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

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Lecture 7: NP Completeness



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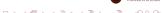




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- They found that certain problems in NP whose complexity is related to that of the entire class
- If a polynomial time algorithm exists for any of these problems, then all problems in NP would be solvable in polynomial time
- These problems are called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

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





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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 ϕ of the example above is satisfiable because the assignment x = 0, y = 1, and z = 0 makes it TRUE





Definition (The Satisfiability Problem)

The Satisfiability Problem is to test whether a given boolean formula ϕ is satisfiable, namely whether it exists an assignment that satisfies it.





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





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

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



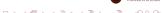


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- When problem A reduces to problem B ($A \le B$), a solution to B can be used to solve A
- We now extend this idea by taking into account computational efficiency
- Intuitively, when a problem A is efficiently reducible to a problem B, an efficient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

 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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- \bullet 3SAT assumes the formula ϕ which we must test the satisfiability of has a particular form
- 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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- \bullet 3SAT assumes the formula ϕ which we must test the satisfiability of has a particular form
- 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:

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





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3*SAT*

Definition (3SAT)

Let ϕ 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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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*)

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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 ϕ be a 3-CNF formula with k clauses as follows:

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





3SAT < P CLIQUE: Proof

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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 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 ϕ_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





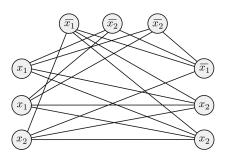
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$$\phi = (x_1 \vee x_1 \vee x_2) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_2}) \wedge (\overline{x_1} \vee x_2 \vee x_2)$$









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3SAT < CLIQUE: Proof

We will show that ϕ is satisfiable iff G has a k-clique

• (\Rightarrow) Suppose that ϕ has a 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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- (i) They cannot be part of the same triple because we select only one node per triple
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- (i) They cannot be part of the same triple because we select only one node per triple
- (ii) They cannot have contradictory labels because the associated literals must be both TRUE in the satisfying assignment
- 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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- (\Leftarrow) Suppose that G has a k-clique
- 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 ϕ 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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- \bullet Such assignment trivially satisfies ϕ because each triple contains a clique node





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





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- 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 ϕ



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

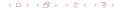




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- At first glance, this may sound odd since the two problems are indeed quite different
- Polynomial time reducibility allows us to link their complexities
- In the following, we show how to link the complexities of an entire class of problems





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A language *B* is *NP*-**complete** if it satisfies the following two conditions:





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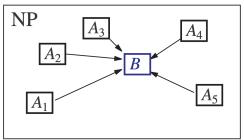




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NP-complete problems are the "most difficult" problems in NP





Polynomial Time Reducibility

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- $\mathbf{0}$ $B \in NP$
- **2** Every $A \in NP$ is polynomial time reducible to B

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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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.

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:

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

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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. 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:

- 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.

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Instead of proving this for every possible $A \in NP$, we can take advantage of knowing that B is NP-complete itself!

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Instead of proving this for every possible $A \in NP$, we can take advantage of knowing that B is NP-complete itself!

Because B is NP-complete, every other $A \in NP$ is polynomial time reducible to it, i.e., $A \leq_P B$, $\forall A \in NP$.

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

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Instead of proving this for every possible $A \in NP$, we can take advantage of knowing that B is NP-complete itself!

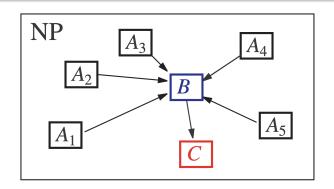
Because B is NP-complete, every other $A \in NP$ is polynomial time reducible to it, i.e., $A \leq_P B$, $\forall A \in NP$.

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

SAT is NP-complete





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SAT is NP-complete

Proof Sketch.

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

 \bullet SAT \in NP





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Proving 1 is straightforward:

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





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

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





Proof Sketch.

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

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Proving 2 is harder!





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- We need to show that A ≤_P SAT
- For every $w \in \Sigma^*$, we want a boolean formula ϕ such that:

 - g f is a polynomial time reduction
- 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 ϕ " \Leftrightarrow "Accepting computation history of N on input w"

• A computation of N (i.e., a list of configurations) on **some** branch of its computation tree is described by a $n^k \times n^k$ tableau





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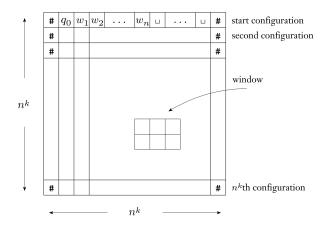




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- The problem of determining whether *N* accepts *w* is **equivalent** to finding if an accepting tableau for *N* on *w* exists

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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- For each $i, j \in \{1, ..., n^k\}$ and for each $s \in C$ we associate a boolean variable $x_{i,j,s}$ of ϕ
- $x_{i,j,s} = 1$ means "cell (i,j) contains s"





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Note

We design ϕ 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 ϕ as the AND of four parts:

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

1 ϕ_{cell} = "for each cell (i, j), there is **exactly one** $s \in C$ with $x_{i,i,s} = 1$ " (cell is well defined)





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- Φ_{accept} = "at least one row of tableau is an accepting configuration of N on w''



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





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- The accepting tableau must meet condition 1





$$\phi_{\text{start}} = x_{1,1,\#} \wedge x_{1,2,q_0} \wedge \\ x_{1,3,w_1} \wedge x_{1,4,w_2} \wedge \dots \wedge x_{1,n+2,w_n} \wedge \\ x_{1,n+3,\sqcup} \wedge \dots \wedge x_{1,n^k-1,\sqcup} \wedge x_{1,n^k,\#}.$$





Polynomial Time Reducibility

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- The accepting tableau must meet condition 2





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

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





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 This formula ensures that each 2x3 window is legal according to N's transition function (proof omitted here)





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- This formula ensures that an accepting configuration occurs in the tableau
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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 ϕ :

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- ϕ is satisfiable if and only if N accepts w
- The subformulas encode the 4 conditions needed there be an accepting tableau for the computation of N on input w
- ullet It only remains to show that the reduction from w to ϕ is computable in polynomial time



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- SAT in NP-complete!





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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





- To prove that 3SAT is NP-complete we must show that:
 - 1 3 $SAT \in NP$
 - \lozenge $\forall A \in NP, A \leq_P 3SAT$





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- To prove that 3SAT is NP-complete we must show that:
 - **1** $3SAT \in NP$
 - **2** \forall *A* ∈ *NP*, *A* ≤_{*P*} 3*SAT*
- Obviously, 3SAT ∈ NP!
- One way to show 2 is to prove that SAT polynomial time reduces to 3SAT
- Instead, we slightly adapt the proof we used to show that SAT is NP-complete to achieve this





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April 7, 2021

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- $\phi_{\rm cell}$ is made of a big AND of subformulas, each one containing a big OR and a big AND of ORs

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ullet Thus, ϕ_{cell} is an AND of **clauses**, therefore already in CNF





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





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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 ℓ 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 $\textit{CLIQUE} \in \textit{NP}$



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- We showed that CLIQUE ∈ NP
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- We showed that CLIQUE ∈ NP
- We showed that 3SAT ≤_P CLIQUE
- We showed that 3*SAT* is *NP*-complete
- Thus, CLIQUE is NP-complete!





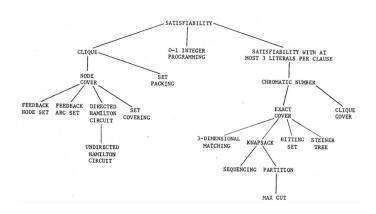


Figure: Karp's 21 NP-complete Problems





Decision problems have YES/NO answers





- Decision problems have YES/NO answers
- Many decision problems have corresponding optimization version





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





Why are NP-complete and NP-hard Important?

- 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!





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

- 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!
- How to deal with an NP-complete or NP-hard problem?
 - Approximation algorithm
 - Probabilistic algorithm
 - Special cases
 - Heuristic





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- Class P comprises problems that can be **decided** in polynomial time (e.g., $PATH \in P$)
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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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• A language B is NP-complete if $B \in NP$ and $A \leq_P B$ for all $A \in NP$





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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)

