

```

import networkx as nx
import itertools
import matplotlib.pyplot as plt
import numpy as np
from networkx.algorithms.community import girvan_newman

def generate_graph(n, k):
    G = nx.Graph()
    G.add_nodes_from(range(1, n+1))

    for i in range(1, n+1):
        for j in range(i+1, n+1):
            if i % k != j % k:
                G.add_edge(i, j)

    return G

def visualize_graph(G, communities=None):
    plt.figure(figsize=(12, 8))

    pos = nx.spring_layout(G, seed=42)

    if communities:
        unique_communities = set(communities.values())
        color_map = plt.cm.rainbow(np.linspace(0, 1,
len(unique_communities)))

        color_dict = {com: color_map[i] for i, com in
enumerate(unique_communities)}
        node_colors = [color_dict[communities[node]] for node in
G.nodes()]

        nx.draw(G, pos, node_color=node_colors, with_labels=True,
node_size=300, font_size=8, alpha=0.7)
    else:
        nx.draw(G, pos, with_labels=True, node_size=300, font_size=8,
alpha=0.7)

    plt.title('Graph Visualization')
    plt.tight_layout()
    plt.show()

```

[30 marks] Generating the graphs and visualizing them with Networkx or GEPHI.

```

graph_sizes = [(40, 7), (20, 5), (10, 3)]

graph_list = []
for n, k in graph_sizes:
    print(f"\n{'='*50}")
    print(f"Graph size (n, k) = ({n}, {k})")

```

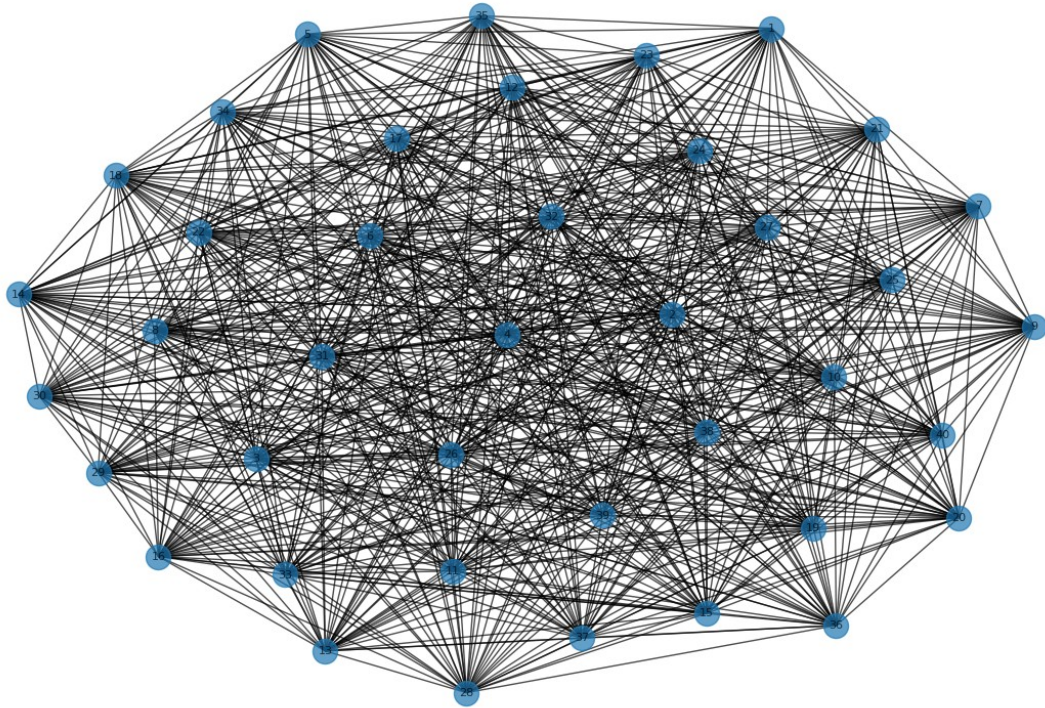
```
print(f"{'='*50}")
```

```
G = generate_graph(n, k)  
graph_list.append(G)  
visualize_graph(G)
```

```
=====  
Graph size (n, k) = (40, 7)  
=====
```

