

NP-Completeness

Yin Tat Lee

Computational Complexity

Goal: Classify problems according to the amount of computational resources used by the best algorithms that solve them

Here we focus on time complexity

Recall: worst-case running time of an algorithm

• **max** # steps algorithm takes on any input of size **n**

Relative Complexity of Problems

- Want a notion that allows us to compare the complexity of problems
- Want to be able to make statements of the form

"If we could solve problem **B** in polynomial time then we can solve problem **A** in polynomial time"

"Problem **B** is at least as hard as problem **A**"

Polynomial Time Reduction

Def $A \leq_P B$: if there is an algorithm for problem A using a 'black box' (subroutine) that solve problem B s.t.,

- Algorithm uses only a polynomial number of steps
- Makes only a polynomial number of calls to a subroutine for B



In words, B is as hard as A (it can be even harder)

 \leq_{n}^{1} Reductions

Here, we often use a restricted form of polynomial-time reduction often called Karp reduction.

 $A \leq_p^1 B$: if and only if there is an algorithm for A given a black box solving B that on input **x**

- Runs for polynomial time computing an input f(x) of B
- Makes one call to the black box for B for input f(x)
- Returns the answer that the black box gave

We say that the function f(.) is the reduction

Let A = bipartite matching. Let B = maxflow.

B

Total Results: 0



Start the presentation to see live content. For screen share software, share the entire screen. Get help at pollev.com/app

6

Answer

Let A = bipartite matching. Let B = maxflow.

We know how to solve bipartite matching by calling maxflow once.

So, it may look like the answer is

Both
$$A \leq_p B$$
 and $A \leq_p^1 B$.

However, since both problems can be solved in polynomial time, one valid reduction would be simply doing nothing.

Hence, all statements are true. So, \leq_p is mainly to distinguish if a problem is in P or not.

Fine-Grained Complexity

There are recent work on distinguishing different polytime.



Figure 1 Partial summary of the implications of the main conjectures. An arrow from problem A to problem B, where A has a(n) next to it, B has b(n) next to it, implies that $A \leq_{a,b} B$. When the inputs are graphs, n stands for the number of nodes. N always stands for the total input size. When both n and N are present for a problem, we assume that $N = n^2$; the meaning is that the reductions are only for dense graphs in which case the input size is quadratic in n. For k-SAT, n denotes the number of variables. For the dynamic problems, different key problems can be reduced to different key problems, and the update/query time tradeoffs vary. References are not comprehensive.

Example 1: Indep Set \leq_p Clique

Indep Set: Given G=(V,E) and an integer k, is there $S \subseteq V$ s.t. $|S| \ge k$ and no two vertices in S are joined by an edge?

Clique: Given G=(V,E) and an integer k, is there $S \subseteq V$, $|U| \ge k$ s.t., every pair of vertices in S is joined by an edge?

Claim: Indep Set \leq_p Clique

Pf: Given G = (V, E) and instance of indep Set. Construct a new graph G' = (V, E') where $\{u, v\} \in E'$ if and only if $\{u, v\} \notin E$.



Example 2: Vertex Cover \leq_p Indep Set



Vertex Cover: Given G=(V,E) and an integer k, is there a vertex cover of size at most k?

Claim: For any graph G = (V, E), S is an independent set iff V - S is a vertex cover

Pf: =>

Let S be an independent set of G Then, S has at most one endpoint of every edge of G So, V - S has at least one endpoint of every edge of G So, V - S is a vertex cover.

 \leq Suppose *V* – *S* is a vertex cover

Then, there is no edge between vertices of S (otherwise, V - S is not a vertex cover)

So, *S* is an independent set.

Example 3: Vertex Cover \leq_p Set Cover

Set Cover: Given a set U, collection of subsets $S_1, ..., S_m$ of U and an integer k, is there a collection of k sets that contain all elements of U?

Claim: Vertex Cover \leq_p Set Cover Pf:

Given (G = (V, E), k) of vertex cover we construct a set cover input f(G, k)

- U = E
- For each $v \in V$ we create a set S_v of all edges connected to v

This clearly is a polynomial-time reduction

So, we need to prove it gives the right answer

Example 3: Vertex Cover \leq_p Set Cover

Claim: Vertex Cover \leq_p Set Cover Pf: Given (G = (V, E), k) of vertex cover we construct a set cover input f(G, k)

- U = E
- For each $v \in V$ we create a set S_v of all edges connected to v

Vertex-Cover (G,k) is yes => Set-Cover f(G,k) is yes

If a set $W \subseteq V$ covers all edges, just choose S_v for all $v \in W$, it covers all U.

