

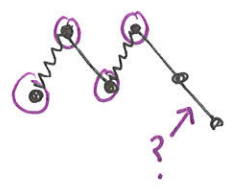
# Sublinear Time Approximation Algorithms:

## Estimating size of maximal matching in degree bounded graph

Why?

• relation to Vertex Cover

- $VC \geq MM$  ← for each edge in matching, ≥ 1 endpoints must be in VC  
these are disjoint
- $VC \leq 2MM$  ← put all MM nodes in VC  
if an edge not covered, then violates maximality



• a step towards approx maximum matching

Note: if  $deg \leq d$ , Maximal matching  $\geq \frac{n}{d}$  ← to see this, run greedy algorithm

## Greedy Sequential Matching Algorithm:

$M \leftarrow \emptyset$

$\forall e = (u, v) \in E,$

if neither  $u$  or  $v$  matched,  
add  $e$  to  $M$

Output  $M$

} output depends only on ordering of input edges

Observe:

$M$  maximal, since if  $e \notin M$  either  $u$  or  $v$  already matched earlier  
"  
 $(u, v)$

# Oracle reduction Framework

assume given deterministic "oracle"  $O(\epsilon)$   
which tells you if  $e \in M$  or not in one step

•  $S \leftarrow S = \frac{8}{\epsilon^2}$  nodes chosen iid.

•  $\forall v \in S$   

$$X_v = \begin{cases} 1 & \text{if any call to } O((v,w)) \text{ for } w \in N(v) \text{ returns "yes"} \\ 0 & \text{o.w.} \end{cases}$$

• Output  $\frac{n}{2s} \sum_{v \in S} X_v + \frac{\epsilon}{2} \cdot n$   
 Since 2 nodes matched for each edge in  $M$  (under the first term)  
 makes an underestimate unlikely (under the second term)

Behavior of output: Why does it work?

$$|M| = \frac{1}{2} \sum_{v \in V} X_v$$

$$E[|output|] = E\left[\frac{n}{2s} \sum_{v \in S} X_v\right] + \frac{\epsilon}{2} \cdot n$$

fraction of matched nodes  
↓

$$= \frac{n}{2s} \sum_{v \in S} E[X_v] + \frac{\epsilon}{2} \cdot n \quad \leftarrow \text{but } E[X_v] = \frac{2|M|}{|V|} = \frac{2|M|}{n}$$

$$= \frac{n}{2s} \cdot s \cdot \frac{2|M|}{n} + \frac{\epsilon}{2} \cdot n = |M| + \frac{\epsilon}{2} \cdot n$$

$$\Pr\left[\left|\frac{n}{2s} \sum_{v \in S} X_v + \frac{\epsilon}{2} \cdot n - E[output]\right| \geq \frac{\epsilon}{2} \cdot n\right]$$

||

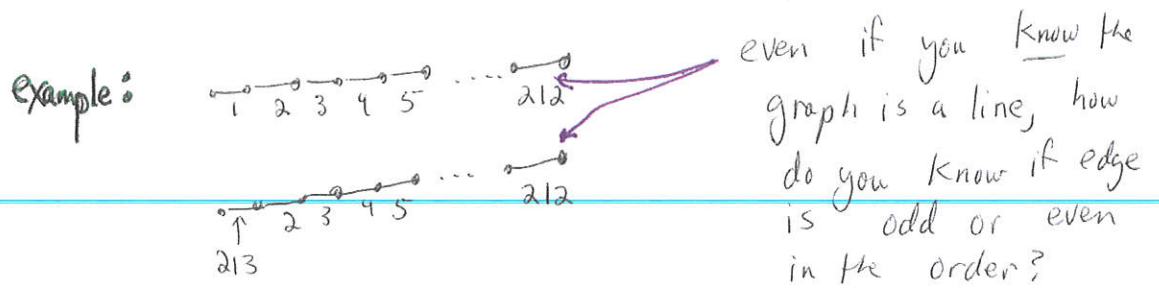
$$\Pr\left[\left|\frac{n}{2s} \sum_{v \in S} X_v - |M|\right| \geq \frac{\epsilon}{2} \cdot n\right] \leq \frac{1}{3}$$

by additive Chernoff-Hoeffding

# Implementing the oracle:

Main idea: figure out "what would greedy do on  $(v, w)$ ?"  
how? according to which input order?