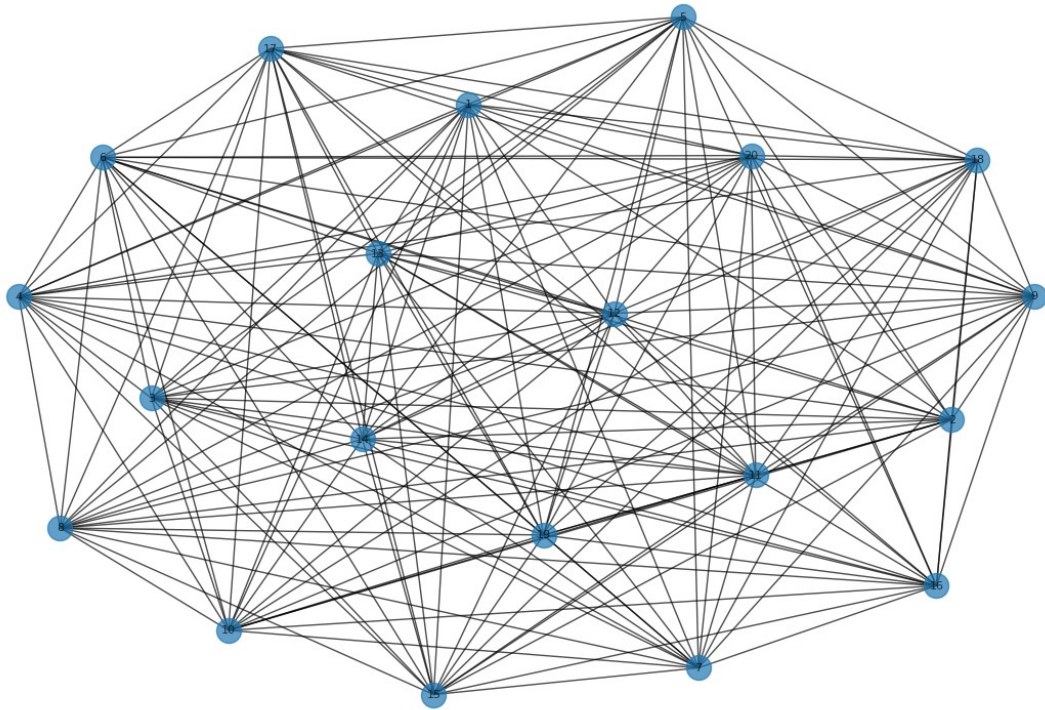
```
<ipython-input-30-640a4c8181d5>:34: UserWarning: This figure includes  
Axes that are not compatible with tight_layout, so results might be  
incorrect.  
plt.tight_layout()
```

