CS420+500: Advanced Algorithm Design and Analysis

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In this lecture we:

- Discussed complexity classes ;
- defined NP, NP-Hard;
- NP-Completeness and reduction.

1 NP-Completeness

Firstly, let's start with some definitions:

Definition 1. Decision Problem: algorithmic questions that can be answered by YES or NO.

Definition 2. P: or Polynomial set is set of decision problems decidable in polynomial time. A decision problem L is in P if there exists a polynomial time algorithm A such that $L = \{x | A \text{ accepts } x\}$. (Note that A is a polynomial time algorithm if there exists a positive integer k such that for all inputs x, A halts on input x and either accepts or rejects x in time $O(|x|^k)$.)

Definition 3. NP: is set of decision problems that have polynomial time "verifications".

Definition 4. Verification: An algorithm V is a polynomial time verifier for a problem L if for every input $x \in L$, there exists a witness w such that V on input (x, w) accepts in time polynomial in |x|, and if $x \notin L$, then for all witnesses w, V on input (x, w) rejects in time polynomial in |x|.

Originally, NP comes from non-deterministic polynomial or more precisely from non-deterministic turing machines. Every problem in P is also in NP since the algorithm A for L acts as a verifier that doesn't require a witness.



1.1 SAT problem

SAT is the set of Boolean formulas in CNF¹ that are satisfiable, that is, there is a truth assignment to the variables in the formula so that the formula evaluates to True.

Theorem 5. $SAT \in NP$

Proof: The string w that specifies the truth assignment is a good witness for ϕ . Verifier V needs to only check that w satisfies ϕ (can be done in polynomial time).

The class Co-NP is the set of decision problems L whose complement is in the class NP. The complement of a decision problem L is the set $\{x | x \notin L\}$.

1.2 CLIQUE problem

 $CLIQUE = \{ \langle G, k \rangle | G \text{ is a graph with clique of size } k \}$. Clique of size k has k vertices that all are adjacent to each other.

Theorem 6. $CLIQUE \in NP$

witness w for $\langle G, k \rangle$ is a set of k vertices of G that form a clique. Verifier can check in polynomial time in $|\langle G, k \rangle|$ that w is a clique or not.

Definition 7. NP-Hard: set of problems L s.t. if L could be solved in polynomial time, then all other problems in NP could also be solved in polynomial time. Formally, $L \in NP - Hard$ means if $L \in P$ then $L' \in P$ for all $L' \in NP$.

Definition 8. NP-Complete: A decision problem L is NP-Complete if:

1. $L \in NP$

2. $L \in NP - Hard$

Theorem 9 (Cook-Levin 1971). SAT is NP-complete.

We are not going to prove theorem 9 in class. NP-Complete contains hardest problems in NP. CLIQUE is an NP-Complete (Yes/No) problem but MAX-CLIQUE is NP-Hard (find maximum size clique in a graph G). Usually, when we convert Yes/No problems to finding problems, they get harder.

2 Reduction to SAT

We know that SAT is NP-Complete problem. It is difficult to prove a problem is NP-hard in the same way that Cook did. However, since we know SAT is NP-hard, we can show that a problem L is NP-hard by showing that a polynomial-time algorithm for L can be used to solve SAT in polynomial time. In other words, by showing how to reduce SAT to L. It is important to do this reduction in a right order.

¹Conjunctive normal form such as $(x \lor y) \land (\overline{x} \lor z)$

Theorem 10. $CLIQUE \in NP$ -Hard.

As a proof, we are going to reduce SAT problem to CLIQUE. So we want to transform a formula ϕ into a graph $\langle G, k \rangle$ so that:

- 1. G contains a clique of size $k \Leftrightarrow \phi$ is satisfiable
- 2. transformation should take polynomial time

So we do the following three steps:

- Create a vertex for every literal in every clause
- Connect a vertex from *i*'th clause to *j*'th clause $(i \neq j)$ unless they are negative of each other (x, \overline{x})
- Let k be number of clauses in ϕ

As a example, let say we have $\phi = (x_1) \land (\overline{x_1} \lor \overline{x_2}) \land (x_1 \lor x_3) \land (x_2 \lor \overline{x_3} \lor x_4)$. Here is a transformed graph:



We claim that $\phi \in SAT$ iff G has k-clique.

 (\Rightarrow) If ϕ has a truth assignment, then every clause has at least one true literal. Thus, we can choose one from each clause of size "k".

(\Leftarrow) If G has a clique "k" then exactly one vertex from each clause is in ϕ . So we can assign one to each literal vertex and as a result, ϕ is satisfiable.