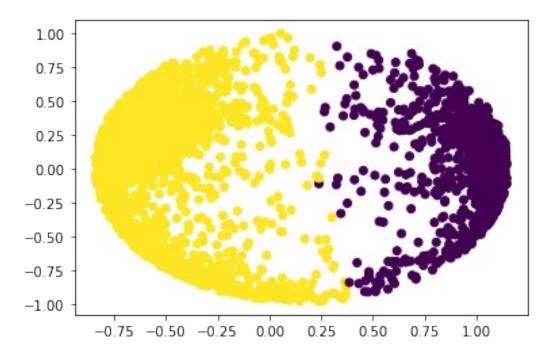
## Machine learning Lab 6

## **Graph Based Clustering**

```
Name: Hasrhvardhan Singh
Roll Number: 1019161
import talib as ta
from pandas datareader import data as pdr
import yfinance as yf
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import SpectralClustering
from sklearn.preprocessing import StandardScaler, normalize
from sklearn.decomposition import PCA
from sklearn.metrics import silhouette score
import warnings
warnings.filterwarnings('ignore')
vf.pdr override()
threshold = 0.75
df = pdr.get_data_yahoo('G00G', '2000-01-01', '2018-01-01')
df = df[df.columns[df.isnull().mean() < threshold]]</pre>
df = df.loc[df.isnull().mean(axis=1) < threshold]</pre>
df = df.dropna()
df = df.iloc[:.:4]
df.head()
 [*****************100%*****************
                                                     1 of 1 completed
                0pen
                          High
                                             Close
                                     Low
Date
2004-08-19 2.490664 2.591785 2.390042 2.499133
2004-08-20 2.515820 2.716817 2.503118 2.697639
2004-08-23 2.758411 2.826406 2.716070 2.724787
2004-08-24 2.770615 2.779581 2.579581 2.611960
2004-08-25 2.614201 2.689918
                               2.587302 2.640104
upper lim = df['Close'].quantile(.95)
lower lim = df['Close'].quantile(.05)
df = df[(df['Close'] < upper lim) & (df['Close'] > lower lim)]
upper lim = df['Open'].quantile(.95)
lower lim = df['Open'].quantile(.05)
df = df[(df [ 'Open'] < upper lim) & (df['Open'] > lower lim)]
```

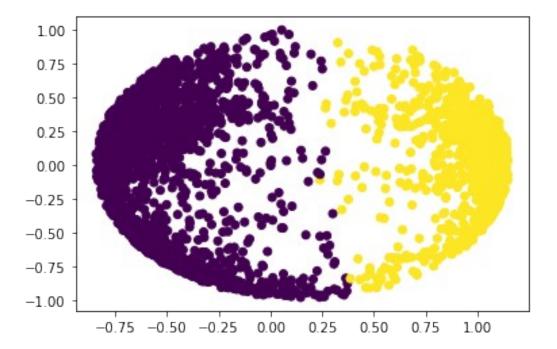
```
df['S 10'] = df['Close'].rolling(window=10).mean()
df['Corr'] = df['Close'].rolling(window=10).corr(df['S 10'])
df['RSI'] = ta.RSI(np.array(df['Close']), timeperiod =10)
df['Open-Close'] = df['Open'] - df['Close'].shift(1)
df[ 'Open-Open'] = df [ 'Open'] - df [ 'Open'].shift(1)
df = df.dropna()
scaler = StandardScaler()
X scaled = scaler.fit transform(df)
X normalized = normalize(X scaled)
X normalized = pd.DataFrame (X_normalized)
pca = PCA(n\_components = 2)
X principal = pca.fit transform(X normalized)
X principal = pd. DataFrame(X principal)
X principal.columns = ['Open-Close', 'Open-Open']
X principal.head(2)
   Open-Close Open-Open
    -0.660956 -0.534079
0
    -0.662805 -0.545855
1
spectral model rbf = SpectralClustering(n clusters = 2, affinity
='rbf')
labels_rbf = spectral_model_rbf.fit_predict(X_principal)
plt.scatter(X principal['Open-Close'], X principal['Open-Open'],c =
SpectralClustering(n clusters = 2, affinity
='nearest neighbors').fit predict(X principal))
```

<matplotlib.collections.PathCollection at 0x7f27ae14e610>

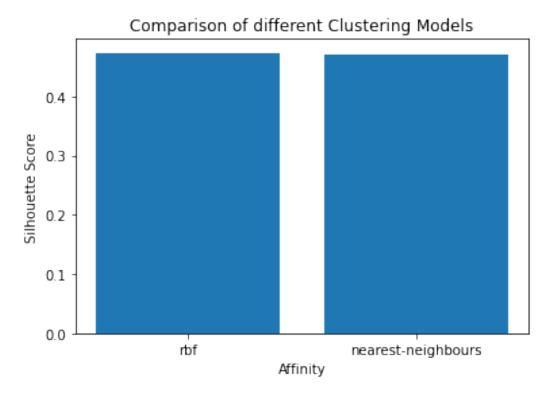


```
spectral_model_nn = SpectralClustering(n_clusters = 2, affinity
='nearest_neighbors')
labels_nn = spectral_model_nn.fit_predict(X_principal)

plt.scatter(X_principal['Open-Close'], X_principal['Open-Open'],
c = SpectralClustering(n_clusters = 2, affinity
='nearest_neighbors').fit_predict(X_principal))
plt.show()
```



```
affinity = ['rbf', 'nearest-neighbours']
S_scores = []
S_scores.append(silhouette_score(df, labels_rbf))
S_scores.append(silhouette_score(df, labels_nn))
plt.bar(affinity, S_scores)
plt.xlabel('Affinity')
plt.ylabel('Silhouette Score')
plt.title('Comparison of different Clustering Models')
plt.show()
print(S_scores)
```



 $\hbox{\tt [0.47319397822466186,\ 0.47107425929598745]}$