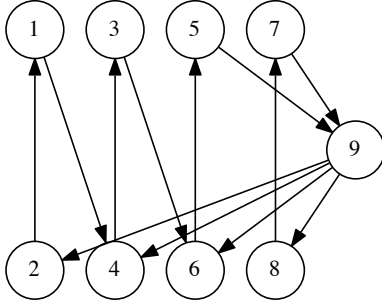


Internet Analytics (COM-308)

Homework Set 3

Exercise 1: Rewiring the Graph



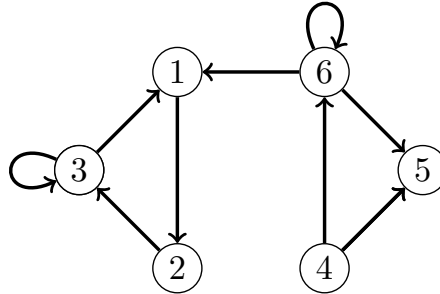
Consider the graph shown on the left, and suppose that you want to maximize the PageRank of node 9. For this, you get to change the head (to) (but not the tail (from)) of any **two** edges of the graph. Sketch the new graph, and give a short explanation about your solution (numerical PageRank values are not necessary.)

Intuitively, in order to maximize the PageRank one has to maximize the time a random walk spends on a particular node. By reducing the long cycles $9 \rightarrow 2 \rightarrow 1 \rightarrow 4 \rightarrow 3 \rightarrow 6 \rightarrow 5 \rightarrow 9$ and $9 \rightarrow 4 \rightarrow 3 \rightarrow 6 \rightarrow 5 \rightarrow 9$ into the short ones $9 \rightarrow 2 \rightarrow 9$ and $9 \rightarrow 4 \rightarrow 9$, the time spent in node 9 is maximized.

Exercise 2: Computing PageRank Scores

In this exercise, we construct the Google matrix and explore the impact of θ on the PageRank score.

(a) Compute the Google Matrix G_g of the graph given below, for $\theta = \{0.9, 1.0\}$.



Graph g

The Google matrix of the graph g , G_g , is

$$G_g = \theta \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1/2 & 0 & 1/2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1/2 & 1/2 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/3 & 0 & 0 & 0 & 1/3 & 1/3 \end{bmatrix} + (1 - \theta) \begin{bmatrix} 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \end{bmatrix} \quad (1)$$

Setting $\theta = 0.85$ we have

$$G_g = \begin{bmatrix} 0.0167 & 0.9167 & 0.0167 & 0.0167 & 0.0167 & 0.0167 \\ 0.0167 & 0.0167 & 0.9167 & 0.0167 & 0.0167 & 0.0167 \\ 0.4667 & 0.0167 & 0.4667 & 0.0167 & 0.0167 & 0.0167 \\ 0.0167 & 0.0167 & 0.0167 & 0.0167 & 0.4667 & 0.4667 \\ 0.1667 & 0.1667 & 0.1667 & 0.1667 & 0.1667 & 0.1667 \\ 0.3167 & 0.0167 & 0.0167 & 0.0167 & 0.3167 & 0.3167 \end{bmatrix} \quad (2)$$

(b) Compute the PageRank vector π of the graph g by solving the equation $x(I - \theta H) = a$. Set a to uniform distribution over all the nodes. Feel free to rely on a tool such as matlab, octave, or mathpy to solve the resulting system of equations.

What qualitative observations can you make about π for different values of θ ?

The matrix H_g for graph g is

$$H_g = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1/2 & 0 & 1/2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1/2 & 1/2 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 0 & 1/3 & 1/3 \end{bmatrix} \quad (3)$$

With $a = [1/6, 1/6, 1/6, 1/6, 1/6, 1/6]$ and $\theta = 0.9$, we have $x = a(I - \theta H)^{-1} = [1.5694, 1.5792, 2.8871, 0.1667, 0.3452, 0.3452]$. Normalizing x gives

$$\pi = \frac{x}{\|x\|_1} = [0.2276, 0.2291, 0.4188, 0.0241, 0.0500, 0.0500].$$

As we increase the value of θ toward one, the random walker will follow the graph edges with higher probability and therefore the nodes 1, 2 and 3 will get high score since they make a dangling set of nodes. Also among these three nodes node 3 will have a higher probability because it has more incoming edges compared to 1 and 2. In the extreme case if we set $\theta = 0.99$, we get the following for page rank vector:

$$\pi = [0.2474, 0.2475, 0.4905, 0.0026, 0.0058, 0.0058].$$

Conversely, if we decrease the value of θ toward zero, there is a higher chance that random walker jumps uniformly at random and hence the score of pages will be more uniform. In the extreme case if we set θ to zero we will get a completely uniform distribution:

$$\pi = [0.16666667, 0.16666667, 0.16666667, 0.16666667, 0.16666667, 0.16666667].$$