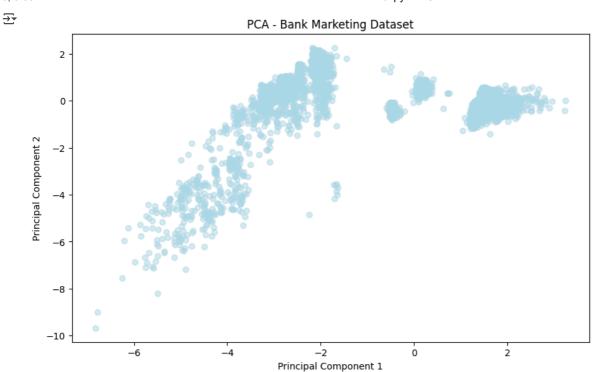
#module6

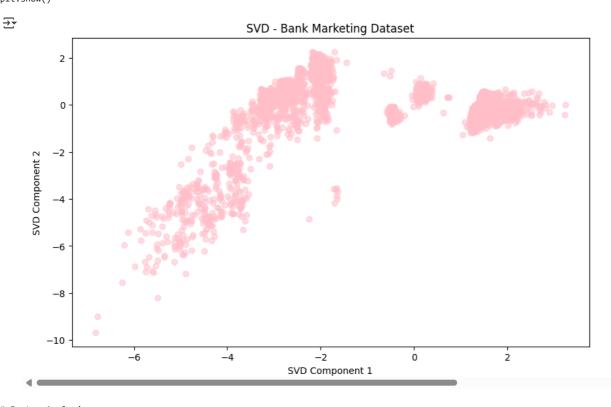
## Module 6: Dimensionality Reduction

Name: Srinivasa Reddy Julakanti Registration Number: 21BDS0220

```
import pandas as pd
from sklearn.preprocessing import StandardScaler
from \ sklearn. decomposition \ import \ PCA, \ Truncated SVD, \ Factor Analysis
from sklearn.manifold import MDS, TSNE
import matplotlib.pyplot as plt
import numpy as np
# Load dataset
# Load the dataset
data = pd.read_csv('bank_marketing_test.csv')
summary = data.describe()
print(summary)
\overline{2}
                            duration
                                        campaign
                                                        pdays
                                                                  previous
                    age
     count 8237.000000 8237.000000 8237.000000 8237.000000 8237.000000
     mean
             40.116547
                         256.007648
                                         2.60471
                                                   962.228724
                                                                  0.174335
     std
             10.465328
                         259.728737
                                         2.91562
                                                   187.533881
                                                                  0.500565
             17.000000
                           4.000000
                                         1.00000
                                                    0.000000
                                                                  0.000000
     25%
              32.000000
                         101.000000
                                         1.00000
                                                   999.000000
                                                                  0.000000
             38.000000
                         179.000000
                                         2.00000
                                                   999.000000
                                                                  0.000000
     50%
             47.000000
     75%
                         316.000000
                                         3.00000
                                                   999.000000
                                                                  0.000000
             89.000000 4918.000000
                                        43.00000
                                                   999.000000
                                                                  6.000000
     max
            emp.var.rate cons.price.idx cons.conf.idx
                                                           euribor3m nr.employed
            8237.000000
                            8237.000000
     count
                                            8237.000000 8237.000000
                                                                      8237.000000
                                                                      5166.589790
     mean
                0.070147
                               93.577806
                                             -40.545320
                                                            3.608206
     std
                1.574685
                                0.582138
                                               4.623626
                                                            1.735931
                                                                        72,470977
     min
               -3.400000
                               92.201000
                                             -50.800000
                                                            0.634000 4963.600000
     25%
               -1.800000
                               93.075000
                                             -42.700000
                                                            1.344000 5099.100000
     50%
               1.100000
                               93.444000
                                             -41.800000
                                                            4.857000
                                                                      5191.000000
                1.400000
                               93.994000
                                             -36.400000
                                                            4.961000 5228.100000
               1.400000
                               94.767000
                                             -26.900000
                                                            5.000000 5228.100000
     max
# Selecting numeric columns and scaling the features
numeric_cols = data.select_dtypes(include=['float64', 'int']).columns
scaler = StandardScaler()
scaled data = scaler.fit transform(data[numeric cols])
pca = PCA(n_components=2)
pca_data = pca.fit_transform(scaled_data)
plt.figure(figsize=(10, 6))
plt.scatter(pca_data[:, 0], pca_data[:, 1], c='lightblue', alpha=0.5)
plt.title("PCA - Bank Marketing Dataset")
plt.xlabel("Principal Component 1")
plt.ylabel("Principal Component 2")
plt.show()
```



