

11/03/24

Big Data - CS483

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6750866741) a) Given: The web has no endsTo Prove:  $w(r') = w(r)$ .

Proof:

$$\begin{aligned}
 w(r') &= \sum_{i=1}^n r'_i \\
 &= \sum_{i=1}^n \left( \sum_{j=1}^n M_{ij} r_j \right) \\
 &= \sum_{j=1}^n \left( \sum_{i=1}^n M_{ij} r_j \right) \\
 &= \sum_{j=1}^n r_j \left( \sum_{i=1}^n M_{ij} \right) \\
 &= \sum_{j=1}^n r_j \\
 &= w(r)
 \end{aligned}$$

Since

$$\sum_{i=1}^n M_{ij} = 1$$

As, sum of  $j$ th column of Matrix  $M$  on the left side. Since the web has no ends  $r_j$  values equal to  $1/r_j$  in the  $j$ th column.

So the sum of each column is therefore equal to 1.

1) b) We teleport to a random node with probability  $1-\beta$ , where  $0 < \beta < 1$ . So,

$$\delta_i = \beta \left( \sum_{j=1}^n M_{ij} r_j \right) + \frac{1-\beta}{n}.$$

So determine when  $w(\delta) = w(r)$  is true.

$$\sum_{i=1}^n \delta_i = \sum_{i=1}^n r_i$$

$$\sum_{i=1}^n \delta_i = \sum_{i=1}^n \left( \beta \left( \sum_{j=1}^n M_{ij} r_j \right) + \frac{1-\beta}{n} \right)$$

$$\sum_{i=1}^n \delta_i = \beta \sum_{i=1}^n \left( \sum_{j=1}^n M_{ij} r_j \right) + \sum_{i=1}^n \frac{1-\beta}{n}$$

$$\sum_{i=1}^n \delta_i = \beta \sum_{j=1}^n \left( \sum_{i=1}^n M_{ij} \right) r_j + 1-\beta$$

$$\sum_{i=1}^n \delta_i = \beta \sum_{j=1}^n r_j + 1-\beta$$

$$w(r) = \beta w(r) + 1-\beta$$

$$w(r) - \beta w(r) = 1-\beta$$

$$w(r)(1-\beta) = 1-\beta$$

$$w(r) = 1$$

Thus  $w(r') = w(r)$  holds true,  
if  $w(r) = 1$

1) c) At each iteration, we teleport to nodes with probability  $1-\beta$  and from dead nodes with probability  $\beta$ . We choose randomly node uniformly to teleport to. Assume  $D$  as set of dead nodes. ③

$$r_i' = \underbrace{\beta \left( \sum_{j \notin D} M_{ij} r_j \right)}_{\textcircled{1}} + \underbrace{\frac{1-\beta}{n} \sum_{j \notin D} r_j}_{\textcircled{2}} + \underbrace{\frac{1}{n} \sum_{j \in D} r_j}_{\textcircled{3}}$$

①  $\Rightarrow$  with probability  $\beta$  a web surfer chooses an out-link on their current page

②  $\Rightarrow \frac{1-\beta}{n}$  represents that a web surfer opens a new page

③  $\Rightarrow$  teleportation from a dead node.

$$\Rightarrow w(r') = \sum_{i=1}^n \left( \beta \sum_{j \notin D} M_{ij} r_j + \frac{1-\beta}{n} \sum_{j \notin D} r_j + \frac{1}{n} \sum_{j \in D} r_j \right)$$

given  $w(r) = \sum_{j=1}^n r_j = 1$  and  $\sum_{i=1}^n M_{ij} = 1$

$$\begin{aligned} w(r') &= \sum_{i=1}^n \left( \beta \sum_{j \notin D} M_{ij} r_j + \frac{1-\beta}{n} \sum_{j \notin D} r_j + \frac{1}{n} \sum_{j \in D} r_j \right) \\ &= \beta \sum_{j \notin D} \left( \sum_{i=1}^n M_{ij} r_j \right) + \frac{1-\beta}{n} \sum_{j \notin D} r_j \left( \sum_{i=1}^n 1 \right) \\ &\quad + \frac{1}{n} \sum_{j \in D} r_j \left( \sum_{i=1}^n 1 \right) \end{aligned}$$

$$= \beta \sum_{j \notin D} r_j + (1 - \beta) \sum_{j \in D} r_j + \sum_{j \in D} r_j$$

$$= \sum_{j \notin D} r_j + \sum_{j \in D} r_j$$

$$= \sum_{j=1}^n r_j$$

$$= 1$$

2) a)

⇒ The top 5 node ids with the PageRank scores:

Node id: 263, PageRank score: 0.002020291181518219  
 Node id: 537, PageRank score: 0.00194334157145315  
 Node id: 965, PageRank score: 0.0019254478071662631  
 Node id: 243, PageRank score: 0.0018526340162417312  
 Node id: 285, PageRank score: 0.0018273721700645142

⇒ The bottom 5 node ids with the PageRank scores:

Node id: 558, PageRank score: 0.0003286018525215297  
 Node id: 93, PageRank score: 0.0003513568937516577  
 Node id: 62, PageRank score: 0.00035314810510596274  
 Node id: 424, PageRank score: 0.00035481538649301454  
 Node id: 408, PageRank score: 0.00038779848719291705



2) b) ⑤

⇒ The 5 node ids with the highest hubbiness scores:

- Node id: 840, hubbiness score: 1.0
- Node id: 155, hubbiness score: 0.9499618624906543
- Node id: 234, hubbiness score: 0.8986645288972264
- Node id: 389, hubbiness score: 0.863417110184379
- Node id: 472, hubbiness score: 0.8632841092495217

⇒ The 5 node ids with the lowest hubbiness scores:

- Node id: 23, hubbiness score: 0.042066854890936534
- Node id: 835, hubbiness score: 0.05779059354433016
- Node id: 141, hubbiness score: 0.06453117646225179
- Node id: 539, hubbiness score: 0.06602659373418492
- Node id: 889, hubbiness score: 0.07678413939216454

⇒ The 5 node ids with the highest authority scores:

- Node id: 893, hubbiness score: 1.0
- Node id: 16, hubbiness score: 0.9635572849634398
- Node id: 799, hubbiness score: 0.9510158161074016
- Node id: 146, hubbiness score: 0.9246703586198444
- Node id: 473, hubbiness score: 0.899866197360405

⇒ The 5 node ids with the lowest authority scores:

- Node id: 19, hubbiness score: 0.05608316377607618
- Node id: 135, hubbiness score: 0.06653910487622794
- Node id: 462, hubbiness score: 0.07544228624641902
- Node id: 24, hubbiness score: 0.08171239406816946
- Node id: 910, hubbiness score: 0.08571673456144878

3) a) To prove:  $C_i$  is a clique for any  $i > 1$

Proof:  $C_i$  is defined as a set of nodes of  $G$  that are divisible by  $i$ ,  $i > 0$ . So every pair of nodes has a common factor  $i$  and are connected.

This implies that there is an edge between every two nodes in  $C_i$ , so  $C_i$  is a clique.

3) b)  $C_i$  is a maximal clique if and only if  $i$  is a prime no.

If  $i$  is a prime, then all no. between 2 and 1000000 that are divisible by  $i$  are in  $C_i$ . If we add a node  $v$  to  $C_i$ , the added node  $v$  is not divisible by  $i$ , since all numbers divisible by  $i$  are already in  $C_i$ . So  $v$  will not connect to all the nodes already in  $C_i$  and  $C_i \cup \{v\}$  will not have the property of a clique. So,  $C_i$  really is a maximal clique and that condition that  $i$  is prime is sufficient.

If  $i$  is not prime, we can write it as  $i = p_1^{h_1} \dots p_{h_k}^{h_k}$ .  
 A node  $p_i$  is not in  $C_i$  since it is not divisible by  $i$ .  
 But if we add it in  $C_i$ , it will connect to all the nodes already in  $C_i$  because  $i$  and  $p_i$  have a common  $p_i$  factor, thus forming a clique.  
 So, for  $i$  not prime,  $C_i$  is not a maximal clique.

3) c)  $C_2$  is the largest clique among all the cliques of form  $C_i$ , So, cardinality of a clique in this form is

$$|C_i| = \left\lfloor \frac{1000000}{i} \right\rfloor$$

Since  $i=2$  is the smallest number,  $|C_2| = 500000$  is the highest of all  $C_i, i > 1$ .

Clique that are not form of  $C_i$  are sets of nodes that all have some common factor, but not all nodes that are divisible by this common factor are necessary in this clique. So cliques of this form are subsets of  $C_i$  for some  $i$  and therefore have smaller cardinality than  $C_2$ .

Thus  $C_2$  is the largest clique.



4) a) i) To prove:

⑧

$$|A(S)| \geq \frac{\epsilon}{1+\epsilon} |S|$$

Proof:

$$\text{Since } \overline{A(S)} = \{i \in S \mid \deg_S(i) > 2(1+\epsilon)P(S)\}$$

$$\text{So, } |\overline{A(S)}| = |S| - |A(S)| \quad \text{--- (1)}$$

W.K.T. Sum of all deg. in a graph =  $2|E(S)|$   
as every edge is counted twice

$\overline{A(S)}$  is a subgraph of  $S$ , So sum of all degrees of vertices in  $\overline{A(S)}$  is at most sum of all degrees of vertices in  $S$ .

So,

$$2|E(S)| \leq \sum_{i \in S} \deg_S(i) \leq \sum_{i \in \overline{A(S)}} \deg_S(i) \quad \text{--- (2)}$$

Now sum of all the degree of vertices in  $\overline{A(S)}$ .

$$\sum_{i \in \overline{A(S)}} \deg_S(i) > \sum_{i \in \overline{A(S)}} 2(1+\epsilon)P(S) = |\overline{A(S)}| \cdot 2(1+\epsilon)P(S)$$

from (2)

$$|\overline{A(S)}| \cdot 2(1+\epsilon)P(S) < 2|E(S)|$$



from ①

⑨

$$(|S| - |A(S)|) \cdot 2(1+\epsilon)P(S) < 2P(S)|S|$$

$$|S| - |A(S)| < \frac{|S|}{1+\epsilon}$$

$$|S| \cdot \left(1 - \frac{1}{1+\epsilon}\right) < |A(S)|$$

$$|S| \cdot \left(\frac{\epsilon}{1+\epsilon}\right) < |A(S)|$$

Thus proved.

ii) To prove:

Algorithm terminates at  $O(\log_{1+\epsilon}(u))$  iteration,  
where  $|S| = u$ .

