

Advanced Algorithm

- sunyajith chillawa

Lecture #1

Assignments 20%

Quiz 1 10%

Mid term 20%

Quiz 2 10%

End sem 30%

Viva 10%

Book: Probability and Computing Mitzenmacher and Upfal

Verify Matrix Multiplication

Given A, B and C all $(n \times n)$ mat.

we need to check if $AB = C_{n \times n}$

→ Multiply A and B and check if

the elements are equal complexity: n^3 ^{mat. mult exp.}
↳ by Vassilis, Almanas - Williams

so far, only deterministic algorithm

- ↳ worst case
- ↳ "fixed" run time
- ↳ No Error (always correct)
- ↓
we want to reduce this
and make bounded run time for
most of the time

Randomised Routing

Randomised algorithm : Primality testing , Robin - Miller

Trend : Hard problem \rightarrow Randomization \rightarrow Approximation.

S^{n^2} [Blasius] \rightarrow Lower Bound on matrix multiplication.

Updated Question: Is there a rand. algorithm that runs in time $O(n^\delta)$ where $\delta < \omega$. (We want a mostly correct.)

If $AB = C$, then \forall vectors v , $ABv = Cv$

else $\exists u$ s.t. $AB \cdot u \neq C \cdot u$ |. u.e. Nullspace of $AB - C$

$$AB\bar{x} = C\bar{x}$$

$K \times L$, $L \times S \rightarrow$ KLS complementarity.

$$\underbrace{AB\bar{x}}_{O(n^2)} \text{ and so } \underbrace{C\bar{x}}_{\uparrow} \Rightarrow \begin{bmatrix} L_1(\bar{x}) \\ \vdots \\ L_n(\bar{x}) \end{bmatrix} = \begin{bmatrix} L'_1(\bar{x}) \\ L'_2(\bar{x}) \\ \vdots \\ L'_n(\bar{x}) \end{bmatrix}$$

$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \Rightarrow \begin{bmatrix} L_i(\bar{x}) - L'_i(\bar{x}) \\ L_n(\bar{x}) - L'_n(\bar{x}) \end{bmatrix}$$

If $AB = C$ then $\forall i \in [1, n]$, $L_i(\bar{x}) - L'_i(\bar{x}) = 0$ for every substitution of \bar{x} from S^n .

If not, \exists a subset for \bar{x} s.t. not all $L_i(\bar{x}) - L'_i(\bar{x})$ are zero at that point.

$$\Pr_{\bar{a} \in S^n} \left[\bigwedge_{i=1}^n (L_i(\bar{a}) - L'_i(\bar{a}) = 0) \right] \leq \max_{\{i \in [n]\}} \Pr \left[L_i(\bar{a}) - L'_i(\bar{a}) = 0 \right]$$

where $\bar{a} \in \text{Nullspace of } A\bar{b} - c$

Bad event: $\forall i, L_i(\bar{a}) \neq L'_i(\bar{a})$

Good Event: $\exists i \in [n] \text{ s.t. } L_i(\bar{a}) \neq L'_i(\bar{a}) \rightarrow \text{prob. of } \bar{a} \text{ being in null space.}$

$$\Pr_{\substack{\bar{a} \in S^n \\ \bar{a} \in \text{Nullspace of } A\bar{b} - c}} \left[\sum_{j=1}^n x_j \cdot a_j = 0 \right] = \frac{|S|^{n-1}}{|S|^n}, \text{ fix } n-1, \text{ nth is fixed} \\ \leq \frac{1}{|S|}$$

Algo:

- 1. Pick \bar{a} uniformly random from S^n
- 2. Compute $A\bar{a}$ and $C\bar{a}$
- 3. Check if pointwise equal.
- 4. If yes, return $A\bar{b} = C \rightarrow$ 100% correct
else return $A\bar{b} \neq C \rightarrow$ correct with prob. $\geq 1 - \frac{1}{|S|}$

Monte Carlo algorithms: correctness not fixed, time fixed

Las Vegas: correctness 1, time unbounded.

Problem: (Polynomial Identity Testing)

Given

$$f(x_1, x_2, \dots, x_n), \quad \{ \text{of degree d} \\ g(x_1, x_2, \dots, x_n)$$

$$x_1^{e_1} x_2^{e_2} \dots x_n^{e_n}$$

$$\sum_{i=1}^n e_i \leq d \rightarrow \text{no. of coeff. } \binom{n+d}{d}$$

if $f(a) = g(a)$ for all points then $f(x) = g(x)$

if $f(x) \neq g(x)$ then $\exists b$ in all distinct evals
 $f(b) \neq g(b)$

Lecture #2

Graph Matching - bipartite

- Matching: M is a subset of edges s.t. for every vertex $v \in V(G)$, $|I(v) \cap M| \leq 1$ where $I(v)$ is the set of edges incident on v .
- Perfect Matching: $\forall v, |I(v) \cap M| = 1$

Edmonds's criterion:

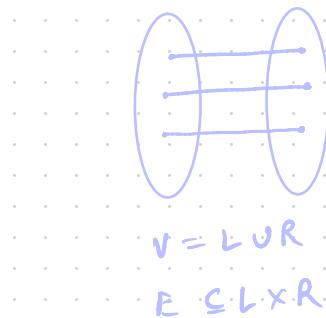
$$(M_G)_{n \times n} |_{ij} = \begin{cases} x_{ij} & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Theorem:

A given bipartite graph has perfect matching if and only if $\det(M_G) \neq 0$.
 (P.M.)

