

COMP20007 Design of Algorithms

Complexity Theory

Daniel Beck

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Complexity Theory

- So far, we have been concerned with the analysis of **algorithms'** running times.
- Complexity theory asks a different question: “What is the inherent difficulty of the **problem?**”
- It does however also uses asymptotic notation, although usually more concerned with lower bounds (Ω notation).
- Tighter (“larger”) lower bounds give us guarantees on best possible algorithms.

Complexity Theory - Examples

Comparison Sorting (worst case)

- A trivial lower bound: $\Omega(n)$.
- A less trivial lower bound: $\Omega(\log n!) \approx \Omega(n \log n)$
 - Can be found by using a technique called Decision Trees.
- Bound is **tight**: we know $\Theta(n \log n)$ algorithms.

Matrix Multiplication

- A trivial lower bound: $\Omega(n^2)$
- Unknown if bound is tight: best algorithms are $O(n^{2.37})$

This lecture: discussion about “hardness” of **problems**.

Decision Problems

- A **decision** problem takes an input and generates a **YES** or a **NO** answer as the output.
 - “Is the integer n a prime number?”
 - “Is there a circuit that visits all nodes in a graph exactly once?” (Hamiltonian Circuit Problem)
- Optimisation problems can be framed as a sequence of decision problems:
 - Knapsack: “Is there a set of items of values at least i and weight at most j ?”

Verification Problems

- A **verification** problem takes an input, a proposed solution and verifies if the solution satisfy the input.
 - “Given n and $\{i_1, i_2, \dots, i_k\}$, check if $\prod i = n$ ”
 - “Given a graph G and a sequence of nodes $\{v_1, v_2, \dots, v_n\}$, verify if there is a path that follows that sequence.

Now we can define what is P and what is NP .

P and NP

- A problem is in P if its decision version has a solution which is polynomial in the input size.
- A problem is in NP if its verification version has a solution which is polynomial in the input size.
- We can turn a verification problem into a decision problem:
 - 1) A **non-deterministic** “machine” generates a candidate.
 - 2) The verification algorithm verifies the solution.
 - 3) Repeat until verified.
- In other words, a problem is in NP if its decision version has a solution which is **non-deterministically polynomial** in the input size.

That's where the N in NP comes from. =)

P and NP

- Any problem in P is in NP ($P \subset NP$).
 - One can verify a solution by solving the decision version and comparing the result.
- The reverse is unknown... ($P \stackrel{?}{\supset} NP$)
 - The non-deterministic step does not guarantee efficiency.

$$P \stackrel{?}{=} NP$$

A Million Dollar Question: Is $P = NP$?

This is one of the seven “millennium problems”: The Clay Institute’s seven most important unsolved mathematical problems.



Reductions

- It's a daunting task to find and prove bounds for every new problem.
- **Reductions** allow us to ease this by, roughly, framing a problem as equivalent to another one we know the class.
- For instance, the Hamiltonian Circuit (HAM) problem can be reduced to the decision version of TSP.
- The **reduction function** is polynomial. Therefore, since HAM is in NP , the decision version of TSP is also in NP .

From HAM to TSP

- Suppose we have a Hamiltonian Circuit in a graph G with n nodes.
- Build a new graph G' where connected nodes in G have an edge of weight 0 and non-connected nodes have weight 1.
 - This can be done in polynomial time.
- Frame Decision-TSP as “Is there a circuit that visit all nodes only once with weight at most 0?”

NP-Completeness

A decision problem D is said to be **NP-Complete** if:

- $D \in NP$ and
 - Every problem in NP has a polynomial reduction to D **or**
 - A polynomial reduction from a known NP -complete problem to D exists.

Key property: if one finds a polynomial time algorithm to solve an NP -complete problem, then $P = NP$.

3-SAT

Proving that every problem in NP has a polynomial reduction to D is hard.

- This feat was accomplished in the 70's by Stephen Cook and Leonid Levin for the **Boolean 3-satisfiability problem**.
- “Given a boolean formula with a maximum of three literals, is there an assignment that results in TRUE?”
 - $(x_1 \vee \bar{x}_2 \vee \bar{x}_3) \wedge (\bar{x}_1 \vee x_2) \wedge (\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_3)$
 - $\{x_1 = \text{true}, x_2 = \text{true}, x_3 = \text{false}\}$

3-SAT

From 3-SAT, one can reduce it to many other problems, all being *NP*-complete as a consequence:

- SAT
- Clique
- Vertex Cover
- Hamiltonian Circuit
- Decision-TSP
- ...

A polynomial time algorithm for any of these would imply that $P = NP$.

Summary

- Complexity Theory deals with bounds for **problems**.
- Decision problems: P contains problems with polynomial time solutions.
- Verification problems: NP contains problems with polynomial time solutions.
- Reductions let us analyse new problems by framing them as existing ones.
- NP -completeness: solving one NP -complete problem implies in $P = NP$ due to reductions.

Last Words

- Some problems are **undecidable**: COMP30026
- Some scientists tried to prove that $P \stackrel{?}{=} NP$ is undecidable.
- Most scientists believe $P \neq NP$.
- While the problem itself still eludes computer scientists, proposed solutions led to advancements in theory, even though they were wrong.