Corso di Laurea Magistrale in Matematica Applicata a.a. 2020-21

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Lecture 7: NP Completeness



#### Table of Contents

- 1 Introduction
- 2 Polynomial Time Reducibility
- NP-completeness
- **4** Summary

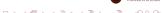




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- They found that certain problems in NP whose complexity is related to that of the entire class
- If a polynomial time algorithm exists for any of these problems, then all problems in NP would be solvable in polynomial time
- These problems are called NP-complete





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- In an attempt to prove that P = NP, one would "only" needs to find a polynomial time algorithm for an NP-complete problem
- Vice versa, trying to prove  $P \neq NP$  would require to find at least a problem that is in NP but not in P; in particular this is true for an *NP*-complete problem
- On the practical side, even if we still don't know if  $P \neq NP$ , showing that a problem is NP-complete is a strong evidence of its non-polynomiality (as most people indeed think  $P \neq NP$ )





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7 / 70



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7 / 70



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- Let's consider variables that can take on only two possible values: 1
   (TRUE) or 0 (FALSE)
- Those are normally called boolean or binary variables
- On those variables, we define **3 operators**:
  - AND: x ∧ y
  - OR: x ∨ y
  - NOT:  $\neg x$  (also denoted as  $\overline{x}$ )





Just to remind how those operators work:

$$0 \land 0 = 0$$
  $0 \lor 0 = 0$   $\overline{0} = 1$   
 $0 \land 1 = 0$   $0 \lor 1 = 1$   $\overline{1} = 0$   
 $1 \land 0 = 0$   $1 \lor 0 = 1$   
 $1 \land 1 = 1$   $1 \lor 1 = 1$ 





A boolean formula is an expression containing boolean variables and the operators above, e.g.,:

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- A boolean formula is satisfiable if some assignment of 0s and 1s to the variables makes the formula evaluate to 1 (i.e., TRUE)
- The boolean formula  $\phi$  of the example above is satisfiable because the assignment x = 0, y = 1, and z = 0 makes it TRUE





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#### Definition (The Satisfiability Problem)

The Satisfiability Problem is to test whether a given boolean formula  $\phi$  is satisfiable, namely whether it exists an assignment that satisfies it. More formally, we define:

$$SAT = \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable boolean formula} \}$$

We now introduce the **Cook-Levin theorem**, which links the complexity of *SAT* with that of **all** problems in *NP* 





#### The Cook-Levin Theorem

#### Theorem (Cook-Levin)

$$SAT \in P \Leftrightarrow P = NP$$





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We (re-)introduce a technique that is crucial to prove the Cook-Levin theorem



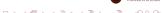


April 7, 2021

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- When problem A reduces to problem B ( $A \le B$ ), a solution to B can be used to solve A
- We now extend this idea by taking into account computational efficiency
- Intuitively, when a problem A is efficiently reducible to a problem B, an efficient solution to B can be used to solve A also efficiently





#### Definition (Polynomial Time Computable Function)

A function  $f: \Sigma^* \mapsto \Sigma^*$  is a **polynomial time computable function** if some polynomial time Turing machine M exists that halts with just f(x) on its tape, when it is given x as input





#### Definition (Polynomial Time Mapping Reduction)

Language A is **polynomial time mapping reducible** to language B (denoted by  $A \leq_P B$ ) if a polynomial time computable function  $f: \Sigma^* \mapsto \Sigma^*$  exists, such that for every x:

$$x \in A \Leftrightarrow f(x) \in B$$

The function f is called the **polynomial time reduction** of A to B



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- Polynomial time reducibility is the efficient-analog to mapping reducibility
- As with any other mapping reduction, a polynomial time reduction of A to B is a way to convert membership testing in A to that in B
- Big plus: the conversion is now done "efficiently" (i.e., in polynomial time)
- If one language is polynomial time reducible to another language B, which we already know a polynomial time solution for, then we obtain an overall polynomial time solution for A





#### Theorem (Polynomial Time Reducibility)

If  $A \leq_P B$  and  $B \in P$  then  $A \in P$ 





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#### Proof.

