ENGR 421 – HW8

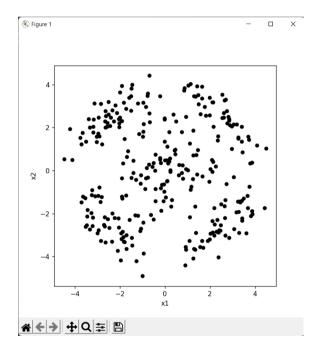
Data Importing:

I imported the data.

```
X = np.genfromtxt("hw08_data_set.csv", delimiter=",")

# sample size
N = X.shape[0]
# cluster count
K = 5

plt.figure(figsize=(6, 6))
plt.plot(X[:, 0], X[:, 1], "k.", markersize=10)
plt.xlabel("x1")
plt.ylabel("x2")
plt.show()
```



B, D, L, Lsymmetric, Z Matrixes

I constructed B matrix according to the rule given in pdf and plot connectivity matrix. I set threshold to 1.25 and used Euclidean distance formula for comparing with threshold.

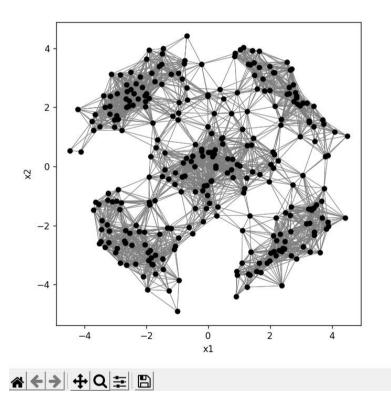
$$b_{ij} = \begin{cases} 1, & \left\| \mathbf{x}_i - \mathbf{x}_j \right\|_2 < \delta \\ 0, & \text{otherwise.} \end{cases}$$

$$b_{ii} = 0$$

```
B = np.zeros((N, N))
Gfor i in range(N):
    for j in range(N):
        if i == j:
            B[i][j] = 0
else:
        # euclidean distance
            distance = np.sqrt((X[j][0] - X[i][0]) ** 2 + (X[j][1] - X[i][1]) ** 2)
        if distance < S:
            B[i][j] = 1
        else:
            B[i][j] = 0

plt.figure(figsize=(6, 6))
Gfor i in range(N):
        if B[i][j] == 1:
            plt.plot([X[i][0], X[j][0]], [X[i][1], X[j][1]], "0.5", linewidth=0.5)
plt.plot(X[:, 0], X[:, 1], '.', markersize=10, color="black")
plt.xlabel("x1")
plt.ylabel("x2")
plt.show()</pre>
```





Then I calculated D, L and Lsymmetric matrixes according to the formula given in the lecture notes. However, I never used the L matrix since I need to use Lsymmetric matrix. I add it since it wants in pdf.

$$\begin{array}{l} \text{dic} = \sum\limits_{j \neq i} \text{bij } \forall i \\ \text{) \# of neighbors} \\ \text{of data point } i \\ \text{dij} = 0 \ \forall (i,j \neq i) \end{array} \qquad \begin{array}{l} L_{symmetric} = I - D^{-1/2}BD^{-1/2} \end{array}$$

```
D = np.zeros((N, N))

for i in range(N):
    total = 0
    for j in range(N):
        total = total + B[i][j]

D[i][i] = total

I = np.eye(N)
L = np.subtract(D, B)
sqrt_inv_D = np.sqrt(np.linalg.inv(D))
L_sym = np.subtract(I, np.matmul(sqrt_inv_D, np.matmul(B, sqrt_inv_D)))
```

I take the eigen values and vectors than take the 2nd smallest, 3rd smallest, 4th smallest, 5th smallest and 6th smallest eigenvalues and form Z matrix with eigen vectors.

```
R = 5
eigen_vals, eigen_vecs = np.linalg.eig(L_sym)
sorted_eigen_vals = eigen_vals.argsort()
Z = eigen_vecs[:, sorted_eigen_vals[1:R + 1]]
```

K-Means Clustering Algorithm

I take the update_centroids, update_memberships, plot_current_centroids, and algorithm iteration part from lab11. I set the initial centroids to 29th, 143rd, 204th, 271st, and 277th rows of Z matrix.

```
idef update_centroids(memberships, X):
    if memberships is None:
        # initialize centroids
        centroids = np.vstack([Z[28], Z[142], Z[203], Z[270], Z[276]])
    else:
        # update centroids
        centroids = np.vstack([np.mean(X[memberships == k, :], axis=0) for k in range(K)])
    return centroids

def update_memberships(centroids, X):
    # calculate distances between centroids and data points
    D = spa.distance_matrix(centroids, X)
    # find the nearest centroid for each data point
    memberships = np.argmin(D, axis=0)
    return memberships
```

```
centroids = None
memberships = None
iteration = 1
3while True:
    print("Iteration#{}:".format(iteration))

    old_centroids = centroids
    centroids = update_centroids(memberships, Z)
    if np.alltrue(centroids == old_centroids):
        break

    old_memberships = memberships
    memberships = update_memberships(centroids, Z)
    if np.alltrue(memberships == old_memberships):
        break

iteration = iteration + 1

centroids = update_centroids(memberships, X)
plot_current_state(centroids, memberships, X)
plt.show()
```

