Lab 2

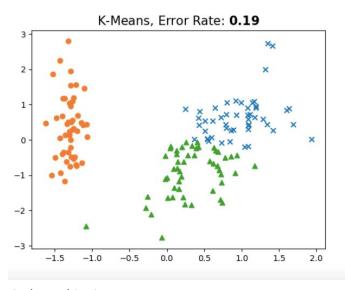
It is an individual programming assignment. This lab assignment is graded based on 100 points and is an individual effort (e.g, no teamwork is allowed).

Part 1 (10x6 = 60 pts) KNN & K-mean

For the datasets provided, complete all the steps requested: **NOTE: You should use both your own NumPy implementation of the algorithms and sklearn.**

Step 1: Load the data points and ground truth values (Some ground truth values may be in a different file)

Step 2 (1 pts): Plot the points according to the ground truth value classes. For example if there are 3 different classes, expected output:



Step 3: Split data into train and test.

Step 4 (2 pts): Use KNN (own NumPy code & sklearn library) to clusters the data.

Step 5 (1 pts): Plot your result in different colors similar to Step 2.

Step 6 (1 pts): Print the confusion matrix to check model performance.

Step 7 (4 pts): Repeat Steps 3 - 6 and use K-means **(own NumPy code & sklearn library)** to perform the same. Additionally, print the centroids.

Step 8 (1pts): Compare the results between KNN and K-means

Datasets:

- 1. Dataset 1/S1.txt: N=5000 vectors and k=15 Gaussian cluster
- 2. Dataset 1/S2.txt: N=5000 vectors and k=15 Gaussian cluster
- 3. **Dataset 1/S3.txt**: N=5000 vectors and k=15 Gaussian cluster
- 4. Dataset 1/S4.txt: N=5000 vectors and k=15 Gaussian cluster

NOTE: Ground Truth and original centroid for S1-4 are present in the **dataset_1/s-originals** folder.

- Dataset_2/dim32.txt: 32 dimensions;
- 6. **Dataset_3/spiral.txt:** N=312, k=3, D=2; Ground truth is the third column present in the same file.

Part 2 (30 pts): Movie Recommender

Context

In this lab, you will be implementing a simple movie recommender system.

Dataset details

You will be using the ml-m1 dataset from the MovieLense website.

You will be using movies.dat and rating.dat for building your recommender.

Steps:

- 1. Create m x u matrix with movies as row and users as column
- 2. Normalize the matrix
- 3. Compute SVD to get U, S, and V.

Use np.linalg.svd()

- 4. From your V.T select 50 components
- 5. Implement a function that takes movieID as input and then implement cosine similarity along with sorting to recommend the top 10 movies.

Project Report (10 pts)

- 1. For each dataset in part 1, show the two plots (original & your predictions). KNN, K-means comparison with a reason for better performance(if any).
- 2. For part2, take any 3 random movie IDs. Compute and report the top 10 recommended movies. Describe if the movies are actually similar.

Deliverables

- 1. Project report
- 2. **Fully commented** Jupyter notebook