Let  $M_B$  be the polynomial time algorithm deciding B and f be the polynomial time reduction from A to B. We can describe a polynomial time algorithm  $M_A$  that decides A as follows:



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 $M_A$  = "On input x:

- **①** Compute f(x);
- **2** Run  $M_B$  on f(x) and output whatever  $M_B$  does."



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- We call a **literal** any boolean variable (x) or its negated  $(\overline{x})$
- A **clause** is several literals connected with  $\forall$ s, e.g.,  $(x_1 \lor \overline{x_2} \lor \overline{x_3} \lor x_4)$





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- We call a **literal** any boolean variable (x) or its negated  $(\overline{x})$
- A **clause** is several literals connected with  $\vee$ s, e.g.,  $(x_1 \vee \overline{x_2} \vee \overline{x_3} \vee x_4)$
- A boolean formula is in conjunctive normal form (CNF) if it comprises several clauses connected with \(\lambda\)s:

$$(x_1 \vee \overline{x_2} \vee \overline{x_3} \vee x_4) \wedge (x_3 \vee \overline{x_5} \vee x_6) \wedge (x_3 \vee \overline{x_6})$$





April 7, 2021

#### 3*SAT*

#### Definition (3SAT)

Let  $\phi$  be a 3-CNF boolean formula, i.e., a CNF boolean formula where each clause has exactly 3 literals, e.g.:

$$\phi = (x_1 \vee \overline{x_2} \vee \overline{x_3}) \wedge (x_3 \vee \overline{x_5} \vee x_6) \wedge (x_3 \vee \overline{x_6} \vee x_4) \wedge (x_4 \vee x_5 \vee x_6)$$





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We define 3SAT as the problem of testing whether a 3-CNF formula is satisfiable, i.e.:

$$3SAT = \{\langle \phi \rangle \mid \phi \text{ is a satisfiable 3-CNF formula}\}$$





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$$3SAT = \{\langle \phi \rangle \mid \phi \text{ is a satisfiable 3-CNF formula}\}$$

#### Note

In a satisfiable CNF formula, each clause must contain **at least one** literal whose assignment equals to 1 (TRUE)

# Polynomial Time Reducibility: $3SAT \leq_P CLIQUE$

Theorem (3*SAT*  $\leq_P$  *CLIQUE*)

3SAT is polynomial time reducible to CLIQUE





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#### Theorem (3*SAT* $\leq_P$ *CLIQUE*)

3SAT is polynomial time reducible to CLIQUE

#### Proof.

A sketch of the proof can be the following.

The polynomial time reduction f that we look for must convert 3-CNF boolean formulas to graphs. Graphs are constructed so as cliques of a specified size correspond to satisfying assignments of the formula. Structures within the graph ar designed to mimic the behavior of literals and clauses.





• Let  $\phi$  be a 3-CNF formula with k clauses as follows:

$$\phi = (a_1 \vee b_1 \vee c_1) \wedge (a_2 \vee b_2 \vee c_2) \wedge \ldots \wedge (a_k \vee b_k \vee c_k)$$





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- The nodes of G are organized into k groups of three nodes each, called **triples**:  $t_1, t_2, \ldots, t_k$





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- The reduction f must generate the string  $\langle G, k \rangle$ , i.e., the encoding of an undirected graph G
- The nodes of G are organized into k groups of three nodes each, called **triples**:  $t_1, t_2, \ldots, t_k$
- Each triple  $t_i$  represents a clause of the original formula  $\phi_i$  and each node in a triple is a literal of the associated clause





## $3SAT <_{P} CLIQUE$ : Proof

• The edges of G connect all but two types of pair of nodes





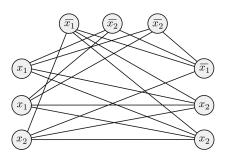
- The edges of G connect all but two types of pair of nodes
- In particular, no edge exists between nodes in the same triple and no edge is present between two nodes having contradictory labels





