

ProjectNotebook

June 19, 2023

Importing All necessary Libraries

```
[1]: import numpy as np
import networkx as nx
import matplotlib.pyplot as plt
```

1 Calculation of centralities

1.1 Calculation of Degree Centrality

1.1.1 Calculation of Degree Centrality using Linear Algebra

```
[2]: def compute_degree_centralitiy(adj_matrix):
    num_nodes = adj_matrix.shape[0]

    degree_centralitiy = np.sum(adj_matrix, axis=1)

    normalization_factor = num_nodes - 1

    degree_centralitiy = degree_centralitiy.astype(float) / normalization_factor

    return dict(enumerate(degree_centralitiy))

G = nx.karate_club_graph()

weighted_adjacency_matrix = nx.to_numpy_array(G)

unweighted_adjacency_matrix = (weighted_adjacency_matrix > 0).astype(int)

print(f"No of Nodes in the Graph is {unweighted_adjacency_matrix.shape[0]}")
print("The Adjacency Matrix corresponding to the graph is:")
print(unweighted_adjacency_matrix)

print("The Graph is")
nx.draw(G,with_labels=True)
plt.show()
```

```

degree centrality_LA = compute_degree centrality(unweighted_adjacency_matrix)

print("Degree Centrality using LA:")
print(degree centrality_LA)

```

No of Nodes in the Graph is 34

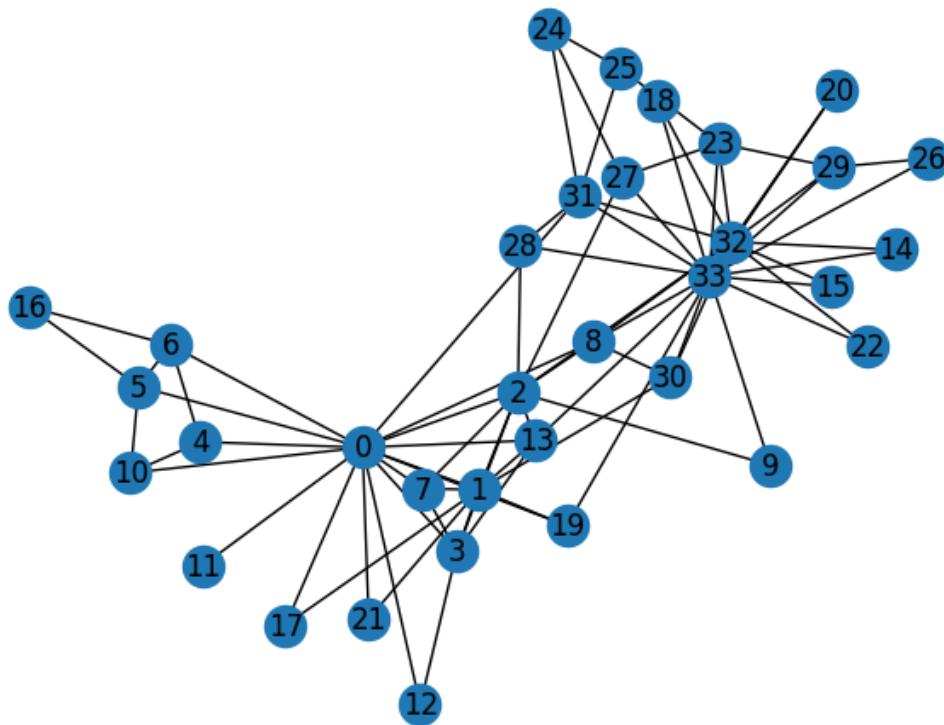
The Adjacency Matrix corresponding to the graph is:

```

[[0 1 1 ... 1 0 0]
 [1 0 1 ... 0 0 0]
 [1 1 0 ... 0 1 0]
 ...
 [1 0 0 ... 0 1 1]
 [0 0 1 ... 1 0 1]
 [0 0 0 ... 1 1 0]]

```

The Graph is



Degree Centrality using LA:

```

{0: 0.48484848484848486, 1: 0.2727272727272727, 2: 0.30303030303030304, 3:
0.18181818181818182, 4: 0.09090909090909091, 5: 0.12121212121212122, 6:
0.12121212121212122, 7: 0.12121212121212122, 8: 0.15151515151515152, 9:

```

```
0.06060606060606061, 10: 0.09090909090909091, 11: 0.030303030303030304, 12:
0.06060606060606061, 13: 0.15151515151515152, 14: 0.06060606060606061, 15:
0.06060606060606061, 16: 0.06060606060606061, 17: 0.06060606060606061, 18:
0.06060606060606061, 19: 0.09090909090909091, 20: 0.06060606060606061, 21:
0.06060606060606061, 22: 0.06060606060606061, 23: 0.15151515151515152, 24:
0.09090909090909091, 25: 0.09090909090909091, 26: 0.06060606060606061, 27:
0.12121212121212122, 28: 0.09090909090909091, 29: 0.12121212121212122, 30:
0.12121212121212122, 31: 0.18181818181818182, 32: 0.36363636363636365, 33:
0.5151515151515151}
```

1.1.2 Calculation of Degree Centrality Using In-Built Function

```
[3]: degree centrality_inBuilt = nx.degree centrality(G)
print("Degree Centrality InBuilt:")
print(degree centrality_inBuilt)
```

```
Degree Centrality InBuilt:
{0: 0.48484848484848486, 1: 0.2727272727272727, 2: 0.30303030303030304, 3:
0.18181818181818182, 4: 0.09090909090909091, 5: 0.12121212121212122, 6:
0.12121212121212122, 7: 0.12121212121212122, 8: 0.15151515151515152, 9:
0.06060606060606061, 10: 0.09090909090909091, 11: 0.030303030303030304, 12:
0.06060606060606061, 13: 0.15151515151515152, 14: 0.06060606060606061, 15:
0.06060606060606061, 16: 0.06060606060606061, 17: 0.06060606060606061, 18:
0.06060606060606061, 19: 0.09090909090909091, 20: 0.06060606060606061, 21:
0.06060606060606061, 22: 0.06060606060606061, 23: 0.15151515151515152, 24:
0.09090909090909091, 25: 0.09090909090909091, 26: 0.06060606060606061, 27:
0.12121212121212122, 28: 0.09090909090909091, 29: 0.12121212121212122, 30:
0.12121212121212122, 31: 0.18181818181818182, 32: 0.36363636363636365, 33:
0.5151515151515151}
```