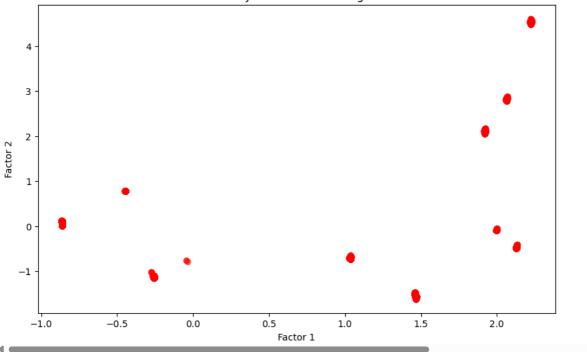
```
# SVD
svd = TruncatedSVD(n_components=2)
svd_data = svd.fit_transform(scaled_data)
plt.figure(figsize=(10, 6))
plt.scatter(svd_data[:, 0], svd_data[:, 1], c='pink', alpha=0.5)
plt.title("SVD - Bank Marketing Dataset")
plt.xlabel("SVD Component 1")
plt.ylabel("SVD Component 2")
plt.show()
```



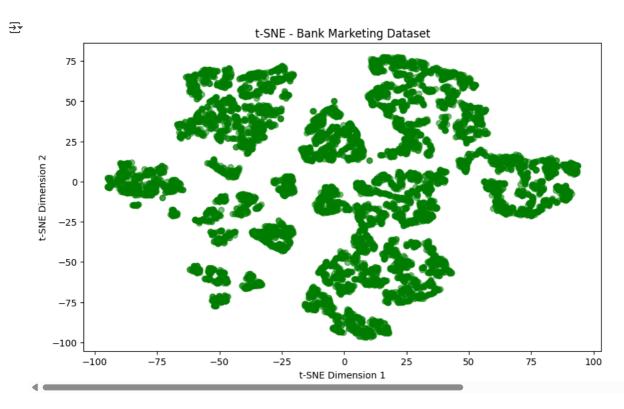
```
# Factor Analysis
fa = FactorAnalysis(n_components=2, random_state=42)
fa_data = fa.fit_transform(scaled_data)
plt.figure(figsize=(10, 6))
plt.scatter(fa_data[:, 0], fa_data[:, 1], c='red', alpha=0.5)
plt.title("Factor Analysis - Bank Marketing Dataset")
plt.xlabel("Factor 1")
plt.ylabel("Factor 2")
plt.show()
```



## Factor Analysis - Bank Marketing Dataset



```
# t-SNE
tsne = TSNE(n_components=2, random_state=42)
tsne_data = tsne.fit_transform(scaled_data)
plt.figure(figsize=(10, 6))
plt.scatter(tsne_data[:, 0], tsne_data[:, 1], c='green', alpha=0.5)
plt.title("t-SNE - Bank Marketing Dataset")
plt.xlabel("t-SNE Dimension 1")
plt.ylabel("t-SNE Dimension 2")
plt.show()
```



!pip install minisom

```
Collecting minisom

Downloading minisom-2.3.5.tar.gz (12 kB)

Preparing metadata (setup.py) ... done

Building wheels for collected packages: minisom

Building wheel for minisom (setup.py) ... done

Created wheel for minisom: filename=MiniSom-2.3.5-py3-none-any.whl size=12031 sha256=75de1f0abfd849d91517d27ba4ab09ced79845fae42e2

Stored in directory: /root/.cache/pip/wheels/19/db/95/5e53bc2b88a328217fdf9f2886cafbe86b0df274f4b601f572
```

Stored in directory: /root/.cache/pip/wheels/19/db/95/5e53bc2b88a328217fdf9f2886cafbe86b0df274f4b601f572
Successfully built minisom

Installing collected packages: minisom

Successfully installed minisom-2.3.5

```
# Self-Organizing Maps (SOM)
from minisom import MiniSom
som = MiniSom(x=10, y=10, input_len=scaled_data.shape[1], sigma=1.0, learning_rate=0.5)
som.random_weights_init(scaled_data)
som.train_random(scaled_data, 100)
plt.figure(figsize=(10, 6))
for i in range(10):
    for j in range(10):
       plt.scatter(i, j, s=100, c='black', alpha=0.3, marker='s')
for i, x in enumerate(scaled_data):
    w = som.winner(x)
    plt.plot(w[0] + 0.5, \ w[1] + 0.5, \ 'o', \ markerfacecolor=plt.cm.rainbow(i \ / \ len(scaled\_data)),
             markeredgecolor='k', markersize=8, alpha=0.7)
plt.title("Self-Organizing Map (SOM) - Bank Marketing Dataset")
plt.xlim([-1, 10])
plt.ylim([-1, 10])
plt.gca().invert_yaxis()
plt.show()
```



## Self-Organizing Map (SOM) - Bank Marketing Dataset