Determinant

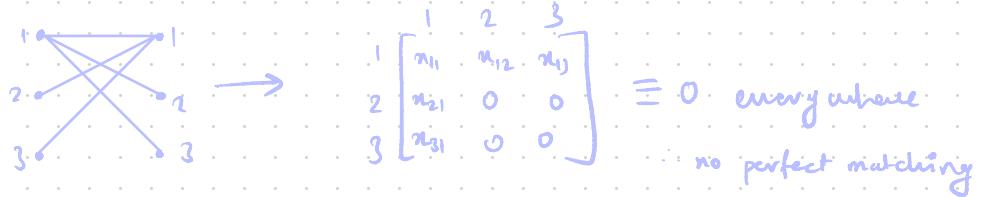
$$\begin{bmatrix} x_{11} & x_{12} \\ x_{21} & x_{22} \end{bmatrix} \rightarrow x_{11}x_{22} - x_{12}x_{21}$$



Homework: Prove this

$$\det(X) = \sum_{\sigma \in S_n} (-1)^{\text{inv}(\sigma)} \prod_{i=1}^n x_{i\sigma(i)}$$

$$\# \text{inv}(\sigma) = |\{(i, j) \mid i < j, \sigma(i) > \sigma(j)\}|$$



$$\text{PM} \Rightarrow \det(M_A) \neq 0 \quad | \quad \det(M_A) \neq 0 \Rightarrow \text{PM}$$

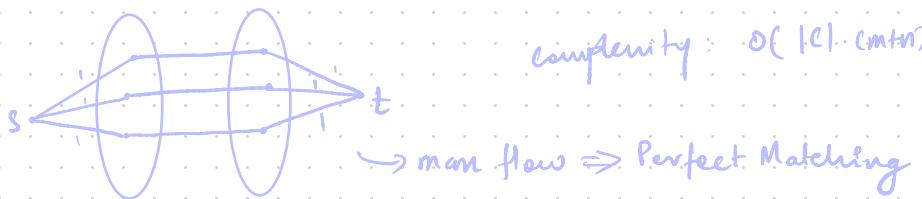
observation: Perfect Matching is indeed a permutation.

Further $\prod_{i \in \sigma(i)} a_{i,i}$ survive iff all $(i, \sigma(i)) \in E$.

Sufficient: To find a point where $\det(M_A) = 0$

σ^* : $\forall i$, set $a_{i, \sigma^*(i)} = 1$

Existence criteria, not how to find not discussed yet.



Problem: sequential

Want a parallel algorithm

Matching in RNC^2

claim: For all $e \in E$: check if $G/\{e\}$ has a PM using EC.

$G \leftarrow G/\{e\}$ if Yes

$M \leftarrow M \cup \{e\}$ if No.

Different Ordering will give different perfect Matching

Non-zeroes of a polynomial

Let $f(n_1, n_2, \dots, n_n)$ be a non-zero polynomial of degree d .

Let $S \subseteq R$ of size $> d+1$

Let (a_1, a_2, \dots, a_n) be chosen s.t. each a_i is picked uniformly at random.

$$\Pr_{\bar{a} \in S^n}[f(a_1, a_2, \dots, a_n)] \leq \frac{d}{|S|}$$

zeros(f) $\leq d \cdot |S|^{n-1}$ over S^n .

Total points = $|S|^n$.

Can be proved using induction. \leftarrow Homework

Algorithm for testing $\text{Det}(M_G) \neq 0$

1. Pick a sets of size $100n$.

2. Sample \bar{a} independently and uniformly at random from $S^{|E|}$

$S^{|E|}$

3. Check if $\text{Det}(M_G(\bar{a})) \stackrel{?}{=} 0$

4. If zero \Rightarrow report $\text{Det}(M_G) = 0$

Else report $\text{Det}(M_G) \neq 0$

If $\text{Det}(M_{G_i}) \equiv 0$ then algorithm makes no error

If $\text{Det}(M_{G_i}) \not\equiv 0$, then the probability of it being correct
is $1 - \frac{1}{100}$

$$\text{Randidness} = n \cdot \log |S|$$

$$\text{Complexity} = O(|E| \log(100n))$$

food for thought for next class:

Having $O(\log(n))$ Randidness = Having no randomness at all

→ Area called pseudo randomness

Here n is the number of random bits sampled

Lecture #3

Graph Matching (bipartite)

There is a random algorithm that finds a ^{perfect} matching in $O(\log^2 n)$ parallel time using $O(m \cdot \text{poly}(n))$ processors.

Outline: Assign wts to edges randomly and ensure and extract unique min. wt. matching

Isolation Lemma: Let S be a finite subset of \mathbb{R} .

Let $T_1, \dots, T_k \subseteq [m]$ ($[m] = \{1, 2, \dots, m\}$) be all distinct sets

For each element $i \in [m]$, let wt_i assign a wt
ind. and u.a.r. from S . wt of a set T_i is given by

$$\sum_{e \in T_i} \text{wt}(e)$$

$$\Pr[\text{Imin.wt.set} = i] \geq 1 - \frac{m}{|S|}$$

Proof: Let E_i be the event that minimum is obtained by two sets U and V and $i \in U$ but not V .

Obs: If min.wt. is not unique then $\exists i, s.t. T_i$ is true

$$\min_{T_j : i \in T_j} \{\text{wt}(T_j)\} = \min_{T_j : i \notin T_j} \{\text{wt}(T_j)\}$$

bad event: $\{\exists i \text{ s.t. } E_i\}$

Good event:

$$\Pr(\bigcap_{i=1}^m \overline{E_i}) = 1 - \Pr(\cup E_i) \geq 1 - m \cdot \max_i \{\Pr(E_i)\}$$

$$\Pr(E_i) \leq \frac{1}{|S|}$$

$$\therefore \Pr(\bigcap_{i=1}^m \overline{E_i}) \geq 1 - \frac{m}{|S|}$$

1. For each edge $e \in E$, assign a random weight ind. U.A.R. frame.

2. Define matrix W s.t.

$$W_{ij} = \begin{cases} 2^{\text{wt}(L(i,j))} & \text{if } (i,j) \in E, \\ 0 & \text{otherwise} \end{cases}$$

Lemma: let M_0 be the unique min int PM in G .

let $r = \text{wt}(M_0)$. Then

$$(i,j) \in M_0 \iff \frac{\text{Det}(w^{(i,j)})}{2^r} \text{ is odd.}$$

3. For each edge $(i,j) \in E$, compute $\frac{\text{Det}(w^{(i,j)})}{2^{\text{wt}(L(i,j))}}$

(in parallel)

$$\text{Det}(W) = \sum_{\sigma \in S_n} (-1)^{\text{sgn}(\sigma)} 2^{\sum_{i=1}^n \text{wt}(i, \sigma(i))}$$
$$\sigma \in \text{PM}(G)$$

M_0, M_1, \dots, M_K

$$\det(w) = \sum_{\sigma \in S_n} (-1)^{\operatorname{sgn}(\sigma)} \prod_{i=1}^n w_{i-\sigma(i)}$$

$$\det(w) = \operatorname{sgn}(M_0) \cdot 2^{\operatorname{wt}(M_0)} + \operatorname{sgn}(M_1) \cdot 2^{\operatorname{wt}(M_1)} + \dots + \operatorname{sgn}(M_k) 2^{\operatorname{wt}(M_k)}$$

$\forall i \neq \operatorname{wt}(M_i)$

Lecture #4

$$w_{ij} = \begin{cases} 2^{w(i,j)} & \text{if } (i,j) \in E \\ 0 & \text{otherwise} \end{cases}$$

wt: $E \rightarrow S$
u.a.r. + Ind.

conditioned on

1. PM existence check (Edmonds's criteria)
2. Unique min wt guarantee (Isolation lemma)

Claim: $\det_{\substack{w \in E \\ w \neq 0}}(W)$ is odd

Lemma: $M_0 \leftarrow \min \text{wt PM}, r \leftarrow \text{wt}(M_0)$

$(i,j) \in M_0$ if and only if $\det_{\substack{w \in E \\ w \neq 0}}(w(i,j))$

Lecture #5

Randomised Routing (Parallel)

Hypercube: $G = \{\{0,1\}^n, \{(u, u \oplus e_i) | u \in \{0,1\}^n, i \in [n]\}\}$

$\xrightarrow{\text{xOR}}$
 $\hookrightarrow (0,0, \dots, 1, \dots, 0) \rightarrow \text{elementary vector}$

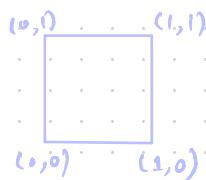
$$\underline{n=2}$$

$$\{0,0\}$$

$$\{0,1\}$$

$$\{1,0\}$$

$$\{1,1\}$$



We want to route packets on this graph from one vertex to other

→ Permutation Routing (oblivious/independent Routing)

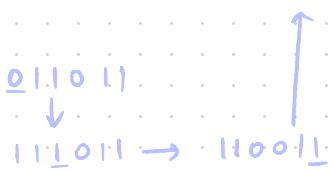
Bit fixing scheme

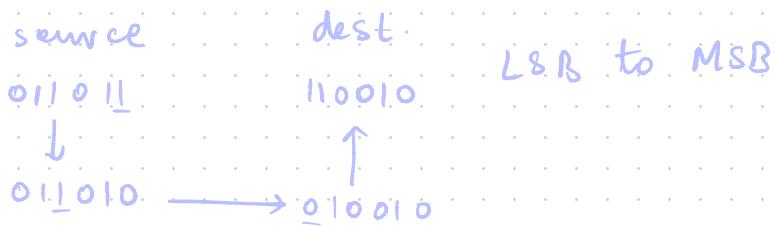
source

011011

destination

110010 MSB to LSB





Task: Route all packets from every source to every dest as early as possible

There could be waiting at the nodes

Every node can hold unlimited packets and each edge carries at most one packet.

Lemma: There are permutations $\Pi: \{0,1\}^n \rightarrow \{0,1\}^n$ for which bit fixing scheme takes $\frac{d^{1/2}}{2}$ time steps to route all the packets.

Possible : Hamming distance = N

$$(a_1, a_2, \dots, a_n) \mapsto (\bar{a}_1, \dots, \bar{a}_n)$$

$$(x_{n/2}, o_{n/2}) \rightarrow (o_{n/2}, x_{n/2}) \vee x_{n/2} \in \{0, 1\}^{n/2} \left\{ \begin{array}{l} 2^{n/2} \text{ many} \\ \text{out of } 2^n \end{array} \right.$$

Claim: Each of these packets need to go through $(0, \frac{1}{n}, 0)$

At each time step at most n packets can be wanted from any node.

Time to route all packets $(x_{n_1}, o_{n_1}) \rightarrow (o_{n_2}, x_{n_2})$ is at least

$$\frac{2^{n/2} - 1}{n}$$

Theorem: For every deterministic protocol, \exists a permutation that needs at least $\sqrt{\frac{2^n}{n}}$ time.

Randomization: For every permutation, all packets can be routed in at most $O(n)$ time steps with high probability.

Lecture # 6

Theorem: There is a randomised protocol that routes all packets to their dest. in at most $15n$ times w.h.p.



claim: $\exists a \in [n]$ s.t.

$$u_i = w_i \quad \forall i > a$$

$$v_i = w_i \quad \forall i \leq a$$

$$v_1, v_2, v_3, \dots, v_n \leftarrow \text{src} = u_1, u_2, \dots, u_n$$

$$v_1, v_2, v_3, \dots, v_n$$

$$v_1, v_2, \dots, v_{n-1}, v_n \rightarrow \text{dest}: v_1, v_2, \dots, v_n$$

$\text{Path}(v) = \text{Path}$ from $v \rightarrow \pi(v)$ using BFS

Lemma 1: For $v = v'$, $\text{Path}(v)$ and $\text{Path}(v')$ do not meet again once they diverge.

Lemma 2: Let S be the set of vertices w s.t. $\text{Path}(w)$ and $\text{Path}(v)$ intersect. Then P_v takes at most $|n + |S||$ time steps to reach $\pi(v)$.

Lemma 3: # of w.s.t. Path(w) intersects with path $v \rightarrow \sigma(v)$
is at most $6n$ w.p. $\geq 1 - 2^{-n}$
similarly, $\sigma(w) \rightarrow \pi(v)$ w.p. $\geq 1 - 2^{-n}$ intersections $\leq 6n$

$$x \begin{cases} \rightarrow y_v = x \oplus e_a \\ \rightarrow y_w = x \oplus e_b \end{cases}$$

Lecture # 7

Chernoff bound:

For some $t \in \mathbb{R}_{\geq 0}$,

$$\begin{aligned} \Pr[X > (1+\delta)ut] &= \Pr[Xt > (1+\delta)ut] \\ &= \Pr[e^{xt} > e^{(1+\delta)ut}] \\ &\leq \frac{\mathbb{E}[e^{xt}]}{e^{(1+\delta)ut}} \end{aligned}$$

$$\mathbb{E}[e^{xt}] = \mathbb{E}\left[e^{t\sum_{i=1}^n X_i}\right] = \prod_{i=1}^n \mathbb{E}[e^{tx_i}]$$

$$\begin{aligned} \mathbb{E}[e^{tx_i}] &= e^{t+1} \cdot p_i + (1-p_i) e^{t+0} \\ &= p_i(e^{t+1}) + 1 \end{aligned}$$

$$\prod_{i=1}^n p_i(e^{t+1}) + 1 \leq \prod_{i=1}^n e^{p_i(e^{t+1})} = e^{\sum p_i(e^{t+1})} = e^{ut(e^{t+1})}$$

$e^{t+1} = 1 + \delta$

$$\begin{aligned} \mathbb{E}[X > (1+\delta)ut] &\leq \left[\frac{e^{(e^{t+1})}}{e^{(1+\delta)t}} \right]^u \\ &= \left[e^{(e^{t+1}) - (1+\delta)t} \right]^u \xrightarrow{\substack{\text{minimize derivative} \\ \text{to get rid of } t}} \\ &\geq \left[\frac{e^\delta}{(1+\delta)^{\frac{1}{1+\delta}}} \right]^u \text{ set } e^{t+1} = 1 + \delta \end{aligned}$$

Homework: Chernoff bound for error reductions

$$x_i = \begin{cases} 1 & 1/2 \text{ prob.} \\ -1 & 1/2 \text{ prob.} \end{cases}$$

$$X = \sum x_i$$

$$\Pr[X > R]$$

$$T_1 \quad T_2$$

$$P_{R_1} = ? \quad P_{R_2} \rightarrow P_{IT}$$

$$\begin{matrix} T_1 \\ 0 \\ x_0 \end{matrix} \quad \begin{matrix} T_2 \\ 0 \\ x_0 \end{matrix}$$

h

$h+1$



$$\prod_{i=1}^k (x_{u_{i+1}} - P_{n_i})$$

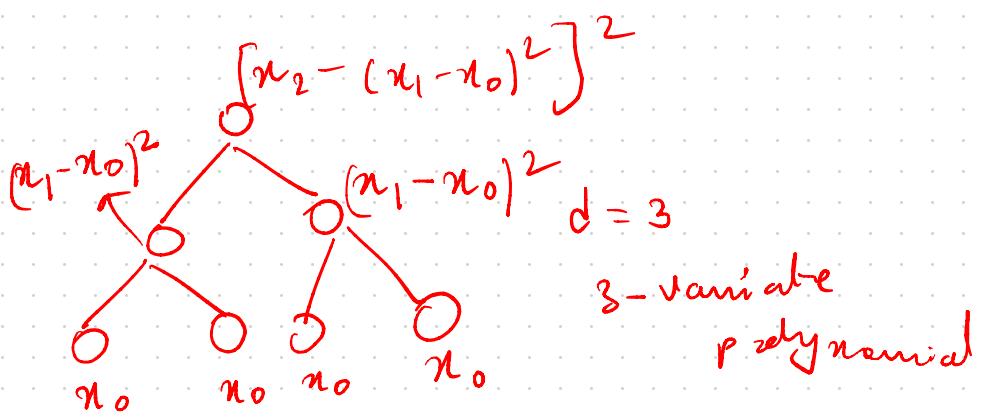
$$\prod_{i=1}^{h+1} (x_{u_{i+1}} - P'_{n_i})$$

$h+1 - \text{vars}$

$$x_0 - x_n \prod_{i=1}^k (\alpha_{h+1} - c_i)$$

$$\prod_{i=1}^k (x_{u_{i+1}} - c_i)$$

$$P[P_{R_1} - P_{R_2} = 0] \leq \frac{s}{|S|}$$



$$x_2^2 - 2x_2(x_1 - n_0)^2 + (x_1 - n_0)^4$$

$$x_2^2 - 2x_2(x_1^2 - 2x_1n_0 + n_0^2) + x_1^4 - 4x_1^3n_0 + 6x_1^2n_0^2 - 4x_1n_0^3 + n_0^4$$

$$x_2^2 - 2x_1^2x_2 + 4x_0x_1x_2 - 2x_0^2x_2 + x_1^2 - 4x_0x_1^3 + 6x_0^2x_1^2 - 4x_0^3x_1 + x_0^4$$

Lecture #8

Satisfiability

Is there an assignment to x_1, \dots, x_n from $\{T, F\}^n$ st. the formulae evaluates to true. (Decision Problem)

Its called SAT

SAT is NP-complete (Cook-Levin Theorem)

ManSAT

Conjunctive Normal Forms: AND of ORs

$$\hookrightarrow c_1 \wedge c_2 \wedge c_3 \dots c_m$$

each c_i is or of literals

$$c_i = (x_{i1} \vee x_{i2} \dots x_{iK})$$

ManSAT: Given a CNF, what is the max. no. of clauses that can be satisfied

\hookrightarrow (optimization problem)

"Solving ManSAT solving SAT"

\hookrightarrow "ManSAT is at least as hard as SAT"

$$z_j = \begin{cases} 1 & \text{if clause } c_j \text{ is sat} \\ 0 & \text{o/w} \end{cases}$$

$$y_j = \begin{cases} 1 & \text{if } x_j \leftarrow T \\ 0 & \text{o/w} \end{cases}$$



$z_j \in \{T, F\}$ w.r.t., independently

Assume each clause has exactly k literals

$$(\quad) \wedge (\quad) \wedge (\quad) \wedge (\quad)$$

$$m(1 - \frac{1}{2^k}) \geq \frac{m}{2}$$

$$\Pr(\sum z_j < \frac{m}{2}) \leq$$

Approximation algorithms

Vertex Cover: subset of V s.t. each edge gets covered

Independent set: subset of vertex s.t. no two vertex have an edge between them

performance ratio: $\frac{\inf_{\mathcal{I}} \frac{\text{max}(I)}{\text{max}_{\text{opt}}(I)}}{\min_{\mathcal{I}} \frac{\text{max}(I)}{\text{max}_{\text{opt}}(I)}}$

Random Assignment gives a $\frac{1}{2}$ -approx. algorithm for SAT

w.p. 8

→ Good event happens w/ non-zero probability.

Lecture #8

K-CNF: $c_1 \wedge c_2 \dots \wedge c_m$

$$C_i = x_{i_1} \vee x_{i_2} \dots \vee x_{i_k}$$

$$u_1, \dots, u_n \in \{T, f\}$$

$$\frac{3m}{4}$$

$$E(\# \text{ clauses sat}) =$$

$$(\pi_1 \vee \pi_2) \wedge (\bar{\pi}_1 \vee \bar{\pi}_2)$$

$$\frac{2m}{3} \quad \frac{2m}{3} \quad \frac{m}{2} \quad \frac{m}{2}$$

→ Method of conditional expectation

$$\frac{m}{4}$$

$$x_1 \cdot x_2 \cdots x_{10}$$

$$m \geq m_* \geq m_{\frac{1}{2}}$$

can find an assignment that sat $\geq \frac{m}{2}$ clauses

\Rightarrow half approximation algorithm.

$$\left(1 - \frac{1}{2^k}\right) \left(1 - \frac{1}{2^k}\right) \dots \left(1 - \frac{1}{2^k}\right) \dots \left(1 - \frac{1}{2^k}\right) \dots \left(1 - \frac{1}{2^k}\right)$$

$\left\{ \frac{n-k}{2^n} \right\}$

$$m\left(1 - \frac{1}{2^k}\right)$$

What is C_j^+ & C_j^- ?

$$\frac{C_1 \wedge C_2 \wedge \dots \wedge C_m}{\mathbb{E}[Z_1]} = P_r[Z_1 = 1] = 1 - \frac{1}{2^k} \geq \frac{1}{2}$$

Randomised Rounding

$$\mathbb{E}[\# \text{ clauses sat} \mid x_i = T]$$

$$y_i = \begin{cases} 1 & \text{w.p. } \hat{q}_i \\ 0 & \text{o/w} \end{cases}$$

$$= \mathbb{E}\left[\sum_j z_j \mid x_i = T\right]$$

$$= \sum_j \mathbb{E}[z_j \mid x_i = T]$$

$$= \sum_j \Pr[z_j = 1 \mid x_i = T]$$

=

$$\mathbb{E}\left[\sum_{j=1}^m z_j\right] = \sum_j \mathbb{E}[z_j] = \sum_{j=1}^m \left(1 - \prod_{i \in C_j^-} \hat{q}_i \prod_{i \in C_j^+} (1 - \hat{q}_i)\right)$$

Lecture #9

$$E\left[\sum_{j=1}^m z_j\right] = \sum E[z_j] = \sum_{j=1}^m \left(1 - \prod_{i \in C_j^+} \tilde{y}_i \prod_{i \in C_j^-} (1 - \tilde{y}_i)\right)$$

Finally, $\tilde{y}_i \in [0,1]$

$\tilde{y}_1, \tilde{y}_2, \dots, \tilde{y}_n, \tilde{z}_1, \dots, \tilde{z}_m$ be solutions of R-LP

Randomized Rounding

$y_i \leftarrow \begin{cases} 1 & \text{w.p. } \tilde{y}_i \\ 0 & \text{o.w.} \end{cases}$

$1 - \left(1 - \frac{\tilde{z}_j}{k}\right)^k \geq 2 \left[1 - \left(1 - \frac{1}{k}\right)^k\right]$

Claim: If clauses sat in F $\leq \frac{m}{k}$

$E[\#\text{of clauses sat}]$

$$\begin{aligned} &= \sum_{j=1}^m E[\text{clause } j \text{ is sat.}] \\ &= \sum_{j=1}^m \Pr[\text{clause } j \text{ is sat.}] \\ &= \sum_{j=1}^m 1 - \left(\prod_{i \in C_j^+} (1 - \tilde{y}_i) \cdot \prod_{i \in C_j^-} \tilde{y}_i\right) \end{aligned}$$

$\prod_{i=1}^k (1 - \tilde{y}_i) \leq \left[\frac{k}{m} \left(1 - \frac{\tilde{z}_j}{k}\right)\right]^k$

$\left[\frac{\sum_{i=1}^k \tilde{y}_i}{m} \geq \tilde{z}_j\right] = \left[k - \left(\frac{\sum_{i=1}^k \tilde{y}_i}{m}\right)\right]^k$

$\leq \left(1 - \frac{\tilde{z}_j}{k}\right)^k$

$> 1 - \left(1 - \frac{\tilde{z}_j}{k}\right)^k \geq \left[1 - \left(1 - \frac{1}{k}\right)^k\right] \tilde{z}_j$

Prob of sat clause \tilde{z}_j

$$1 - \left(1 - \frac{\tilde{z}_j}{k}\right)^k = \left[1 - \left(1 - \frac{1}{k}\right)^k\right] \tilde{z}_j \quad z_j \in [0,1]$$

$$E(\#\text{ clauses sat.}) \geq (1 - c^{-1}) \sum_{j=1}^m \tilde{z}_j$$

Lecture #10

Minimum Cut

Glarry Johnson

Find s^* such that s^* minimizes $\max_{S \subseteq V} |\text{cut}(s, \bar{s})|$

Problem is NP-hard.