Function for comparison of two dictionaries based on some precision

```
[4]: def compare_dictionaries(dict1, dict2, precision):
    for key in dict1.keys():
        if key not in dict2:
            return False
        value1 = dict1[key]
        value2 = dict2[key]
        if (round(value1, precision) != round(value2, precision)):
            return False
    return True
```

1.1.3 Comparison of the Degree Centrality Obtained by calculation using Linear Algebra and using Inbuilt function of networkx

```
[5]: DecimalPlaces = 15
```

```

if
    ↪compare_dictionaries(degree centrality_inBuilt,degree centrality_LA,DecimalPlaces):
    ↪
        print("The both dictionaries are equal")
else:
    print("The both dictionaries are not equal")

```

The both dictionaries are equal

1.2 Calculation of Betweenness Centrality

1.2.1 Calculation of Betweenness Centrality using LA

```

[6]: def compute_betweenness centrality(graph):
    nodes = graph.nodes()
    betweenness centrality = {node: 0.0 for node in nodes}

    for source in nodes:
        betweenness = {node: 0.0 for node in nodes}
        stack = []
        distance = {node: -1 for node in nodes}
        sigma = {node: 0 for node in nodes}
        distance[source] = 0
        sigma[source] = 1
        queue = [source]

        while queue:
            node = queue.pop(0)
            stack.append(node)

            for neighbor in graph.neighbors(node):
                if distance[neighbor] == -1:
                    queue.append(neighbor)
                    distance[neighbor] = distance[node] + 1

                if distance[neighbor] == distance[node] + 1:
                    sigma[neighbor] += sigma[node]

        dependency = {node: 0.0 for node in nodes}
        while stack:
            node = stack.pop()
            for predecessor in graph.neighbors(node):
                if distance[predecessor] == distance[node] - 1:
                    dependency[predecessor] += (1 + dependency[node]) *
            ↪sigma[predecessor] / sigma[node]

            if node != source:
                betweenness[node] += dependency[node]

```

```

        for node in nodes:
            if node != source:
                betweenness centrality [node] += betweenness [node]

num_nodes = len(nodes)
normalization_factor = ((num_nodes - 1) * (num_nodes - 2))

for node in nodes:
    betweenness centrality [node] /= normalization_factor

return betweenness centrality

G = nx.karate_club_graph()

weighted_adjacency_matrix = nx.to_numpy_array(G)

print(f"The Weighted Adjacency matrix is\n{weighted_adjacency_matrix}")

betweenness centrality_LA = compute_betweenness centrality(G)

print("Betweenness Centrality computed using brandes algorithm")
print(betweenness centrality_LA)

```

The Weighted Adjacency matrix is

```

[[0. 4. 5. ... 2. 0. 0.]
 [4. 0. 6. ... 0. 0. 0.]
 [5. 6. 0. ... 0. 2. 0.]
 ...
 [2. 0. 0. ... 0. 4. 4.]
 [0. 0. 2. ... 4. 0. 5.]
 [0. 0. 0. ... 4. 5. 0.]]

```

Betweenness Centrality computed using brandes algorithm

```

{0: 0.43763528138528146, 1: 0.053936688311688304, 2: 0.14365680615680618, 3:
0.011909271284271283, 4: 0.0006313131313131313, 5: 0.02998737373737374, 6:
0.029987373737373736, 7: 0.0, 8: 0.055926827801827804, 9: 0.0008477633477633478,
10: 0.0006313131313131313, 11: 0.0, 12: 0.0, 13: 0.04586339586339586, 14: 0.0,
15: 0.0, 16: 0.0, 17: 0.0, 18: 0.0, 19: 0.0324750481000481, 20: 0.0, 21: 0.0,
22: 0.0, 23: 0.017613636363636363, 24: 0.0022095959595959595, 25:
0.0038404882154882154, 26: 0.0, 27: 0.022333453583453577, 28:
0.0017947330447330447, 29: 0.0029220779220779218, 30: 0.014411976911976912, 31:
0.13827561327561322, 32: 0.14524711399711399, 33: 0.3040749759499759}

```

1.2.2 Calculation of Betweenness Centrality using In-Built function

```
[7]: betweenness centrality_inBuilt= nx.betweenness centrality(G)
print("Betweenness Centrality computed using inBuilt function of networkx:")
print(betweenness centrality_inBuilt)
```

```
Betweenness Centrality computed using inBuilt function of networkx:
{0: 0.43763528138528146, 1: 0.053936688311688304, 2: 0.14365680615680618, 3:
0.011909271284271283, 4: 0.0006313131313131313, 5: 0.02998737373737374, 6:
0.029987373737373736, 7: 0.0, 8: 0.05592682780182781, 9: 0.0008477633477633478,
10: 0.0006313131313131313, 11: 0.0, 12: 0.0, 13: 0.04586339586339586, 14: 0.0,
15: 0.0, 16: 0.0, 17: 0.0, 18: 0.0, 19: 0.03247504810004811, 20: 0.0, 21: 0.0,
22: 0.0, 23: 0.017613636363636363, 24: 0.0022095959595959595, 25:
0.0038404882154882154, 26: 0.0, 27: 0.02233345358345358, 28:
0.0017947330447330447, 29: 0.0029220779220779218, 30: 0.014411976911976909, 31:
0.13827561327561325, 32: 0.145247113997114, 33: 0.30407497594997596}
```

