Ex 2: Pov

Cho dữ liệu như pov_12.csv

- 1. Chuẩn hóa dữ liệu X chứa cột 0, 1
- 2. Tìm số cụm k phù hợp
- 3. Áp dụng thuật toán GMM để giải bài toán phân cụm với số cụm = k ở câu 2
- 4. Cho X_test = np.array([[0.1, 0.1], [0.8,0.8], [0.5, 0.5]]), cho biết những phần tử này thuộc cụm nào?
- 5. Vẽ hình, xem kết quả

```
In [1]: from sklearn import metrics
    from scipy.spatial.distance import cdist
    import numpy as np
    import matplotlib.pyplot as plt
    import pandas as pd
```

```
In [2]: data = pd.read_csv("pov_12.csv", sep=" ", header=None)
    print(data.shape)
    data.head()
```

(150, 3)

```
        Out[2]:
        0
        1
        2

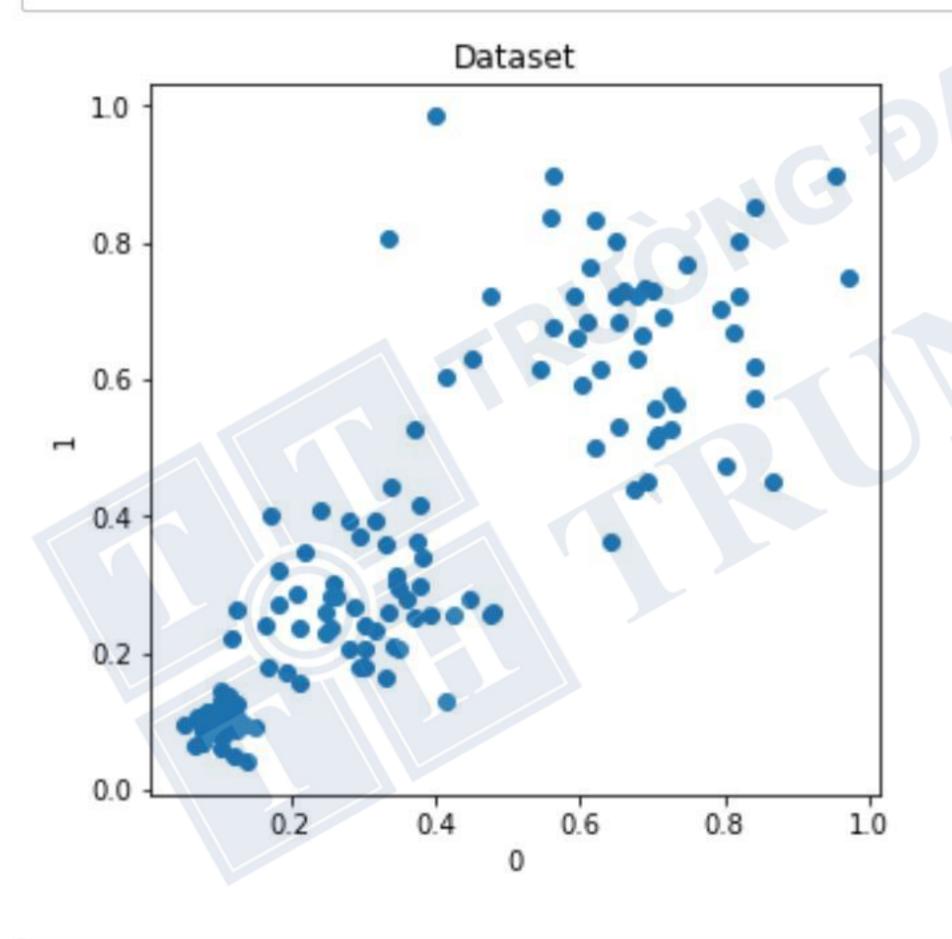
        0
        0.109393
        0.085409
        Cluster1

        1
        0.082571
        0.101796
        Cluster1

        2
        0.084990
        0.113641
        Cluster1
```

3 0.114611 0.115524 Cluster1 4 0.097356 0.095484 Cluster1

```
In [3]: plt.figure(figsize=(5,5))
    plt.scatter(data[0], data[1])
    plt.title('Dataset')
    plt.xlabel("0")
    plt.ylabel("1")
    plt.show()
```



```
In [4]: X_train = data[[0,1]]
```

In [5]: from sklearn.mixture import GaussianMixture

```
In [6]: from sklearn import metrics
list_sil = [] # chua danh sach cac gia tri sil
K = range(2,8) # chua danh sach cac k
for k in K:
    gmm = GaussianMixture(n_components=k) # 2, 3, 4...
    gmm.fit(X_train)
    labels = gmm.predict(X_train)
    # k = 2 => 0,1
    # k = 3 => 0, 1, 2
    sil = metrics.silhouette_score(X_train, labels, metric='euclidean')
    list_sil.append(sil)
```

```
plt.plot(K, list_sil, 'bx-')
         plt.xlabel('k')
         plt.ylabel('sil_score')
         plt.title('The silhouette_score & k')
         plt.show()
                             The silhouette_score & k
            0.65
            0.60 -
          SCO7
0.55
            0.50
            0.45
 In [8]: # Select k = 2
 In [9]: import numpy as np
         from sklearn.mixture import GaussianMixture
         gmm = GaussianMixture(n_components=2)
         gmm.fit(X_train)
 Out[9]: GaussianMixture(covariance_type='full', init_params='kmeans', max_iter=100,
                          means_init=None, n_components=2, n_init=1, precisions_init=None,
                          random_state=None, reg_covar=1e-06, tol=0.001, verbose=0,
                          verbose_interval=10, warm_start=False, weights_init=None)
In [10]: print(gmm.weights_)
          [0.66606181 0.33393819]
         print(gmm.means_)
In [11]:
          [[0.20020692 0.18647424]
           [0.66682326 0.66125669]]
In [12]: print(gmm.covariances_)
          [[[ 0.01339299  0.00880714]
             0.00880714 0.01067513]]
           [[ 0.01872054 -0.00121377]
            [-0.00121377 0.01794757]]]
In [13]: X_{\text{test}} = \text{np.array}([[0.1, 0.1], [0.8, 0.8], [0.4, 0.4]])
         pred = gmm.predict(X_test)
         pred
Out[13]: array([0, 1, 0], dtype=int64)
In [14]: types = gmm.predict(X_train)
In [16]: # plot mixture of Gaussians
         plt.figure(figsize=(6,4))
         X, Y = np.meshgrid(np.linspace(0,1), np.linspace(0,1))
         XX = np.array([X.ravel(), Y.ravel()]).T
         Z = gmm.score_samples(XX)
         Z = Z.reshape((50,50))
         plt.contour(X, Y, Z)
         plt.scatter(X_train[0], X_train[1], c=types)
         plt.scatter(X_test[:,0], X_test[:,1], marker="s", c='b')
         plt.scatter(gmm.means_[:,0], gmm.means_[:,1], color="red")
         plt.show()
           0.6
                     0.2
                                                0.8
                              0.4
                                       0.6
```

In [7]: # Plot

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