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$$\phi = (x_1 \vee x_1 \vee x_2) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_2}) \wedge (\overline{x_1} \vee x_2 \vee x_2)$$









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#### 3SAT < CLIQUE: Proof

We will show that  $\phi$  is satisfiable iff G has a k-clique

•  $(\Rightarrow)$  Suppose that  $\phi$  has a satisfying assignment





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- In each triple of *G* we select one node corresponding to a TRUE literal in the existing satisfying assignment
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- The nodes just selected form a k-clique!









Why do the selected nodes form a k-clique?

First of all, we select k nodes, i.e., one for each of the k triples





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- (ii) They cannot have contradictory labels because the associated literals must be both TRUE in the satisfying assignment





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- Therefore, G has a k-clique









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- Therefore, each of the k triples contains exactly one of the k nodes which the k-clique is made of





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- Thus, no two of the clique's nodes occur in the same triple because nodes in the same triples are not connected by construction
- Therefore, each of the k triples contains exactly one of the k nodes which the k-clique is made of
- $\bullet$  We assign truth values to the variables of  $\phi$  so that each literal labeling a clique node is set to TRUE









Why can we make such assignment?

• Two nodes labeled in a contradictory way are not connected and therefore cannot be part of the *k*-clique!





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- Hence, each clause contains at least a literal that is assigned TRUE





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- $\bullet$  Such assignment trivially satisfies  $\phi$  because each triple contains a clique node
- Hence, each clause contains at least a literal that is assigned TRUE
- Therefore,  $\phi$  is satisfiable!





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- Suppose we are given with a 3-CNF formula  $\phi$  having k clauses and m variables
- The graph G = (V, E) of  $\langle G, k \rangle$  we build has:
  - |V| = 3k nodes
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  - |V| = 3k nodes
  - $|E| < {3k \choose 2} = \frac{3k(3k-1)}{2} = O(k^2)$
- $\bullet$  The size of the graph  ${\it G}$  is polynomial in the size of the 3-CNF formula  $\phi$



 The last theorem we just proved tells us that, if CLIQUE is solvable in polynomial time so is 3SAT

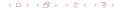




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- At first glance, this may sound odd since the two problems are indeed quite different
- Polynomial time reducibility allows us to link their complexities
- In the following, we show how to link the complexities of an entire class of problems





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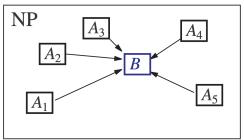




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NP-complete problems are the "most difficult" problems in NP





Polynomial Time Reducibility

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- $\mathbf{0}$   $B \in NP$
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#### Note

- NP-complete problems are the "most difficult" problems in NP
- If we omit first requirement (i.e.,  $B \in NP$ ), then we say that B is NP-hard





#### Theorem

If B is NP-complete and  $B \in P$  then P = NP





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From the definition above, if B is NP-complete it means that  $B \in NP$  and **every** other language/problem  $A \in NP$  is polynomially reducible to B.

Polynomial Time Reducibility

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#### Proof.

From the definition above, if B is NP-complete it means that  $B \in NP$  and **every** other language/problem  $A \in NP$  is polynomially reducible to B. Now, if we know that a solver for B exists and it runs in polynomial time (i.e.,  $B \in P$ ) then we can solve **every** other problem  $A \in NP$  by:

 $oldsymbol{0}$  applying the polynomial time reduction from A to B

Polynomial Time Reducibility

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- 2 running the polynomial time solver for B

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- 1 applying the polynomial time reduction from A to B
- 2 running the polynomial time solver for B

Since the process above is a composition of polynomial time algorithms and it holds for all  $A \in NP$ , we can state that  $\forall A \in NP, A \in P \Leftrightarrow P = NP.$ 

#### Theorem

If B is NP-complete and  $B \leq_P C$  for  $C \in NP$ , then C is NP-complete





Polynomial Time Reducibility

#### Theorem

If B is NP-complete and  $B \leq_P C$  for  $C \in NP$ , then C is NP-complete

#### Proof.

We know that  $C \in NP$ , so we must show that every other problem in NPis polynomial time reducible to C.

Polynomial Time Reducibility

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Polynomial Time Reducibility

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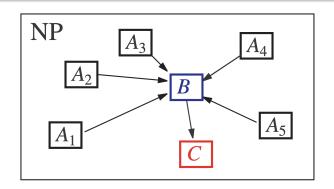
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Plus, we know from the hypothesis that  $B \leq_P C$ , and due to composition of polynomial time reductions  $A \leq_P B \ \forall A \in NP$  and  $B \leq_P C$  then  $A \leq_P C \ \forall A \in NP$ . Therefore C is NP-complete!



#### **Theorem**

If B is NP-complete and  $B \leq_P C$  for  $C \in NP$ , then C is NP-complete







Polynomial Time Reducibility

 Once we have one NP-complete problem, we may obtain others by polynomial time reduction from it





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- Once we have one NP-complete problem, we may obtain others by polynomial time reduction from it
- However, finding such first NP-complete problem is really hard!
- Historically, we do so by showing that our original problem of boolean satisfiability (SAT) is NP-complete





#### Theorem

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#### Proof Sketch.

In order to show that *SAT* is *NP*-complete we must prove that:

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Proving 1 is straightforward:

A polynomial time NTM can guess assignment to a boolean formula  $\phi$  and accept if that assignment satisfies  $\phi$ 





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In alternative, given  $\phi$  along with a certificate (i.e., an assignment) we can design a polynomial time verifier that checks if the assignment satisfies φ





### Proof Sketch.

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Proving 2 is harder!





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- Let N be polynomial time NTM that decides A in time at most  $n^k$ , where n = |w|





### Outline of the basic approach:

 $w \in A \Leftrightarrow \mathsf{NTM}\ N$  accepts input w

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#### Note

The basic intuition is to be able to show that any algorithm can be encoded as a boolean formula!





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"Satisfying assignment of  $\phi$ "  $\Leftrightarrow$  "Accepting computation history of N on input w"

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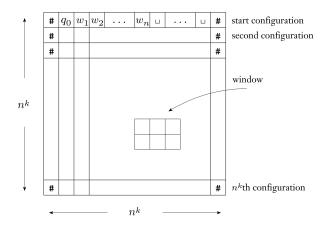




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- The problem of determining whether *N* accepts *w* is **equivalent** to finding if an accepting tableau for *N* on *w* exists

Polynomial Time Reducibility







Polynomial Time Reducibility

**Step 1:** Describe computations of NTM N on w by boolean variables using the tableau

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- $x_{i,j,s} = 1$  means "cell (i,j) contains s"





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#### Note

We design  $\phi$  so that a satisfying assignment to its variables  $x_{i,i,s}$ corresponds to an accepting tableau for N on w



**Step 2:** Express conditions for an accepting sequence of configurations of NTM N on w by a boolean formula  $\phi$  as the AND of four parts:

$$\phi = \phi_{\mathsf{cell}} \land \phi_{\mathsf{start}} \land \phi_{\mathsf{move}} \land \phi_{\mathsf{accept}}$$

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$$\phi_{\text{cell}} = \bigwedge_{1 \le i, j \le n^k} \left[ \left( \bigvee_{s \in C} x_{i,j,s} \right) \land \left( \bigwedge_{\substack{s,t \in C \\ s \ne t}} \left( \overline{x_{i,j,s}} \lor \overline{x_{i,j,t}} \right) \right) \right].$$





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Polynomial Time Reducibility