1.2.3 Comparison of the Betweenness Centrality Obtained by calculation using Linear Algebra and using Inbuilt function of networkx

```
[8]: if
    ↳compare_dictionaries(betweenness centrality_LA,betweenness centrality_inBuilt,DecimalPlaces
    ↳
    print("Both the dictionaries are equal")
else:
    print("Both the dictionaries are not equal")
```

Both the dictionaries are equal

1.3 Calculation of Closeness Centrality

1.3.1 Calculation of Closeness Centrality using LA

```
[9]: def compute_closeness centrality(graph):
    num_nodes = graph.number_of_nodes()

    closeness_scores = {}

    for node in graph.nodes():
        shortest_paths = nx.shortest_path_length(graph, source=node)

        total_distance = sum(shortest_paths.values())

        closeness centrality = (num_nodes - 1) / total_distance

        closeness_scores[node] = closeness centrality

    return closeness_scores
```

```
closeness centrality_LA = compute_closeness centrality(G)
```

```
print("Closeness Centrality computed using LA:")  
print(closeness centrality_LA)
```

Closeness Centrality computed using LA:

```
{0: 0.5689655172413793, 1: 0.4852941176470588, 2: 0.559322033898305, 3:  
0.4647887323943662, 4: 0.3793103448275862, 5: 0.38372093023255816, 6:  
0.38372093023255816, 7: 0.44, 8: 0.515625, 9: 0.4342105263157895, 10:  
0.3793103448275862, 11: 0.36666666666666664, 12: 0.3707865168539326, 13:  
0.515625, 14: 0.3707865168539326, 15: 0.3707865168539326, 16:  
0.28448275862068967, 17: 0.375, 18: 0.3707865168539326, 19: 0.5, 20:  
0.3707865168539326, 21: 0.375, 22: 0.3707865168539326, 23: 0.39285714285714285,  
24: 0.375, 25: 0.375, 26: 0.3626373626373626, 27: 0.4583333333333333, 28:  
0.4520547945205479, 29: 0.38372093023255816, 30: 0.4583333333333333, 31:  
0.5409836065573771, 32: 0.515625, 33: 0.55}
```

1.3.2 Calculation of Closeness Centrality using In-Built function

```
[10]: closeness centrality_inBuilt = nx.closeness centrality(G)  
print("Closeness Centrality using InBuilt function:")  
print(closeness centrality_inBuilt)
```

Closeness Centrality using InBuilt function:

```
{0: 0.5689655172413793, 1: 0.4852941176470588, 2: 0.559322033898305, 3:  
0.4647887323943662, 4: 0.3793103448275862, 5: 0.38372093023255816, 6:  
0.38372093023255816, 7: 0.44, 8: 0.515625, 9: 0.4342105263157895, 10:  
0.3793103448275862, 11: 0.36666666666666664, 12: 0.3707865168539326, 13:  
0.515625, 14: 0.3707865168539326, 15: 0.3707865168539326, 16:  
0.28448275862068967, 17: 0.375, 18: 0.3707865168539326, 19: 0.5, 20:  
0.3707865168539326, 21: 0.375, 22: 0.3707865168539326, 23: 0.39285714285714285,  
24: 0.375, 25: 0.375, 26: 0.3626373626373626, 27: 0.4583333333333333, 28:  
0.4520547945205479, 29: 0.38372093023255816, 30: 0.4583333333333333, 31:  
0.5409836065573771, 32: 0.515625, 33: 0.55}
```

1.3.3 Comparison of the Closeness Centrality Obtained by calculation using Linear Algebra and using Inbuilt function of networkx

```
[11]: if compare_dictionaries(closeness centrality_LA ,closeness centrality_inBuilt,  
    ↪DecimalPlaces):  
    print("Both the dictionaries are equal")  
else:  
    print("Both the dictionaries are not equal")
```

Both the dictionaries are equal

1.4 Calculation of EigenVector Centrality

1.4.1 Calculation of Eigenvector Centrality using LA

```
[12]: def eigenvector_centrality(adjacency_matrix, max_iterations=1000):
    A = np.array(adjacency_matrix)
    arr = np.ones(adjacency_matrix.shape[0])
    arr = arr.T
    for i in range(max_iterations):
        arr = A @ arr
        arr = arr/np.linalg.norm(arr)

    centrality_scores = dict(enumerate(arr))
    return centrality_scores

G = nx.karate_club_graph()

weighted_adjacency_matrix = nx.to_numpy_array(G)

unweighted_adjacency_matrix = (weighted_adjacency_matrix > 0).astype(int)

eigenvector_centrality_LA = eigenvector_centrality(unweighted_adjacency_matrix)
print("Eigenvector Centrality Computed Using LA:")
print(eigenvector_centrality_LA)
```

Eigenvector Centrality Computed Using LA:

```
{0: 0.35549144452456655, 1: 0.26595991955249165, 2: 0.31719250448643155, 3:
0.21117972037789023, 4: 0.07596881818306894, 5: 0.07948304511709947, 6:
0.07948304511709947, 7: 0.17095974804479633, 8: 0.22740390712540018, 9:
0.10267425072358623, 10: 0.07596881818306894, 11: 0.05285569749352132, 12:
0.08425462871671373, 13: 0.22647272014248127, 14: 0.10140326218952461, 15:
0.10140326218952461, 16: 0.023635628104591376, 17: 0.09239953819570258, 18:
0.10140326218952461, 19: 0.1479125102933875, 20: 0.10140326218952461, 21:
0.09239953819570258, 22: 0.10140326218952461, 23: 0.15011857186115288, 24:
0.05705244054116565, 25: 0.05920647491677849, 26: 0.07557941348827213, 27:
0.13347715338024013, 28: 0.13107782298371085, 29: 0.13496081926232786, 30:
0.17475830231435285, 31: 0.19103384140654378, 32: 0.30864421979104717, 33:
0.37336347029148315}
```

