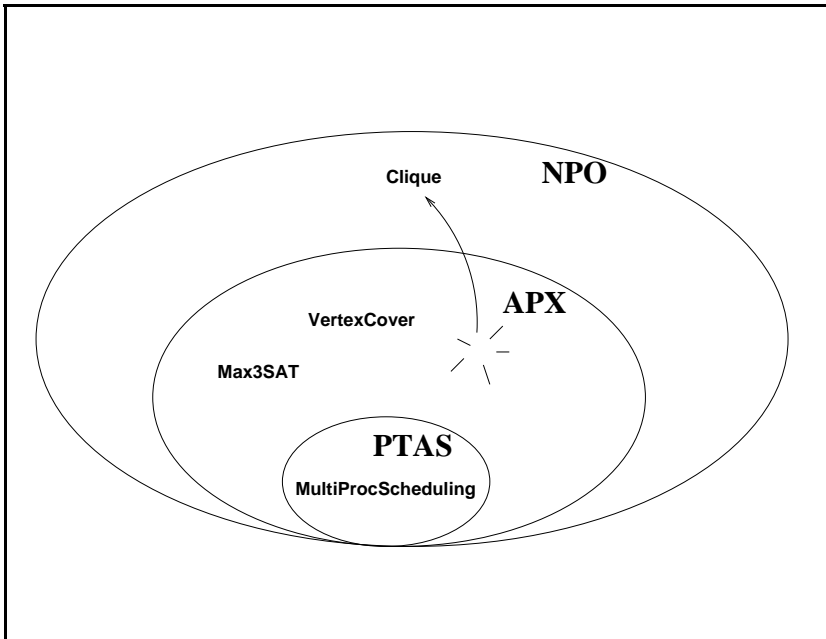


1

For any $\epsilon > 0$ approximating CLIQUE with ratio $1 + \epsilon$ is NP-hard, where n is the number of vertices in the input graph.

There is an $\epsilon > 0$ such that approximating CLIQUE with ratio n^ϵ is NP-hard, where n is the number of vertices in the input graph.

3



2

Significance

(Rather negative result) If ϵ was $\frac{1}{2}$ and the largest cliques in graphs on n vertices had size $O(\sqrt{n})$ we would not be guaranteed to find (in polynomial time) a clique with more than $O(1)$ vertices.

(Better picture in practice) On average the largest cliques of an n vertex graph chosen at random have size $O(\log n)$ (therefore exhaustive search should give us, in polynomial time, some good candidates).

4

Beyond APX, the CLIQUE problem

The hardness result for CLIQUE relies upon its interesting *self-improvement* behavior when we take graph products. The next example describes this behavior.

For a graph $G = (V, E)$ let \hat{E} denote E with all self-loops added, that is, $\hat{E} = E \cup \{(u, u) : u \in V\}$. For graphs $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$, their *product* $G_1 \times G_2$ is the graph whose vertex set is the set $V_1 \times V_2$, and edge set is

$$\{((u_1, v_1), (u_2, v_2)) : (u_1, u_2) \in \hat{E}_1 \wedge (v_1, v_2) \in \hat{E}_2\}$$

5

Self-improvement technique

Now suppose a reduction h exists from SAT to CLIQUE, such that the graph produced by the reduction has clique number either l , or $(1 - \epsilon)l$, depending on whether or not the SAT formula was satisfiable or not.

Claim: It is NP-hard to approximate CLIQUE with any constant ratio.

Suppose we had a c -approximation algorithm A for CLIQUE.

Choose k so that $c < (1 - \epsilon)^{-k}$ (e.g. $k = \lceil \log_{1/(1-\epsilon)} c \rceil$ would do).

To decide SAT reduce it to CLIQUE, then compute Let H , be the product of $h(G)$ with itself k times.

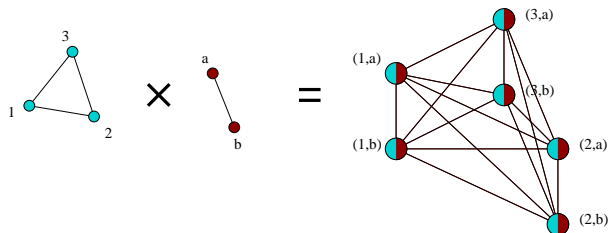
$\omega(H)$ is either l^k or $(1 - \epsilon)^k l^k$.

But $c < (1 - \epsilon)^{-k}$, hence you can use A to verify whether the optimal cliques in $h(G)^k$ are “small” or “large”.

This decides SAT deterministically in polynomial time!

7

Generalisation of G^k we saw last week.



Let $\omega(G)$ be the size of the largest clique in a graph. It is easily checked that $\omega(G_1 \times G_2) = \omega(G_1) \cdot \omega(G_2)$.

6

Key property. (Self-improvement) The gap in clique numbers, $(1 - \epsilon)^{-k}$, can be made arbitrarily large by increasing k enough.

Note however that H has size $n^{O(k)}$, so k must remain $O(1)$ if the above construction has to work in polynomial time.

8

The rapid increase in problem size when using self-improvement may seem hard to avoid. Surprisingly, the following combinatorial object often allows us to do just that.

Let n be an integer. A (n, k, α) *booster* is a collection \mathcal{S} of subsets of $\{1, 2, \dots, n\}$, each of size k . For every subset $A \subseteq \{1, 2, \dots, n\}$, the sets in the collection that are subsets of A constitute a fraction^a between $(\frac{|A|}{n} - \alpha)^k$ and $(\frac{|A|}{n} + \alpha)^k$ of all sets in \mathcal{S} .

^a When $\frac{|A|}{n} < 1.1\alpha$, the quantity $(\frac{|A|}{n} - \alpha)^k$ should be considered to be 0.

More specifically, take $n = 7$ and $k = 3$. The following collection of subsets of $\{1, 2, 3, 4, 5, 6, 7\}$ is a $(3, 7, 0)$ -booster.

$\{1, 2, 3\}, \{1, 2, 4\}, \{1, 3, 4\}, \{2, 3, 4\}, \{1, 2, 5\},$
 $\{1, 3, 5\}, \{2, 3, 5\}, \{1, 4, 5\}, \{2, 4, 5\}, \{3, 4, 5\},$
 $\{1, 2, 6\}, \{1, 3, 6\}, \{2, 3, 6\}, \{1, 4, 6\}, \{2, 4, 6\},$
 $\{3, 4, 6\}, \{1, 5, 6\}, \{2, 5, 6\}, \{3, 5, 6\}, \{4, 5, 6\},$
 $\{1, 2, 7\}, \{1, 3, 7\}, \{2, 3, 7\}, \{1, 4, 7\}, \{2, 4, 7\},$
 $\{3, 4, 7\}, \{1, 5, 7\}, \{2, 5, 7\}, \{3, 5, 7\}, \{4, 5, 7\},$
 $\{1, 6, 7\}, \{2, 6, 7\}, \{3, 6, 7\}, \{4, 6, 7\}, \{5, 6, 7\}$

Take set $A = \{1, 2, 5, 6, 7\}$. There are $\binom{5}{3} = 10$ elements of the booster that are subsets of A :

$\{1, 2, 5\}, \{1, 2, 6\}, \{1, 5, 6\}, \{2, 5, 6\}, \{1, 2, 7\},$
 $\{1, 5, 7\}, \{2, 5, 7\}, \{1, 6, 7\}, \{2, 6, 7\}, \{5, 6, 7\}$

and $0.2857 = \frac{10}{35} \sim \left(\frac{|A|}{n}\right)^3 = 0.3644$.

Example

The set $\mathcal{P}(n)$ of all subsets of $\{1, 2, \dots, n\}$ of size k is a booster with $\alpha \sim 0$. For any $A \subseteq \{1, 2, \dots, n\}$, with $|A| = O(n)$, the fraction of subsets of A in $\mathcal{P}(n)$ is $\binom{|A|}{k} / \binom{n}{k}$, which is approximately $(|A|/n)^k$.

