# ENGINEERING DIVISION | NYU ABU DHABI

# UH-3332 - Applied Machine Learning

Clustering

Due Date: Refer to Brightspace

#### ENGINEERING DIVISION | NYU ABU DHABI

## K-means clustering

#### Introduction

K-means is one of the widely used unsupervised learning algorithms that solve the well-known clustering problem. The procedure follows a simple and easy way to classify a given data set into a certain number of clusters (assume k clusters). The main idea is to define k centers, one for each cluster. These centers should be placed in a cunning way because different locations will result in different results.

#### **Dataset**

Test your algorithm on two separate datasets (1) Use make\_blobs function to generate synthetic data set from sklearn library. (2) Use an RGB image to cluster the R,G,B data into K clusters to demonstrate image compression. Display images before and after

# Requirements

- 1. Use sklearn library to generate the synthetic data for k-means clustering.
  - α. We set the total number of instances to be 300
  - β. The number of centers is 4 with the standard deviation 0.6
- 2. Plot the generated data with labels by using matplotlib
- 3. Implement the K-means function return the labels and centers
- 4. Fit the model on the dataset (default seed) and plot the figure
- 5. Fit the model on the dataset (seed=2) and plot the figure
- 6. Implement the K-means++ function return the labels and centers
- 7. Fit the model on the dataset (default seed) and plot the figure
- 8. Fit the model on the dataset (seed=2) and plot the figure
- 9. Compare the results from 4,5,7 and 8. State your observations

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# **Hierarchical clustering**

#### Introduction

Hierarchical clustering involves creating clusters that have a predetermined ordering. For example, all files and folders on the hard disk are organized in a hierarchy.

### **Dataset**

In this project you will work on the Mall Customer dataset (Mall Customers.csv)

# Requirements

Implement a hierarchical clustering model using Ward distance and plot the dendrogram.

## **Deliverables**

A .ipynb file containing the following:

- 1. Source code
- 2. Detailed description of the project if needed

Before submitting your project, please make sure to test your program on the given dataset.

#### **Notes**

You may discuss the general concepts in this project with other students, but you must implement the program on your own. **No sharing of code or report is allowed.** Violation of this policy can result in a grade penalty.

Late submission is acceptable with the following penalty policy:

10 points deduction for every day after the deadline