friends

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1 import libraries

```
[7]: import numpy as np
    from sklearn.cluster import KMeans
    import matplotlib.pyplot as plt
    import random

import warnings
warnings.filterwarnings("ignore")
```

2 perform spectral clustering

```
[8]: A = np.zeros((1495, 1495), dtype=int)
     file name = 'friends.txt'
     with open(file_name, 'r') as file:
         for line in file:
             node1, node2 = map(int, line.strip().split(','))
             A[node1 - 1][node2 - 1] = 1
     L = np.diag(np.sum(A, axis=1)) - A
     eigenvalues, eigenvectors = np.linalg.eig(L)
     sorted_indices = np.argsort(eigenvalues)
     sorted_eigenvalues = eigenvalues[sorted_indices]
     sorted_eigenvectors = eigenvectors[:, sorted_indices]
     print(np.round(sorted_eigenvalues[:12],3))
     for i in range(6):
         v = sorted_eigenvectors[:,i]
         threshold = np.max(np.abs(v)) * 0.9 # dynamically set threshold base on the_
      →values of the eigenvector
         nodes = np.where(np.abs(v) > threshold)
         print("number of nodes in cluster", i+1, ":", len(nodes[0]))
```

[-0. -0. 0. 0. 0. 0. 0.014 0.054 0.074 0.081

```
0.12 0.133]
number of nodes in cluster 1 : 2
number of nodes in cluster 2 : 3
number of nodes in cluster 3 : 2
number of nodes in cluster 4 : 2
number of nodes in cluster 5 : 2
number of nodes in cluster 6 : 1484
```

2.1 interpret the output

- There are 6 zero eigenvalues, which means the graph has 6 connected components
- For each cluster, we can find the number of nodes within that cluster

3 cluster nodes with certain constrains

- 1. $150 < \text{node_size} < 750$
- 2. cluster the data into at least 3 groups
- 3. the conductance between groups must smaller than 0.1
- 4. The conductance of $S \subset V$ is defined as

$$conductance = \frac{E}{D}$$

where E = number of boundary edges of S, D = total degree of S

```
[9]: def compute_conductance(A, S):
         n = A.shape[0]
         S_complement = np.setdiff1d(np.arange(n), S) # find the complement of S
         boundary_edges = np.sum(A[np.ix_(S, S_complement)]) # find the number of_
      ⇔boundary edges
         degree S = np.sum(np.sum(A[S, :], axis=1)) # find the degree of S
         if degree_S == 0:
             return 1
         return boundary_edges / degree_S
     def find_low_conductance_sets(A, min_size=150, max_size=750, max_conductance=0.
      →1):
         # we need to try the proper combination of eigenvectors to find the best _{\sqcup}
      \rightarrowpartition
         # try different number of clusters are also important to avoid some
      ⇔clusters are too small or too large
         selected_eigenvectors = sorted_eigenvectors[:, 6:15]
         kmeans = KMeans(n_clusters=50, random_state=0)
         labels = kmeans.fit_predict(selected_eigenvectors)
         sets = []
         for i in range(np.max(labels) + 1):
             S = np.where(labels == i)[0]
             if min_size <= len(S) <= max_size:</pre>
                 conductance = compute_conductance(A, S)
                 if conductance <= max_conductance:</pre>
                      sets.append(S)
         return sets
     sets = find_low_conductance_sets(A)
```

```
for i, S in enumerate(sets):
    print(f"S{i + 1}:")
    print(f"Size: {len(S)}")
    print(f"first 10 members: {S[:10]}")
    print(f"conductance: {compute_conductance(A, S)}")
    print()
```

S1:

Size: 198

first 10 members: [1 5 7 12 16 20 54 59 66 70]

conductance: 0.008555133079847909

S2:

Size: 550

first 10 members: [4 6 11 14 22 23 26 29 33 41]

conductance: 0.012405699916177704

S3:

Size: 506

first 10 members: [0 2 8 9 13 15 18 19 21 25]

conductance: 0.042727618220315976