Proof:

Let's denote  $S_i$  as a subgraph obtained  
in  $i$ th iteration. Its cardinality

$$|S_i| = |S_{i-1}| - |A(S_{i-1})| \leq |S_{i-1}| - \frac{\epsilon}{1+\epsilon} |S_{i-1}| \leq |S_{i-1}| \cdot \left(\frac{1}{1+\epsilon}\right)$$

Cardinality of  $S$  at the beginning is  $|S_0| = u$ .  
after  $k$  iteration  $S_k$  has cardinality  $S_k$   
 $S_k \leq u \cdot \left(\frac{1}{1+\epsilon}\right)^k$  we need to find highest  
 $k$ , for which  $|S_k|$   
is still nonzero.

So,

$$0 < |S_k| \leq n \cdot \left(\frac{1}{1+\epsilon}\right)^k$$

$$1 \leq n \cdot \left(\frac{1}{1+\epsilon}\right)^k$$

$$\frac{1}{n} \leq (1+\epsilon)^{-k}$$

$$\log_{1+\epsilon} \left(\frac{1}{n}\right) \leq -k$$

$$-\log_{1+\epsilon}(n) \leq -k$$

$$\log_{1+\epsilon}(n) \geq k$$

Thus it's proved that the algorithm takes at most  $\log_{1+\epsilon}(n)$  steps.

4) b) i) To prove:

~~$$P(S) \geq \frac{|S|}{1+\epsilon}$$~~

$$\deg_{S^*}(v) \geq P^*(G).$$

Proof:

$$P(S) = \frac{|E(S)|}{|S|}$$

$P(S^*)$  is the highest among all densities of subgraph,

it has to include all possible edges between nodes in  $S^*$  that are in  $G$ .

So

$$p^*(G) = p(s^*) \quad \text{--- (1)}$$

There exists a vertex  $v \in S^+$

$$\deg_{S^+}(v) < p^*(G) \quad \text{--- (2)}$$

Since  $\bar{S} = S^+ \setminus \{v\}$ , So

$$p(s^*) = \frac{E[S^+]}{|S^+|}$$

$$= \frac{E[\bar{S}] + \deg_{S^+}(v)}{|S^+|}$$

$$= \frac{|S^+| - 1}{|S^+|} \frac{E[\bar{S}]}{|S^+| - 1} + \frac{\deg_{S^+}(v)}{|S^+|}$$

$$p(s^*) = \left(1 - \frac{1}{|S^+|}\right) p(\bar{S}) + \frac{1}{|S^+|} \deg_{S^+}(v)$$

From (1) + (2) it follows, that

$$\deg_{S^+}(v) < p^*(s) = p(s^*)$$

So,

$$p(\bar{S}) > p(s^*)$$

But this is a contradiction with the fact that  $S^+$  is the densest subgraph.



ii) To prove:

$$2(1+\epsilon)P(S) \geq P^*(G)$$

Proof: Assuming there exists a node

$v \in S^* \cap A(S)$ , we can prove this

As  $v \in A(S)$ , from 4(a)

$$\deg_S(v) \leq 2(1+\epsilon)P(S).$$

As  $v \in S^*$ , from 4(b)(i)

$$P^*(G) \geq \deg_{S^*}(v)$$

From  $S^*$ , we know that  $S^* \subset S$ , so each node in  $S^*$  has smaller degree than the same node in  $S$ :  $\deg_{S^*}(v) \leq \deg_S(v)$

So,

$$2(1+\epsilon)P(S) \geq \deg_S(v) \geq \deg_{S^*}(v) \geq P^*(G)$$

(iii) To prove:

(13)

$$P(\tilde{S}) \geq \frac{1}{2(1+\epsilon)} P^*(G)$$

Proof: In every iteration we remove all the nodes from  $A(S)$  if  $P(S) > P(\tilde{S})$ . From some step forward  $P(S) \leq P(\tilde{S})$  will be true. Therefore  $\tilde{S} \leftarrow S$  will never again be executed. While  $P(S)$  will become smaller with each iteration  $P(\tilde{S})$  will stay same.

In final iteration we get  $P(S) \leq P(\tilde{S})$

From 4(b) (i)

$$P^*(G) \leq 2(1+\epsilon) P(S) \leq 2(1+\epsilon) P(\tilde{S})$$

Thus proved.

5) a)

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group 01

(10)g ... (12)g

```
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```

(2)g, 2(3)g ...

(10)g ...

(3)g(2+1) ...

Group 01



5) b).

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```
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