

## Maximal independent set

trivial sequential algorithm

- inherently sequential
- from node point of view: each thinks can join MIS if others stay out
- randomization breaks this symmetry

Randomized idea

- each node joins with some probability
- all neighbors excluded
- many nodes join
- few phases needed

Algorithm:

- all degree 0 nodes join
- node  $v$  joins with probability  $1/2d(v)$
- if edge  $(u, v)$  has both ends marked, unmark lower degree vertex
- put all marked nodes in IS
- delete all neighbors

Intuition:  $d$ -regular graph

- vertex vanishes if it or neighbor gets chosen
- mark with probability  $1/2d$
- prob (no neighbor marked) is  $(1 - 1/2d)^d$ , constant
- so const prob. of neighbor of  $v$  marked—destroys  $v$
- const fraction of neighbors vanish:  $O(\log n)$  iters
- what about unmarking?
- prob(unmarking forced) only constant.
- So just changes constants

Implementing a phase trivial in  $O(\log n)$ .

Prob chosen for IS, given marked, exceeds  $1/2$

- suppose  $w$  marked. only unmarked if higher degree neighbor marked

- higher degree neighbor marked with prob.  $\leq 1/2d(w)$
- only  $d(w)$  neighbors
- prob. any marked at most  $1/2$ .
- deduce prob. good vertex killed exceeds  $(1 - e^{-1/6})/2$

Good vertices

- good: at least  $1/3$  neighbors have lower degree
- prob. no neighbor of good marked  $\leq (1 - 1/2d(v))^{d(v)/3} \leq e^{-1/6}$ .

Good edges

- any edge with a good neighbor
- has const prob. to vanish
- show half edges good
- deduce  $O(\log n)$  iterations.

Proof

- Let  $V_B$  be bad vertices; we count edges with both ends in  $V_B$ .
- direct edges from lower to higher degree  $d_i$  is indegree,  $d_o$  outdegree
- if  $v$  bad, then  $d_i(v) \leq d(v)/3$

- deduce

$$\sum_{V_B} d_i(v) \leq \frac{1}{3} \sum_{V_B} d(v) = \frac{1}{3} \sum_{V_B} (d_i(v) + d_o(v))$$

- so  $\sum_{V_B} d_i(v) \leq \frac{1}{2} \sum_{V_B} d_o(v)$
- which means indegree can only “catch” half of outdegree; other half must go to good vertices.

- more carefully,

- $d_o(v) - d_i(v) \geq \frac{1}{3}(d(v)) = \frac{1}{3}(d_o(v) + d_i(v))$ .

- Let  $V_G, V_B$  be good, bad vertices

- degree of bad vertices is

$$\begin{aligned} 2e(V_B, V_B) + e(V_B, V_G) + e(V_G, V_B) &= \sum_{v \in V_B} d_o(v) + d_i(v) \\ &\leq 3 \sum (d_o(v) - d_i(v)) \\ &= 3(e(V_B, V_G) - e(V_G, V_B)) \\ &\leq 3(e(V_B, V_G) + e(V_G, V_B)) \end{aligned}$$

Deduce  $e(V_B, V_B) \leq e(V_B, V_G) + e(V_G, V_B)$ . result follows.

Derandomization:

- Analysis focuses on edges,
- so unsurprisingly, pairwise independence sufficient
- not immediately obvious, but again consider  $d$ -uniform case
- prob vertex marked  $1/2d$
- neighbors  $1, \dots, d$  in increasing degree order
- Let  $E_i$  be event that  $i$  is marked.
- Let  $E'_i$  be  $E_i$  but no  $E_j$  for  $j < i$
- $A_i$  event no neighbor of  $i$  chosen
- Then prob eliminate  $v$  at least

$$\begin{aligned} \sum \Pr[E'_i \cap A_i] &= \sum \Pr[E'_i] \Pr[A_i | E'_i] \\ &\geq \sum \Pr[E'_i] \Pr[A_i] \end{aligned}$$

