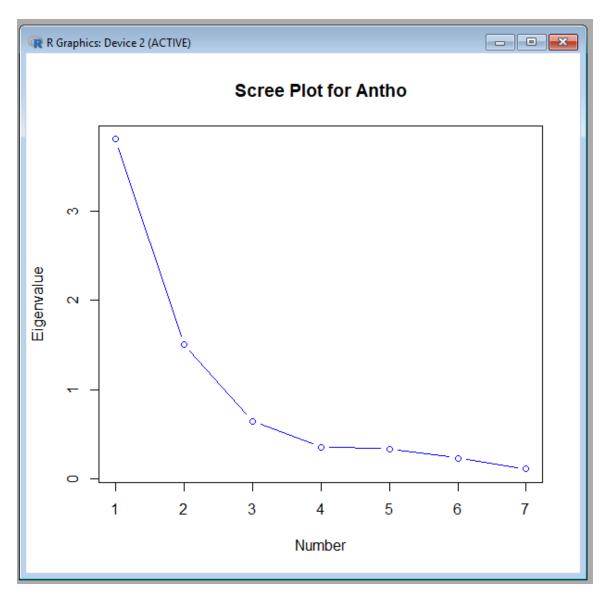
Height 0.340 0.183 0.345 0.661 0.800 0.736 1.000

- > eigenAntho<-eigen(antho)
- > std<-sqrt(eigenAntho\$values)
- > proportionVar<-eigenAntho\$values/sum(eigenAntho\$values)
- > cumulativeProp<-cumsum(eigenAntho\$values)/sum(eigenAntho\$values)
- > Loadings<-eigenAntho\$vectors
- >> #Importance of components:
- > #Standard deviation
- > std
- [1] 1.9492241 1.2256950 0.8061063 0.6000474 0.5823766 0.4850290 0.3375164
- > #Proportion of Variance

- > proportionVar
- $[1]\ 0.54278208\ 0.21461832\ 0.09282963\ 0.05143670\ 0.04845178\ 0.03360759\ 0.01627391$
- > #Cumulative Proportion
- > cumulativeProp
- [1] 0.5427821 0.7574004 0.8502300 0.9016667 0.9501185 0.9837261 1.0000000
- > #Loadings:
- > Loadings
 - [,1] [,2] [,3] [,4] [,5] [,6] [,7]
- $[1,] 0.2763037 0.3647677 \quad 0.882274766 0.08573946 0.06740350 \quad 0.005384671 0.01638732$
- [2,] -0.2118636 -0.6392041 -0.257527788 0.68707351 0.08129399 0.034955657 0.01762744
- $[3,] 0.2951449 0.5123928 0.381447691 0.69856220 0.10071831 \ 0.033740772 0.07462604$
- [4,] -0.4375581 0.2349399 -0.069924234 0.10160027 -0.61923662 0.318242311 0.50339046
- [5,] -0.4557045 0.2766674 -0.036669136 0.11311530 -0.03907675 0.290305975 -0.78475748
- [6,] -0.4502341 0.1784374 -0.059124621 0.05299938 -0.03440885 -0.870489463 0.01445146
- [7,] -0.4356893 0.1795404 -0.006212105 -0.08162701 0.76976550 0.233030117 0.35269900
- >> eigenAntho\$values
- [1] 3.7994745 1.5023283 0.6498074 0.3600569 0.3391625 0.2352531 0.1139173

#The first two eigenvalues account for 0.757 of the total correlation and are the only eigenvalues greater than 1.0

- > layout(1)
- > plot(1:7, eigenAntho\$values, main='Scree Plot for Antho',
- + xlab='Number', ylab='Eigenvalue', col='blue', typ='b')



The usable number of principal components can be chosen by the screening criteria. The 'scree' plot is very sharp in slope for the first two components. The distinction is not due to random effects.

Use MDS to analyze this matrix. Show the output. (10 pts)

How many coordinates are need to account for at least 80% of the sum of absolute eigenvalues? (1 pt)

Using the first two coordinates, plot the location of the flowers relative to each other. (4 pts)

- > flowers<-read.csv('flowers.csv', header=T)
- > rownames(flowers)<-colnames(flowers)
- > flowersMDS<- cmdscale(flowers, k=3, eig=TRUE)
- > flowersMDS

\$`points`

[,1] [,2] [,3]

Bgn 0.441626810 -0.13910199 -0.024866489

Brm -0.411002097 -0.10646858 0.112134733

Cml 0.135998713 -0.05878727 -0.300985873

Dhl -0.117861814 -0.32989431 -0.146573607

Fmn 0.267792041 0.12142878 -0.000862644

Fch 0.295445879 -0.10140928 -0.284987056

Grn 0.167256180 -0.34946203 -0.053768686

Gld -0.061150521 -0.25197021 0.182807089

Hth 0.131374192 0.27948118 0.188840138

Hyd -0.173171040 0.35681051 -0.212378459

Irs 0.267620864 0.18687636 0.084102966

Lly -0.001753092 0.22650611 0.071416733

Lvy 0.051226880 0.34595178 0.113183286

Pny 0.015761013 0.11587345 -0.058215025

Pnc -0.178928108 0.09234363 0.155484871

Rdr -0.427237687 -0.04816267 -0.218569105

Scr -0.491784734 -0.01604355 0.006881662

Tlp 0.088786521 -0.32397190 0.386355466

\$eig

[1] 1.173085e+00 8.944353e-01 5.732009e-01 4.843006e-01 2.638516e-01 2.298165e-01 8.383417e-02 6.645861e-02 2.984017e-02 -2.775558e-17

[11] -2.147959e-02 -4.094094e-02 -4.366468e-02 -8.867401e-02 -1.045334e-01 -1.275889e-01 -1.714295e-01 -2.045124e-01

\$x

NULL

\$ac

[1] 0

\$GOF

- [1] 0.5738645 0.6951420
- >> cumulativeProp<-cumsum(abs(flowersMDS\$eig))/sum(abs(flowersMDS\$eig))
- > cumulativeProp
- $[1] \ 0.2549273 \ 0.4493002 \ 0.5738645 \ 0.6791096 \ 0.7364481 \ 0.7863903 \ 0.8046086 \ 0.8190510 \ 0.8255356 \ 0.8255356 \ 0.8302034 \ 0.8391005 \ 0.8485894 \ 0.8678594$
- [15] 0.8905760 0.9183028 0.9555567 1.0000000

7 coordinates are need to account for at least 80% of the sum of absolute eigenvalues.

- > layout(1)
- > plot(flowersMDS\$points[,1], flowersMDS\$points[,2], type='n', main='MDS 2-D Plot of flowers', xlab='EW', ylab='NS')
- > text(flowersMDS\$points[,1], flowersMDS\$points[,2], labels=colnames(flowers), col='blue')