Graph Visualization



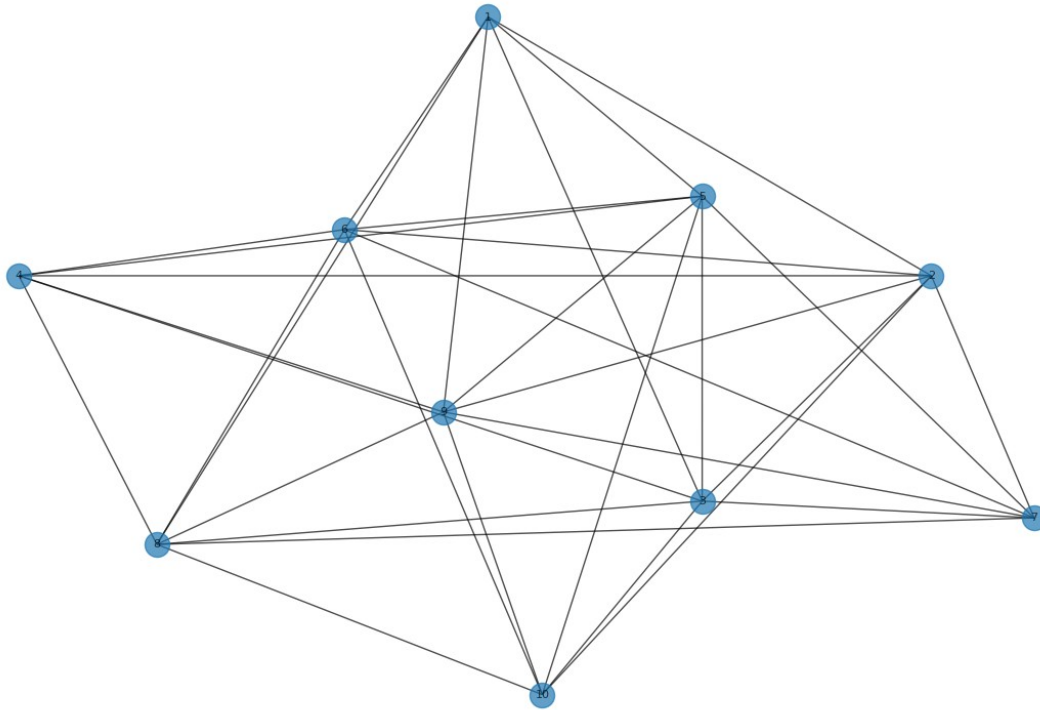
```
=====  
Graph size (n, k) = (20, 5)  
=====
```

# Graph Visualization



=====  
Graph size (n, k) = (10, 3)  
=====

Graph Visualization



[30 marks] Count the number of cliques of each size in each graph ...  
Plot their histograms

```
graph_sizes = [(40, 7), (20, 5), (10, 3)]

for i, tup in enumerate(graph_sizes):
    n, k = tup
    G = graph_list[i]
    cliques = list(nx.find_cliques(G))

    # Count number of cliques
    clique_counts = {}
    for clique in cliques:
        size = len(clique)
        clique_counts[size] = clique_counts.get(size, 0) + 1