1.4.2 Calculation of Eigenvector Centrality using In-Built function

```
[13]: eigenvector_centrality_inBuilt = nx.eigenvector_centrality(G)
print("Eigenvector Centrality using InBuilt function:")
print(eigenvector_centrality_inBuilt)
```

Eigenvector Centrality using InBuilt function:

```
{0: 0.35548349418519426, 1: 0.2659538704545024, 2: 0.3171893899684447, 3:
0.21117407832057056, 4: 0.0759664588165738, 5: 0.07948057788594245, 6:
```



```
0.07948057788594245, 7: 0.1709551149803543, 8: 0.22740509147166046, 9:
0.10267519030637756, 10: 0.0759664588165738, 11: 0.05285416945233646, 12:
0.08425192086558085, 13: 0.22646969838808145, 14: 0.10140627846270832, 15:
0.10140627846270832, 16: 0.02363479426059687, 17: 0.0923967566684595, 18:
0.10140627846270832, 19: 0.14791134007618667, 20: 0.10140627846270832, 21:
0.0923967566684595, 22: 0.10140627846270832, 23: 0.15012328691726787, 24:
0.057053735638028055, 25: 0.0592082025027901, 26: 0.07558192219009324, 27:
0.13347932684333308, 28: 0.13107925627221215, 29: 0.13496528673866567, 30:
0.17476027834493088, 31: 0.191036269797917, 32: 0.3086510477336959, 33:
0.37337121301323506}
```

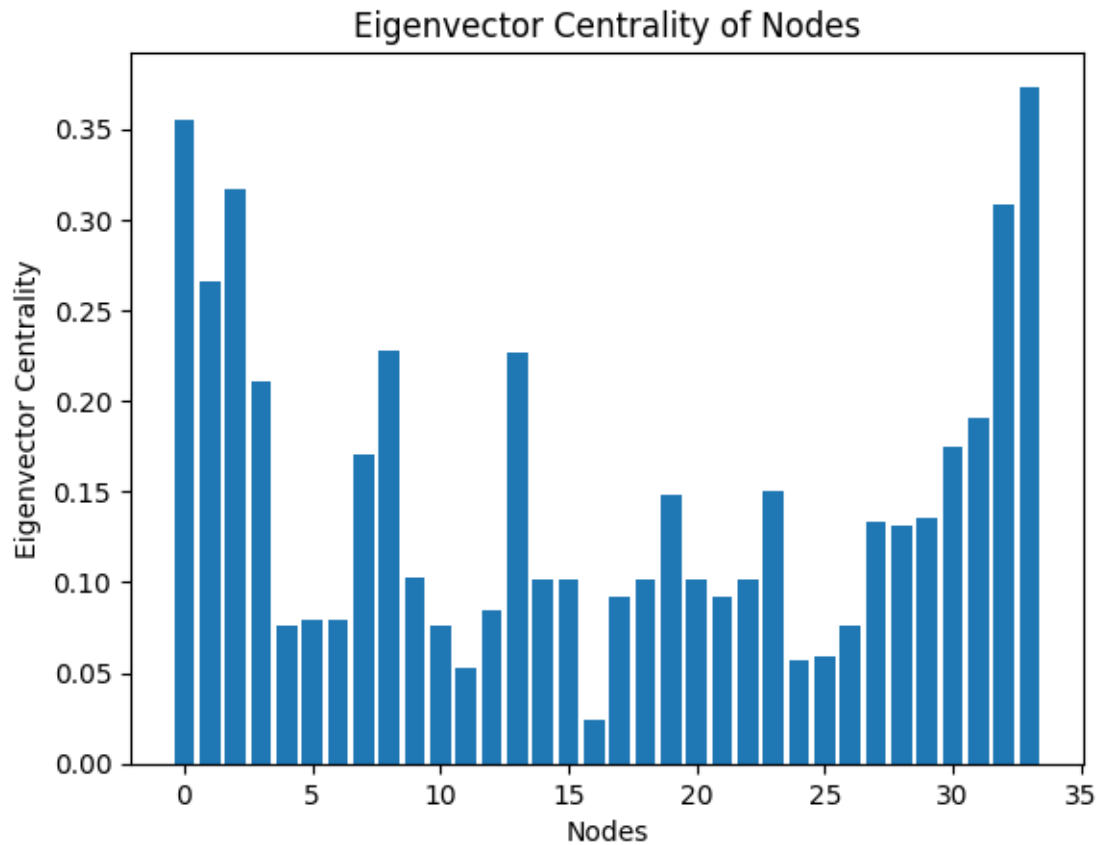
1.4.3 Comparison of the Eigenvector Centrality Obtained by calculation using Linear Algebra and using Inbuilt function of networkx

```
[14]: if compare_dictionaries(eigenvector_centrality_inBuilt,
    ↪, eigenvector_centrality_inBuilt, 4):
    print("Both the dictionaries are equal")
else:
    print("Both the dictionaries are not equal")
```

