Lecture 21: NP-Completeness I

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Introduction

Last few weeks: slower and slower algorithms for harder and harder problems

- From O(m + n) time algorithms for BFS/DFS/topological sort/SCCs, to O(m²n) for max flow
- Today: start of two lectures on NP-completeness: the (or at least one) line between tractability and intractability

Definition

An algorithm runs in *polynomial time* if its (worst-case) running time is $O(n^c)$ for some constant $c \ge 0$, where n is the size of the input.

Think of polynomial time as "fast", super-polynomial time as "slow"

Question: When do polynomial-time algorithms exist?

Decision Problems

Definition

A *decision problem* is a computational problem in which the output is either YES or NO.

Examples:

- Max-Flow: Input is G = (V, E), c: E → ℝ_{≥0}, s, t ∈ V, k ∈ ℝ⁺. Output YES if there is an (s, t)-flow of value at least k, otherwise output NO.
- Shortest s t path: Input is G = (V, E), ℓ: E → ℝ, s, t ∈ V, k ∈ ℝ. Output YES if d(s, t) ≤ k, otherwise output NO.

Some problems naturally decision, others naturally optimization, but can turn any optimization problem into a decision problem.

If can solve decision, can almost always solve optimization.

Note: Can divide instances (inputs) of any decision problem into YES-instances and NO-instances

Definition

P is the set of decision problems that can be solved in polynomial time.

Note: problems are in P, not algorithms

Question: Are all problems in **P**? **Answer:** No!

- By time hierarchy theorem there are problems that require super-polynomial time!
- Undecidability: there are problems which cannot be solved by any algorithm at all!

Verification

Different Setting: If in addition to the input we're given a purported solution, can we check that this solution is valid/feasible (in polynomial time)?

• Max-Flow: given $f: E \to \mathbb{R}_{\geq 0}$, check that value $\geq k$, flow conservation at all nodes other than s, t, and capacity constraints obeyed

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Definition (3-Coloring)
Input: Undirected graph G = (V, E)
Output: YES if \exists coloring f : V \rightarrow \{R, G, B\} such that f(u) \neq f(v) for all \{u, v\} \in E. NO otherwise
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Verification: Given **f**,

- Check that $f(u) \in \{R, G, B\}$ for all $u \in V$, and
- Check each edge $\{u, v\}$ to make sure that $f(u) \neq f(v)$

NP

NP: decision problems where solutions can be *verified* in polynomial time.

Definition

A decision problem **Q** is in **NP** (*nondeterministic polynomial time*) if there exists a polynomial time algorithm V(I, X) (called the *verifier*) such that

- If I is a YES-instance of Q, then there is some X (usually called the *witness*, *proof*, or *solution*) with size polynomial in |I| so that V(I,X) = YES.
- 2. If I is a NO-instance of Q, then V(I, X) = NO for all X.

Examples:

- ▶ 3-coloring: Witness X is a coloring $f: V \rightarrow \{R, B, G\}$, verifier checks each edge $\{u, v\}$ to make sure $f(u) \neq f(v)$
 - If I is a YES instance, then there is a coloring so verifier will return YES
 - If I is a NO instance, then no valid coloring exists. Whatever X is, verifier returns NO.

Max-Flow: Witness **X** is a flow $\mathbf{f} : \mathbf{E} \to \mathbb{R}_{\geq 0}$, verifier checks that it's feasible of value $\geq \mathbf{k}$

If I is a YES instance, then there is a feasible flow of value at least k so verifier (on this flow) will return YES

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P vs NP

Theorem		
P ⊆ NP		

Proof.

Let $\mathbf{Q} \in \mathbf{P}$. $\mathbf{V}(\mathbf{I}, \mathbf{X})$: Ignore \mathbf{X} , solve on instance \mathbf{I} .

Question: Does P = NP, i.e., is $NP \subseteq P$?

- Almost everyone thinks no, but we don't know for sure!
- Not even particularly close to a proof.
- Think about what P = NP would mean...

Reductions

Question: How could we prove that P = NP or $P \neq NP$?

- **P** = **NP**: Need to show that *every* problem in **NP** is also in **P**!
- P ≠ NP: Need to prove that some problem in NP not in P. What is the "hardest" problem in NP?

Definition

Problem **A** is *polytime reducible* to problem **B** (written $A \leq_p B$) if, given a polynomial-time algorithm for **B**, we can use it to produce a polynomial-time algorithm for **A**.

Means that B is "at least as hard" as A: if B is in P, then so is A.

• So "hardest" problems in **NP** are problems that many other problems reduce to.

Many-One (Karp) Reductions

Almost always (and always in this course), use a special type of reduction.

Definition

A *Many-one* or *Karp* reduction from A to B is a function f which takes arbitrary instances of A and transforms them into instances of B so that

- 1. If x is a YES-instance of A then f(x) is a YES-instance of B.
- 2. If x is a NO-instance of A then f(x) is a NO-instance B.
- 3. f can be computed in polynomial time.

So given instance x of A, compute f(x) and use polytime algorithm for B on f(x)

- ${\scriptstyle \blacktriangleright}$ Polytime, since f in polytime and algorithm for B in polytime
- Correct by first two properties of many-one reduction.

NP-Completeness

So what is "hardest problem" in NP?

Definition

Problem **Q** is **NP**-hard if $\mathbf{Q}' \leq_{\mathbf{p}} \mathbf{Q}$ for all problems \mathbf{Q}' in **NP**.

Definition

Problem **Q** is **NP**-complete if it is **NP**-hard and in **NP**.

So suppose **Q** is **NP**-complete.

- ► To prove P ≠ NP: Hardest problem in NP! If anything in NP is not in P, then Q is not in P
- To prove P = NP: Just need to prove that $Q \in P$.

Is anything NP-complete?

Circuit-SAT

Definition

Circuit-SAT: Given a boolean circuit with a single output and no loops (some inputs might be hardwired), is there a way of setting the inputs so that the output of the circuit is **1**?



Circuit-SAT

Theorem

Circuit-SAT is **NP**-complete.

Sketch of proof here. See book for details.

Lemma

Circuit-SAT is in NP.

Proof.

Witness is a T/F (or 1/0) assignment to inputs. Verifier simulates circuit on assignment, checks that it outputs **1**.

- If input is a YES instance then there is some assignment so circuit outputs 1. When verifier run on that assignment, returns YES.
- In input is a NO instance then in every assignment circuit outputs 0. So verifier returns NO on every witness.

Circuit-SAT is **NP**-hard

Let $A \in NP$. Want to show $A \leq_p$ Circuit-SAT (construct a many-one reduction).

Where to start? What do we know about A?

- In NP, so has verifier algorithm V
- V algorithm runs on a computer (or Turing machine)!

Computer: memory + circuit for modifying memory!



Not a boolean circuit in Circuit-SAT sense: loops (feedback)

Fix: "Unroll" circuit using fact that **V** runs in polynomial time

Reduction



Reduction: given instance I of A, construct this circuit for V, hardwire I. Combined circuit f(I)

- Polytime since V runs in polytime
- If I YES of A: there is some X so that V(I, X) = YES
- $f_{\text{reality}} \xrightarrow{\text{fire } f} \implies \text{ some } X \text{ so that when } X \text{ input to } f(I),$ with outputs 1
 - \implies f(I) YES instance of Circuit-SAT.
 - If I NO of A: For every X, know that V(I, X) = NO

 \implies for every X, when X input to f(I), outputs

 \implies **f(I)** NO instance of Circuit-SAT

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