    print("\nClique Analysis:")
    print("Total number of cliques:", len(cliques))
    print("Clique Counts:", clique_counts)

    plt.figure(figsize=(10, 6))

    if clique_counts:
        max_size = max(clique_counts.keys())
```

```

    sizes = list(range(1, max_size + 1))
    counts = [clique_counts.get(size, 0) for size in sizes]
else:
    sizes = []
    counts = []

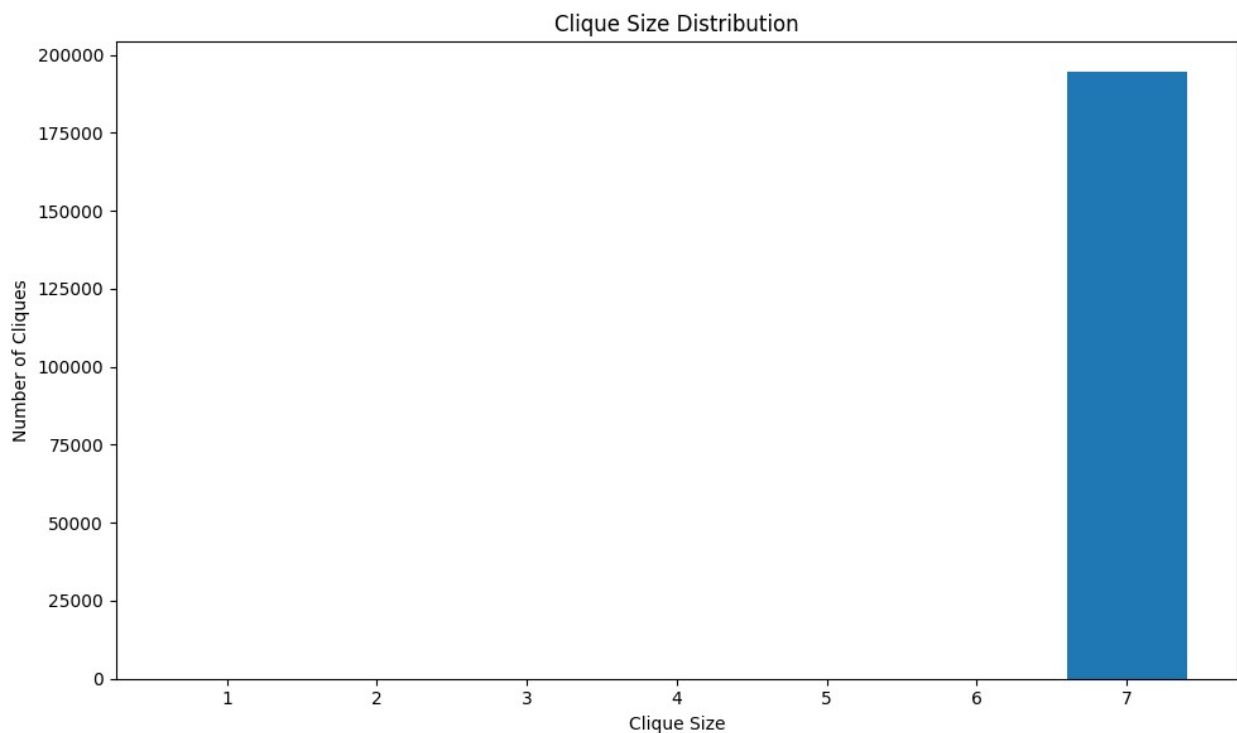
plt.bar(sizes, counts)
plt.title('Clique Size Distribution')
plt.xlabel('Clique Size')
plt.ylabel('Number of Cliques')
plt.xticks(sizes)
plt.tight_layout()
plt.show()

