

Lecture 19: NP and NP-completeness

Harvard SEAS - Fall 2024

2024-11-07

1 Announcements

- Salil OH 11-12pm; Anurag zoom OH Fri 1:30-2:30 pm
- Next SRE moved to *Thursday* 11/14.

Recommended Reading:

- MacCormick §12.0–12.3, Ch. 13

2 Polynomial-Time Reductions

Definition 2.1. For computational problems Π and Γ , we write $\Pi \leq_p \Gamma$ if there is a *polynomial-time* reduction R from Π to Γ . That is, there is a constant $c \geq 0$ such that R runs in time at most $O(N^c)$ on inputs of length N , counting oracle calls as one time step. Equivalently, there is a constant d such that $\Pi \leq_{O(N^d), O(N^d) \times O(N^d)} \Gamma$.

Some examples of polynomial-time reduction that we've seen include:

- 3-Coloring \leq_p SAT (Lecture 15)
- LongPath \leq_p SAT (SRE 5)
- IntervalScheduling-Decision \leq_p Sorting (Lecture 4). In this case a simpler polynomial time reduction is to

Using polynomial-time reductions to compare problems fits nicely with the study of the classes P_{search} and P , since they are “closed” under such reductions:

Lemma 2.2. *Let Π and Γ be computational problems such that $\Pi \leq_p \Gamma$. Then:*

1.

2.

Proof. 1.

2. Contrapositive of Item 1

□

This lemma means that we can use polynomial-time reductions both positively—to show that problems are in P_{search} —and negatively—to give evidence that problems are not in P_{search} . For example, under the *assumption* that 3-Coloring is not in P_{search} , it follows that SAT is not in P_{search} , by the above lemma and the fact that 3-Coloring \leq_p SAT (SRE5). As always, *the direction of the reduction is crucial!*

Another very useful feature of polynomial-time reductions is that they compose with each other:

Lemma 2.3. *If $\Pi \leq_p \Gamma$ and $\Gamma \leq_p \Theta$ then $\Pi \leq_p \Theta$.*

This follows from Problem 2 in Problem Set 2, and then using the definition of polynomial time reduction.

3 NP



Figure 1: Can you find a cat?

Roughly speaking, NP_{search} consists of the computational problems where valid outputs can be *verified* in polynomial time. This is a very natural requirement; what's the point in searching for something if we can't recognize when we've found it?

Definition 3.1. A computational problem $\Pi = (\mathcal{I}, \mathcal{O}, f)$ is in NP_{search} if the following conditions hold:

1. All valid outputs are of polynomial length:
2. All valid outputs are verifiable in polynomial time:

(Remark on terminology: $\text{NP}_{\text{search}}$ is often called **FNP** in the literature, and is closely related to, but slightly more restricted than, the class **PolyCheck** defined in the MacCormick text.)

Examples:

1. Satisfiability:
2. GraphColoring:
3. IndependentSet-ThresholdSearch:

Potential non-example:

1. IndependentSet-OptimizationSearch:

Even though this problem does not appear to be in $\text{NP}_{\text{search}}$ (its still an open question in the theory of computing!), it reduces in polynomial time to IndependentSet-ThresholdSearch, which is in $\text{NP}_{\text{search}}$ (to be discussed next week in the course).

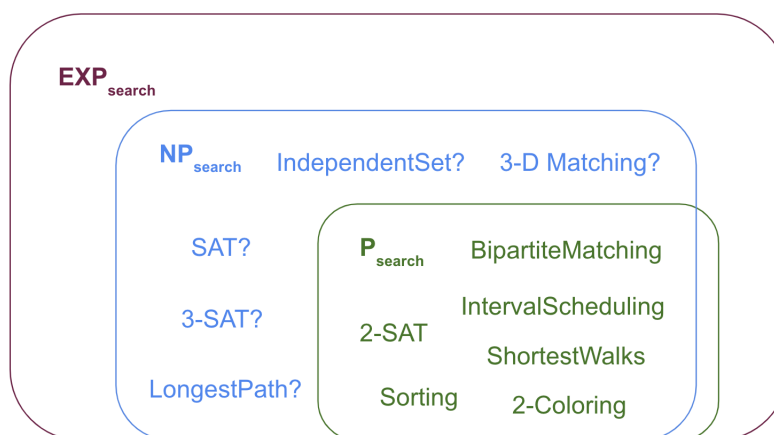
The following proposition shows that every problem in $\text{NP}_{\text{search}}$ can be solved in exponential time.

Proposition 3.2. $\text{NP}_{\text{search}} \subseteq \text{EXP}_{\text{search}}$.

Proof.

□

So now our diagram of complexity classes looks like this:



Remarks:

- P_{search} vs NP_{search} : Somewhat counterintuitively, $P_{search} \not\subseteq NP_{search}$. This due to artificial examples that you may see later in the course, but most of the natural problems in P_{search} are also in NP_{search} (like all of the green problems in the above diagram).
- **Class NP:** Every problem in NP_{search} has a corresponding decision problem (deciding whether or not there is a solution). The class of such decision problems is called **NP**. We will discuss the class **NP** more next week.

We still have question marks next to all of the blue problems; we don't know whether they (and thousands of other important problems in NP_{search}) are in P_{search} or not. We will now try to get a handle on these questions.

