## Problem 8

## Loading the data

```
import pandas as pd
from sklearn.cluster import SpectralClustering
from sklearn.metrics import adjusted rand score,
adjusted_mutual_info_score, silhouette score
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
data = pd.read csv('/content/drive/MyDrive/sem 7/ID5055/Assignment
3/Problem 8/gene_expression.csv')
data.head()
   Gene One Gene Two Cancer Present
0
        4.3
                  3.9
1
        2.5
                  6.3
                                    0
2
                  3.9
                                    1
        5.7
3
                  6.2
        6.1
                                    0
        7.4
                  3.4
df = data
```

Implement spectral clustering using Python and scikit-learn to identify clusters of co-expressed genes within the dataset.

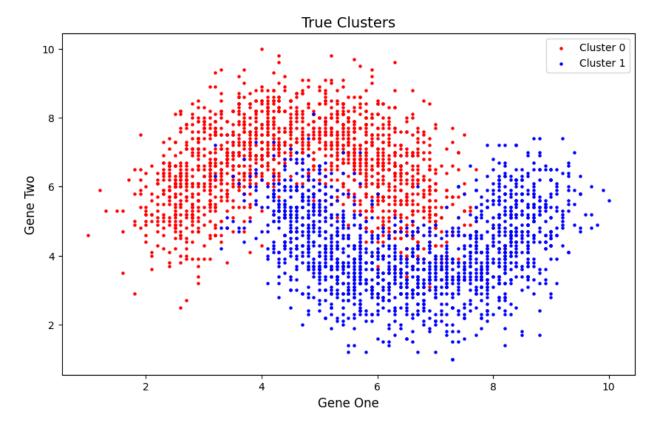
```
n clusters = 2
X = df.iloc[:, :2]
X scaled = StandardScaler().fit transform(X)
spectral = SpectralClustering(n clusters=n clusters,
affinity='nearest_neighbors', random_state=0)
df['Cluster'] = spectral.fit predict(X scaled)
df.head()
   Gene One Gene Two Cancer Present Cluster
        4.3
                  3.9
0
                                     1
1
                                              1
        2.5
                  6.3
                                     0
2
        5.7
                  3.9
                                              0
                                     1
3
                  6.2
                                              1
        6.1
                                     0
        7.4
                  3.4
                                     1
                                              0
```

Create visualizations for the true clusters based on the information in the 3rd column of the dataset.

```
plt.figure(figsize=(10, 6))
colors = ['red', 'blue']

for i in range(n_clusters):
    cluster_data = df[df['Cancer Present'] == i]
    plt.scatter(cluster_data['Gene One'], cluster_data['Gene Two'],
color=colors[i], label=f'Cluster {i}', s = 5)

plt.title('True Clusters', fontsize = 14)
plt.xlabel('Gene One', fontsize = 12)
plt.ylabel('Gene Two', fontsize = 12)
plt.legend()
plt.show()
```



Evaluate and provide insights on the outcomes, including a comprehensive report on performance metrics such as Adjusted Rand Index, Adjusted Mutual Information, and Silhouette Score.

```
ari = adjusted_rand_score(df['Cancer Present'], df['Cluster'])
ami = adjusted_mutual_info_score(df['Cancer Present'], df['Cluster'])
silhouette = silhouette_score(X_scaled, df['Cluster'])
```

```
print(f'Adjusted Rand Index: {ari}')
print(f'Adjusted Mutual Information: {ami}')
print(f'Silhouette Score: {silhouette}')

Adjusted Rand Index: 0.5153624060863754
Adjusted Mutual Information: 0.413040052343147
Silhouette Score: 0.4540574993371185
```

- 1. Adjusted Rand Index (ARI): 0.515 -
- The ARI measures the similarity between the true clusters (Cancer Present) and the clusters generated by the spectral clustering algorithm.
- An ARI score of 0.515 indicates a moderate degree of similarity between the true clusters and the clusters identified by the algorithm.
- This suggests that while spectral clustering is capturing some underlying structure in the data, there may still be room for improvement.
- 1. Adjusted Mutual Information (AMI): 0.413
- The AMI is another measure of the agreement between the true clusters and the clusters produced by the algorithm.
- An AMI score of 0.413 suggests a moderate level of mutual information between the true clusters and the algorithm's clusters.
- Similar to the ARI, this indicates that the spectral clustering algorithm is providing some meaningful clustering, but it may not be capturing all of the underlying patterns in the data.
- 1. Silhouette Score: 0.454
- The Silhouette Score measures the quality of the clusters themselves. It assesses how well-separated the clusters are and how similar the data points within each cluster are to each other.
- A Silhouette Score of 0.454 is relatively high, indicating that the clusters are reasonably well-separated and that data points within each cluster are similar to each other.
- This suggests that the algorithm is successful in creating meaningful clusters, and the clusters are relatively distinct from each other.