Set-Cover f(G,k) is yes => Vertex-Cover (G,k) is yes If $(S_{v_1}, ..., S_{v_k})$ covers all U, the set $\{v_1, ..., v_k\}$ covers all edges of G.

Decision Problems

A decision problem is a computational problem where the answer is just yes/no

Here, we study computational complexity of decision Problems.

Why?

- Simpler to deal with
- Decision version is not harder than Search version, so it gives a lower bound for Decision version
- usually, you can use decider multiple times to find an answer.

Polynomial Time

Define P (polynomial-time) to be the set of all decision problems solvable by algorithms whose worst-case running time is bounded by some polynomial in the input size.

Do we understand P?

- We can prove that a problem is in P by exhibiting a polynomial time algorithm
- It is in most cases very hard to prove a problem is not in P.

Beyond P?

We have seen many problems that seem hard

- Independent Set
- 3-coloring
- Vertex Cover
- 3-SAT



Given a formula $(x_1 \lor \overline{x_2} \lor x_9) \land (\overline{x_2} \lor x_3 \lor x_7) \land \cdots$, is there a satisfying assignment?

Common Property: If the answer is yes, there is a "short" proof (a.k.a., certificate), that allows you to verify (in polynomial-time) that the answer is yes.

• The proof may be hard to find

Decision Problems

A decision problem is a computational problem where the answer is just yes/no.

We can define a problem by a set $X \subset \{0,1\}^n$. The answer for the input *s* is YES iff $s \in X$.

Certifier: Algorithm C(s, t) is a certifier for problem A if $s \in X$ if and only if (There is a *t* such that C(s, t) = YES))

NP: Set of all decision problems for which there exists a polytime certifier.

Co-NP: $X \in co - NP$ if and only if $\overline{X} \in NP$.

Example: 3SAT is in NP

Given a 3-CNF formula, is there a satisfying assignment? (conjunctive normal form (CNF) is AND of ORs)

Certificate: An assignment of truth values to the n boolean variables.

Verifier: Check that each clause has at least one true literal.

Ex:
$$(x_1 \lor \overline{x_3} \lor x_4) \land (x_2 \lor \overline{x_4} \lor x_3) \land (x_2 \lor \overline{x_1} \lor x_3)$$

Certificate: $x_1 = T, x_2 = F, x_3 = T, x_4 = F$

Conclusion: 3-SAT is in NP

Question: Is Maxflow is in NP?

Decision problem: Is the maximum flow value = k?

Answer 1:

Certificate: A flow f and a cut (S, \overline{S}) Verifier: Check if $val(f) = cap(S, \overline{S})$

Answer 2:

Certificate: None Verifier: Any polynomial time maxflow algo.

What do we know about NP?

- Nobody knows if all problems in NP can be done in polynomial time, i.e. does P=NP?
 - one of the most important open questions in all of science.
 - Huge practical implications specially if answer is yes
- Every problem in P is in NP one doesn't even need a certificate for problems in P so just ignore any hint you are given
- Every problem in NP is in exponential time
- Some problems in NP seem really hard
 - nobody knows how to prove that they are really hard to solve, i.e. P ≠ NP

NP Completeness

Complexity Theorists Approach: We don't know how to prove any problem in NP is hard. So, let's find hardest problems in NP.

NP-hard: A problem B is NP-hard iff for any problem $A \in NP$, we have $A \leq_p B$

NP-Completeness: A problem B is NP-complete iff B is NP-hard and $B \in NP$.

Motivations:

- If P ≠ NP, then every NP-Complete problems is not in P. So, we shouldn't try to design Polytime algorithms
- To show P = NP, it is enough to design a polynomial time algorithm for just one NP-complete problem.

Cook-Levin Theorem

Theorem (Cook 71, Levin 73): 3-SAT is NP-complete, i.e., for all problems $A \in NP$, $A \leq_p 3$ -SAT.