4 NP_{search} -Completeness

Unfortunately, although it is widely conjectured, we do not know how to prove that $NP_{search} \not\subseteq P_{search}$. As we will see next week, this is an equivalent formulation of the famous **P** vs. **NP** problem,

considered one of the most important open problems in computer science and mathematics. However, even without resolving the P vs. NP conjecture, we can give strong evidence that problems are not solvable in polynomial time by showing that they are $\text{NP}_{\text{search}}$ -complete:

Definition 4.1 (NP-completeness, search version). A problem Γ is $\text{NP}_{\text{search}}$ -complete if:

- 1.
- 2.

We can think of the NP-complete problems as the “hardest” problems in NP. Indeed:

Proposition 4.2. Suppose Γ is $\text{NP}_{\text{search}}$ -complete. Then $\Gamma \in \text{P}_{\text{search}}$ iff $\text{NP}_{\text{search}} \subseteq \text{P}_{\text{search}}$.

Proof. We first show that if $\Gamma \in \text{P}_{\text{search}}$, then $\text{NP}_{\text{search}} \subseteq \text{P}_{\text{search}}$. For every problem $\Pi \in \text{NP}_{\text{search}}$, we have that $\Pi \leq_p \Gamma$. Lemma 2.2 now ensures that $\Pi \in \text{P}_{\text{search}}$. Thus, $\text{NP}_{\text{search}} \subseteq \text{P}_{\text{search}}$.

On the other hand, if $\text{NP}_{\text{search}} \subseteq \text{P}_{\text{search}}$, then $\Gamma \in \text{P}_{\text{search}}$, using the fact that $\Gamma \in \text{NP}_{\text{search}}$. This completes the proof. \square

In other words, if any $\text{NP}_{\text{search}}$ -complete problem is in P_{search} , then all problems in $\text{NP}_{\text{search}}$ are in P_{search} . Remarkably, there are natural NP-complete problems. The first one is CNF-Satisfiability:

Theorem 4.3 (Cook–Levin Theorem). SAT is $\text{NP}_{\text{search}}$ -complete.

This can be interpreted as strong evidence that SAT is not solvable in polynomial time. If it were, then every problem in $\text{NP}_{\text{search}}$ would be solvable in polynomial time. We will return to a proof of the Cook–Levin Theorem later in the course.

5 More $\text{NP}_{\text{search}}$ -complete Problems

Once we have one $\text{NP}_{\text{search}}$ -complete problem, we can get others via reductions from it. Consider the computational problem 3-SAT, which is obtained when we restrict the number of literals in each clause of SAT.

Input	: A CNF formula φ on n variables z_0, \dots, z_{n-1} in which each clause has width at most 3 (i.e. contains at most 3 literals)
Output	: An $\alpha \in \{0, 1\}^n$ such that $\varphi(\alpha) = 1$ (if one exists)

Computational Problem 3-SAT

Theorem 5.1. 3-SAT is $\text{NP}_{\text{search}}$ -complete.

Proof. The full proof is deferred to Lecture 20. The proof follows in two steps.

1. 3SAT is in $\text{NP}_{\text{search}}$:

2. 3SAT is $\text{NP}_{\text{search}}$ -hard: Since every problem in $\text{NP}_{\text{search}}$ reduces to SAT (Theorem 4.3), all we need to show is $\text{SAT} \leq_p 3\text{SAT}$ (since reductions compose - Lemma 2.3).

The reduction algorithm from SAT to 3SAT has the following components (Figure 2). First, we give an algorithm R which takes a SAT instance φ to a 3SAT instance φ' .

$$\text{SAT instance } \varphi \xrightarrow{\text{polytime R}} \text{3SAT instance } \varphi'$$

Then we feed the instance φ' to our 3SAT oracle and obtain a satisfying assignment β to φ' or \perp if none exists. If we get \perp from the oracle, we return \perp , else we transform β into a satisfying assignment to φ using another algorithm S.

$$\text{SAT assignment } \alpha \xleftarrow{\text{polytime S}} \text{3SAT assignment } \beta$$

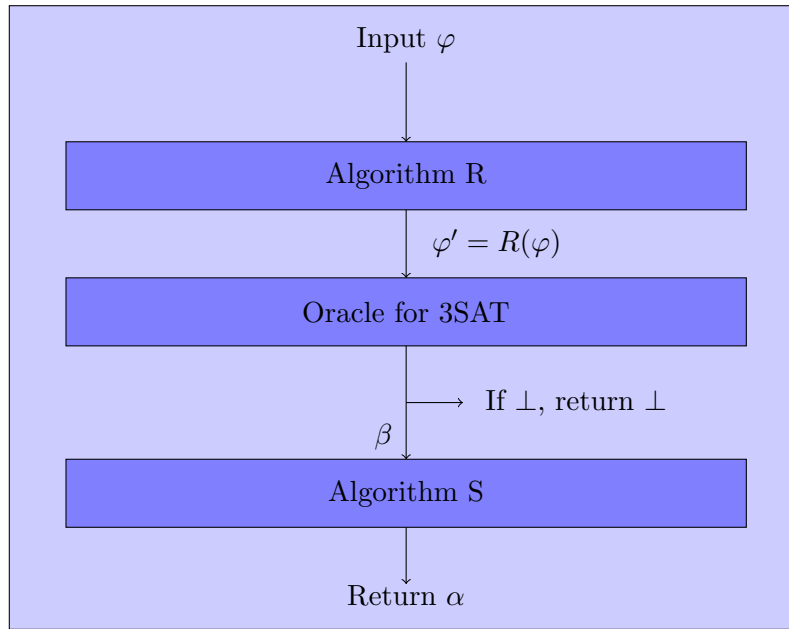


Figure 2: Reduction algorithm from SAT to 3SAT.

□