Attempt - 1:

For each vertex v put v in s w.p. $1/2$

$$x_i = \begin{cases} 1 & \text{if } e_i \in E(s, \bar{s}) \\ 0 & \text{otherwise} \end{cases}, e_1, e_2, \dots, e_m$$

$$\begin{aligned} E[|\text{cut}(s, \bar{s})|] &= \sum_{i=1}^m E(x_i) \\ &= \sum_{i=1}^m \frac{1}{2} \times \frac{1}{2} + \frac{1}{2} \times \frac{1}{2} \\ &= \sum_{i=1}^m \frac{1}{2} = \frac{m}{2} \end{aligned}$$

Lecture #11

$$\max \sum_{(i,j) \in E} \left(\frac{x_i - x_j}{2} \right)^2 \xrightarrow{\text{or}} 1 - x_i^T x_j$$

subject to

$$x_i, x_j \in \{-1, 1\}$$

Lecture #12

Min cut - Karger's algorithm

Min-cut from Max flow algorithm: $\leq n^2 \cdot \underbrace{O(m \cdot m + n)}_{\text{LCI}}$

claim: G has at least $Kn/2$ edges if $|\text{min cut}| = K$

sub claim: Every cut has a degree at least K.

↳ if not, cut size < K.

C: What is the probability of obtaining this specific cut? (c is specified by K edges)

$E_i \rightarrow$ Event that edges from C are not picked in the i^{th} step, $1 \leq i \leq n-2$

$$\Pr[\bigcap_{i=1}^{n-2} E_i] = 1 - \Pr[\bigcup_{i=1}^n \bar{E}_i] \xrightarrow{\substack{\text{union bound} \\ \text{may not} \\ \text{give good} \\ \text{results}}}$$
$$= \Pr[E_1] \cdot \Pr[E_2 | E_1] \dots \Pr[E_{n-2} | \bigcap_{i=1}^{n-3} E_i]$$

$$\Pr[E_1] = 1 - \frac{K}{m} \geq 1 - \frac{2K}{Kn} = 1 - \frac{2}{n}$$

super nodes: Set of Nodes

super edges: set of edges

Init: $T \leftarrow$ super nodes, $F \leftarrow$ super edges

$T \leftarrow V$, $F \leftarrow E$

$$\Pr [E_j \mid \bigwedge_{i=1}^{j-1} E_i] \geq 1 - \frac{k_j}{\frac{k_j(n-j+1)}{2}}$$

$$\Pr \left[\bigwedge_{i=1}^n E_i \right] \geq \prod_{j=1}^{n-2} \left(1 - \frac{2}{n-j+1} \right)$$

$$= \prod_{j=1}^{n-2} \left(\frac{n-j-1}{n-j+1} \right) = \frac{2}{n(n-1)} \geq \frac{2}{n^2}$$

Lecture #13

while $|V| > 2$:

1. Pick an edge at random and merge the end-points
2. Repeat 1. (Pick an edge from the edges that have survived)

Return the edges between the last two nodes.

E_i : Event that edges from cut are not merged after i^{th} step

$$\text{want } \Pr \left[\bigcap_{i=1}^k E_i \right]$$

claim: If a graph has a cut of size K then $|E| \geq \frac{Kn}{2}$

cut: e_1, e_2, \dots, e_K

$$\Pr [E_1] = 1 - \frac{2}{|E|} \geq 1 - \frac{2}{n}$$

$$\Pr [E_2 | E_1]$$

$E^{(i)}$: Edges surviving after i^{th} step

$$|E^{(i)}| \geq \frac{K(n-i)}{2}$$

$$|E^{(d)}| \geq \frac{K(n-d)}{2}$$

$$\Pr [E_d | E_1, E_2, \dots, E_{d-1}] = 1 - \frac{K}{|E^{(d)}|}$$

$$\Pr(\varepsilon_i | \varepsilon_1 \cap \varepsilon_2 \cap \dots \cap \varepsilon_{i-1}) \geq 1 - \frac{2}{n-(i-1)}$$

$$\Pr\left[\bigcap_{i=1}^{n-1} \varepsilon_i\right] = \prod_{j=1}^{n-1} \left(1 - \frac{2}{n-(j-1)}\right) = \frac{2}{n(n-1)}.$$

$$\left(1 - \frac{2}{n(n-1)}\right)^{t-1} \frac{2}{n(n-1)}$$

$$\left(1 - \frac{2}{n^2}\right)^t \rightarrow \left(1 - \frac{2}{n^2}\right)^{\frac{n^2}{2}} \cdot \frac{2t}{n^2} \rightarrow e^{-2t/n^2}$$

Algorithm	Runtime	Prob
Karger	$t \cdot n^2$	$1 - e^{-2t/n^2}$
Max flow	$n^2(m+n) C $	1

Lecture #14

Counting Problems

- Count the # of Hamiltonian cycles in graphs
- Count the # of solution to a SAT instance
- Count the # of PMS in a given graphs
- Count the # of K-cliques in a graph

$$\text{#P} \stackrel{\text{dec}}{\in} \text{NP} \longleftrightarrow \text{Counting Version}$$

