# Spectral Graph Clustering Algorithm

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Group: Assignment Group 8

# Clearing previous variables and command window

```
clc; clear ; close all;
```

# Reading dataset from ./data and visualising the dataset

File path to the dataset expample1.dat or expample2.dat

```
dataset_path = fullfile('.','data','example1.dat');
```

Reading the csv data from the file path

```
dataset = csvread(dataset_path);
```

Getting the source nodes and destination node

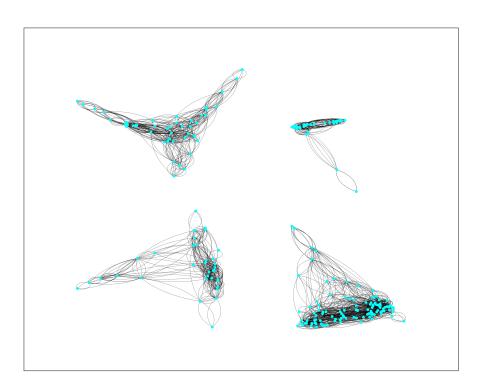
```
source_nodes = dataset(:,1);
destination_nodes = dataset(:,2);
```

Adding edges to graph

```
G = graph(source_nodes, destination_nodes);
```

#### Displaying the Graph

```
figure('Name','Original graph without clusters');
p = plot(G);
p.LineWidth= 0.2500;
p.LineStyle= '-';
p.Marker = '.';
p.NodeColor = 'c';
p.EdgeColor = 'k';
p.MarkerSize = 10;
```



# **Algorithm**

Construct the adjacency matrix (affinity matrix)

```
max_ids = max(max(source_nodes, destination_nodes));
As = sparse(source_nodes, destination_nodes, 1, max_ids, max_ids);
A = full(As);
```

#### Construct diagonal matrix

```
Di = diag(sum(A,2));
```

Construct normalized laplacian matrix ( $L = D^{-1/2}AD(-1/2)$ )

```
L = (Di^{(-0.5)}) * A * (Di^{(-0.5)});
```

Calculating k largest eigenvectors of L and stack them into X

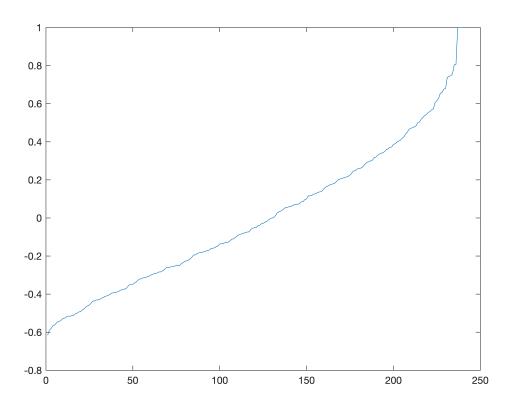
```
% [X,D1] = eigs(L, k);
[X, D1] = eigs(L, size(L,1)-1);
% size(L) - 1 to use eigs() instead of eig()
```

Construct eigenvalues for D1

```
eigenvalues = diag(D1);
eig_gaps = -1*diff(eigenvalues);
% -1* used because eigenvalues is sorted
% in desc. order so diffs are negatives.
```

#### Plot Eignevalues of sorted normalized of Laplacian matrix

```
figure('Name', 'Eignevalues of sorted normalized of Laplacian matrix '); plot(sort(eigenvalues)); % plotting the eignevalues (ascending order) and normalized (
```



#### Calculating optimal value of k

```
[~,k]=min(diff(flipud(sort(eigenvalues))))
k = 4
```

### Constructing Matrix X by stacking eigen values

```
X = X(:, (1:k));
```

## Create Y by renormalizing each of X's rows to have unit length

```
Y = X./(sum(X.^2,2)).^(0.5);
```

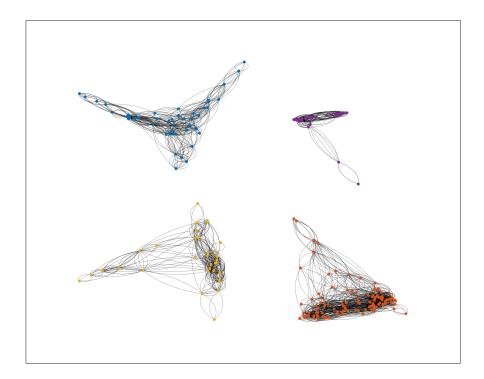
### Apply K-means to rows of Y

```
Idx = kmeans(Y, k);
```

### Display graphs with clusters

```
figure('Name','Graph with clusters');
p = plot(G);
p.EdgeColor = 'k';
p.LineWidth= 0.2500;
p.LineStyle= '-';
p.Marker = '.';
p.MarkerSize = 10;
cmap = lines();

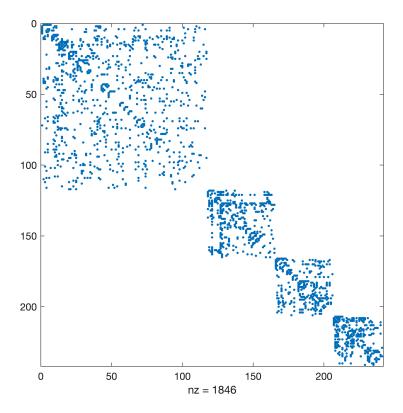
for i=1:size(Idx,1)
    cluster = Idx(i,1);
    highlight(p,i,'NodeColor',cmap(cluster,:));
end
```



# Find communities using the Fiedler Vector

Display sparsity pattern

```
figure('Name', 'Sparsity Pattern');
```



### Construct laplacian matrix

```
H = simplify(G);
new_L = laplacian(H);
```

## Get eigenvalues from laplacian matrix

```
[v,new_D] = eigs(new_L,k,'smallestreal');
```

#### Construct Fiedler Vector

```
fv = v(:,2);
```

#### Plot the sorted Fiedler vector

```
figure('Name','Sorted Fiedler Vector');
plot(sort(fv));
```

