

Assignment - 9

BSAN 450 (Spring 2023)

This assignment is due on April 27, 2022 at 9 PM Central. The total points possible are 100 and there is one (1) problem with several sub-parts. Each sub-part carries equal points. You can form groups to attempt these problems. Each group will submit one copy of the assignment on Canvas, either in word or pdf, and the assignment should clearly include the names of the group members.

1. In this example we will analyze the NCI60 cancer cell line microarray data, which consists of 6,830 gene expression measurements on 64 cancer cell lines. To read this data in R, you will need the following code:

```
library(ISLR) nci.labs=NCI60$labs  
nci.data=NCI60$data
```

The data `nci.data` has 64 rows and 6,830 columns. Each row is a cell line for which we have gene expressions recorded for 6,830 genes. The `nci.labs` data has the cancer type for each of these 64 cell lines. Your analysis will rely primarily on `nci.data` and not on `nci.labs` as the latter are just labels.

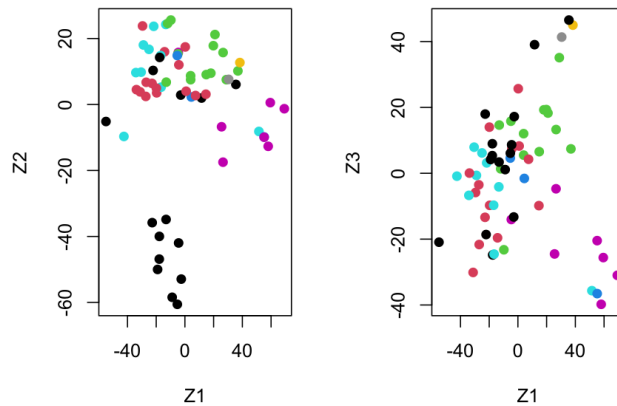
a. Perform PCA on `nci.data` with `scale=TRUE` in the `prcomp` function.

```
pr.out1 = prcomp(nci.data, scale=TRUE)
```

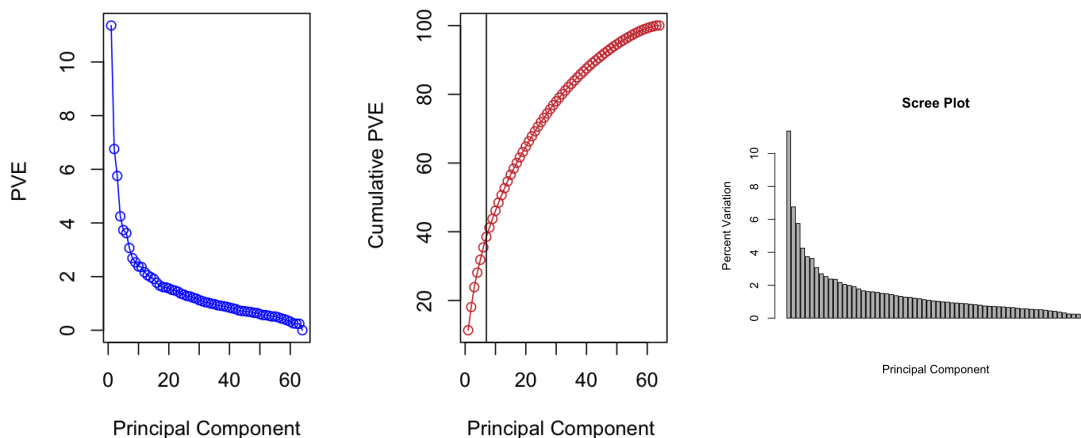
b. Plot the first two principal components. You may require the following code wherein `pr.out` is the name of the output from the `prcomp` function that you used in part (a).

```
par(mfrow=c(1,2))  
  
plot(pr.out$x[,1:2], col=as.factor(nci.labs), pch=19, xlab="Z1", ylab="Z2")  
plot(pr.out$x[,c(1,3)], col=as.factor(nci.labs), pch=19,  
xlab="Z1", ylab="Z3")
```

You will notice from your plot that observations belonging to a single cancer type tend to lie near each other in this low-dimensional space.



- c. As demonstrated in the lecture, plot the PVE of each principal component (i.e. a scree plot) and the cumulative PVE of each principal component. What is the % of variance explained by the first seven principal components? Based on the scree plot, how many principal components should we choose?



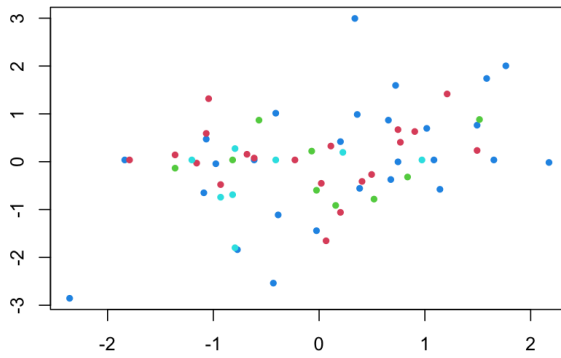
About 40% of the variance is explained by the first seven principal components. Based on the scree plot, we would recommend using the first four principal components.

- d. Scale the nci.data such that the features have zero mean and 1 standard deviation. You will have to use the `scale()` function in R for this purpose. Then, on the scaled data, perform K-means clustering with $K = 4$. What are the cluster sizes that you obtain? Calculate the Between Cluster sum of squares as a % of total sum of squares.

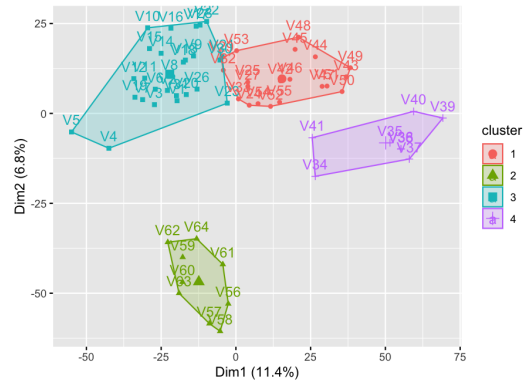
Cluster sizes: 20, 9, 27, 8

Between cluster ss: 19.92218%

K-Means Clustering Results with K=2



Cluster plot



Maria Emma Arcia, Corinne Scales, Olin Yoder