```

Clique Analysis:

Total number of cliques: 194400

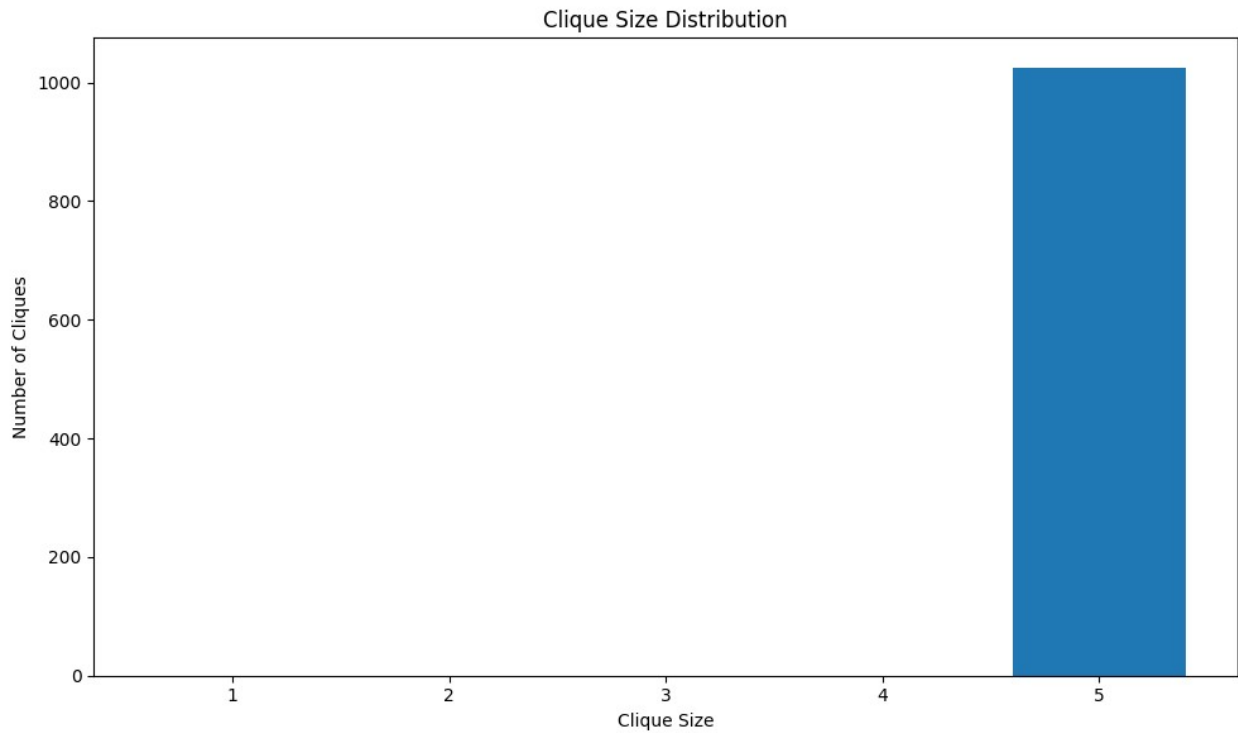
Clique Counts: {7: 194400}



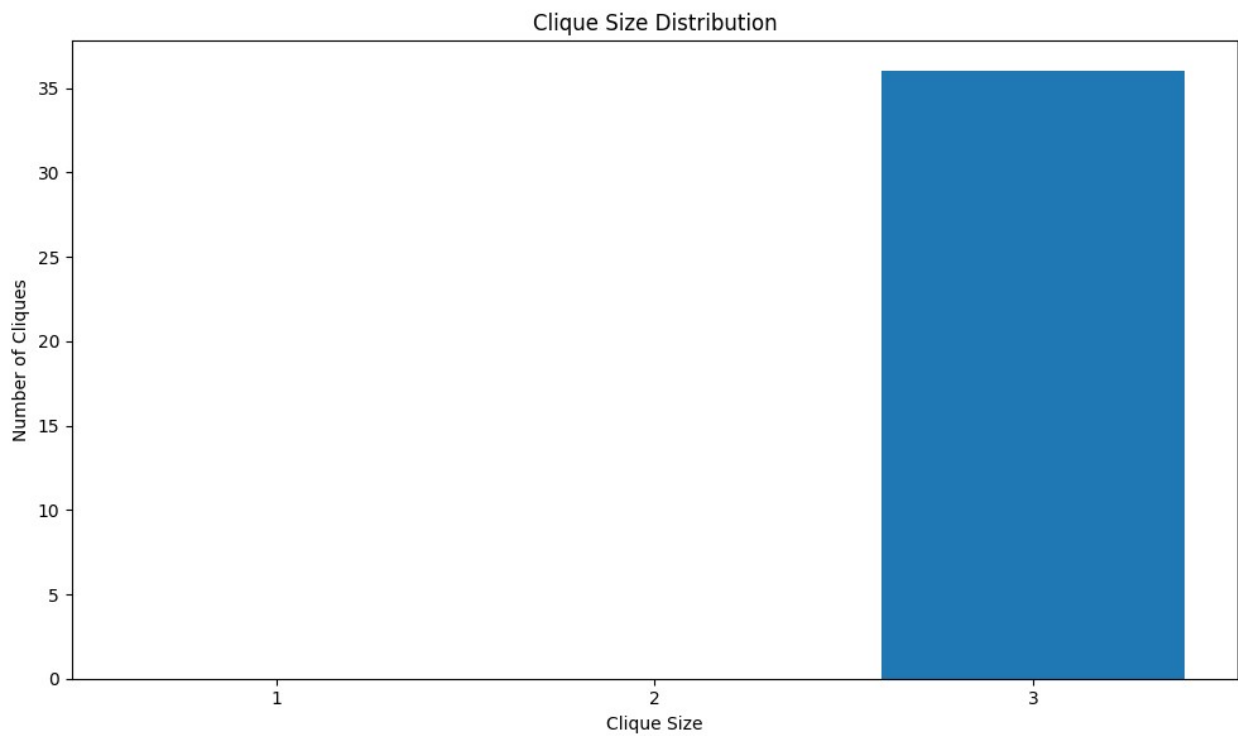
Clique Analysis:

Total number of cliques: 1024

Clique Counts: {5: 1024}



Clique Analysis:  
Total number of cliques: 36  
Clique Counts: {3: 36}



[30 marks] Identify the main communities in the graphs generated

```
graph_sizes = [(40, 7), (20, 5), (10, 3)]
```

```
for i, tup in enumerate(graph_sizes):
    n, k = tup
    G = graph_list[i]
    comp = girvan_newman(G)
    first_partition = next(comp)
    communities = [list(c) for c in first_partition]

    partition = {}
    for community_id, nodes in enumerate(communities):
        for node in nodes:
            partition[node] = community_id

    # Count communities
    community_counts = {}
    for community_id in set(partition.values()):
        community_size = sum(1 for v in partition if partition[v] ==
community_id)
        community_counts[community_id] = community_size

    print("\nCommunity Analysis:")
    print("Number of Communities:", len(set(partition.values())))
    print("Community Sizes:", dict(sorted(community_counts.items(),
key=lambda x: x[1], reverse=True)))
    visualize_graph(G)
```

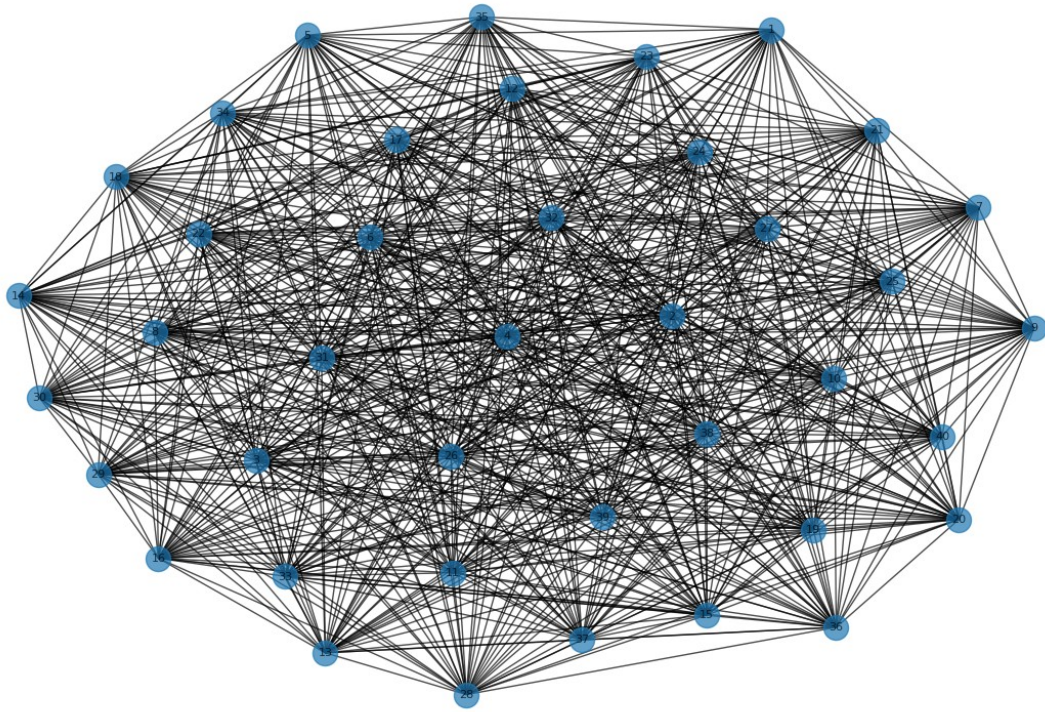
```
Community Analysis:
Number of Communities: 2
Community Sizes: {0: 39, 1: 1}
```

```
<ipython-input-30-640a4c8181d5>:34: UserWarning: This figure includes
Axes that are not compatible with tight_layout, so results might be
incorrect.
```

```
plt.tight_layout()
```