Both the dictionaries are equal

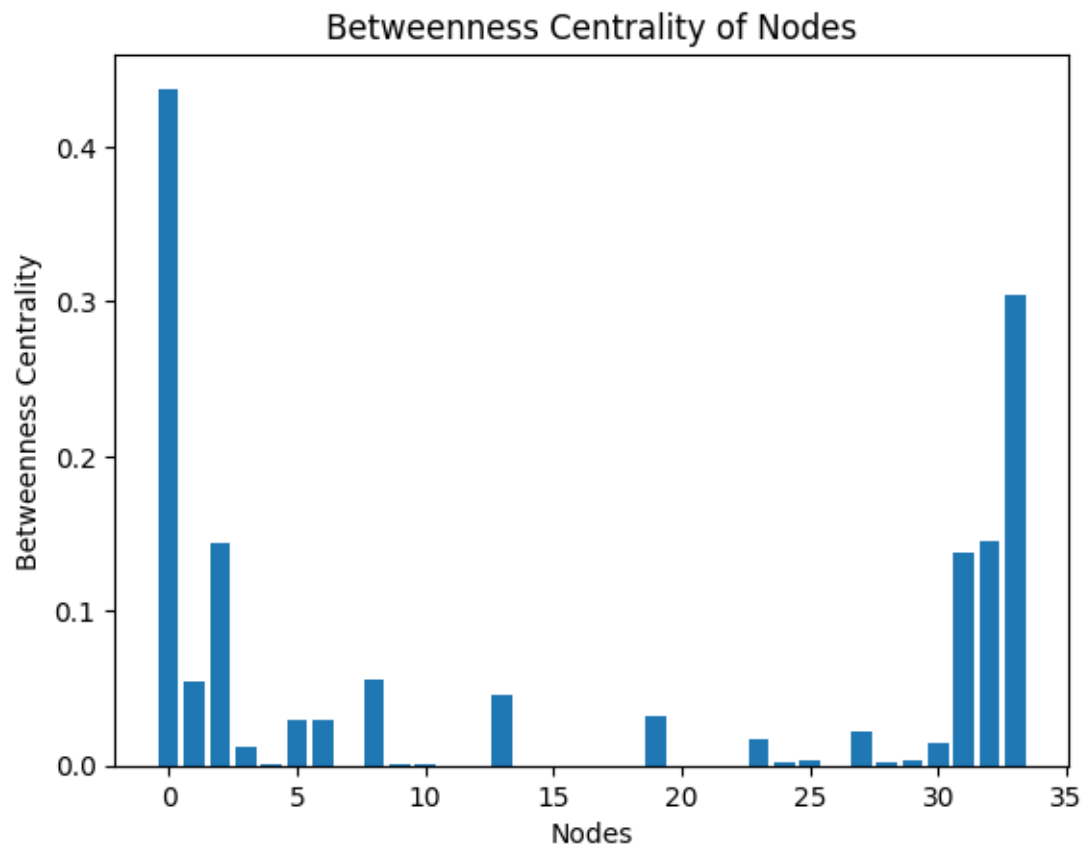
```
[23]: x = eigenvector_centrality_inBuilt.keys()
y = eigenvector_centrality_inBuilt.values()

plt.bar(x,y)
plt.title("Eigenvector Centrality of Nodes")
plt.xlabel("Nodes")
plt.ylabel("Eigenvector Centrality")
plt.show()
```



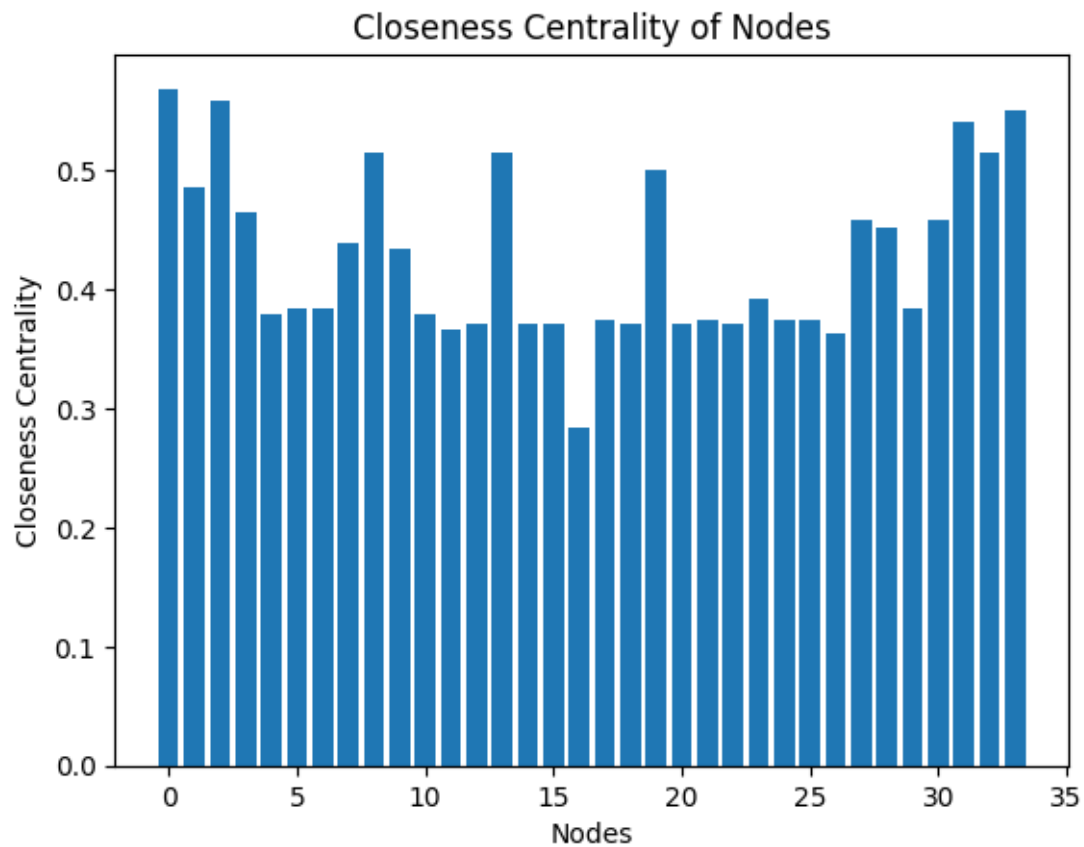
```
[24]: x = betweenness centrality_inBuilt.keys()
      y = betweenness centrality_inBuilt.values()

      plt.bar(x,y)
      plt.title("Betweenness Centrality of Nodes")
      plt.xlabel("Nodes")
      plt.ylabel("Betweenness Centrality")
      plt.show()
```