- Wait: show  $\Pr[A_i | E'_i] \geq \Pr[A_i]$ 
  - true if independent
  - measure  $\Pr[\neg A_i | E'_i] \leq \sum \Pr[E_w | E'_i]$
  - measure

$$\begin{aligned} \Pr[E_w | E'_i] &= \frac{\Pr[E_w \cap E'_i]}{\Pr[E'_i]} \\ &= \frac{\Pr[E_w \cap \neg E_1 \cap \dots | E_i]}{\Pr[\neg E_1 \cap \dots | E_i]} \\ &\leq \frac{\Pr[E_w | E_j]}{1 - \sum \Pr[E_j | E_i]} \\ &= \Theta(\Pr[E_i]) \end{aligned}$$

- But expected marked neighbors  $1/2$ , so by Markov  $\Pr[A_i] > 1/2$
- so prob eliminate  $v$  exceeds  $\sum \Pr[E'_i] = \Pr[\cup E_i]$
- lower bound as  $\sum \Pr[E_i] - \sum \Pr[E_i \cap E_j] = 1/2 - d(d-1)/8d^2 > 1/4$
- so  $1/2d$  prob.  $v$  marked but no neighbor marked, so  $v$  chosen
- Generate pairwise independent with  $O(\log n)$  bits
- try all polynomial seeds in parallel

- one works
- gives deterministic  $NC$  algorithm

with care,  $O(m)$  processors and  $O(\log n)$  time (randomized)  
LFMIS P-complete.

## Perfect Matching

We focus on bipartite; book does general case.  
Detecting one easy in  $NC$ :

- Tutte matrix
- Determinant nonzero iff PM
- Replace vars with values  $1, \dots, 2^m$ , same holds
- Matrixu Mul, Determinant in  $NC$
- Wait: big numbers?
- Who cares: poly bits,  $NC$  to multiply etc

How about finding one?

- If unique, no problem
- Remove each edge, see if still PM in parallel
- multiplies processors by  $m$
- still  $NC$
- generalizes to polynomial number of matchings

Idea:

- make unique minimum weight perfect matching
- find it

Isolating lemma:

- Family of distinct sets over  $x_1, \dots, x_m$
- assign random weights in  $1, \dots, 2m$
- $\Pr(\text{unique min-weight set}) \geq 1/2$
- Odd: no dependence on number of sets!
- (of course  $< 2^m$ )

Proof:

- Fix item  $x_i$
- $Y$  is min-sets containing  $x_i$
- $N$  is min-sets no containing  $x_i$
- true min-sets are either those in  $Y$  or in  $N$
- how decide? Value of  $x_i$
- For  $x_i = -\infty$ , min-sets are  $Y$
- For  $x_i = +\infty$ , min-sets are  $N$
- As increase from  $-\infty$  to  $\infty$ , single transition value when both  $X$  and  $Y$  are min-weight
- If only  $Y$  min-weight, then  $x_i$  in every min-set
- If only  $X$  min-weight, then  $x_i$  in no min-set
- If both min-weight,  $x_i$  is *ambiguous*
- Suppose no  $x_i$  ambiguous. Then min-weight set unique!
- Exactly one value for  $x_i$  makes it ambiguous given remainder
- So  $\text{pr}(\text{ambiguous}) = 1/2m$
- So  $\text{pr}(\text{any ambiguous}) < m/2m = 1/2$

Usage:

- Consider tutte matrix  $A$
- Assign random value  $2^{w_i}$  to  $x_i$ , with  $w_i \in 1, \dots, 2m$
- Weight of matching is  $2^{\sum w_i}$
- Let  $W$  be minimum sum
- Unique w/pr  $1/2$
- If so, determinant is odd multiple of  $2^W$
- Try removing edges one at a time
- Edge in PM iff new determinant/ $2^W$  is odd.

$NC$  algorithm open.

For exact matching,  $P$  algorithm open.

## Upcoming

Vempala: “An Eye for Elegance”

- More markov chains
- convex volume estimation
- geometric embeddings
- 11-12:30

Joel Spencer

- 9:30-11
- Probabilistic method
- List of people who took it last time

Spielman advanced complexity  
Next year: advanced algorithms.  
Bring your research problems