$\ensuremath{\text{\# 1.}}$ Importing the necessary libraries and Modules

import pandas as pd
import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

import math
import random

2. Importing the Dataset

Read the Excel file without specifying usecols to inspect the actual column names
dataset = pd.read_excel("/content/default_of_credit_card_clients.xls", skiprows = 1)
dataset

	ID	LIMIT_BAL	SEX	EDUCATION	MARRIAGE	AGE	PAY_0	PAY_2	PAY_3	PAY_4	•••	BILL_AMT3	BILL_AMT4	BILL_AMT5	BILL_AMT6
0	1	20000	2	2	1	24	2	2	-1	-1		689	0	0	0
1	2	120000	2	2	2	26	-1	2	0	0		2682	3272	3455	3261
2	3	90000	2	2	2	34	0	0	0	0		13559	14331	14948	15549
3	4	50000	2	2	1	37	0	0	0	0		49291	28314	28959	29547
4	5	50000	1	2	1	57	-1	0	-1	0		35835	20940	19146	19131
29995	29996	220000	1	3	1	39	0	0	0	0		208365	88004	31237	15980
29996	29997	150000	1	3	2	43	-1	-1	-1	-1		3502	8979	5190	(
29997	29998	30000	1	2	2	37	4	3	2	-1		2758	20878	20582	19357
29998	29999	80000	1	3	1	41	1	-1	0	0		76304	52774	11855	48944
29999	30000	50000	1	2	1	46	0	0	0	0		49764	36535	32428	15313
30000 rc	ws × 24	columns													
4															

3. Geting the Statistics of a Dataset

dataset.describe()



	PAY_3	PAY_2	PAY_0	AGE	MARRIAGE
3	30000.000000	30000.000000	30000.000000	30000.000000	00.000000
	-0.166200	-0.133767	-0.016700	35.485500	1.551867
	1.196868	1.197186	1.123802	9.217904	0.521970
	-2.000000	-2.000000	-2.000000	21.000000	0.000000
	-1.000000	-1.000000	-1.000000	28.000000	1.000000
	0.000000	0.000000	0.000000	34.000000	2.000000
	0.000000	0.000000	0.000000	41.000000	2.000000
	8.000000	8.000000	8.000000	79.000000	3.000000

4. Pre-process the data

def Feature_Normalization(X):

X = (X - np.mean(X)) / np.std(X) # Calculate mean and std across the entire 1D array return X

x = dataset.iloc[: , :].values

 $\mbox{\tt\#}$ Initialize $x_{\tt norm}$ as a list to store normalized features $x_{\tt norm}$ = []

for i in range(x.shape[1]): # Iterate through columns of x
 norm_feature = Feature_Normalization(x[:, i])
 x_norm.append(norm_feature) # Append normalized feature to the list

```
# 5. Find the the no. of Clusters
# Randomly initialize the centroids by selecting 3 unique points from the dataset
# `random.sample(range(0, len(dataset)), 3)` will select 3 random indices without replacement from the dataset.
# Each selected index represents a row in the dataset, which will serve as an initial centroid.
init_centroids = random.sample(range(0, len(dataset)), 2)
# Create an empty list to store the initial centroid points.
centroids = []
for i in init_centroids:
    # Append the data point at each randomly chosen index to the centroids list.
    centroids.append(dataset.iloc[i])
# Convert the list of centroids to a NumPy array for easier manipulation in further calculations.
centroids = np.arrav(centroids)
# Print the initialized centroids
centroids
⇒ array([[-1.49343194, 0.63598104, -1.23432296, 0.18582826, 0.85855728,
              0.48976158, \quad 0.90471219, \quad -1.55887596, \quad -1.53219171, \quad -1.52194355,
              -1.53004603, -1.48604076, -0.69564183, -0.69098343, -0.67792868,
              -0.67249727, -0.66305853, -0.65272422, -0.34194162, -0.25698952,
              -0.29680127, -0.30806256, -0.31413612, -0.29338206],
             \hbox{$[-0.96746585,\ -0.90549825,\ 0.81016074,\ 0.18582826,\ -1.05729503,}
              0.59824792, 1.79456386, 1.78234817, 1.8099213, 0.18874609, 0.23491652, 0.25313738, -0.20227344, -0.17518087, -0.26818499,
             -0.3855629 , -0.39861741, -0.44403656, -0.1363026 , -0.26045141, -0.26593284, -0.29476192, -0.29000694]])
                                                                    -0.25698952.
# 5. Preprocessing the Data
X=np.array(x norm).T
# 6. Function to Claculate the Euclidian Distance
def Calculate Distance(i,j):
  return np.sqrt(np.sum((i-j)**2))
# 7. Function to find the Nearest Centroid
def Find_Nerest_Centroid(centroids,X):
  assigned_cluster=[]
  for i in X:
    dist=[]#list of calced distances
    for j in centroids:
      dist.append(Calculate_Distance(i,j))
    assigned_cluster.append(np.argmin(dist))
  return assigned_cluster
# 8. Function For Getting the New Centroids i.e. Updating the Old Centroids
def Calculating_Centroids(cluster_number,X):
  new centorids=[]
  new_df=pd.concat([pd.DataFrame(X),pd.Series(cluster_number,name='cluster')],axis=1)
  for c in set(new_df['cluster']):
    current_cluster=new_df[new_df['cluster']==c][new_df.columns[:-1]]
    cluster_mean=current_cluster.mean(axis=0)
    new_centorids.append(cluster_mean)
    return new_centorids
# 9. Training the Model
epochs=10
for i in range(epochs):
  get_centroids=Find_Nerest_Centroid(centroids,X)
  centroids=Calculating_Centroids(get_centroids,X)
  if i == 0 or i == epochs - 1:
    plt.figure(figsize=(10,10))
    plt.scatter(np.array(centroids)[:,0],np.array(centroids)[:,1],color="black")
    plt.scatter(X[:,0],X[:,1],c=get\_centroids,alpha = 0.3)
    plt.show()
```