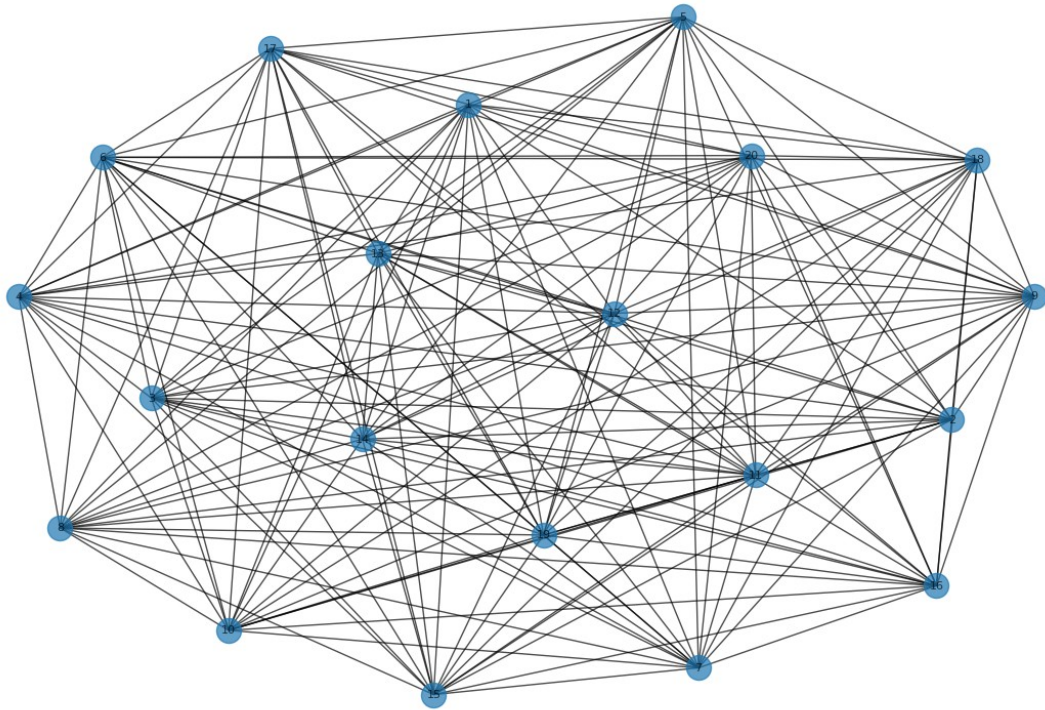
Graph Visualization



Community Analysis:  
Number of Communities: 2  
Community Sizes: {1: 19, 0: 1}

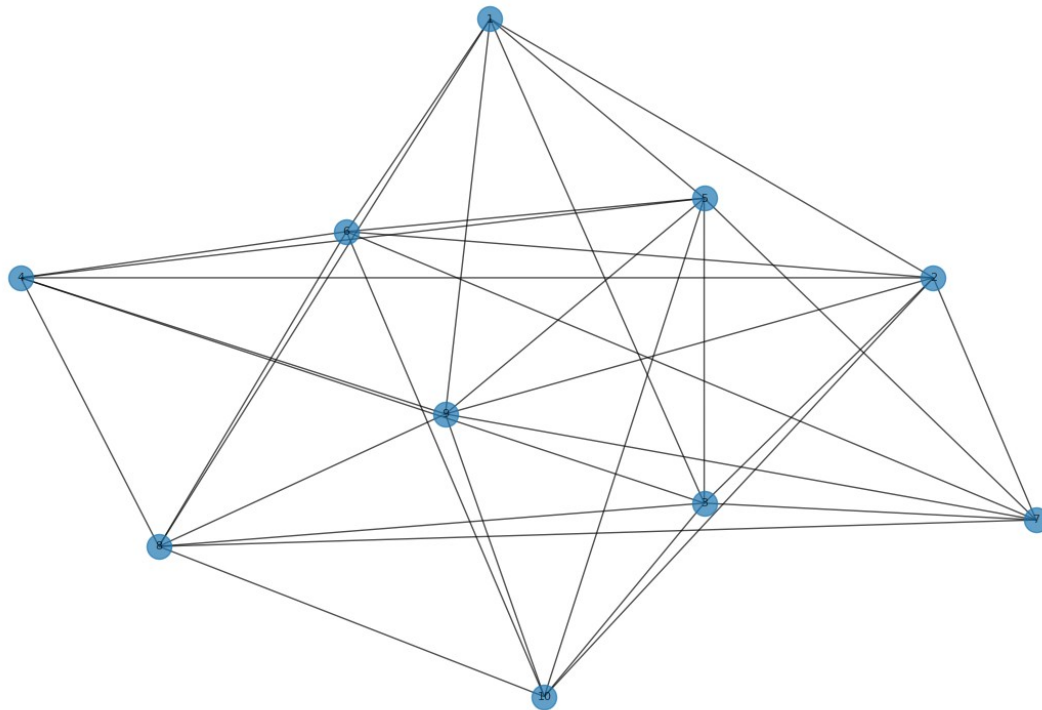


Graph Visualization



Community Analysis:  
Number of Communities: 2  
Community Sizes: {0: 9, 1: 1}

Graph Visualization



[30 marks] Based on Example 10.10,

You are given a graph say  $(n, k)$  where  $n$  represents no. of nodes numbered 1 to  $n$  &  $k$  represents an arbitrary number using which we will generate our graph. Now as per the example 10.10 In such a graph, a link exists between nodes if 2 nodes numbered  $i$  &  $j$  when divided by  $k$  doesn't leave the same remainder. Generate such a graph, calculate the number of edges present in it & compare it with the approximation offered by the fraction  $(k-1)/k$  for the following values of  $n$  &  $k$

```
def generate_graph(n, k):
    graph = {i: [] for i in range(1, n+1)}
    edge_count = 0

    for i in range(1, n+1):
        for j in range(i+1, n+1):
            if (i % k) != (j % k):
                graph[i].append(j)
                graph[j].append(i)
                edge_count += 1

    return graph, edge_count

def analyze_graph(n, k):