Pf (Draft. Take CSE 431 for more.):

Since $A \in NP$, there is a polytime certifier *C* such that $s \in A$ iff C(s, t) = 1 for some *t*

To solve the problem A, it suffices to find t.

Since C is polytime, we can

- convert *C* to a poly size circuit (of AND OR NOT)
- Some input are the given s.
- Some input are t.
- Our goal is to find *t* to make the output is TRUE.



Cook-Levin Theorem

Theorem (Cook 71, Levin 73): 3-SAT is NP-complete, i.e., for all problems $A \in NP$, $A \leq_p 3$ -SAT. Pf (Draft. Take CSE 431 for more.): To find an input such that output is true, we convert the circuit to 3CNF $(x_1 \lor \overline{x_2} \lor x_9) \land (\overline{x_2} \lor x_3 \lor x_7) \land \cdots$

Example:

- An OR gate with input a,b and output <u>c</u> can be represented by $(a \lor b \lor \overline{c}) \land (\overline{a} \lor c) \land (\overline{b} \lor c)$
- A NOT gate with input a and output c can be represented by $(a \lor c) \land (\overline{a} \lor \overline{c})$

Suppose the circuit gate C_1, C_2, \dots, C_q with final output Z

Then, the 3CNF is $\overline{C}_1 \wedge \overline{C}_2 \wedge \cdots \wedge \overline{C}_q \wedge Z$ where \overline{C}_i are the 3CNF version of C_i .

Cook-Levin Theorem

Theorem (Cook 71, Levin 73): 3-SAT is NP-complete, i.e., for all problems $A \in NP$, $A \leq_p 3$ -SAT.

• So, 3-SAT is the hardest problem in NP.

What does this say about other problems of interest? Like Independent set, Vertex Cover, ...

Fact: If $A \leq_p B$ and $B \leq_p C$ then, $A \leq_p C$ Pf idea: Just compose the reductions from A to B and B to C

So, if we prove 3-SAT \leq_p Independent set, then Independent Set, Clique, Vertex cover, Set cover are all NP-complete 3-SAT \leq_p Independent Set \leq_p Vertex Cover \leq_p Set Cover

$3\text{-SAT} \leq_p \text{Independent Set}$

Map a 3-CNF to (G,k). Say m is number of clauses

- Create a vertex for each literal
- Joint two literals if
 - They belong to the same clause (blue edges)
 - The literals are negations, e.g., x_i , $\overline{x_i}$ (red edges)
- Set k be the # of clauses.

$$(x_1 \lor \overline{x_3} \lor x_4) \land (x_2 \lor \overline{x_4} \lor x_3) \land (x_2 \lor \overline{x_1} \lor x_3)$$



Correctness of 3-SAT \leq_p Indep Set

<u>F satisfiable => An independent of size k</u>

Given a satisfying assignment, Choose one node from each clause where the literal is satisfied

 $(x_1 \lor \overline{x_3} \lor x_4) \land (x_2 \lor \overline{x_4} \lor x_3) \land (x_2 \lor \overline{x_1} \lor x_3)$

Satisfying assignment: $x_1 = T$, $x_2 = F$, $x_3 = T$, $x_4 = F$



- S has exactly one node per clause => No blue edges between S
- S follows a truth-assignment => No red edges between S
- S has one node per clause => |S|=k

Correctness of 3-SAT \leq_p Indep Set

An independent set of size $k \Rightarrow A$ satisfying assignment Given an independent set S of size k. S has exactly one vertex per clause (because of blue edges) S does not have $x_i, \overline{x_i}$ (because of red edges) So, S gives a satisfying assignment



Satisfying assignment: $x_1 = F$, $x_2 = ?$, $x_3 = T$, $x_4 = T$ $(x_1 \lor \overline{x_3} \lor x_4) \land (x_2 \lor \overline{x_4} \lor x_3) \land (x_2 \lor \overline{x_1} \lor x_3)$

Summary

- If a problem is NP-hard it does not mean that all instances are hared, e.g., Vertex-cover has a polynomial-time algorithm in trees
- We learned the crucial idea of polynomial-time reduction. This can be even used in algorithm design, e.g., we know how to solve max-flow so we reduce image segmentation to max-flow
- NP-Complete problems are the hardest problem in NP
- NP-hard problems may not necessarily belong to NP.
- Polynomial-time reductions are transitive relations