Problem: Greedy is "sequential"  
Can have long dependency chains



How to implement oracle based on greedy?

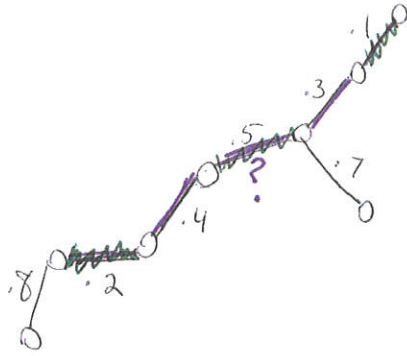
- To decide if  $e$  in matching,
  - need to know decisions for adjacent edges that came before  $e$  in ordering
  - do not need to know anything about any edge that comes after  $e$  in ordering since not considered by greedy algorithm before  $e$

so, if any adjacent edge before  $e$  in ordering matched,  
 $e$  is not matched  
 otherwise  $e$  is matched

How to break length of dependency chains?

assign random ordering to edges

example



is edge .5 in  $M$ ?

- recurse on .3

- recurse on .1

- no other adjacent edges ~~to~~

- .1 is matched

- therefore .3 is not matched

- no need to recurse on .7

- don't know yet about .5. so recurse on .4

- recurse on .2

- .8 comes after .2 in order  
so doesn't affect Greedy's  
behavior

- same for .4

- so .2 is matched

- .4 is not matched

- .5 is matched

Implementation of oracle: assume ranks  $r_e$  assign to each edge  $e$

to check if  $e \in M$ :

$\forall e'$  neighboring  $e$ ,

• if  $r_{e'} < r_e$ , recursively check  $e'$   $\dagger$

$\dagger$  if  $e' \in M$ , return " $e \notin M$ " + halt

else continue

return " $e \in M$ "

$\uparrow$  since no  $e'$  of lower rank than  $e$   
is in  $M$

Correctness: follows from correctness of greedy

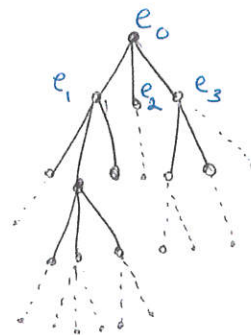
Query complexity:

Claim expected # queries to graph per  
oracle query is  $2^{O(d)}$

Claim  $\Rightarrow$  total query complexity is  $\frac{2^{O(d)}}{\epsilon^2}$

## Pf of Claim

- Consider QueryTree where root node labelled by original query edge, children of each node are edges adjacent to it.



- will only query paths that are monotone decreasing in rank

- $\Pr[\text{given path of length } k \text{ explored}] = \frac{1}{(k+1)!}$

- # edges in original graph at dist  $\leq k$  in tree  $\leq d^k$

- $E[\text{\# edges explored at dist } \leq k] \leq \frac{d^k}{(k+1)!}$

- $E[\text{total \# edges explored}] \leq \sum_{k=0}^{\infty} \frac{d^k}{(k+1)!}$   
 $\leq \frac{e^d}{d}$

- $E[\text{query complexity}] \leq d \cdot \frac{e^d}{d} = e^d = 2^{O(d)}$

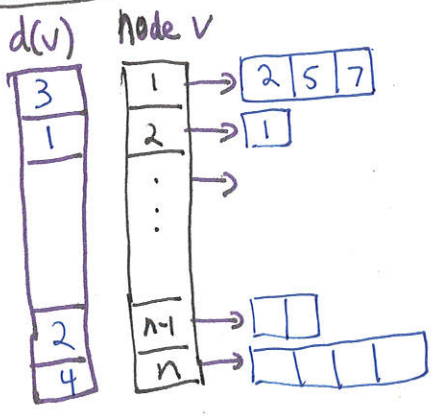


Approximating the Average degree

def Average degree  $\bar{d} = \frac{\sum_{u \in V} d(u)}{|V|}$

$G$ : simple (no parallel edges, self-loops)  
 $\Omega(n)$  edges (not "ultra-sparse")