Unfortunately $|\mathcal{P}(n)| = \binom{n}{k} = O(n^k)$, hence k must be $O(1)$ if the booster has to be used in polynomial-time reductions.

In the following treatment we would like k to be as large as possible.

Technical result

For any $k = O(\log n)$ and $\alpha > 0$, an (n, k, α) booster of size $\text{poly}(n)$ can be constructed in $\text{poly}(n)$ time.

The proof of this result is beyond the scope of this module, but it has very important consequences in our main argument.

Booster product

Let G be a graph on n vertices. The *booster product* of G , $\mathcal{B}(G, n, k, \alpha)$ is a graph whose vertices are the sets of a (n, k, α) -booster \mathcal{S} , and there is an edge between sets S_i and S_j if and only if $S_i \cup S_j$ is a clique in G .

What is this? Why do we need all this?

13

Argument

Let $A \subseteq \{1, 2, \dots, n\}$ be a clique of size $\omega(G)$ in graph G .

Then the number of sets from \mathcal{S} that are subsets of A is between $(\frac{\omega(G)}{n} - \alpha)^k |\mathcal{S}|$ and $(\frac{\omega(G)}{n} + \alpha)^k |\mathcal{S}|$.

Clearly, all such sets form a clique in the booster product.

Conversely, given the largest clique B in the booster product, let A be the union of all sets in the clique.

Then A is a clique in G , and hence must have size at most $\omega(G)$.

The booster property implies that the size of B is as claimed.

15

For any graph G , and any (n, k, α) booster, the clique number of the booster product of G lies between $(\frac{\omega(G)}{n} - \alpha)^k |\mathcal{S}|$ and $(\frac{\omega(G)}{n} + \alpha)^k |\mathcal{S}|$.

In other words:

Option 1 Using graph products we go from $\omega(G)$ to $\omega(G)^k$.

Option 2 Using booster products we go from $\omega(G)$ to $\omega(G)^k n^{O(1)}$

Using option 2 we “inflate” the gap even more!

14

Finally we come to the most important consequence of the argument so far:

There is an $\epsilon > 0$ such that approximating CLIQUE with ratio n^ϵ is NP-hard, where n is the number of vertices in the input graph.

16

(1) Let G be the graph obtained from the usual 3SAT to CLIQUE reduction, and suppose it has N vertices (remember that N is three times the number of clauses in the original formula).

(2) The reduction ensures, for some fixed $\beta > 0$, that $\omega(G)$ is either at least $N/3$ or at most $N(1 - \beta)/3$, and it is NP-hard to decide which case holds.

(3) Now construct a $(N, \log N, \alpha)$ booster, \mathcal{S} , by choosing $\alpha = \beta/9$.

(4) Construct the booster product of G . The number of vertices in the booster product is $|\mathcal{S}|$, and Lemma above says the clique number is either at least $((3 - \beta)/9)^{\log N} |\mathcal{S}|$ or at most $((3 - 2\beta)/9)^{\log N} |\mathcal{S}|$. Hence the gap is now N^γ for some $\gamma > 0$, and further, $|\mathcal{S}| = N^{O(1)}$, so this gap is $|\mathcal{S}|^\epsilon$ for some $\epsilon > 0$.

17

What should you keep of this?

- The definition of greedy + dynamic programming algorithms, and the “feeling” that these paradigms can lead to very good quality solutions in some cases.
- Advises on how to design a program in several different cases.
- The knowledge of what an approximation algorithm is and how can you prove results about them.

Many big industrial projects in Europe (e.g. a recent railway optimisation project in Switzerland, Germany, Italy, Greece).

19

That’s all folks!

- Two major design paradigms: greedy + dynamic programming.
- A number of application areas. We looked more deeply into string algorithms and graph matching algorithms.
- A quick glimpse in space complexity theory.
- The broad and very widely used area approximation algorithms.

18