## Stat 432 HW 07

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Include the R code for this HW.

```
knitr::opts_chunk$set(echo = TRUE)
library(ISLR2)
library(GGally)
library(tibble)
library(dplyr)
library(knitr)
library(kableExtra)
#add more libraries as needed.
```

### Question 1 (K-means clustering)

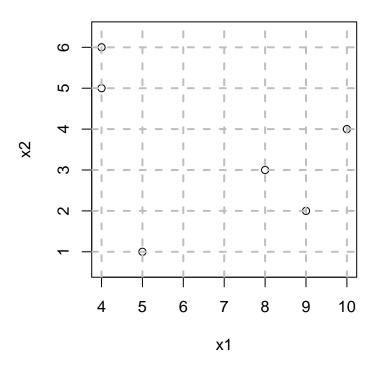
In this question, we will replicate the image pixels K-means clustering example in our lectures. Perform the following:

- (a) Choose a colored .jpg image by your own. The resolution should be at least  $200 \times 200$  but no larger than  $800 \times 800$ . You may reduce the resolution of your image before loading it into R. Plot this image in R.
- (b) "vectorize' this image dataset by converting it to a matrix with dimension: Number of Pixels  $\times$  3, where 3 represents the dimension of (RGB) colors.
- (c) Perform k-means with k = 2, 5, 10, 30, 50, 100 by treating the pixels as observations. For each k setting, after obtaining the clusters, reconstruct a new image that replaces the pixel colors in each cluster by its cluster mean color. Plot these six new images. Use par(mfrow = c(2, 3)) to organize your 6 new figures with two rows of three images per row. Label each figure with their corresponding k.

## Question 2 (K-means clustering)

In this problem, you will perform K-means clustering manually, with K=2, on a small example. The observations are as follows:

Obs	x1	x2	initial.cluster
1	5	1	1
2	9	2	1
3	8	3	2
4	4	6	2
5	4	5	2
6	10	4	2

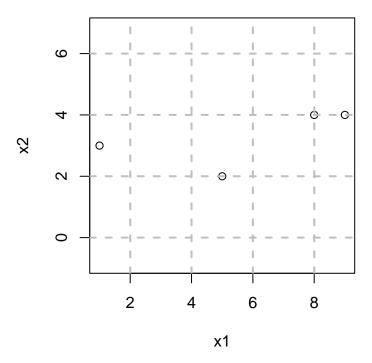


- (a) Compute the centroid for the initial cluster.
- (b) Assign each observation to the centroid to which it is closest, in terms of Euclidean distance. Report the cluster labels for each observation.
- (c) Repeat (a) and (b) until the answers obtained stop changing.
- (d) Plot the observations according to the cluster labels obtained.

# Question 3 (Hierarchical clustering)

In this problem, you will perform hierarchical clustering manually on a small example. The observations are as follows:

Obs	x1	x2
1	1	3
2	5	2
3	9	4
4	8	4



- (a) Conduct the hierarchical clustering based on complete linkage. For each step of the hierarchical clustering, report the new fusion and the corresponding dissimilarity (distance).
- (b) Conduct the hierarchical clustering based on single linkage. For each step of the hierarchical clustering, report the new fusion and the corresponding dissimilarity (distance).

# Question 4 (PCA)

(a) Generate a simulated data set with 20 observations in each of three classes (i.e. 60 observations total), and 50 variables.

Hint: There are a number of functions in R that you can use to generate data. One example is the rnorm() function; runif() is another option. Be sure to add a mean shift to the observations in each class so that there are three distinct classes.

(b) Perform PCA on the 60 observations and plot the first two principal component score vectors. Use a different color to indicate the observations in each of the three classes.

If the three classes do not appear separated in this plot, then return to part (a) and modify the simulation so that there is greater separation between the three classes.