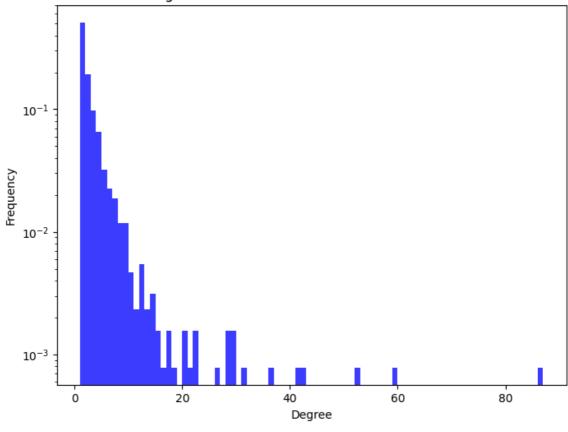
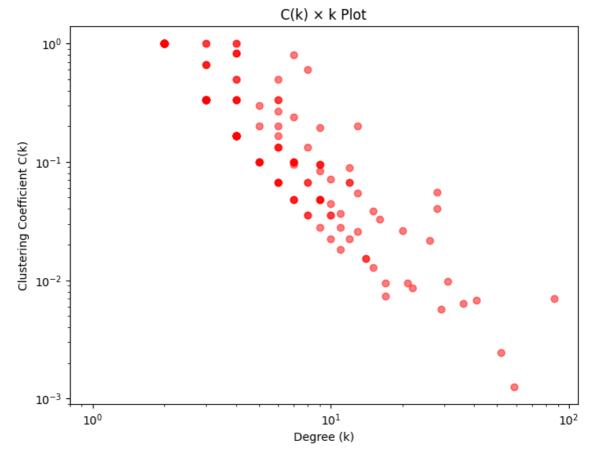
Ques5. Part A - Construct the Yeast Protein Interactome (YPI) using data available from Yeast Interactome Database (http://interactome.dfci.harvard.edu/S_cerevisiae/). Plot its degree distribution as well as C(k) × k plot. Visualize the network using Cytoscape and highlight importance of a protein in terms of its degree/connectivity.

```
In [ ]: import networkx as nx
        import matplotlib.pyplot as plt
        import numpy as np
        import pandas as pd
        file path = 'CCSB-Y2H.txt'
        df = pd.read_csv(file_path, sep='\t', header=None, names=['Protein1', 'Pr
        # Construct the Yeast Protein Interactome (YPI) network
        G = nx.Graph()
        G.add_edges_from(zip(df['Protein1'], df['Protein2']))
        num nodes = G.number of nodes()
        num edges = G.number of edges()
        print(f"Number of nodes (proteins): {num nodes}")
        print(f"Number of interactions: {num edges}")
        # Ploting degree distribution
        degrees = [deg for _, deg in G.degree()]
        plt.figure(figsize=(8, 6))
        plt.hist(degrees, bins=range(1, max(degrees)+1), density=True, alpha=0.75
        plt.xlabel('Degree')
        plt.ylabel('Frequency')
        plt.title('Degree Distribution of Yeast Protein Interactome')
        plt.yscale('log')
        plt.show()
        # Computing clustering coefficient and ploting C(k) \times k
        ck = nx.average clustering(G, count zeros=True)
        k_values = np.array(degrees)
        c k values = np.array([nx.clustering(G, node) for node in G.nodes()])
        plt.figure(figsize=(8, 6))
        plt.scatter(k_values, c_k_values, alpha=0.5, color='r')
        plt.xlabel('Degree (k)')
        plt.ylabel('Clustering Coefficient C(k)')
        plt.title('C(k) x k Plot')
        plt.xscale('log')
        plt.yscale('log')
        plt.show()
        #YPI Network Construction
        output_path = "/home/akash/Music/NS_A2/A2_submission/yeast_protein_intera
        nx.write graphml(G, output path)
        print(f"Graph exported as '{output path}' for Cytoscape visualization.")
        # Highlighting the importance of protein in terms of degree
        most connected = max(G.degree, key=lambda x: x[1])
        print(f"The most connected protein is {most_connected[0]} with {most_conn
        average_degree = np.mean(degrees)
        print(f"The average degree of the network is {average degree:.2f}")
```

Number of nodes (proteins): 1278 Number of interactions: 1809







Graph exported as '/home/akash/Music/NS_A2/A2_submission/yeast_protein_i nteractome.graphml' for Cytoscape visualization.

The most connected protein is YLR291C with 87 interactions.

The average degree of the network is 2.83

YLR291C is a hub protein, meaning it has significantly more interactions than the average protein in the network.

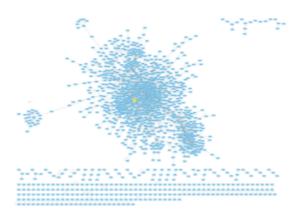
In the Yeast Protein Interactome, the degree of a protein represents its connectivity, indicating how many other proteins it interacts with.

