

# Homework 1

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**Problem 1:** In order to let Katz centrality converge, the term  $(\mathbf{I} - \alpha\mathbf{A})^{-1}$  must have value, which means  $\mathbf{I} - \alpha\mathbf{A}$  is invertible. So we need  $\det(\mathbf{I} - \alpha\mathbf{A}) \neq 0$ . Note that if  $\mathbf{A}$  is a singular matrix, we have

$$\det(\mathbf{I} - \alpha\mathbf{A}) = 0 \Rightarrow \det(\mathbf{A} - \frac{1}{\alpha}\mathbf{I}) = 0 \Rightarrow \alpha = \frac{1}{\lambda}$$

where  $\lambda$  is the eigen value of matrix  $\mathbf{A}$ .  $\alpha$  takes the minimum value when  $\lambda$  is the largest eigen value. So to ensure  $\mathbf{A}$  is not singular, we should let  $\alpha$  smaller than this minimum value, in other words:

$$\alpha < \frac{1}{\lambda_1}$$

Where  $\lambda_1$  is the leading eigen value of matrix  $\mathbf{A}$ .

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**Problem 2:** Using the "walk" concept, you will start from node  $v_i$  and go two hops, then count the distinct paths that end in node  $v_j$ , which should be equal to the number of common neighbours. Mathematically: We have adjacency matrix  $\mathbf{A}$ , we want to check the common neighbourhood of node  $v_i$  and  $v_j$ . The number of common neighbour equals to the element  $\mathbf{A}_{ij}^2$  which can be calculated as:

$$|N(v_i) \cap N(v_j)| = [\mathbf{A}^2]_{ij} = \sum_k \mathbf{A}_{ik} \mathbf{A}_{kj}$$

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### Problem 3:

A. Code written in python and pushed to github

First, we calculate the Jaccard similarity of each pair of nodes.

```
# calculate the Jaccard similarity between every pair of nodes
S = np.zeros((len(nodes),len(nodes)))
for i in range(len(nodes)):
    for j in range(len(nodes)):
        NinNj = np.array(A.todense()[i,:]*np.array(A.todense()[j,:]) # calculate the
                                                                    intersection of N(vi) and N(vj)
        NiuNj = np.array(A.todense()[i,:]+np.array(A.todense()[j,:]) # calculate the
                                                                    union of N(vi) and N(vj)
        S[i,j] = len(np.nonzero(NinNj)[0])/len(np.nonzero(NiuNj)[0]) # put into S
                                                                    matrix
```

Then we create another graph where the edges will be node "Ginori" connecting to all the nodes in the graph. The value on the edges will be the corresponding similarity.

```
# creating a copy of original graph
G2 = nx.create_empty_copy(G, with_data=True)
old_edges = copy.deepcopy(G.edges())
new_edges, metric = [], []

# set new edges according to Similarity to node "Ginori"
for i in range(len(G2.nodes)):
    G2.add_edge('Ginori', list(nodes)[i])
    # print(f"({u}, {v}) -> {p:.8f}")
    new_edges.append(('Ginori', list(nodes)[i]))
    metric.append(S_dict[list(nodes)[i]])
```

B. This is the plot:

