

KNAPSACK Has an Approximation Threshold of Zero^a

Theorem 78 *For any ϵ , there is a polynomial-time ϵ -approximation algorithm for KNAPSACK.*

- We have n weights $w_1, w_2, \dots, w_n \in \mathbb{Z}^+$, a weight limit W , and n values $v_1, v_2, \dots, v_n \in \mathbb{Z}^+$.^b
- We must find an $S \subseteq \{1, 2, \dots, n\}$ such that $\sum_{i \in S} w_i \leq W$ and $\sum_{i \in S} v_i$ is the largest possible.

^aIbarra and Kim (1975).

^bIf the values are fractional, the result is slightly messier, but the main conclusion remains correct. Contributed by Mr. Jr-Ben Tian (R92922045) on December 29, 2004.

The Proof (continued)

- Let

$$V = \max\{v_1, v_2, \dots, v_n\}.$$

- Clearly, $\sum_{i \in S} v_i \leq nV$.
- Let $0 \leq i \leq n$ and $0 \leq v \leq nV$.
- $W(i, v)$ is the minimum weight attainable by selecting only from the first i items and with a total value of v .
 - It is an $(n + 1) \times (nV + 1)$ table.
- Set $W(0, v) = \infty$ for $v \in \{1, 2, \dots, nV\}$ and $W(i, 0) = 0$ for $i = 0, 1, \dots, n$.^a

^aContributed by Mr. Ren-Shuo Liu (D98922016) and Mr. Yen-Wei Wu (D98922013) on December 28, 2009.

The Proof (continued)

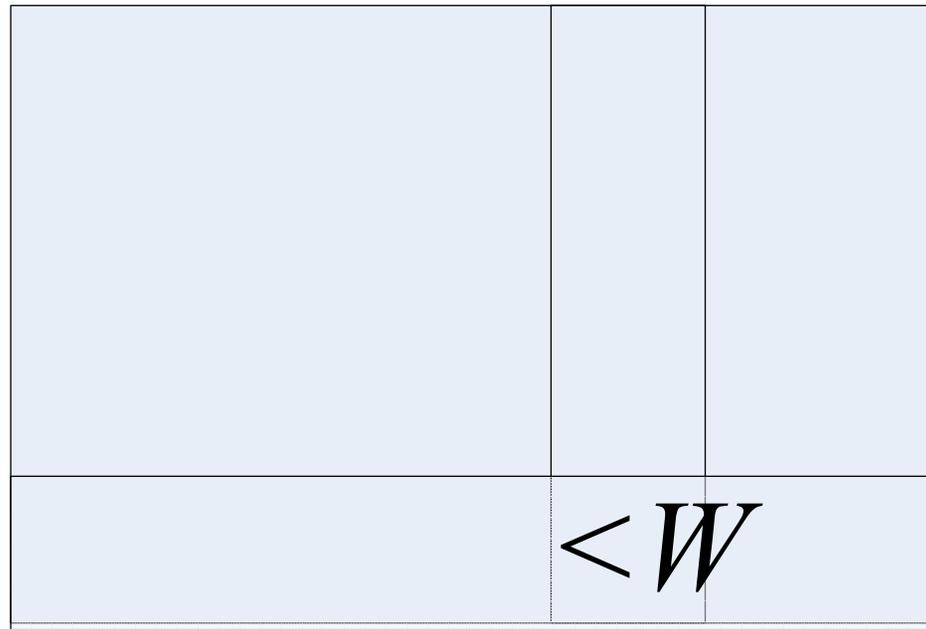
- Then, for $0 \leq i < n$,

$$W(i + 1, v) = \min\{W(i, v), W(i, v - v_{i+1}) + w_{i+1}\}.$$

- Finally, pick the largest v such that $W(n, v) \leq W$.^a
- The running time is $O(n^2V)$, not polynomial time.
- Key idea: Limit the number of precision bits.

^aLawler (1979).

0 v nV



The Proof (continued)

- Define

$$v'_i = 2^b \left\lfloor \frac{v_i}{2^b} \right\rfloor.$$

- This is equivalent to zeroing each v_i 's last b bits.

- Call the original instance

$$x = (w_1, \dots, w_n, W, v_1, \dots, v_n).$$

- Call the approximate instance

$$x' = (w_1, \dots, w_n, W, v'_1, \dots, v'_n).$$

The Proof (continued)

- Solving x' takes time $O(n^2V/2^b)$.
 - The algorithm only performs subtractions on the v_i -related values.
 - So the b last bits can be *removed* from the calculations.
 - That is, use $v_i'' = \lfloor \frac{v_i}{2^b} \rfloor$ and $V = \max(v_1'', v_2'', \dots, v_n'')$ in the calculations.
 - Then multiply the returned value by 2^b .
 - It is an $(n + 1) \times (nV + 1)/2^b$ table.

The Proof (continued)

- The solution S' is close to the optimum solution S :

$$\sum_{i \in S'} v_i \geq \sum_{i \in S'} v'_i \geq \sum_{i \in S} v'_i \geq \sum_{i \in S} (v_i - 2^b) \geq \sum_{i \in S} v_i - n2^b.$$

- Hence

$$\sum_{i \in S'} v_i \geq \sum_{i \in S} v_i - n2^b.$$

- Without loss of generality, assume $w_i \leq W$ for all i .
 - Otherwise, item i is redundant.
- V is a lower bound on OPT.
 - Picking an item with value V is a legitimate choice.

The Proof (concluded)

- The relative error from the optimum is $\leq n2^b/V$:

$$\frac{\sum_{i \in S} v_i - \sum_{i \in S'} v_i}{\sum_{i \in S} v_i} \leq \frac{\sum_{i \in S} v_i - \sum_{i \in S'} v_i}{V} \leq \frac{n2^b}{V}.$$

- Suppose we pick $b = \lfloor \log_2 \frac{\epsilon V}{n} \rfloor$.
- The algorithm becomes ϵ -approximate.^a
- The running time is then $O(n^2V/2^b) = O(n^3/\epsilon)$, a polynomial in n and $1/\epsilon$.^b

^aSee Eq. (16) on p. 683.

^bIt hence depends on the *value* of $1/\epsilon$. Thanks to a lively class discussion on December 20, 2006. If we fix ϵ and let the problem size increase, then the complexity is cubic. Contributed by Mr. Ren-Shan Luoh (D97922014) on December 23, 2008.

Comments

- INDEPENDENT SET and NODE COVER are reducible to each other (Corollary 40, p. 348).
- NODE COVER has an approximation threshold at most 0.5 (p. 691).
- But INDEPENDENT SET is unapproximable (see the textbook).
- INDEPENDENT SET limited to graphs with degree $\leq k$ is called k -DEGREE INDEPENDENT SET.
- k -DEGREE INDEPENDENT SET is approximable (see the textbook).

On P vs. NP

If 50 million people believe a foolish thing,
it's still a foolish thing.
— George Bernard Shaw (1856–1950)

Density^a

The **density** of language $L \subseteq \Sigma^*$ is defined as

$$\text{dens}_L(n) = |\{x \in L : |x| \leq n\}|.$$

- If $L = \{0, 1\}^*$, then $\text{dens}_L(n) = 2^{n+1} - 1$.
- So the density function grows at most exponentially.
- For a unary language $L \subseteq \{0\}^*$,

$$\text{dens}_L(n) \leq n + 1.$$

– Because $L \subseteq \{\epsilon, 0, 00, \dots, \overbrace{00 \cdots 0}^n, \dots\}$.

^aBerman and Hartmanis (1977).

Sparsity

- **Sparse languages** are languages with polynomially bounded density functions.
- **Dense languages** are languages with superpolynomial density functions.

Self-Reducibility for SAT

- An algorithm exhibits **self-reducibility** if it finds a certificate by exploiting algorithms for the *decision* version of the same problem.
- Let ϕ be a boolean expression in n variables x_1, x_2, \dots, x_n .
- $t \in \{0, 1\}^j$ is a **partial** truth assignment for x_1, x_2, \dots, x_j .
- $\phi[t]$ denotes the expression after substituting the truth values of t for $x_1, x_2, \dots, x_{|t|}$ in ϕ .

An Algorithm for SAT with Self-Reduction

We call the algorithm below with empty t .

```
1: if  $|t| = n$  then  
2:   return  $\phi[t]$ ;  
3: else  
4:   return  $\phi[t_0] \vee \phi[t_1]$ ;  
5: end if
```

The above algorithm runs in exponential time, by visiting all the partial assignments (or nodes on a depth- n binary tree).^a

^aThe same idea was used in the proof of Proposition 71 on p. 583.

NP-Completeness and Density^a

Theorem 79 *If a unary language $U \subseteq \{0\}^*$ is NP-complete, then $P = NP$.*

- Suppose there is a reduction R from SAT to U .
- We use R to find a truth assignment that satisfies boolean expression ϕ with n variables if it is satisfiable.
- Specifically, we use R to prune the exponential-time exhaustive search on p. 727.
- The trick is to keep the already discovered results $\phi[t]$ in a table H .

^aBerman (1978).

```
1: if  $|t| = n$  then
2:   return  $\phi[t]$ ;
3: else
4:   if  $(R(\phi[t]), v)$  is in table  $H$  then
5:     return  $v$ ;
6:   else
7:     if  $\phi[t_0] = \text{“satisfiable”}$  or  $\phi[t_1] = \text{“satisfiable”}$  then
8:       Insert  $(R(\phi[t]), \text{“satisfiable”})$  into  $H$ ;
9:       return  $\text{“satisfiable”}$ ;
10:    else
11:      Insert  $(R(\phi[t]), \text{“unsatisfiable”})$  into  $H$ ;
12:      return  $\text{“unsatisfiable”}$ ;
13:    end if
14:  end if
15: end if
```

The Proof (continued)

- Since R is a reduction, $R(\phi[t]) = R(\phi[t'])$ implies that $\phi[t]$ and $\phi[t']$ must be both satisfiable or unsatisfiable.
- $R(\phi[t])$ has polynomial length $\leq p(n)$ because R runs in log space.
- As R maps to unary numbers, there are only polynomially many $p(n)$ values of $R(\phi[t])$.
- How many nodes of the complete binary tree (of invocations/truth assignments) need to be visited?

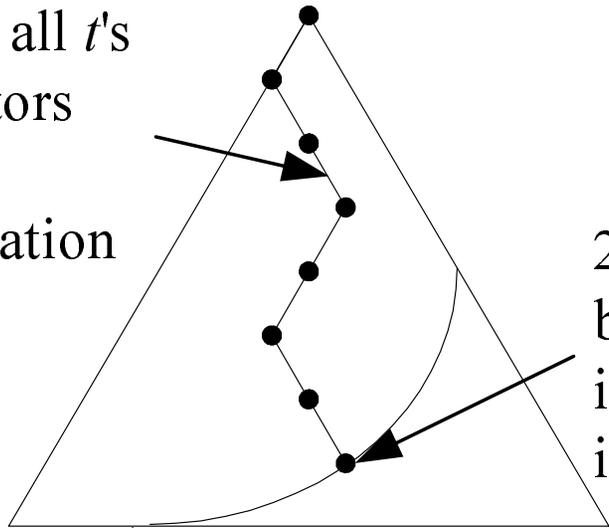
The Proof (continued)

- A search of the table takes time $O(p(n))$ in the random-access memory model.
- The running time is $O(Mp(n))$, where M is the total number of invocations of the algorithm.
- If that number is a polynomial, the overall algorithm runs in polynomial time and we are done.
- The invocations of the algorithm form a binary tree of depth at most n .

The Proof (continued)

- There is a set $T = \{t_1, t_2, \dots\}$ of invocations (partial truth assignments, i.e.) such that:
 1. $|T| \geq (M - 1)/(2n)$.
 2. All invocations in T are recursive (nonleaves).
 3. None of the elements of T is a prefix of another.

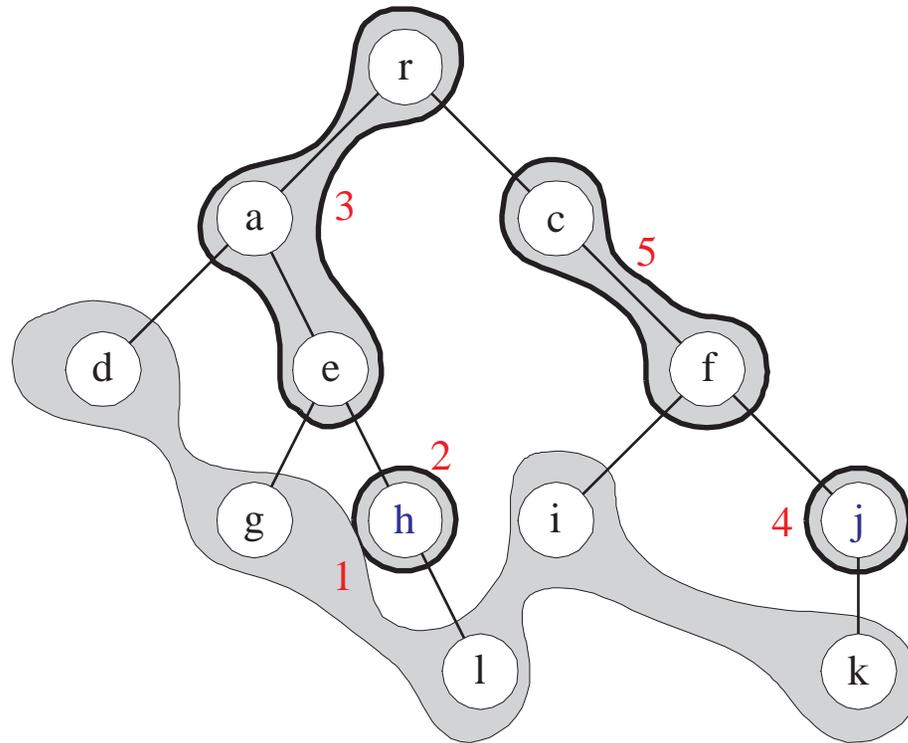
3rd step: Delete all t 's
at most n ancestors
(prefixes) from
further consideration



2nd step: Select any
bottom undeleted
invocation t and add
it to T

1st step: Delete
leaves; $(M - 1)/2$
nonleaves remaining

An Example



$T = \{h, j\}$; none of h and j is a prefix of the other.

The Proof (continued)

- All invocations $t \in T$ have different $R(\phi[t])$ values.
 - The invocation of one started after the invocation of the other had terminated.
 - If they had the same value, the one that was invoked later would have looked it up, and therefore would not be recursive, a contradiction.
- The existence of T implies that there are at least $(M - 1)/(2n)$ different $R(\phi[t])$ values in the table.

The Proof (concluded)

- We already know that there are at most $p(n)$ such values.
- Hence $(M - 1)/(2n) \leq p(n)$.
- Thus $M \leq 2np(n) + 1$.
- The running time is therefore $O(Mp(n)) = O(np^2(n))$.
- We comment that this theorem holds for any sparse language, not just unary ones.^a

^aMahaney (1980).

coNP-Completeness and Density

Theorem 80 (Fortung (1979)) *If a unary language $U \subseteq \{0\}^*$ is coNP-complete, then $P = NP$.*

- Suppose there is a reduction R from SAT COMPLEMENT to U .
- The rest of the proof is basically identical except that, now, we want to make sure a formula is unsatisfiable.

The Power of Monotone Circuits

- Monotone circuits can only compute monotone boolean functions.
- They are powerful enough to solve a P-complete problem, MONOTONE CIRCUIT VALUE (p. 294).
- There are NP-complete problems that are not monotone; they cannot be computed by monotone circuits at all.
- There are NP-complete problems that are monotone; they can be computed by monotone circuits.
 - HAMILTONIAN PATH and CLIQUE.

CLIQUE $_{n,k}$

- CLIQUE $_{n,k}$ is the boolean function deciding whether a graph $G = (V, E)$ with n nodes has a clique of size k .
- The input gates are the $\binom{n}{2}$ entries of the adjacency matrix of G .
 - Gate g_{ij} is set to true if the associated undirected edge $\{i, j\}$ exists.
- CLIQUE $_{n,k}$ is a monotone function.
- Thus it can be computed by a monotone circuit.
- This does not rule out that nonmonotone circuits for CLIQUE $_{n,k}$ may use fewer gates, however.

Crude Circuits

- One possible circuit for $\text{CLIQUE}_{n,k}$ does the following.
 1. For each $S \subseteq V$ with $|S| = k$, there is a circuit with $O(k^2)$ \wedge -gates testing whether S forms a clique.
 2. We then take an OR of the outcomes of all the $\binom{n}{k}$ subsets $S_1, S_2, \dots, S_{\binom{n}{k}}$.
- This is a monotone circuit with $O(k^2 \binom{n}{k})$ gates, which is exponentially large unless k or $n - k$ is a constant.
- A **crude circuit** $\text{CC}(X_1, X_2, \dots, X_m)$ tests if *any* of $X_i \subseteq V$ forms a clique.
 - The above-mentioned circuit is $\text{CC}(S_1, S_2, \dots, S_{\binom{n}{k}})$.

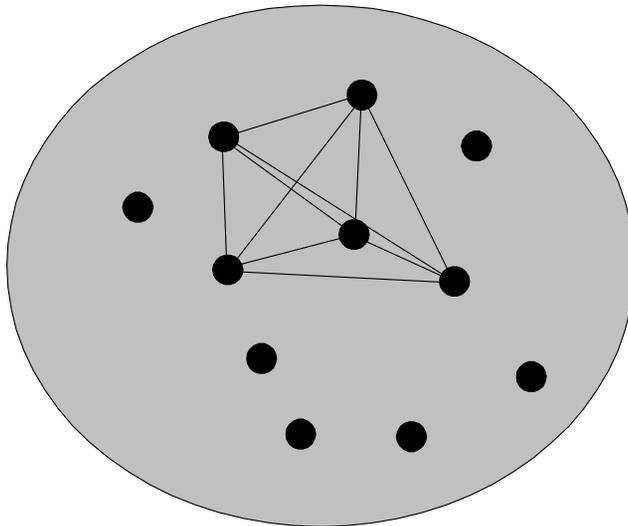
The Proof: Positive Examples

- Analysis will be applied to only **positive examples** and **negative examples** as inputs.
- A positive example is a graph that has $\binom{k}{2}$ edges connecting k nodes in all possible ways.
- There are $\binom{n}{k}$ such graphs.
- They all should elicit a true output from $\text{CLIQUE}_{n,k}$.

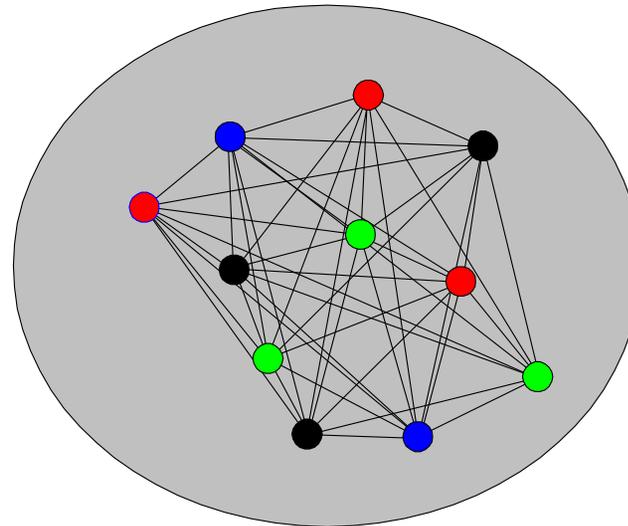
The Proof: Negative Examples

- Color the nodes with $k - 1$ different colors and join by an edge any two nodes that are colored differently.
- There are $(k - 1)^n$ such graphs.
- They all should elicit a false output from $\text{CLIQUE}_{n,k}$.
 - Each set of k nodes must have 2 identically colored nodes; hence there is no edge between them.

Positive and Negative Examples with $k = 5$



A positive example



A negative example

A Warmup to Razborov's Theorem

Lemma 81 (The birthday problem) *The probability of collision, $C(N, q)$, when q balls are thrown randomly into $N \geq q$ bins is at most*

$$\frac{q(q-1)}{2N}.$$

Lemma 82 *If crude circuit $CC(X_1, X_2, \dots, X_m)$ computes $\text{CLIQUE}_{n,k}$, then $m \geq n^{n^{1/8}/20}$ for n sufficiently large.*

The Proof (continued)

- Let $k = n^{1/4}$.
- Let $\ell = \sqrt{k}/10$.
- Let $X \subseteq V$.
- Suppose $|X| \leq \ell$.

The Proof (continued)

- A random $f : X \rightarrow \{1, 2, \dots, k - 1\}$ has collisions with probability less than 0.01 (see Lemma 81 on p. 744).
- Hence f is one-to-one with probability 0.99.
- When f is one-to-one, f is a coloring of X with $k - 1$ colors without repeated colors.
- As a result, when f is one-to-one, it generates a clique on X .

The Proof (continued)

- Note that a random negative example is simply a random $g : V \rightarrow \{1, 2, \dots, k - 1\}$.
- So our random $f : X \rightarrow \{1, 2, \dots, k - 1\}$ is simply a random g restricted to X .
- In summary, the probability that X is not a clique when supplied with a random negative example is at most 0.01.

The Proof (continued)

- Now suppose $|X| > \ell$.
- Consider the probability that X is a clique when supplied with a positive example.
- It is the probability that X is part of the clique.
- Hence the desired probability is $\binom{n-\ell}{k-\ell} / \binom{n}{k}$.

The Proof (continued)

- Now,

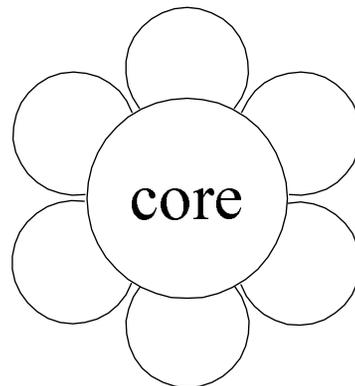
$$\begin{aligned}\frac{\binom{n-\ell}{k-\ell}}{\binom{n}{k}} &= \frac{k(k-1)\cdots(k-\ell+1)}{n(n-1)\cdots(n-\ell+1)} \\ &\leq \left(\frac{k}{n}\right)^\ell \\ &\leq n^{-(3/4)\ell} \\ &\leq n^{-\sqrt{k}/20} \\ &= n^{-n^{1/8}/20}.\end{aligned}$$

The Proof (concluded)

- In summary, the probability that X is a clique when supplied with a random positive example is at most $n^{-n^{1/8}/20}$.
- So we need at least $n^{n^{1/8}/20}$ X s in the crude circuit.

