NP-completeness

CPSC 320 2024W1

Up to now...

- We've been learning techniques to design or analyze efficient algorithms
- Most problems we've seen can be solved in O(n), or O(n log n), or O(n²) or maybe O(n³) time
- Every problem has been "easy" to solve that is, has an algorithm that runs in O(n^k) time for some constant k

Now:

- We will look at problems that (we think) are hard to solve
- By this we mean: we don't know of any efficient algorithms for them
- Examples:
 - Find a way to schedule n exams in k time slots without any students having an exam conflict
 - Find the cheapest route for a traveling business person to visit n cities and come back to her starting point
 - Find k students in a class of size n that (already) all know one another
 - Divide a group of **n** people into two teams of equal total weight to play tug-of-war

Decision vs. optimization problems

- Two types of problems:
 - Optimization problem: we want to find the solution s that maximizes or minimizes a function f(s)
 - Decision problem: given a parameter k, we want to know if there exists a solution s for which f(s) ≥ k (maximization) or f(s) ≤ k (minimization)
- The two are almost equivalent

Decision vs. optimization problems

Example: Independent Set

- An independent set in a graph G is a subset W of V such that no two vertices in W are joined by an edge in G
- Optimization problem: Given G = (V, E), find the largest independent set in G.

Decision problem: Given G = (V, E) and an integer k, does G have an independent set with at least k vertices?

Independent set in blue.

Image credit:

https://en.wikipedia.org/wiki/lndependent_set_(graph_theory)

Decision vs. optimization problems

- The two problems are similar in complexity:
 - If we have a solution to the optimization problem, then it gives us an answer to the decision problem
 - If we can solve the decision problem efficiently, then we can use binary search on k to find the answer to the optimization problem
- So we will look at decision problems only

P and NP

- We will consider two sets of problems:
 - P: the set of all problems for which we have an efficient solver:
 - Given a problem instance, the solver decides in polynomial time if the answer is Yes or No
 - NP: the set of all problems for which we have an efficient verifier:
 - For every problem instance whose answer is Yes, there is a "proof" that a verifier can check in polynomial time

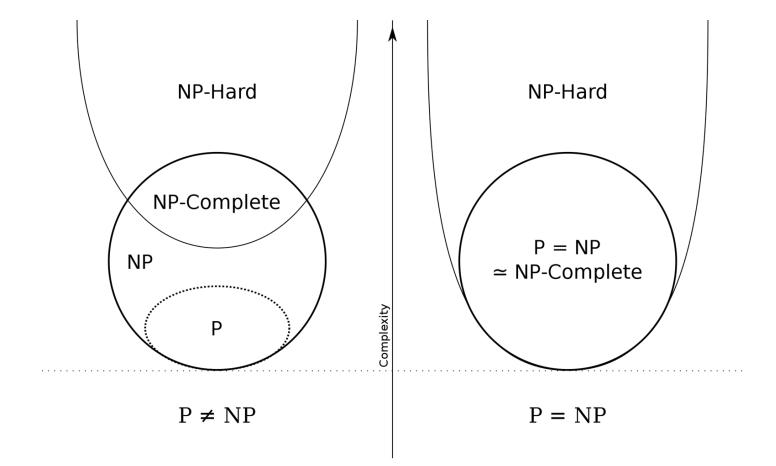
P and NP

Open question: Is P = NP?

- Nobody knows
- Most people think the answer is No.

P and NP

- We have:
 - Known easy problems (in P)
 - Problems that we think are hard (in NP, but maybe not P)'
- Cook's Theorem: if SAT can be solved in polynomial time, then every problem in NP can be solved in polynomial time
- A problem that belongs to NP and is as hard as SAT (more on what we mean by this later) is called NP-complete

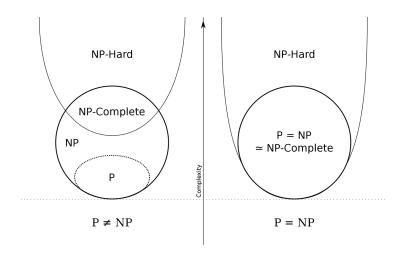


Familiar NP-complete problems

- We've already encountered some NP-complete problems:
 - SAT, 3-SAT (reductions slides), but not 2-SAT
 - Tour Grouping problem (A1 actually graph colouring in disguise)
 - Vertex cover, dominating set (midterm 1)
 - Subset sum (A4)