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$$\phi_{\text{move}} = \bigwedge_{1 \leq i < n^k, \ 1 < j < n^k} \text{(the } (i, j)\text{-window is legal)}.$$

$$\bigvee_{\substack{a_1,\dots,a_6 \text{ is a legal window}}} \left( x_{i,j-1,a_1} \wedge x_{i,j,a_2} \wedge x_{i,j+1,a_3} \wedge x_{i+1,j-1,a_4} \wedge x_{i+1,j,a_5} \wedge x_{i+1,j+1,a_6} \right)$$





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49 / 70



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• Given a non-deterministic Turing machine N and some input w we have shown that we can build a propositional formula  $\phi$ :

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- Rather than using SAT, typically a reduction is shown from one of its variant, i.e., 3SAT or 3-CNF formulas
- Before being able to do that, we need to show that 3SAT is also NP-complete





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- One way to show 2 is to prove that SAT polynomial time reduces to 3SAT





- To prove that 3SAT is NP-complete we must show that:
  - **1**  $3SAT \in NP$
  - $\mathbf{2} \ \forall A \in NP, A \leq_P 3SAT$
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- Instead, we slightly adapt the proof we used to show that SAT is NP-complete to achieve this





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$$\phi_{\text{cell}} = \bigwedge_{1 \le i, j \le n^k} \left[ \left( \bigvee_{s \in C} x_{i,j,s} \right) \land \left( \bigwedge_{\substack{s,t \in C \\ s \ne t}} (\overline{x_{i,j,s}} \lor \overline{x_{i,j,t}}) \right) \right].$$





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ullet Thus,  $\phi_{\mathsf{cell}}$  is an AND of **clauses**, therefore already in CNF





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- Let's see how each of  $\phi$ 's subformulas are organized
- $\phi_{\text{start}}$  is just a big AND of variables

$$\begin{split} \phi_{\text{start}} &= x_{1,1,\sharp} \wedge x_{1,2,q_0} \wedge \\ &\quad x_{1,3,w_1} \wedge x_{1,4,w_2} \wedge \ldots \wedge x_{1,n+2,w_n} \wedge \\ &\quad x_{1,n+3,\sqcup} \wedge \ldots \wedge x_{1,n^k-1,\sqcup} \wedge x_{1,n^k,\sharp} \,. \end{split}$$





Polynomial Time Reducibility

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- Each variable can be considered as a "degenerate" clause of size 1 with a single literal
- Therefore,  $\phi_{\text{start}}$  is also in CNF





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- It is a big AND of subformulas, each containing an OR of ANDs describing all the possible windows
- Using the distributive law, however, we can transform any OR of ANDs into an equivalent AND of ORs (i.e., CNF)





## Converting to CNF

 Every propositional formula can be converted into an equivalent formula that is in CNF





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**3** distributive law:

$$P \lor (Q \land R) \Leftrightarrow (P \lor Q) \land (P \lor R); P \land (Q \lor R) \Leftrightarrow (P \land Q) \lor (P \land R)$$





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- In each clause containing less than 3 literals, we just replicate one of the literals until getting a 3-literal clause
- In each clause that has more than 3 literals, we need to split them into multiple 3-literal clauses preserving the satisfiability





#### Example

Suppose our clause is made of 4 literals:

$$c = (a_1 \lor a_2 \lor a_3 \lor a_4)$$





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Here, z is a new variable (literal) and if some assignment of the  $a_i$ 's satisifies c we can also find a setting of z that satisfies c'.



#### Example

More generally, if the clause contains  $\ell$  literals:

$$c = (a_1 \vee a_2 \vee \ldots \vee a_\ell)$$

We can replace it with  $\ell-2$  clauses as follows:

$$c' = \left( a_1 \vee a_2 \vee z_1 \right) \wedge \left( \overline{z_1} \vee a_3 \vee z_2 \right) \wedge \left( \overline{z_2} \vee a_4 \vee z_3 \right) \ldots \wedge \left( \overline{z_{\ell-3}} \vee a_{\ell-1} \vee a_{\ell} \right) \right)$$





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3SAT is NP-complete!