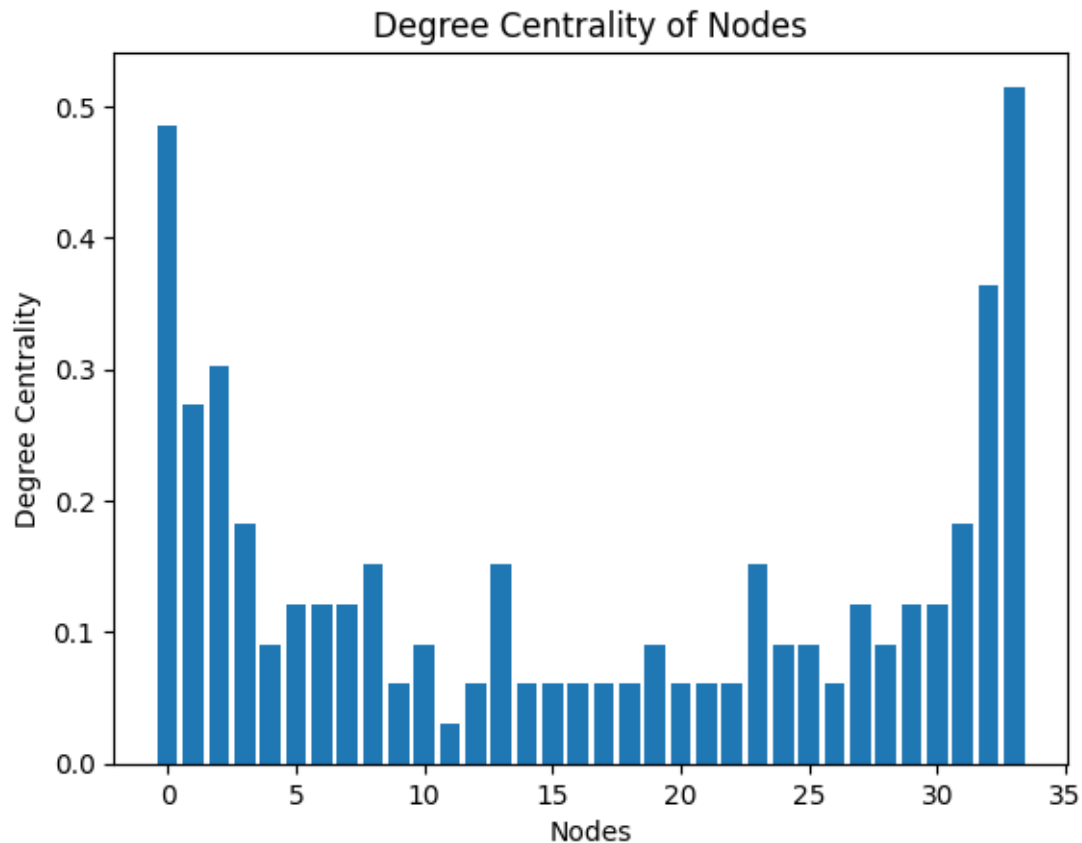
```
[25]: x = closeness centrality_inBuilt.keys()
      y = closeness centrality_inBuilt.values()

      plt.bar(x,y)
      plt.title("Closeness Centrality of Nodes")
      plt.xlabel("Nodes")
      plt.ylabel("Closeness Centrality")
      plt.show()
```



```
[26]: x = degree centrality_inBuilt.keys()
      y = degree centrality_inBuilt.values()

      plt.bar(x,y)
      plt.title("Degree Centrality of Nodes")
      plt.xlabel("Nodes")
      plt.ylabel("Degree Centrality")
      plt.show()
```



2 Finding All Centralities of Facebook Data

2.1 Taking 100 Edges

```
[15]: G = nx.Graph()

NumberofNodes = 100

with open("data.txt","r") as f:
    for i in range(NumberofNodes):
        data = f.readline()
        node1,node2 = data.strip().split(' ')
        G.add_edge(node1,node2)

degree centrality = nx.degree centrality(G)
betweenness centrality = nx.betweenness centrality(G)
closeness centrality = nx.closeness centrality(G)
```

```

eigenvector_centrality = nx.eigenvector_centrality(G)

max_degree_centrality = max(degree_centrality,key=degree_centrality.get)
max_betweenness_centrality = _
    ↪max(betweenness_centrality,key=betweenness_centrality.get)
max_closeness_centrality = max(closeness_centrality,key=closeness_centrality.
    ↪get)
max_eigenvector_centrality = _
    ↪max(eigenvector_centrality,key=eigenvector_centrality.get)

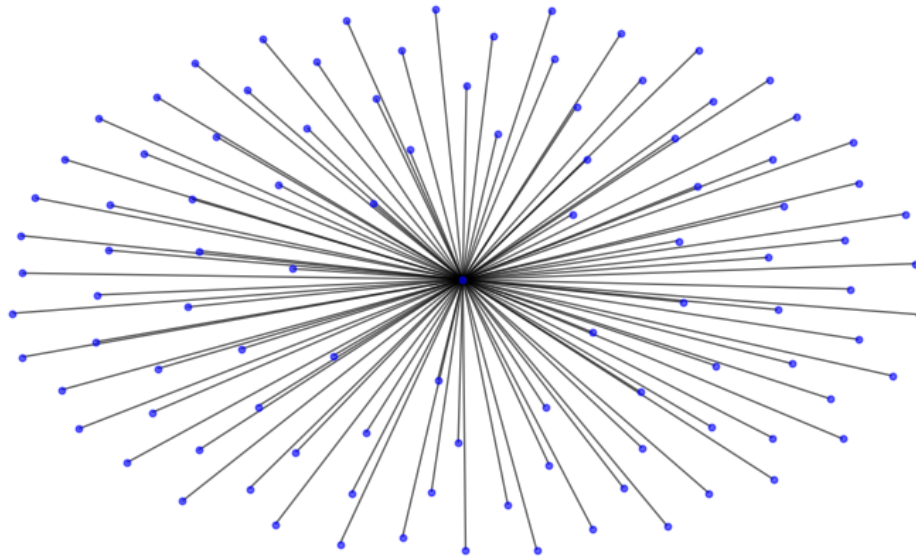
print(f"Max Degree Centrality Node:{max_degree_centrality} and It's Centrality:
    ↪{degree_centrality[max_degree_centrality]}")
print(f"Max Betweenness Centrality Node:{max_betweenness_centrality} and It's _
    ↪Centrality:{betweenness_centrality[max_betweenness_centrality]}")
print(f"Max Closeness Centrality Node:{max_closeness_centrality} and It's _
    ↪Centrality:{closeness_centrality[max_closeness_centrality]}")
print(f"Max Eigenvector Centrality Node:{max_eigenvector_centrality} and It's _
    ↪Centrality:{eigenvector_centrality[max_eigenvector_centrality]}")
print()

import matplotlib.pyplot as plt
plt.figure(figsize=(10, 6))
pos = nx.spring_layout(G, seed=42)
nx.draw_networkx(G, pos, with_labels=False, node_size=10, node_color='blue', _
    ↪alpha=0.6)
plt.title("Facebook Social Circles Graph")
plt.axis("off")
plt.show()

