

PSG COLLEGE OF TECHNOLOGY

**DEPARTMENT OF COMPUTER SCIENCE AND
ENGINEERING**



19OH01 – SOCIAL AND ECONOMIC NETWORK ANALYSIS

TOPIC: GITHUB RECOMMENDER SYSTEM

TEAM MEMBERS:

18Z335 - NAVEEN P

19Z433 - GOPISANKAR G

19Z461 - RAHUL R

19Z464 - MOULEESWARAN S

Problem Statement:

By using this recommending system, we analyze user similarity and recommend repositories to users using the concept of bipartite graphs. In our project, there are two sets of nodes (Projects or Repositories and Users). By finding similar projects and similarity score between given users, we recommend repositories to users.

Dataset Description:

Our data set is in the form of adjacency list, whereas the first word is the head node and the following nodes are the neighbours of first node. The word begins with 'u' is considered as users in our project and word begin with 'p' is considered as projects or repositories in our project. Total number of nodes in the dataset is 22353 and edges are 20052. In that nodes 10721 are users and 11632 nodes are projects.

Tools Used:

1. API Used:

→ **Google Colaboratory** - Anybody to write and execute arbitrary python code through the browser, and is especially well suited to machine learning, data analysis and education.

2. Library:

→ **Networkx** - NetworkX is a Python library for studying graphs and networks.

Challenges faced:

1. Dataset is in the format of adjacency list. It is hard to modify the dataset to convert for our convenience.

2. We are new to this Google colab API and networkx. It takes time to understand colab environment and some functions in networkx package.

3. Initially, we have faced difficulties in loading the dataset with our concept in Gephi. So, we moved to Google Colab API because of easy implementation.

Contribution of Team Members:

S.NO	Roll.no	Name	Contribution	Report Preparation (Commonly done by using Google Meet)
1)	18Z335	Naveen P	Dataset Modification,Coding	
2)	19Z433	Gopisankar G	Code Part,Documentation	
3)	19Z461	Rahul R	Project Ideas,Documentation	
4)	19Z464	Mouleeswaran S	Graph Visualization,Coding	

Annexure I:

```
→ import networkx as nx
→ G=nx.Graph()
→ G=nx.read_adjlist("dataset.txt")
→ type(G)
→ len(G.nodes())
→ len(G.edges())

→ for n in G.nodes():
    if(n[1]=='p'):
        G.nodes[n]['bipartite'] = 'projects'
    else:
        G.nodes[n]['bipartite'] = 'users'

→ #To count number of users and projects in our dataset
def get_nodes_from_partition(G,partition):
    nodes = []
    for n in G.nodes():
        if G.nodes[n]['bipartite'] == partition:
            #if the node belongs to given partition, then the node is added to nodes list
            nodes.append(n)
    return nodes
print(len(get_nodes_from_partition(G, 'users')))
print(len(get_nodes_from_partition(G, 'projects'))))

→ def shared_partition_nodes(G, n1, n2):
    #Checking whether passed nodes belong to users partition or not
    assert G.nodes[n1]['bipartite'] == 'users'
    assert G.nodes[n2]['bipartite'] == 'users'

    # Getting neighbors of node1
    nbr1 = G.neighbors(n1)
    # Getting neighbors of node2
    nbr2 = G.neighbors(n2)

    # To get shared projects between users
    common = set(nbr1).intersection(nbr2)
    return common

→ def user_similarity(G, u1, u2, proj_nodes):
```

```

#Checking whether passed nodes belong to users partition or not
assert G.nodes[u1]['bipartite'] == 'users'
assert G.nodes[u2]['bipartite'] == 'users'

# To get shared projects between given users
shared_nodes = shared_partition_nodes(G, u1, u2)
# return similarity score
return len(shared_nodes) / len(proj_nodes)

```

```

→ from collections import defaultdict
def most_similar_users(G, user, user_nodes, proj_nodes):
    #Checking whether passed nodes belong to users partition or not
    assert G.nodes[user]['bipartite'] == 'users'

    # Getting other user nodes
    user_nodes = set(user_nodes)
    user_nodes.remove(user)

    # Creating dictionary
    similarities = defaultdict(list)
    for n in user_nodes:
        similarity = user_similarity(G, user, n, proj_nodes)
        similarities[similarity].append(n)

    # computing maximum similarity
    max_similarity = max(similarities.keys())

    # returning maximum similarity users
    return similarities[max_similarity]
user_nodes = get_nodes_from_partition(G, 'users')
project_nodes = get_nodes_from_partition(G, 'projects')

```

```

→ def recommend_repositories(G, from_user, to_user):
    # repositories of user1
    from_repos = set(G.neighbors(from_user))
    #repositories of user2
    to_repos = set(G.neighbors(to_user))

    print("Suggesting repositories to user1")
    print(to_repos.difference(from_repos))
    print("Suggesting repositories to user2")
    print(from_repos.difference(to_repos))

```

```

→ G1=nx.Graph()

→def draw_graph(user1,user2):
    G1.add_node(user1)
    G1.add_node(user2)
    repos1=G.neighbors(user1)
    repos2=G.neighbors(user2)
    for n in repos1:
        G1.add_node(n)
        G1.add_edge(user1,n)
    for n in repos2:
        G1.add_node(n)
        G1.add_edge(user2,n)
    #nx.draw(G1,with_labels=True)
    nx.draw(G1,node_color="red",with_labels=True,node_size=1000)

→user1='u21'
user2='u2509'
print("Number of Shared repositories")
print(len(shared_partition_nodes(G,user1,user2)))
print("Similarity Score")
project_nodes = get_nodes_from_partition(G, 'projects')
similarity_score = user_similarity(G,user1,user2, project_nodes)
print(similarity_score)
print("Recommending repositories")
recommend_repositories(G,user1,user2)
draw_graph(user1,user2)

```

Annexure II:

```

[4] type(G)
networkx.classes.graph.Graph

[5] len(G.nodes())
22353

[6] len(G.edges())
20052

```


References:

1. <https://towardsdatascience.com/github-recommender-system-python-c8ff64dc83f4>
2. <https://networkx.org/documentation/stable/tutorial.html>
3. https://networkx.org/documentation/latest/_modules/networkx/algorithms/operators/product.html
4. <https://campus.datacamp.com/courses/intermediate-network-analysis-in-python/bipartite-graphs-product-recommendation-systems?ex=12>
5. <https://campus.datacamp.com/courses/intermediate-network-analysis-in-python/bipartite-graphs-product-recommendation-systems?ex=9>
6. [https://towardsdatascience.com/customizing-networkx-graphs-f80b4e69bedf#:~:text=Altering%20node%20size%20globally%20is,\(\)%20method%20%E2%80%94just%20specify%20node_size!](https://towardsdatascience.com/customizing-networkx-graphs-f80b4e69bedf#:~:text=Altering%20node%20size%20globally%20is,()%20method%20%E2%80%94just%20specify%20node_size!)