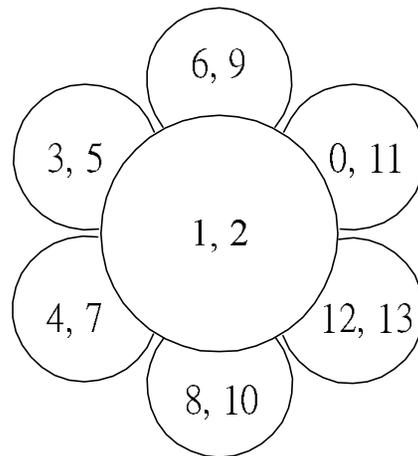
Sunflowers

- Fix $p \in \mathbb{Z}^+$ and $\ell \in \mathbb{Z}^+$.
- A **sunflower** is a family of p sets $\{P_1, P_2, \dots, P_p\}$, called **petals**, each of cardinality at most ℓ .
- Furthermore, all pairs of sets in the family must have the same intersection (called the **core** of the sunflower).



A Sample Sunflower

$\{\{1, 2, 3, 5\}, \{1, 2, 6, 9\}, \{0, 1, 2, 11\},$
 $\{1, 2, 12, 13\}, \{1, 2, 8, 10\}, \{1, 2, 4, 7\}\}.$



The Erdős-Rado Lemma

Lemma 83 *Let \mathcal{Z} be a family of more than $M = (p - 1)^\ell \ell!$ nonempty sets, each of cardinality ℓ or less. Then \mathcal{Z} must contain a sunflower (with p petals).*

- Induction on ℓ .
- For $\ell = 1$, p different singletons form a sunflower (with an empty core).
- Suppose $\ell > 1$.
- Consider a *maximal* subset $\mathcal{D} \subseteq \mathcal{Z}$ of *disjoint* sets.
 - Every set in $\mathcal{Z} - \mathcal{D}$ intersects some set in \mathcal{D} .

The Proof of the Erdős-Rado Lemma (continued)

For example,

$$\begin{aligned}\mathcal{Z} &= \{\{1, 2, 3, 5\}, \{1, 3, 6, 9\}, \{0, 4, 8, 11\}, \\ &\quad \{4, 5, 6, 7\}, \{5, 8, 9, 10\}, \{6, 7, 9, 11\}\}, \\ \mathcal{D} &= \{\{1, 2, 3, 5\}, \{0, 4, 8, 11\}\}.\end{aligned}$$

The Proof of the Erdős-Rado Lemma (continued)

- Suppose \mathcal{D} contains at least p sets.
 - \mathcal{D} constitutes a sunflower with an empty core.
- Suppose \mathcal{D} contains fewer than p sets.
 - Let C be the union of all sets in \mathcal{D} .
 - $|C| < (p - 1)\ell$.
 - C intersects every set in \mathcal{Z} by \mathcal{D} 's maximality.
 - There is a $d \in C$ that intersects more than $\frac{M}{(p-1)\ell} = (p - 1)^{\ell-1}(\ell - 1)!$ sets in \mathcal{Z} .
 - Consider $\mathcal{Z}' = \{Z - \{d\} : Z \in \mathcal{Z}, d \in Z\}$.

The Proof of the Erdős-Rado Lemma (concluded)

- (continued)

- \mathcal{Z}' has more than $M' = (p - 1)^{\ell - 1}(\ell - 1)!$ sets.
- M' is just M with ℓ replaced with $\ell - 1$.
- \mathcal{Z}' contains a sunflower by induction, say

$$\{P_1, P_2, \dots, P_p\}.$$

- Now,

$$\{P_1 \cup \{d\}, P_2 \cup \{d\}, \dots, P_p \cup \{d\}\}$$

is a sunflower in \mathcal{Z} .

Paul Erdős (1913–1996)



Comments on the Erdős-Rado Lemma

- A family of more than M sets must contain a sunflower.
- **Plucking** a sunflower means replacing the sets in the sunflower by its core.
- By *repeatedly* finding a sunflower and plucking it, we can reduce a family with more than M sets to a family with at most M sets.
- If \mathcal{Z} is a family of sets, the above result is denoted by $\text{pluck}(\mathcal{Z})$.
- Note: $\text{pluck}(\mathcal{Z})$ is not unique.

An Example of Plucking

- Recall the sunflower on p. 752:

$$\mathcal{Z} = \{\{1, 2, 3, 5\}, \{1, 2, 6, 9\}, \{0, 1, 2, 11\}, \\ \{1, 2, 12, 13\}, \{1, 2, 8, 10\}, \{1, 2, 4, 7\}\}$$

- Then

$$\text{pluck}(\mathcal{Z}) = \{\{1, 2\}\}.$$

Razborov's Theorem

Theorem 84 (Razborov (1985)) *There is a constant c such that for large enough n , all monotone circuits for $\text{CLIQUE}_{n,k}$ with $k = n^{1/4}$ have size at least $n^{cn^{1/8}}$.*

- We shall approximate any monotone circuit for $\text{CLIQUE}_{n,k}$ by a restricted kind of crude circuit.
- The approximation will proceed in steps: one step for each gate of the monotone circuit.
- Each step introduces few errors (false positives and false negatives).
- But the final crude circuit has exponentially many errors.

The Proof

- Fix $k = n^{1/4}$.
- Fix $\ell = n^{1/8}$.
- Note that^a

$$2 \binom{\ell}{2} \leq k - 1.$$

- p will be fixed later to be $n^{1/8} \log n$.
- Fix $M = (p - 1)^\ell \ell!$.
 - Recall the Erdős-Rado lemma (p. 753).

^aCorrected by Mr. Moustapha Bande (D98922042) on January 05, 2010.

The Proof (continued)

- Each crude circuit used in the approximation process is of the form $CC(X_1, X_2, \dots, X_m)$, where:
 - $X_i \subseteq V$.
 - $|X_i| \leq \ell$.
 - $m \leq M$.
- It answers true if any X_i is a clique.
- We shall show how to approximate any circuit for $CLIQUE_{n,k}$ by such a crude circuit, inductively.
- The induction basis is straightforward:
 - Input gate g_{ij} is the crude circuit $CC(\{i, j\})$.

The Proof (continued)

- Any monotone circuit can be considered the OR or AND of two subcircuits.
- We shall show how to build approximators of the overall circuit from the approximators of the two subcircuits.
 - We are given two crude circuits $CC(\mathcal{X})$ and $CC(\mathcal{Y})$.
 - \mathcal{X} and \mathcal{Y} are two families of at most M sets of nodes, each set containing at most ℓ nodes.
 - We construct the approximate OR and the approximate AND of these subcircuits.
 - Then show both approximations introduce few errors.

The Proof: OR

- $\text{CC}(\mathcal{X} \cup \mathcal{Y})$ is *equivalent* to the OR of $\text{CC}(\mathcal{X})$ and $\text{CC}(\mathcal{Y})$.
 - A set of nodes $\mathcal{C} \in \mathcal{X} \cup \mathcal{Y}$ is a clique if and only if $\mathcal{C} \in \mathcal{X}$ is a clique or $\mathcal{C} \in \mathcal{Y}$ is a clique.
- Violations in using $\text{CC}(\mathcal{X} \cup \mathcal{Y})$ occur when $|\mathcal{X} \cup \mathcal{Y}| > M$.
- Such violations can be eliminated by using

$$\text{CC}(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$$

as the approximate OR of $\text{CC}(\mathcal{X})$ and $\text{CC}(\mathcal{Y})$.

The Proof: OR

- If $CC(\mathcal{Z})$ is true, then $CC(\text{pluck}(\mathcal{Z}))$ must be true.
 - The quick reason: If Y is a clique, then a subset of Y must also be a clique.
 - For each $Y \in \mathcal{X} \cup \mathcal{Y}$, there must exist at least one $X \in \text{pluck}(\mathcal{X} \cup \mathcal{Y})$ such that $X \subseteq Y$.
 - If Y is a clique, then this X is also a clique.
- We now bound the number of errors this approximate OR makes on the positive and negative examples.

The Proof: OR (concluded)

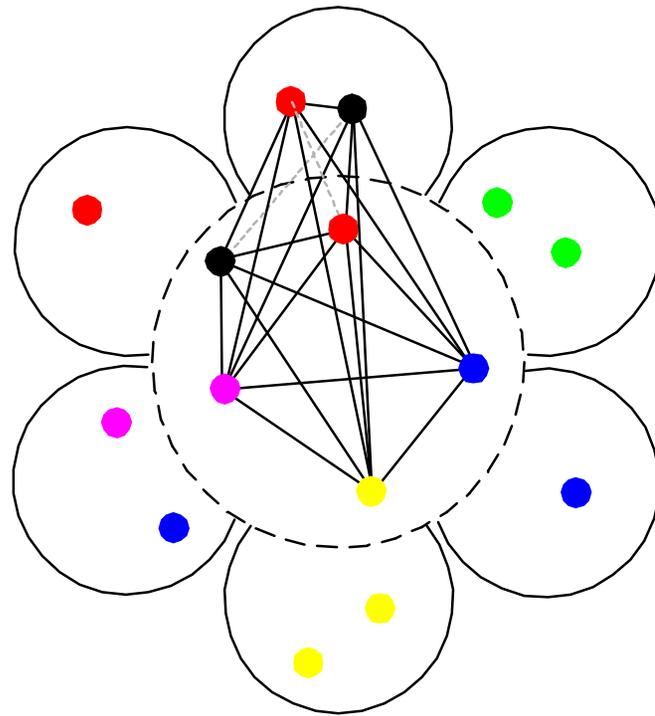
- $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ *introduces* a **false positive** if a negative example makes both $CC(\mathcal{X})$ and $CC(\mathcal{Y})$ return false but makes $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ return true.
- $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ *introduces* a **false negative** if a positive example makes either $CC(\mathcal{X})$ or $CC(\mathcal{Y})$ return true but makes $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ return false.
- How many false positives and false negatives are introduced by $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$?

The Number of False Positives

Lemma 85 $\text{CC}(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ introduces at most $\frac{M}{p-1} 2^{-p} (k-1)^n$ false positives.

- A plucking replaces the sunflower $\{Z_1, Z_2, \dots, Z_p\}$ with its core Z .
- A false positive is *necessarily* a coloring such that:
 - There is a pair of identically colored nodes in each petal Z_i (and so both crude circuits return false).
 - But the core contains distinctly colored nodes.
 - * This implies at least one node from each same-color pair was plucked away.
- We now count the number of such colorings.

Proof of Lemma 85 (continued)



Proof of Lemma 85 (continued)

- Color nodes V at random with $k - 1$ colors and let $R(X)$ denote the event that there are repeated colors in set X .
- Now $\text{prob}[R(Z_1) \wedge \cdots \wedge R(Z_p) \wedge \neg R(Z)]$ is at most

$$\begin{aligned} & \text{prob}[R(Z_1) \wedge \cdots \wedge R(Z_p) | \neg R(Z)] \\ &= \prod_{i=1}^p \text{prob}[R(Z_i) | \neg R(Z)] \leq \prod_{i=1}^p \text{prob}[R(Z_i)]. \quad (19) \end{aligned}$$

- First equality holds because $R(Z_i)$ are independent given $\neg R(Z)$ as Z contains their only common nodes.
- Last inequality holds as the likelihood of repetitions in Z_i decreases given no repetitions in $Z \subseteq Z_i$.

Proof of Lemma 85 (continued)

- Consider two nodes in Z_i .
- The probability that they have identical color is $\frac{1}{k-1}$.
- Now $\text{prob}[R(Z_i)] \leq \frac{\binom{|Z_i|}{2}}{k-1} \leq \frac{\binom{\ell}{2}}{k-1} \leq \frac{1}{2}$.
- So the probability^a that a random coloring is a new false positive is at most 2^{-p} by inequality (19).
- As there are $(k-1)^n$ different colorings, each plucking introduces at most $2^{-p}(k-1)^n$ false positives.

^aProportion, i.e.

Proof of Lemma 85 (concluded)

- Recall that $|\mathcal{X} \cup \mathcal{Y}| \leq 2M$.
- $\text{pluck}(\mathcal{X} \cup \mathcal{Y})$ ends the moment the set system contains $\leq M$ sets.
- Each plucking reduces the number of sets by $p - 1$.
- Hence at most $\frac{M}{p-1}$ pluckings occur in $\text{pluck}(\mathcal{X} \cup \mathcal{Y})$.
- At most

$$\frac{M}{p-1} 2^{-p} (k-1)^n$$

false positives are introduced.^a

^aNote that the numbers of errors are added not multiplied. Recall that we count how many *new* errors are introduced by each approximation step. Contributed by Mr. Ren-Shuo Liu (D98922016) on January 5, 2010.

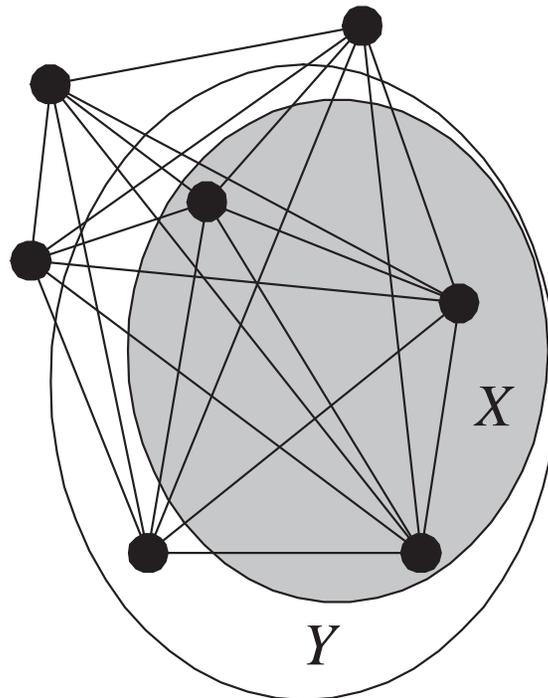
The Number of False Negatives

Lemma 86 $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ *introduces no false negatives.*

- A plucking replaces sets in a crude circuit by their (common) subset.
- This makes the test for cliqueness less stringent (p. 765).^a

^aRecall that $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ introduces a false negative if a positive example makes either $CC(\mathcal{X})$ or $CC(\mathcal{Y})$ return true but makes $CC(\text{pluck}(\mathcal{X} \cup \mathcal{Y}))$ return false.

The Number of False Negatives (concluded)



The Proof: AND

- The approximate AND of crude circuits $CC(\mathcal{X})$ and $CC(\mathcal{Y})$ is

$$CC(\text{pluck}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}, |X_i \cup Y_j| \leq \ell\})).$$

- We now count the number of errors this approximate AND makes on the positive and negative examples.

The Proof: AND (concluded)

- The approximate AND *introduces* a **false positive** if a negative example makes either $CC(\mathcal{X})$ or $CC(\mathcal{Y})$ return false but makes the approximate AND return true.
- The approximate AND *introduces* a **false negative** if a positive example makes both $CC(\mathcal{X})$ and $CC(\mathcal{Y})$ return true but makes the approximate AND return false.
- How many false positives and false negatives are introduced by the approximate AND?

The Number of False Positives

Lemma 87 *The approximate AND introduces at most $M^2 2^{-p} (k-1)^n$ false positives.*

- We prove this claim in stages.
- $\text{CC}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}\})$ introduces no false positives.
 - If $X_i \cup Y_j$ is a clique, both X_i and Y_j must be cliques, making both $\text{CC}(\mathcal{X})$ and $\text{CC}(\mathcal{Y})$ return true.
- $\text{CC}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}, |X_i \cup Y_j| \leq \ell\})$ introduces no additional false positives because we are testing fewer sets for cliqueness.

Proof of Lemma 87 (concluded)

- $|\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}, |X_i \cup Y_j| \leq \ell\}| \leq M^2$.
- Each plucking reduces the number of sets by $p - 1$.
- So $\text{pluck}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}, |X_i \cup Y_j| \leq \ell\})$ involves $\leq M^2/(p - 1)$ pluckings.
- Each plucking introduces at most $2^{-p}(k - 1)^n$ false positives by the proof of Lemma 85 (p. 767).
- The desired upper bound is

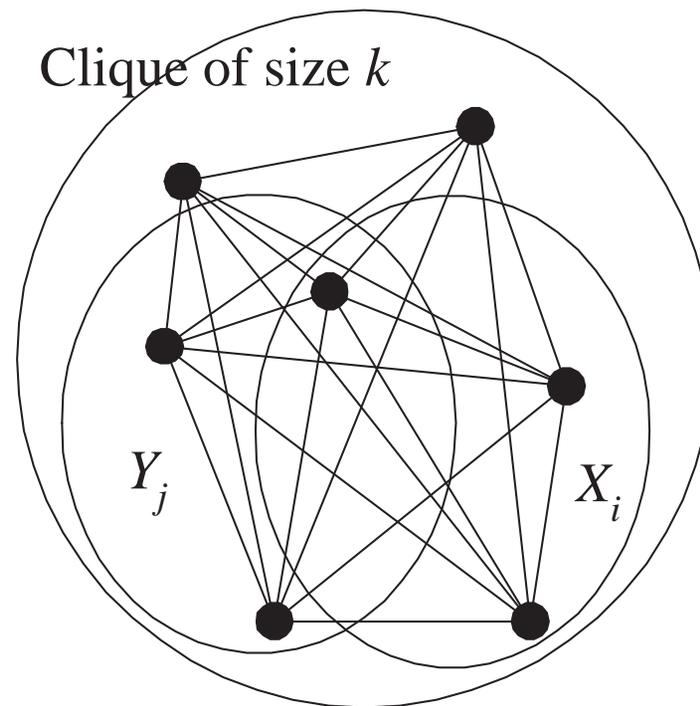
$$\lceil M^2/(p - 1) \rceil 2^{-p}(k - 1)^n \leq M^2 2^{-p}(k - 1)^n.$$

The Number of False Negatives

Lemma 88 *The approximate AND introduces at most $M^2 \binom{n-\ell-1}{k-\ell-1}$ false negatives.*

- We again prove this claim in stages.
- $\text{CC}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}\})$ introduces no false negatives.
 - Suppose both $\text{CC}(\mathcal{X})$ and $\text{CC}(\mathcal{Y})$ accept a positive example with a clique of size k .
 - This clique must contain an $X_i \in \mathcal{X}$ and a $Y_j \in \mathcal{Y}$.
 - * This is why both $\text{CC}(\mathcal{X})$ and $\text{CC}(\mathcal{Y})$ return true.
 - As the clique contains $X_i \cup Y_j$, the new circuit returns true.

Proof of Lemma 88 (continued)



Proof of Lemma 88 (continued)

- $\text{CC}(\{X_i \cup Y_j : X_i \in \mathcal{X}, Y_j \in \mathcal{Y}, |X_i \cup Y_j| \leq \ell\})$ introduces $\leq M^2 \binom{n-\ell-1}{k-\ell-1}$ false negatives.
 - Deletion of set $Z = X_i \cup Y_j$ larger than ℓ introduces false negatives *only if* Z is part of a clique.
 - There are $\binom{n-|Z|}{k-|Z|}$ such cliques.
 - * It is the number of positive examples whose clique contains Z .
 - $\binom{n-|Z|}{k-|Z|} \leq \binom{n-\ell-1}{k-\ell-1}$ as $|Z| > \ell$.
 - There are at most M^2 such Z s.

Proof of Lemma 88 (concluded)

- Plucking introduces no false negatives.
 - Recall that if $CC(\mathcal{Z})$ is true, then $CC(\text{pluck}(\mathcal{Z}))$ must be true (p. 765).

Two Summarizing Lemmas

From Lemmas 85 (p. 767) and 87 (p. 776), we have:

Lemma 89 *Each approximation step introduces at most $M^2 2^{-p} (k-1)^n$ false positives.*

From Lemmas 86 (p. 772) and 88 (p. 778), we have:

Lemma 90 *Each approximation step introduces at most $M^2 \binom{n-\ell-1}{k-\ell-1}$ false negatives.*

The Proof (continued)

- The above two lemmas show that each approximation step introduces “few” false positives and false negatives.
- We next show that the resulting crude circuit has “a lot” of false positives or false negatives.

The Final Crude Circuit

Lemma 91 *Every final crude circuit is:*

1. *Identically false—thus wrong on all positive examples.*
2. *Or outputs true on at least half of the negative examples.*
 - Suppose it is not identically false.
 - By construction, it accepts at least those graphs that have a clique on some set X of nodes, with $|X| \leq \ell$, which at $n^{1/8}$ is less than $k = n^{1/4}$.
 - The proof of Lemma 85 (p. 767ff) shows that at least half of the colorings assign different colors to nodes in X .
 - So half of the negative examples have a clique in X and are accepted.

The Proof (continued)

- Recall the constants on p. 761: $k = n^{1/4}$, $\ell = n^{1/8}$, $p = n^{1/8} \log n$, $M = (p - 1)^\ell \ell! < n^{(1/3)n^{1/8}}$ for large n .
- Suppose the final crude circuit is identically false.
 - By Lemma 90 (p. 782), each approximation step introduces at most $M^2 \binom{n-\ell-1}{k-\ell-1}$ false negatives.
 - There are $\binom{n}{k}$ positive examples.
 - The original monotone circuit for $\text{CLIQUE}_{n,k}$ has at least

$$\frac{\binom{n}{k}}{M^2 \binom{n-\ell-1}{k-\ell-1}} \geq \frac{1}{M^2} \left(\frac{n-\ell}{k} \right)^\ell \geq n^{(1/12)n^{1/8}}$$

gates for large n .

The Proof (concluded)

- Suppose the final crude circuit is not identically false.
 - Lemma 91 (p. 784) says that there are at least $(k - 1)^n / 2$ false positives.
 - By Lemma 89 (p. 782), each approximation step introduces at most $M^2 2^{-p} (k - 1)^n$ false positives
 - The original monotone circuit for $\text{CLIQUE}_{n,k}$ has at least

$$\frac{(k - 1)^n / 2}{M^2 2^{-p} (k - 1)^n} = \frac{2^{p-1}}{M^2} \geq n^{(1/3)n^{1/8}}$$

gates.

Alexander Razborov (1963–)



$P \neq NP$ Proved?

- Razborov's theorem says that there is a monotone language in NP that has no polynomial monotone circuits.
- If we can prove that all monotone languages in P have polynomial monotone circuits, then $P \neq NP$.
- But Razborov proved in 1985 that some monotone languages in P have no polynomial monotone circuits!

Finis