Representation of  $G$ :



Adjacency list + degrees:

- degree queries:
- neighbor queries:

on  $v$  return  $d(v)$   
 for  $(v, j)$  return  $j$ th nbr of  $v$

Naive sampling:Pick ?? random nodes  $v_1 \dots v_s$ Output  $\frac{1}{s} \sum_i d(v_i)$  (ave degree of sample)straight forward use of Chernoff/Hoeffding needs  $\Omega(\frac{1}{\epsilon^2})$  samplesDegree sequences are special? $(n-1, 0, 0, \dots, 0)$  not possible $(n-1, 1, 1, \dots, 1)$  is possibleSome lower bounds for approximation:

"Ultra sparse" case: need linear time to get any multiplicative

approx:

graph with 0 edges  
ave deg = 0

vs.

graph with 1 edge  
ave deg =  $1/n$ requires  $\Omega(n)$   
queries to  
distinguish

queries to



Ave deg  $\geq 2$  case :

n-cycle  $\bar{d} = 2$



n -  $cn^{1/2}$  cycle +  $cn^{1/2}$ -clique  $\bar{d} \approx 2 + c^2$



need  $\Omega(n^{1/2})$  queries to find a clique node

### Algorithm

idea: group nodes of similar degrees  
estimate average w/in each group

doesn't work for estimating ave of arbitrary numbers, why should it work here?

buckets:

set  $\beta = \epsilon/c$   
 $t = O(\frac{\log n}{\epsilon})$  # buckets

$$B_i = \{v \mid (1+\beta)^{i-1} < d(v) \leq (1+\beta)^i\} \quad \text{for } i \in \{0 \dots t-1\}$$

Note that total degree of nodes in  $B_i$

$$(1+\beta)^{i-1} |B_i| \leq d_{B_i} \leq (1+\beta)^i |B_i|$$

+ total degree of graph  $\sum_i (1+\beta)^{i-1} |B_i| \leq d_{\text{total}} \leq \sum_i (1+\beta)^i |B_i|$

First idea:

• Take sample  $S$

•  $S_i \leftarrow S \cap B_i$

(samples that fell in  $i$ th bucket)

• estimate average degree of  $B_i$

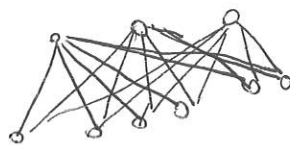
using  $S_i$   
i.e.  $p_i \leftarrow \frac{|S_i|}{|S|}$

note:  $\forall i,$   
 $E[p_i] = E\left[\frac{|S_i|}{|S|}\right] = \frac{|B_i|}{n}$

• output  $\sum_i p_i (1+\beta)^{i-1}$

Problem:  $\left. \begin{matrix} i \text{ st. } |S_i| \text{ is small} \\ \text{" " } |B_i| \text{ " " } \end{matrix} \right\}$  for these, estimate of  $p_i$  will be "off"

example



$\leftarrow$  3 nodes, degree  $n-3$

$\leftarrow$   $n-3$  nodes, degree 3

$f \leftarrow i \text{ st. } (1+\beta)^{i-1} \leq 3 \leq (1+\beta)^i$

$g \leftarrow j \text{ st. } (1+\beta)^{j-1} \leq n-3 \leq (1+\beta)^j$

$|B_f| = n-3 \leftarrow$  contributes  $m$   
 $\text{+ all other } |B_i| = 0$

$|B_g| = 3$

$\leftarrow$  will never be hit, but contributes  $m$   
is it good enough for 2-approximation?