```

Max Degree Centrality Node:0 and It's Centrality:1.0
 Max Betweenness Centrality Node:0 and It's Centrality:1.0
 Max Closeness Centrality Node:0 and It's Centrality:1.0
 Max Eigenvector Centrality Node:0 and It's Centrality:0.7071033667005829

Facebook Social Circles Graph



2.2 Taking 1000 Edges

```
[16]: G = nx.Graph()

NumberofNodes = 1000

with open("data.txt","r") as f:
    for i in range(NumberofNodes):
        data = f.readline()
        node1,node2 = data.strip().split(' ')
        G.add_edge(node1,node2)

degree centrality = nx.degree centrality(G)
betweenness centrality = nx.betweenness centrality(G)
closeness centrality = nx.closeness centrality(G)
eigenvector centrality = nx.eigenvector centrality(G)

max_degree centrality = max(degree centrality,key=degree centrality.get)
max_betweenness centrality = max(betweenness centrality,key=betweenness centrality.get)
max_closeness centrality = max(closeness centrality,key=closeness centrality.get)
```

```

max_eigenvector_centrality =
    ↪max(eigenvector_centrality,key=eigenvector_centrality.get)

print(f"Max Degree Centrality Node:{max_degree_centrality} and It's Centrality:
    ↪{degree_centrality[max_degree_centrality]}")
print(f"Max Betweenness Centrality Node:{max_betweenness_centrality} and It's
    ↪Centrality:{betweenness_centrality[max_betweenness_centrality]}")
print(f"Max Closeness Centrality Node:{max_closeness_centrality} and It's
    ↪Centrality:{closeness_centrality[max_closeness_centrality]}")
print(f"Max Eigenvector Centrality Node:{max_eigenvector_centrality} and It's
    ↪Centrality:{eigenvector_centrality[max_eigenvector_centrality]}")
print()

import matplotlib.pyplot as plt
plt.figure(figsize=(10, 6))
pos = nx.spring_layout(G, seed=42)
nx.draw_networkx(G, pos, with_labels=False, node_size=10, node_color='blue',
    ↪alpha=0.6)
plt.title("Facebook Social Circles Graph")
plt.axis("off")
plt.show()

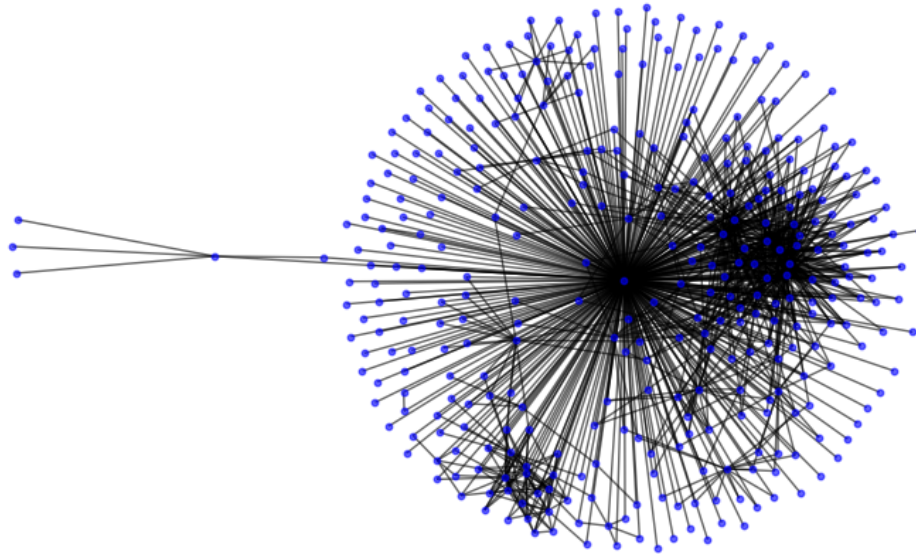
```

```

Max Degree Centrality Node:0 and It's Centrality:0.9914285714285714
Max Betweenness Centrality Node:0 and It's Centrality:0.927374266793833
Max Closeness Centrality Node:0 and It's Centrality:0.9915014164305949
Max Eigenvector Centrality Node:0 and It's Centrality:0.559416676755645

