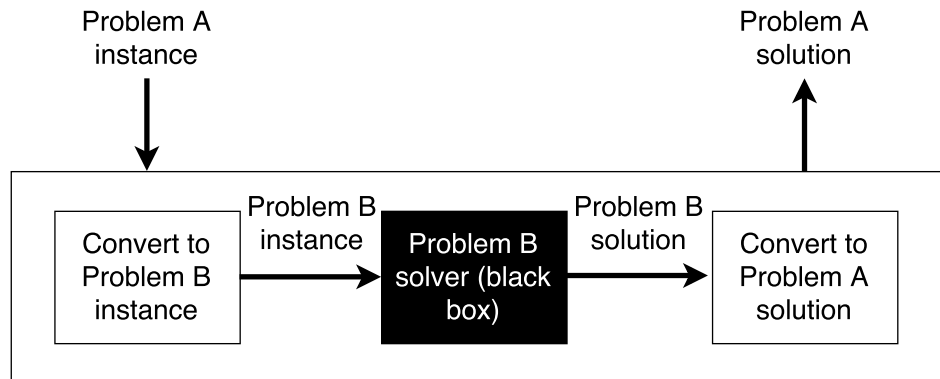


## CPSC 320: NP-completeness, SAT and 3SAT\*

We often use reductions to solve new problems based on problems we can already solve. For example, in an earlier worksheet, we saw a reduction from the Resident Hospital problem to the Stable Matching problem.



But... there's another way to use reductions. A more **sinister** way.<sup>1</sup> We'll illustrate this using a famous problem from logic: Satisfiability.

### 1 Boolean Satisfiability

Boolean satisfiability (SAT) is—as far as Computer Scientists know—a hard problem, in the sense that no-one knows of an algorithm to solve SAT that has worst-case polynomial runtime. In more technical terms, SAT is *NP-complete*. Recall that this problem is defined by:

The input is a collection of  $m$  clauses over  $n$  boolean variables  $X_1, X_2, \dots, X_n$ . Each clause is a disjunction of some of the variables or their complements.

The problem consists in answering the question “Is there a way to assign truth values to each variable that makes **every** clause of the instance TRUE?”

If there is an assignment of truth values that makes every clause of a SAT instance TRUE, then we call the instance *satisfiable*. Here is a sample instance of SAT:

$$\begin{aligned} &X_1 \vee \overline{X}_2 \vee X_3 \vee X_4 \\ &X_5 \\ &\overline{X}_1 \\ &X_2 \vee \overline{X}_3 \vee \overline{X}_5 \\ &\overline{X}_2 \vee X_3 \end{aligned}$$

For convenience, we will insist on using the variables  $X_1, X_2, \dots, X_n$  for some  $n$ , without skipping any. Given an instance  $I$  of SAT, we want to know: is the instance  $I$  satisfiable or not? The answer is either Yes or No, so we call this a *decision problem*.

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<sup>1</sup>Well, OK. Just **another** way.

1. Is the example SAT instance above satisfiable? If not, explain why not. If so, prove it by giving an assignment that makes the statement true.

2. For a SAT instance  $I$ , a truth assignment is a potential solution, and the truth assignment is a *good* solution if it satisfies instance  $I$ .

Suppose in addition to instance  $I$  you were given a truth assignment, say represented as an array  $T[1..n]$  where  $T[i]$  is true if and only if  $X_i$  is set to true. How long would it take to certify that truth assignment  $T$  is good?

3. A brute force algorithm could make a list of the variables  $X_1, \dots, X_n$  in the problem, try every assignment of truth values to these variables, and return **YES** if any satisfies the expression or **NO** otherwise. Asymptotically, how many truth assignments might this algorithm try (in terms of  $n$ )?

## 2 3-SAT and SAT

The 3-SAT problem is just like SAT, except that **every** clause must be **exactly** of length 3. Because it's a bit more restricted than SAT, 3SAT is a better problem to use for NP-completeness reductions (recall that we will reduce **from** 3SAT **to** the problem we want to prove is NP-complete). Let's build a reduction from SAT to 3-SAT (so we're solving SAT in terms of 3-SAT). We'll map an instance  $I$  of SAT to an instance  $I'$  of 3-SAT, working on one clause at a time. Importantly, for our reduction to work,  $I$  should be satisfiable if and only if  $I'$  is satisfiable.

