## Notes: Pagerank II

## 1 Conceptual

**Problem 1.** The purpose of the pagerank vector  $\pi$  is to provide a ranking of of how important a node is. There are many alternative ways to provide such a ranking. One simple alternative is to rank nodes by their in-degree. For "typical" graphs, the in-degree ranking and the pagerank ranking will be similar, but there are graphs for which the two rankings can be arbitrarily different from each other.

Draw a graph such that the top ranked node according to pagerank is the bottom ranked node according to in-degree.

**Problem 2.** Either prove or give a counterexample to the following claims.

1. The node with the largest out-degree can never have the highest pagerank.

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2. Two nodes in a graph cannot have the same pagerank value.

pagerank.			

3. Assume that all nodes in the graph have an in-degree of 2 and an out-degree of 2. Then all nodes will

## 2 Runtimes

**Problem 3.** In this question you will calculate the runtime of the power method for computing pagerank. Assume that P is a sparse matrix and that  $\pi$  is dense.

1. Equation 5.1 shows the power method iteration for solving for  $\pi$ . It is reproduced below

$$\mathbf{x}^{(k)T} = \alpha \mathbf{x}^{(k-1)T} \mathbf{P} + (\alpha \mathbf{x}^{(k-1)T} \mathbf{a} + (1-\alpha)) \mathbf{v}^{T}.$$
 (1)

What is the runtime of computing  $\mathbf{x}^{(k)}$  from  $\mathbf{x}^{(k-1)}$ ?

2. Given only  $\mathbf{x}^{(0)}$ , what is the runtime of computing  $\mathbf{x}^{(K)}$  by iterating Equation (1) K times?

3.	When computing pagerank, we typically do not know the final number of iterations	K	in	advance.
	Instead, we continue our computation until the following condition is met:			

until the following condition is met: 
$$\|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\|_2 \le \epsilon, \tag{2}$$

where  $\epsilon$  is a "small" number that controls how accurate we want our solution to be. The expression  $\|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\|_2$  is often called the *residual*.

Compute a formula for the number of iterations K required to achieve a residual less than  $\epsilon$ .

HINT: See the discussion on page 346.

4. Substitute your answer for part 3 into your answer for part 2 to get a formula for the overall runtime in terms of the final desired residual  $\epsilon$ .

**Problem 4.** Repeat Problem 3 assuming P is stored as a dense matrix instead of a sparse matrix.

**Problem 5.** How do the results from problems and 4 above compare to the LAPACK runtimes for computing the top-eigenvalue of a matrix?

## 3 Implementation Details

**Problem 6.** Why does it never make sense to store  $\pi$  as a sparse vector?

