

Homework 1

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Problem 1: Katz centrality

$$\mathbf{c}_{Katz} = \beta(\mathbf{I} - \alpha\mathbf{A})^{-1}\mathbf{1}.$$

α is a positive constant. When we let $\alpha \rightarrow 0$, then all the vertices have the same centrality β . As we increase α from 0, the centrality calculated will increase and then comes to a divergence point, where $(\mathbf{I} - \alpha\mathbf{A})^{-1}$ diverges. That is when

$$\det(\mathbf{I} - \alpha\mathbf{A}) = \det(\mathbf{A} - \alpha^{-1}\mathbf{I}) = 0.$$

As α increases, the determinant first crosses 0 when $\alpha = 1/k_1$, where k_1 is the largest eigenvalue of \mathbf{A} . Therefore, when α is less than $1/k_1$, the expression for Katz centrality will converge.

Problem 2: Use "walk" to compute the total number of common neighbors $|N(v_i) \cap N(v_j)|$ between nodes v_i and v_j .

$$|N(v_i) \cap N(v_j)| = N_{ij}^{(2)},$$

where $N_{ij}^{(2)}$ is the number of walks of length 2 from v_i to v_j ,

$$N_{ij}^{(2)} = \sum_{k=1}^n A_{ik}A_{kj} = [A^2]_{ij},$$

and A is the adjacency matrix.

Problem 3: See Appendix for code. The similarity plot is shown as follows:

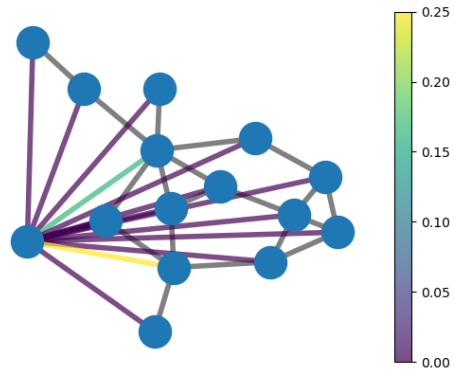


Figure 1: Jaccard's similarity between "Ginori" family and other families in the Florentine Families graph, edge colored by the corresponding similarity values.

Appendix

```
1
2 """
3 ACMS 80770-03: Deep Learning with Graphs
4 Instructor: Navid Shervani-Tabar
5 Fall 2022
6 University of Notre Dame
7
8 Homework 1: Programming assignment
9 """
10
11 from operator import le
12 from platform import node
13 import networkx as nx
14 import matplotlib.pyplot as plt
15 import numpy as np
16 from networkx.algorithms import bipartite
17 from networkx.generators.random_graphs import erdos_renyi_graph
18 import copy
19
20
21 # -- Initialize graphs
22 seed = 30
23 G = nx.florentine_families_graph()
24 nodes = G.nodes()
25 layout = nx.spring_layout(G, seed=seed)
26
27
28 # -- compute jaccard's similarity
29 """
30     This example is using NetworX's native implementation to compute similarities.
31     Write a code to compute Jaccard's similarity and replace with this function.
32 """
33 # pred = nx.jaccard_coefficient(G)
34 def my_jaccard_similarity(G):
35     nodes = list(G.nodes()) # the node names list
36     A = nx.to_numpy_array(G)
37     # matrix of total number of shared neighbors (intersection)
38     A_cap = np.matmul(A,A)
39     # matrix of total number of neighbors (union)
40     A_cup = np.zeros_like(A)
41     for i in range(len(A)):
42         for j in range(len(A)):
43             A_cup[i][j] = sum(A[i])+sum(A[j])-A_cap[i][j]
44
45     # Jaccard's similarity matrix
46     S = A_cap/A_cup
47
48     return ((nodes[i],nodes[j],S[i][j]) for i in range(len(A)) for j in range(len(A))
49 )
50 pred = my_jaccard_similarity(G)
51
```

```
52
53 # -- keep a copy of edges in the graph
54 old_edges = copy.deepcopy(G.edges())
55
56 # -- add new edges representing similarities.
57 new_edges, metric = [], []
58 for u, v, p in pred:
59     G.add_edge(u, v)
60     print(f"({u}, {v}) -> {p:.8f}")
61     new_edges.append((u, v))
62     metric.append(p)
63
64 # -- plot Florentine Families graph
65 nx.draw_networkx_nodes(G, nodelist=nodes, label=nodes, pos=layout, node_size=600)
66 nx.draw_networkx_edges(G, edgelist=old_edges, pos=layout, edge_color='gray', width=4)
67
68 # -- plot edges representing similarity
69 """
70     This example is randomly plotting similarities between 8 pairs of nodes in the
71     graph.
72     Identify the      Ginori
73 """
74 ## Identify the      Ginori
75 Ginori_edge_ls = []
76 Ginori_metric_ls = []
77 for i in range(len(new_edges)):
78     if new_edges[i][0] == 'Ginori' and new_edges[i][1] != 'Ginori':
79         Ginori_edge_ls.append(new_edges[i])
80         Ginori_metric_ls.append(metric[i])
81 ## plot
82 ne = nx.draw_networkx_edges(G, edgelist=Ginori_edge_ls, pos=layout, edge_color=np.
83     asarray(Ginori_metric_ls), width=4, alpha=0.7)
84 plt.colorbar(ne)
85 plt.axis('off')
86 plt.show()
```

Listing 1: Python code