1. Suppose that  $I$  has a clause with two literal, say  $(\bar{X}_2 \vee X_3)$ . To obtain  $I'$  from  $I$ , we want to replace this clause by one or more clauses, while ensuring that  $I$  is satisfiable if and only if  $I'$  is. How can we do this? Hint: introduce a new variable  $Y$ .

2. What if  $I$  has a clause with only one literal, say  $(X_5)$ ? Hint: introduce two new variables  $Y$  and  $Z$ .
3. Now suppose that  $I$  has a clause with four literals, say  $(X_1 \vee \overline{X_2} \vee X_3 \vee X_4)$ . What 3-SAT clauses will you put in  $I'$  to replace this clause, so that  $I'$  is satisfiable if and only if  $I$  is? Hints: Break the clause up somehow. Don't try using de Morgan's laws. Instead, create a brand new variable, say  $Y$ , and integrate that into your new clauses.
4. For your construction of part 3, show that if  $I$  is satisfiable then  $I'$  must also be satisfiable (and modify your construction if needed to ensure this).

5. Also for your construction of part 3, show that if  $I'$  is satisfiable then  $I$  must also be satisfiable.

6. Extend your 4-literal clause plan above to a 5-literal clause like  $(X_1 \vee \bar{X}_2 \vee X_3 \vee X_4 \vee \bar{X}_5)$ .

7. Show, by filling in the blanks below, how you would transform any clause with  $k > 3$  literals

$$(l_1 \vee l_2 \vee \dots \vee l_k)$$

into clauses with three literals (keeping in mind overall reduction correctness). You can use new variables that have not already been "used up", starting with  $X_{i+1}$  (where  $i \geq n$ ). How many clauses do you get? What would be the runtime of an algorithm to do this, as a function of  $k$ ?

$$\begin{aligned} & (l_1 \vee l_2 \vee X_{i+1}) \wedge (\bar{X}_{i+1} \vee l_3 \vee \text{_____}) \wedge (\bar{X}_{i+2} \vee l_4 \vee X_{i+3}) \wedge \dots \\ & \dots \wedge (\bar{X}_{i+(j-2)} \vee l_j \vee X_{i+(j-1)}) \wedge \dots \\ & \wedge (\bar{X}_{i+(k-4)} \vee l_{k-2} \vee \text{_____}) \wedge (\bar{X}_{i+(k-3)} \vee l_{k-1} \vee \text{_____}). \end{aligned}$$

8. Let's use the name TRANSFORM-CLAUSE to refer to the algorithm for transforming a clause, as described in part 7. Suppose that  $I'$  is obtained from  $I$  by transforming clause  $(l_1 \vee l_2 \dots \vee l_k)$  using algorithm TRANSFORM-CLAUSE. Explain why  $I$  is satisfiable if and only if  $I'$  is satisfiable.

9. Give a reduction from SAT to 3-SAT. Recall that a reduction consists of two algorithms that "connect" one problem to another, as in the diagram from page 1.

**Transform instance algorithm:**

**Transform solution algorithm:**

10. Why is the reduction correct?

### 3 What does a reduction tell us?

Here, consider a reduction from problem A to problem B, as illustrated in the figure of page 1.

1. **SCENARIO #1 (how we've used reductions prior to this worksheet):** Say our reduction's two algorithms take  $O(f(n))$  time and the black box solver for B also takes  $O(f(n))$  time. What can we say about the running time to solve problem A?
2. **SCENARIO #2 (what we usually think of NP-completeness as meaning):** Say our reduction's two algorithms take  $O(g(n))$  time and we know that there is **no algorithm** for problem A that runs in  $O(g(n))$  time. What do we know about the running time for problem B? Why?
3. **SCENARIO #3 (what NP-completeness technically means):** Say that we know (which we do) that if SAT can be solved in polynomial time, then **any** problem in the large set called "NP" can also be solved in polynomial time. What does our reduction from SAT to 3-SAT tell us? Why?