```


Facebook Social Circles Graph



2.3 Taking 10000 Edges

```
[17]: G = nx.Graph()

NumberofNodes = 10000

with open("data.txt","r") as f:
    for i in range(NumberofNodes):
        data = f.readline()
        node1,node2 = data.strip().split(' ')
        G.add_edge(node1,node2)

degree centrality = nx.degree centrality(G)
betweenness centrality = nx.betweenness centrality(G)
closeness centrality = nx.closeness centrality(G)
eigenvector centrality = nx.eigenvector centrality(G)

max_degree centrality = max(degree centrality,key=degree centrality.get)
max_betweenness centrality = max(betweenness centrality,key=betweenness centrality.get)
max_closeness centrality = max(closeness centrality,key=closeness centrality.get)
```

```

max_eigenvector_centrality =
    ↪max(eigenvector_centrality,key=eigenvector_centrality.get)

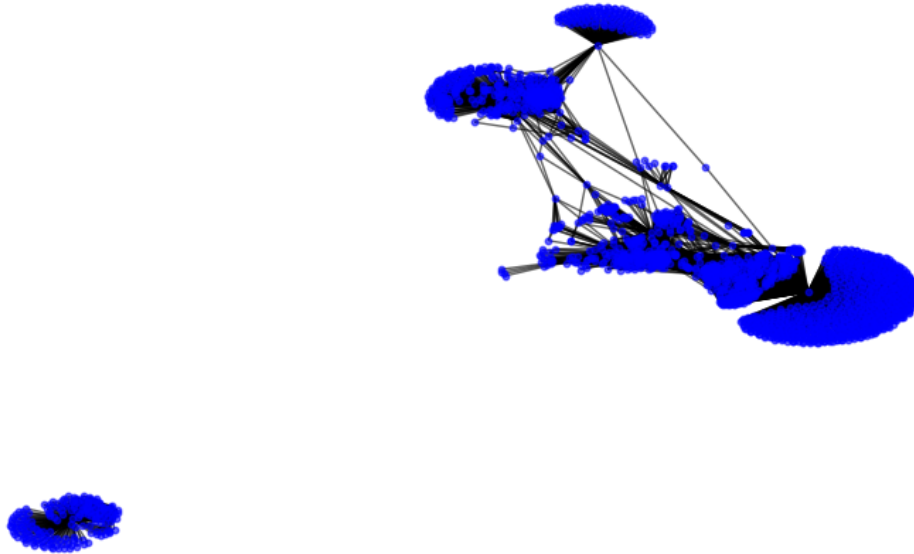
print(f"Max Degree Centrality Node:{max_degree_centrality} and It's Centrality:
    ↪{degree_centrality[max_degree_centrality]}")
print(f"Max Betweenness Centrality Node:{max_betweenness_centrality} and It's
    ↪Centrality:{betweenness_centrality[max_betweenness_centrality]}")
print(f"Max Closeness Centrality Node:{max_closeness_centrality} and It's
    ↪Centrality:{closeness_centrality[max_closeness_centrality]}")
print(f"Max Eigenvector Centrality Node:{max_eigenvector_centrality} and It's
    ↪Centrality:{eigenvector_centrality[max_eigenvector_centrality]}")
print()

import matplotlib.pyplot as plt
plt.figure(figsize=(10, 6))
pos = nx.spring_layout(G, seed=42)
nx.draw_networkx(G, pos, with_labels=False, node_size=10, node_color='blue',
    ↪alpha=0.6)
plt.title("Facebook Social Circles Graph")
plt.axis("off")
plt.show()

```

Max Degree Centrality Node:107 and It's Centrality:0.5222388805597201
 Max Betweenness Centrality Node:107 and It's Centrality:0.6663327965109708
 Max Closeness Centrality Node:107 and It's Centrality:0.6121482053413674
 Max Eigenvector Centrality Node:348 and It's Centrality:0.1837055147302193

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2.4 Taking all Edges

```
[18]: G = nx.Graph()

NumberofNodes = 88234

with open("data.txt","r") as f:
    for i in range(NumberofNodes):
        data = f.readline()
        node1,node2 = data.strip().split(' ')
        G.add_edge(node1,node2)

degree centrality = nx.degree centrality(G)
betweenness centrality = nx.betweenness centrality(G)
closeness centrality = nx.closeness centrality(G)
eigenvector centrality = nx.eigenvector centrality(G)

max_degree centrality = max(degree centrality,key=degree centrality.get)
max_betweenness centrality = max(betweenness centrality,key=betweenness centrality.get)
max_closeness centrality = max(closeness centrality,key=closeness centrality.get)
```

```

max_eigenvector_centrality =
    ↪max(eigenvector_centrality,key=eigenvector_centrality.get)

print(f"Max Degree Centrality Node:{max_degree_centrality} and It's Centrality:
    ↪{degree_centrality[max_degree_centrality]}")
print(f"Max Betweenness Centrality Node:{max_betweenness_centrality} and It's
    ↪Centrality:{betweenness_centrality[max_betweenness_centrality]}")
print(f"Max Closeness Centrality Node:{max_closeness_centrality} and It's
    ↪Centrality:{closeness_centrality[max_closeness_centrality]}")
print(f"Max Eigenvector Centrality Node:{max_eigenvector_centrality} and It's
    ↪Centrality:{eigenvector_centrality[max_eigenvector_centrality]}")
print()

import matplotlib.pyplot as plt
plt.figure(figsize=(10, 6))
pos = nx.spring_layout(G, seed=42)
nx.draw_networkx(G, pos, with_labels=False, node_size=10, node_color='blue',
    ↪alpha=0.6)
plt.title("Facebook Social Circles Graph")
plt.axis("off")
plt.show()

```

Max Degree Centrality Node:107 and It's Centrality:0.258791480931154
 Max Betweenness Centrality Node:107 and It's Centrality:0.4805180785560152
 Max Closeness Centrality Node:107 and It's Centrality:0.45969945355191255
 Max Eigenvector Centrality Node:1912 and It's Centrality:0.09540696149067629

Facebook Social Circles Graph