## Proving *NP*-completeness: Summary

- Following the definition may be tedious as we need to show that:

  - **2** C is NP-hard, i.e.,  $\forall A \in NP, A \leq_P C$





## Proving *NP*-completeness: Summary

- Following the definition may be tedious as we need to show that:
  - $\mathbf{0}$   $C \in NP$  ("easy")
  - **2** C is NP-hard, i.e.,  $\forall A \in NP$ ,  $A \leq_P C$
- Recall that we proved that if B is NP-complete and  $B \leq_P C$  then C is NP-complete





## Proving *NP*-completeness: Summary

- We therefore need to show that:
  - $\mathbf{0} \ C \in NP$ 
    - ② a well-known NP-complete problem B polynomial time reduces to C (e.g.,  $SAT \leq_P C$  or  $3SAT \leq C$ )
    - the reduction actually takes polynomial time









This results follows directly from the previous findings

ullet We showed that  $\textit{CLIQUE} \in \textit{NP}$ 



64 / 70



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- We showed that 3SAT ≤<sub>P</sub> CLIQUE
- We showed that 3*SAT* is *NP*-complete
- Thus, CLIQUE is NP-complete!





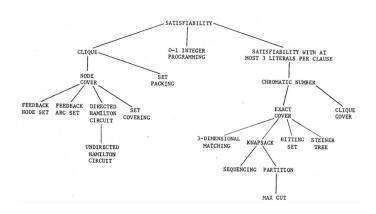


Figure: Karp's 21 NP-complete Problems





Decision problems have YES/NO answers





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## NP-hard Optimization Problems

- Decision problems have YES/NO answers
- Many decision problems have corresponding optimization version
- Optimization version of NP-complete problems are NP-hard

Problem	Decision Version	Optimization Version
CLIQUE	Does a graph $G$ have	Find largest clique
	a clique of size $k$ ?	
ILP	Does $\exists$ integer vector $y$	Find integer vector $y$ to
	such that $Ay \leq b$ ?	$\left  max \ d^{\top} y \right  \; s.t. \; Ay \leq b \left  \right $
TSP	Does a graph $G$ have tour	Find min length tour
	of length $\leq d$ ?	
Scheduling	Given set of tasks and constraints,	Find min time schedule
	can we finish all tasks in time $d$ ?	





## Why are NP-complete and NP-hard Important?

Polynomial Time Reducibility

 Suppose you are faced with a problem and you can't come up with an efficient algorithm for it





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- Suppose you are faced with a problem and you can't come up with an efficient algorithm for it
- If you can prove the problem is NP-complete or NP-hard, then there is no known efficient algorithm to solve it
  - No known polynomial-time algorithms for NP-complete and NP-hard problems!





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  - No known polynomial-time algorithms for NP-complete and NP-hard problems!
- How to deal with an NP-complete or NP-hard problem?
  - Approximation algorithm
  - Probabilistic algorithm
  - Special cases
  - Heuristic





### Table of Contents

- 1 Introduction
- Polynomial Time Reducibility
- NP-completeness
- 4 Summary





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- NP includes all problems that are in P plus, for example, HAMPATH, CLIQUE, SUBSET-SUM, 3SAT, etc.
- P vs. NP question:
  - **1** We know that  $P \subseteq NP$
  - **2** We **don't** know whether  $P \neq NP$  or P = NP





$$w \in A \Leftrightarrow f(w) \in B$$





• Polynomial-time mapping reducibility:  $A \leq_P B$  if exists polynomial-time computable function f such that:

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- If B is NP-complete and  $B \leq_P C$  for  $C \in NP$ , then C is NP-complete
- Cook-Levin Theorem: *SAT* is *NP*-complete
- 3SAT, CLIQUE, SUBSET-SUM, HAMPATH, etc. are all NP-complete (via polynomial time reduction)