#P — "Sharp" P

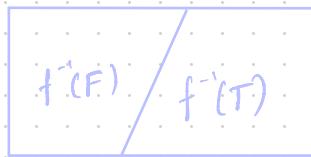
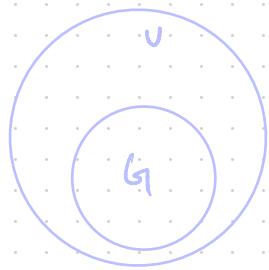
↳ At least as hard as NP

↳ Existence of PTIME algos is questionable

$$\text{Det}(X) = \sum_{\sigma \in S_n} (-1)^{\text{sign}(\sigma)} \prod_{i=1}^n x_{i\sigma(i)}$$

$$\text{Perm}(X) = \sum_{\sigma \in S_n} \prod_{i=1}^n x_{i\sigma(i)}$$

Count the no. of sat instance "approximately" and in polynomial time



$U \rightarrow$ Universe

Set of interest G

If $a \in U$ is picked up, computing $f(a)$ is easy

NP : is $|f^{-1}(T)| \geq 1$

#P: $|f^{-1}(T)|$?

a_1, \dots, a_N w.r.t, ind. from U .

$$\Pr(a_i \in G) = \frac{|G|}{|U|}$$

$$Y_i = \begin{cases} 1, & a_i \in G \\ 0, & \text{o/w} \end{cases}$$

$$\text{If } z = \frac{|U|}{N} \sum_{i=1}^N Y_i, \text{ then } E(z) = |G|$$

Estimator R.V.

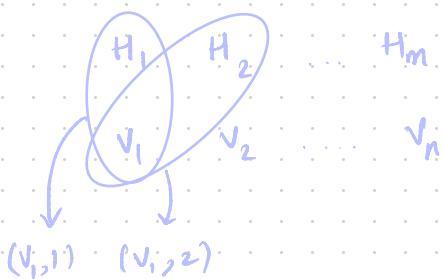
$$E(z) = |G|$$

PR AS
(z, δ) - RAS

DNF \rightarrow ORs of AND

Lecture #15

DNF \leftrightarrow CNF



Claim : $\frac{|\tilde{U}|}{|\bigcup_{i=1}^m H_i|} \leq m \rightarrow$ At most m repetitions possible.

$$\tilde{U} = \bigcup_{i=1}^m H_i$$

$$G = \{(v, i) \mid f(v, i) = 1\}$$

where $f(v, i) = \begin{cases} 1 & \text{if } i = \min \{j \mid (v, j) \in \tilde{U}\} \\ 0 & \text{o/w} \end{cases}$

$$|G| = |\bigcup_{i=1}^m H_i|$$

Uniform sampling from \tilde{U}

- First pick $i \in [m]$ at random w.p. $\frac{|H_i|}{|\tilde{U}|}$

$> \epsilon$

Lecture #16

Problem: Count the no. of Independent set in a given graph

NP-complete decision Problem: $\exists |I| \geq K ?$

$e_1, e_2, \dots, e_m \leftarrow$ Enumeration of edges in arbitrary order

$$G_1 = (V, E \setminus \{e_1\})$$

$$G_2 = (V, E \setminus \{e_2\})$$

$$\vdots$$

$$G_m = (V, E \setminus \{e_m\})$$

$|I(G)| \rightarrow$ No. independent sets in Graphs G_i

2^n

$$|I(G)| = \frac{|I(G)|}{|I(G_1)|} \times \frac{|I(G_2)|}{|I(G_2)|} \times \dots \times \frac{|I(G_{m-1})|}{|I(G_{m-1})|} \times |I(G_m)|$$

$$E(G_i) \subset E(G_{i-1})$$

$I(G_{i-1}) \subset I(G_i)$: contains independent sets with u, v where

$$e_i^u = (u, v)$$

$$|I(G_i) \setminus I(G_{i-1})| \leq |I(G_{i-1})|$$

$$r_i = \frac{|I(h_{i-1})|}{|I(h_i)|}$$

$$I(G_p) = 2^n \prod_{i=1}^m r_i$$

$$U \leftarrow I(h_i)$$

$$G \leftarrow I(h_{i-1})$$

$x_a = 1$ if the a^{th} sample $\in G$

$$\left| \Pr(X_a = 1) - \frac{|I(G_{i-1})|}{|I(h_i)|} \right| \leq \frac{\epsilon}{6m}$$

Overall, we want (ϵ, δ) -approximation

For each r_i , if we have $(\frac{\epsilon}{2m}, \frac{\delta}{2m})$ -approx., it is sufficient

$$\Pr(|\hat{r}_i - r_i| \geq \frac{\epsilon}{2m} r_i) \leq \frac{\delta}{2m}$$

$$1 - \frac{\epsilon}{2m} \leq \frac{\hat{r}_i}{r_i} \leq 1 + \frac{\epsilon}{2m}$$

w.p. $\geq 1 - \delta$,

$$(1 - \frac{\epsilon}{2m})^m \leq \prod \frac{\hat{r}_i}{r_i} \leq (1 + \frac{\epsilon}{2m})^m$$

Chapter 10/11 - Mitzenmacher Upfal

Lecture #17

マルコフ チャンス

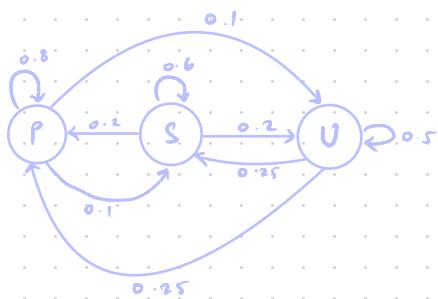
なんでもない

$$G = (V, E)$$

Lecture #18

0 1 2 3 4

	0	1	2	3	4
0	0	0	0	0	1
1	0	0	0	1-p	p
2	0	0	1-p	p	0
3	0	1-p	p	0	0
4	1-p	p	0	0	0



$$\begin{matrix}
 & P & S & U \\
 P & 0.8 & 0.1 & 0.1 \\
 S & 0.2 & 0.6 & 0.2 \\
 U & 0.25 & 0.25 & 0.5
 \end{matrix}$$

$$\begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.2 & 0.6 & 0.2 \\ 0.25 & 0.25 & 0.5 \end{bmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.2 & 0.6 & 0.2 \\ 0.25 & 0.25 & 0.5 \end{bmatrix}$$

$$= \begin{bmatrix} \end{bmatrix}$$

$$\begin{aligned}
 & 0.25 \times 0.8 + 0.25 \times 0.2 + 0.5 \times 0.25 \\
 & 0.2 + 0.05 + 0.125 \\
 & 0.375
 \end{aligned}$$