Proving a new problem is NP-complete

- Two steps for proving that a new problem *P* is NP-complete:
 - Prove that P belongs to NP
 - Prove that P belongs to NP-hard



Proving a new problem NP-complete, step 1

- To prove that P is in NP, we must prove that we can efficiently verify a certificate to the problem
 - Certificate is a "proof" of a Yes answer; a verifier is an algorithm that checks if a certificate is actually a correct proof
 - We can ask an optimization problem as: Does there exist an X such that Y holds?
 - Example: Does there exist a <u>subset of nodes in G</u> such that <u>no two of the</u> nodes share an edge and the size of the subset is at least k?
 - Certificate = whatever goes in <u>X</u> (in CPSC 320 vernacular: this is what a "valid solution" looks like)
 - Verifier = an algorithm that takes an X and checks whether Y holds (in CPSC 320 vernacular: this is an algorithm that checks if a "valid solution" is, in fact, a "good solution")

Proving a new problem NP-complete, step 1

- To prove that P is in NP, we must prove that we can efficiently verify a certificate to the problem
- If a certificate can be verified in **polynomial time**, then *P* is in NP
- Examples:
 - For independent set: the certificate is a subset of the vertices in G. The verifier checks that the subset has size at least k and that no two vertices in the subset share an edge (quadratic time)
 - For SAT: the certificate is an assignment of true/false values to variables.
 The verifier goes through the clauses and checks that each one contains a true literal (linear time)

Recall the Tour Group Problem (FGP) from assignment 1: given a set of students and a matrix K where K[i][j] = 1 if and only if student i knows student j, can we find a partitioning of students into three groups such that no two students in the same group know each other?

Which of the following is the **best** certificate for the friend group problem?

- A. A list of all the students
- B. A subset of the students
- C. A subset of the rows of *K*
- D. An assignment of each student to one of the three groups

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Proving a new problem NP-complete, part 2

- Next step: prove that P is at least as hard as any other problem in NP (i.e., that it's in NP-hard)
- We prove "at least as hard" via polynomial-time reduction:
 - Pick a known NP-complete problem P_{NPC}
 - \circ Give a polynomial-time algorithm that transforms an instance of P_{NPC} into an instance of P with the same Yes/No answer
- If you could solve *P* efficiently, then you could solve *P*_{NPC} as follows:
 - Transform it into an instance of P
 - Solve the instance of P and return the same answer
- Since P_{NPC} is NP-complete, this means you could solve every problem in NP in polynomial time!

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 - \circ Give a polynomial-time algorithm that transforms an instance of P_{NPC} into an instance of P with the same Yes/No answer
- **Intuition**: since we can use a solver for P to solve $P_{NPC.}$, this tells us that P is at least as hard as P_{NPC} (its solver is powerful enough to handle $P_{NPC.}$!)

On the midterm exam, we reduced Vertex Cover and Dominating Set to SAT (correctly and in polynomial time). SAT is known to be NP-complete. Does this mean Vertex Cover and Dominating Set are in NP-hard?

- A. Yes
- B. No

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- A. Yes
- B. No

Somewhat trick question: VC/DS *are* NP-hard, but the reduction to SAT doesn't prove it. We've also seen some **non-**NP-hard problems that we've reduced to SAT (e.g., bipartite graph problem).

To prove VC/DS are NP-hard, we would need to reduce **from SAT** (or another NP-complete problem) **to VC/DS**.

How do we solve NP-complete problems?

- Exactly solving an arbitrary instance will take exponential time. Some workarounds:
 - Approximation algorithms: these don't return an optimal solution, but we have a bound on how bad the solution is (e.g., using a minimum spanning tree to solve TSP gives a solution with no more than twice the optimal cost)
 - Randomization: get a faster running time by allowing the algorithm to fail with some probability
 - Restriction to certain problem types: sometimes certain subtypes of a problem are solvable in polynomial time (e.g., SAT is NP-complete, but we can still solve 2-SAT in poly-time)
 - Heuristics: algorithms (e.g., search-based) that often lack theoretical guarantees but work well enough in practice