    graph, actual_edges = generate_graph(n, k)
```

```

    theoretical_edges = n * (n-1) * (k-1) / (2 * k)

    approximation_accuracy = abs(actual_edges - theoretical_edges) /
actual_edges * 100

    return {
        'n': n,
        'k': k,
        'actual_edges': actual_edges,
        'theoretical_edges': theoretical_edges,
        'approximation_accuracy': approximation_accuracy
    }

test_cases = [(40, 7), (20, 5), (10, 3)]
results = [analyze_graph(n, k) for n, k in test_cases]
for result in results:
    print(f"\nGraph Analysis for n={result['n']}, k={result['k']}:")
    print(f"Actual Number of Edges: {result['actual_edges']}")
    print(f"Theoretical Approximation:
{result['theoretical_edges']:.2f}")
    print(f"Approximation Accuracy:
{result['approximation_accuracy']:.2f}%")

```

```

Graph Analysis for n=40, k=7:
Actual Number of Edges: 685
Theoretical Approximation: 668.57
Approximation Accuracy: 2.40%

```

```

Graph Analysis for n=20, k=5:
Actual Number of Edges: 160
Theoretical Approximation: 152.00
Approximation Accuracy: 5.00%

```

```

Graph Analysis for n=10, k=3:
Actual Number of Edges: 33
Theoretical Approximation: 30.00
Approximation Accuracy: 9.09%

```