$\pi = \pi P \rightarrow$ eigenvector with value 1

Stationary distribution

↳ "Limiting" distribution (if it exists)

$$\pi_i = \sum_{k=1}^n \pi_k \cdot p_{k,i}$$

Doubly stochastic matrices have unique stationary distribution

Row & column sums both are 1

↓
It may not have a
limiting

Lecture # 20

$$h_j = \frac{1}{3}(1 + h_{j+1}) + \frac{2}{3}(1 + h_{j-1})$$

$$2(h_j - h_{j-1}) = (h_{j+1} - h_j) + 3$$

$$h_j = 2^{n+2} - 2^{j+2} - 3(n-j)$$

Lecture #21

Q:
- can I take MFDS?

Markov Chains

- Random walks on undirected graphs

$$P_{v_1 \rightarrow v_2} = \begin{cases} 1/d_{v_1} & \text{if } (v_1 \rightarrow v_2) \in E \\ 0 & \text{o/w} \end{cases}$$

$$\pi = \pi P$$

↳

$$\pi_i = \sum_{k=1}^n \pi_k P_{ki}, \forall i \in [n]$$

$$\pi_i = \sum_{k: v_i \in N(v_k)} \pi_k \cdot \frac{1}{d_k}$$

If the MC is "strongly connected" (irreducible)

"finite" and "aperiodic" then we have unique stationary distribution.

$$r_{ij}^t = \text{Prob. that } X_t = j; X_0 = i$$

and $\forall s \in [1, t-1]$

$$X_s \neq j$$

$$\pi_i = \frac{1}{h_{ii}}$$

h_{ij} = Exp. time to first visit state j for the first time starting from state i .

$$h_{ij} = \sum_{t>0} t \cdot r_{ij}^t$$

① ② . . . (n)

$$P_{i \rightarrow i+1} = \frac{i}{i+1}$$

$$P_{i \rightarrow 1} = \frac{1}{i+1}$$

$$\gamma_1^t = \left(\prod_{i=1}^{t-1} \frac{i}{i+1} \right) \cdot \frac{1}{t+1}$$

$$h_1^t = \sum_{t>0} \frac{1}{t+1} = \infty$$

$P_{ij}^t \neq 1$ → This is not correct acc. to sir, check this once

For a state s :

Period(s)

$$= \text{gcd}(t \mid (P^t)_{s,s} > 0)$$

State is aperiodic if $\text{period}(s) = 1$.

- If $|q| < n$ customers
 - w.p. λ , a new customer joins
- If $|q| > 0$, the w.p. μ the head of the queue is served
- w.r.t., q is unchanged

Gambler's ruin:

Player 1 cannot loss more than b_1 dollars

Player 2 cannot loss more than b_2 dollars

what is the prob. that P_1 gains b_2 dollars before P_2 gains b_1 dollars?

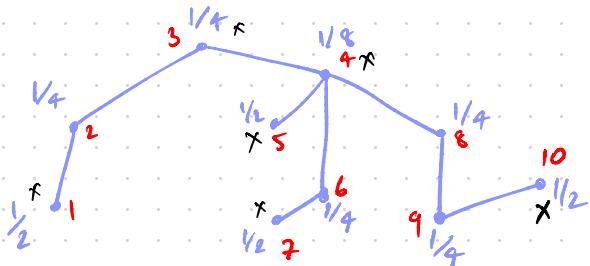
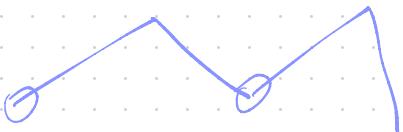
side question:

Prove a non bipartite graph (SC, finite) is aperiodic.

Lecture #22

Minimal Independent sets (MIS) - huby's Algorithm

Can this be parallelized



If v is marked, then include it in
 S

update $I \leftarrow I \cup S$

update $U : V \leftarrow V \setminus (S \cup N(S))$

Claim: At every stage, at least half of the edges are removed

Lemma: In an iteration, if a vertex w is marked, then w is added to s w.p.

Lemma: Let v be a good vertex. Then prob. that a neighbour of v gets marked is at least $(1 - e^{-1/6})$

A vertex v is good if $| \{w \in N(v) \mid \deg(w) > \deg(v)\} |$ is at most $\frac{2}{3} d_v$

Good Edge: At least one end point is good.

Lemma: # good Edge $\geq \frac{|E|}{2}$

Lemma 1 + Lemma 2 $\Rightarrow \geq \frac{(1 - e^{-1/6})}{2}$ with which a good vertex $\in N(s)$

Lecture #23

MIS

1. $I \leftarrow \{\}$ 2. while V is not empty:2a. For all $v \in V$ (in parallel)If $\deg(v) = 0$, then $I \leftarrow I \cup \{v\}$, $V \leftarrow V \setminus \{v\}$ else mark V w.p. $\frac{1}{2\deg(v)}$ 2b. For all $(u, v) \in E$ (in pair)

If both end points are marked,

2c. For all $v \in V$ (in parallel)Add $v \in S$ if v is marked2d. $I \leftarrow I \cup S$ 3. Return I "Good" vertex $\geq \frac{d_v}{3}$ neighbours have $\deg \leq d_v$ Book for probabilistic Methods: Alon & Spencer (2nd chapter)

PCP Theorems - NP certificates