High-degree proteins, known as hub proteins, play crucial roles in cellul ar processes, often acting as central regulators or essential components in biological pathways.

Identifying these hub proteins can help in understanding critical interactions, potential drug targets, and overall network stability.

```
In [39]: #CYTOSCAPE VISUALIZATION
    from PIL import Image
    import matplotlib.pyplot as plt
    image = Image.open("CCSB-Y2H.txt.png")

plt.imshow(image)
    plt.axis("off")
    plt.show()
```



Part B - Identify key proteins in YPI based on their number of interactions. Compute and plot the fraction of proteins that are essential for a given degree (k). (Find the information of essential genes from existing data sources from the internet.)

In [7]: # AS PER THE ANNOUNCEMENT WE HAVE SKIPPED THIS PART OF THE QUESTION

Part C - Implement the random and targeted node deletion strategies (Refer to Question 3 above) for YPI. Compute and plot the robustness (S and < s >) of YPI for these strategies.

```
import random
import networkx as nx
import numpy as np
import matplotlib.pyplot as plt

#Random and targeted node deletion strategies for YPI.
def gen_BA_mod(n, m, alpha=1, seed=None):
    if seed:
```

```
random.seed(seed)
        np.random.seed(seed)
    G = nx.complete graph(m)
    for nn in range(m, n):
        degrees = np.array([G.degree(n) for n in G.nodes()]) ** alpha
        prob = degrees / degrees.sum()
        targets = np.random.choice(list(G.nodes()), min(m, len(G)), repla
        G.add edges from((nn, t) for t in targets)
    return G
def comp robust(G, remove order):
    #Computes robustness metrics S and <s> for a given node removal order
    S values, s values = [], []
    G copy = G.copy()
    for node in remove order:
        G copy remove node(node)
        if len(G copy) == 0:
            S values.append(0)
            s values.append(0)
            break
        largest cc = max(nx.connected components(G copy), key=len, defaul)
        S values.append(len(largest cc))
        left ccs = []
        for cc in nx.connected components(G copy):
            if len(cc) < len(largest cc):</pre>
                left ccs.append(len(cc))
        if left ccs:
            s values.append(np.mean(left ccs))
        else:
            s values.append(0)
    return S values, s values
def plot_robust(S_random, s_random, S_targeted, s_targeted):
    #Plots robustness metrics for random and targeted attacks.
    plt.figure(figsize=(10, 5))
    plt.subplot(1, 2, 1)
    plt.plot(S_random, label='Random Attack', color='blue')
    plt.plot(S targeted, label='Targeted Attack', color='red')
    plt.xlabel('Fraction of Nodes Removed')
    plt.ylabel('Largest Component Size (S)')
    plt.legend()
    plt.title('Robustness of YPI: Largest Component Size')
    plt.subplot(1, 2, 2)
    plt.plot(s random, label='Random Attack', color='blue')
    plt.plot(s_targeted, label='Targeted Attack', color='red')
    plt.xlabel('Fraction of Nodes Removed')
    plt.ylabel('Average Component Size ((s))')
    plt.legend()
    plt.title('Robustness of YPI: Avg Component Size')
    plt.tight_layout()
    plt.show()
# Load YPI network from the dataset
def load_G(file_path):
    G = nx.Graph()
```

```
with open(file path, 'r') as f:
        for line in f:
            nodes = line.strip().split('\t')[:2]
            G.add edge(nodes[0], nodes[1])
    return G
# Load graph
file path = 'CCSB-Y2H.txt'
G_ypi = load_G(file_path)
# Generate modified BA model for comparison
n mod = len(G ypi.nodes())
m \mod = 3
G ba = gen BA mod(n mod, m mod, alpha=2, seed=42)
# Random node deletion
remove rand ypi = list(G ypi.nodes())
random.shuffle(remove rand ypi)
S random ypi, s random ypi = comp robust(G ypi, remove rand ypi)
remove rand ba = list(G ba.nodes())
random.shuffle(remove rand ba)
S_random_ba, s_random_ba = comp_robust(G_ba, remove_rand_ba)
# Targeted node deletion (highest-degree first)
nodes sorted ypi = sorted(G ypi.nodes(), key=G ypi.degree, reverse=True)
nodes sorted ba = sorted(G ba.nodes(), key=G ba.degree, reverse=True)
S targeted ypi, s targeted ypi = comp robust(G ypi, nodes sorted ypi)
S targeted ba, s targeted ba = comp robust(G ba, nodes sorted ba)
# Plot robustness
plot_robust(S_random_ypi, s_random_ypi, S_targeted_ypi, s_targeted_ypi)
plot robust(S random ba, s random ba, S targeted ba, s targeted ba)
```

