Step 1: Import necessary libraries

```
In [9]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import load_iris
from sklearn.cluster import KMeans, SpectralClustering, DBSCAN
from sklearn.metrics import silhouette_score
```

Step 2: Load the dataset

```
In [16]:
        # Load the Iris dataset
        X = pd.read_csv("C:/Users/SMIT YENKAR/OneDrive/Desktop/T.Y .Sub/ML LAB/Iris.
        print(X.head())
        # Perform encoding for the 'Species' column
        from sklearn.preprocessing import LabelEncoder
        label_encoder = LabelEncoder()
        X['Species'] = label_encoder.fit_transform(X['Species'])
          Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                        Species
                       5.1
                                    3.5
                                                  1.4
                                                               0.2 Iris-setosa
          1
                                                               0.2 Iris-setosa
       1
         2
                       4.9
                                    3.0
                                                  1.4
       2 3
                       4.7
                                   3.2
                                                  1.3
                                                               0.2 Iris-setosa
                       4.6
                                   3.1
                                                  1.5
                                                               0.2 Iris-setosa
                       5.0
                                                               0.2 Iris-setosa
                                    3.6
                                                  1.4
```

Step 3: Apply K-means clustering

```
In [17]: # Apply K-means clustering
kmeans = KMeans(n_clusters=3)
kmeans_labels = kmeans.fit_predict(X.drop(columns=['Species']))
```

Step 4: Apply Spectral clustering

```
In [18]: # Apply Spectral clustering
spectral = SpectralClustering(n_clusters=3, affinity='nearest_neighbors')
spectral_labels = spectral.fit_predict(X.drop(columns=['Species']))
```

Step 5: Apply DBSCAN clustering

1 of 2 4/27/24, 12:41 AM

```
In [19]: # Apply DBSCAN clustering
    dbscan = DBSCAN(eps=0.5, min_samples=5)
    dbscan_labels = dbscan.fit_predict(X.drop(columns=['Species']))
```

Step 6: Evaluate clustering performance using silhouette score

```
In [27]:
         from sklearn.metrics import silhouette_score
         from sklearn.preprocessing import StandardScaler
         # Normalize the data
         scaler = StandardScaler()
         X_scaled = scaler.fit_transform(X.drop(columns=['Species']))
         # Adjust hyperparameters and apply DBSCAN clustering
         dbscan = DBSCAN(eps=0.4, min_samples=5)
         dbscan_labels = dbscan.fit_predict(X_scaled)
         # Evaluate clustering performance using silhouette score
         unique_labels = np.unique(dbscan_labels)
         num_clusters = len(unique_labels) - (1 if -1 in dbscan_labels else 0) # Exc
         if num_clusters > 1:
             silhouette_dbscan = silhouette_score(X_scaled, dbscan_labels)
             print("Silhouette Score (DBSCAN):", silhouette_dbscan)
         else:
             print("DBSCAN failed to identify distinct clusters.")
         print("\nSilhouette Score:")
         print("K-means:", silhouette_kmeans)
         print("Spectral Clustering:", silhouette_spectral)
        Silhouette Score (DBSCAN): 0.10816320348375333
        Silhouette Score:
        K-means: 0.5821934246576435
        Spectral Clustering: 0.5514935658161797
 In [ ]:
```

2 of 2 4/27/24, 12:41 AM