

NP-Complete problems

NP-complete problems (NPC):

- A subset of NP.
- If *any* NP-complete problem can be solved in polynomial time, then *every* problem in NP has a polynomial time solution.

NP-complete languages are the “hardest” languages in NP.

Formal definition of NP-complete languages is based on the concept of **polynomial time reducibility**.

From P to NPC

- Examples of problems that belong to P:
 1. Find the *shortest* path between two vertices in a directed graph.
 2. Does a directed graph have an Euler tour. i.e. a cycle that visits all edges once?
 3. Is a Boolean formula in 2-conjunctive normal form satisfiable?

From P to NPC

- However, their slight variants are in NPC:
 1. Find the *longest* path between two vertices in a directed graph.
 2. Does a directed graph have a Hamiltonian cycle: a cycle that visits all vertices once?
 3. Is a Boolean formula in 3-conjunctive normal form satisfiable?

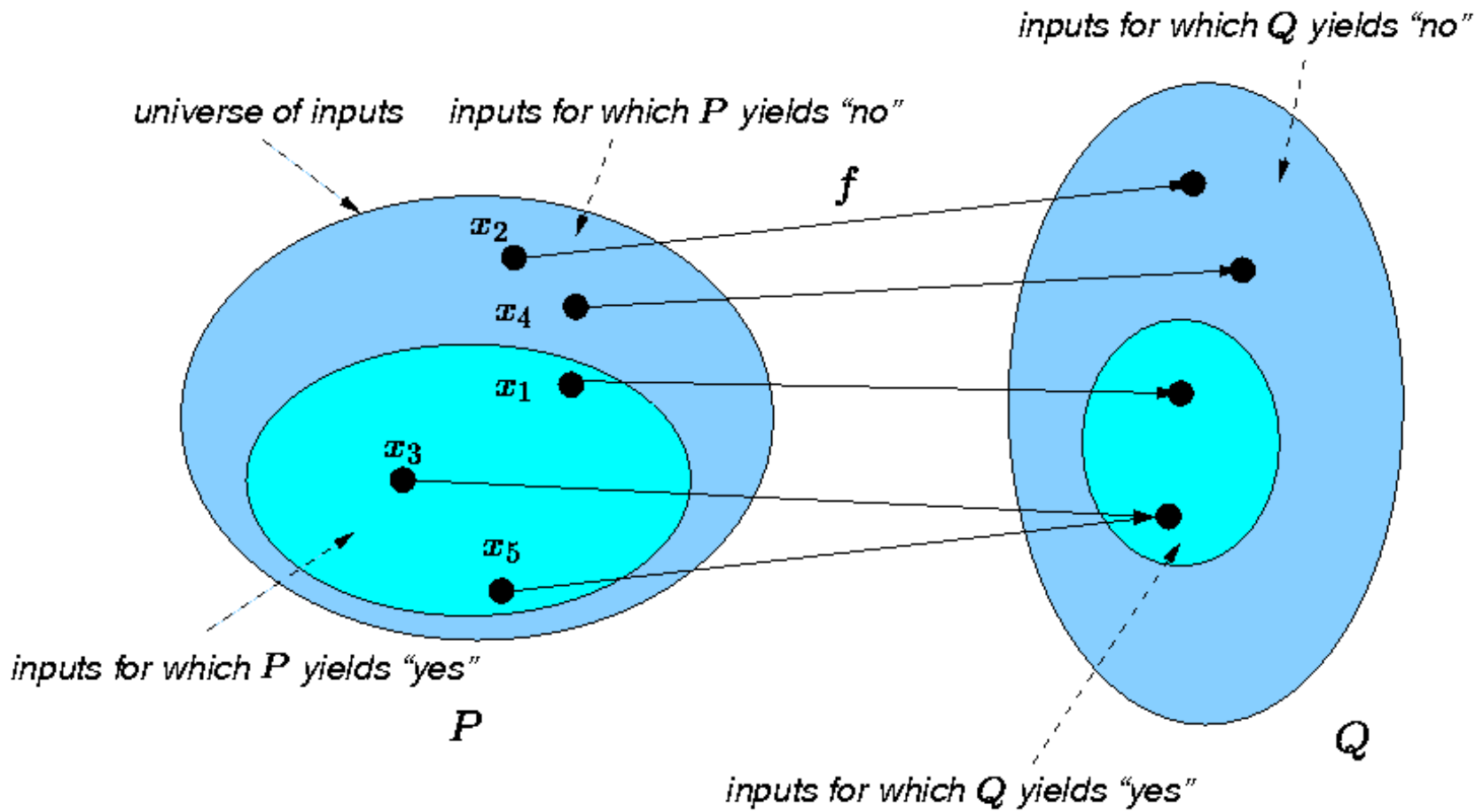
Polynomial Time Reducibility

Definition: A decision problem A is polynomial-time reducible to a decision problem B (written $A \leq_p B$) if:

- There exists a polynomial-time algorithm F that transforms any instance α of A into some instance $\beta = F(\alpha)$ of B ,
- The answer of A for α is “yes” iff the answer of B for β is “yes” .

Polynomial reductions

Polynomial reductions



A Formal Language framework: Reducibility

Every decision problem has a corresponding language = the maximal set of input strings that produce “yes” answers.

Let $L_A, L_B \subseteq \{0, 1\}^*$ be the languages corresponding to the two decision problems A and B, respectively.

Definition: L_A is **polynomial-time reducible** to L_B (written $L_A \leq_p L_B$) if:

- there exists a poly-time computable function

$$f: \{0, 1\}^* \rightarrow \{0, 1\}^*$$

- such that for all $\alpha \in \{0, 1\}^*$

$$\alpha \in L_A \text{ if and only if } f(\alpha) \in L_B.$$

Implication of $A \leq_p B$

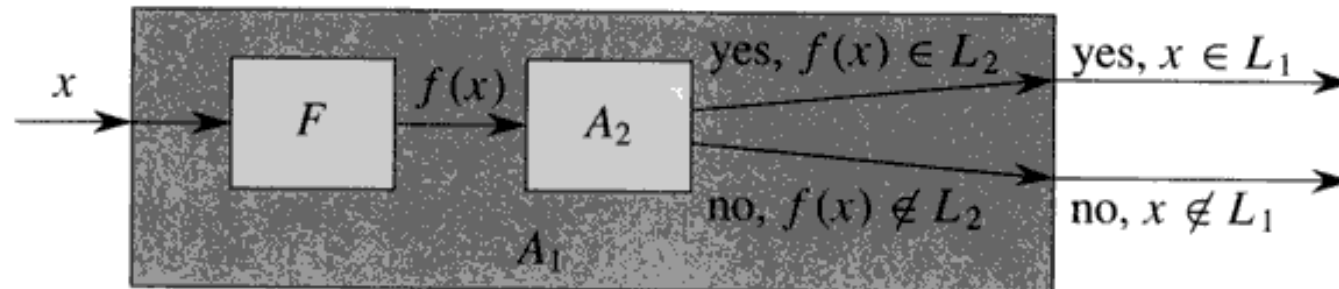
- f is called a **reduction function** and the poly-time algorithm F that computes f is called a **reduction algorithm**.
- We can use B to solve A :
 - Providing an answer to whether $f(\alpha) \in L_B$ directly provides the answer to whether $\alpha \in L_A$. Hence:
 - Solving A is no “harder” than solving B .

Implication of $A \leq_p B$

Lemma 34.3 If $L_1 \leq_p L_2$ and $L_2 \in P$, then $L_1 \in P$.

- Proof:

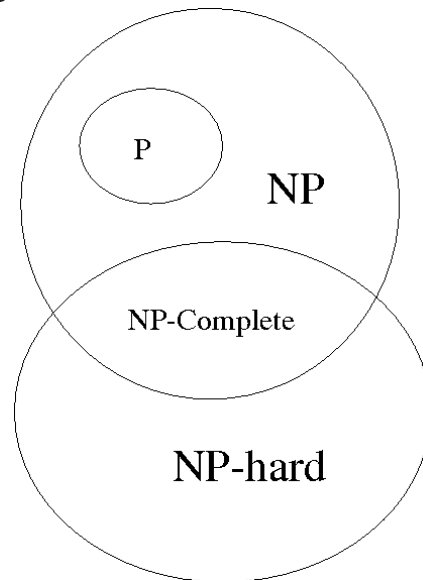
- Let A_2 be a poly-time algorithm that decides L_2 .
- Let F be a poly-time reduction algorithm that does the reduction.
- We construct a poly-time algorithm A_1 that decides L_1 .



NP-hard and NP-Complete

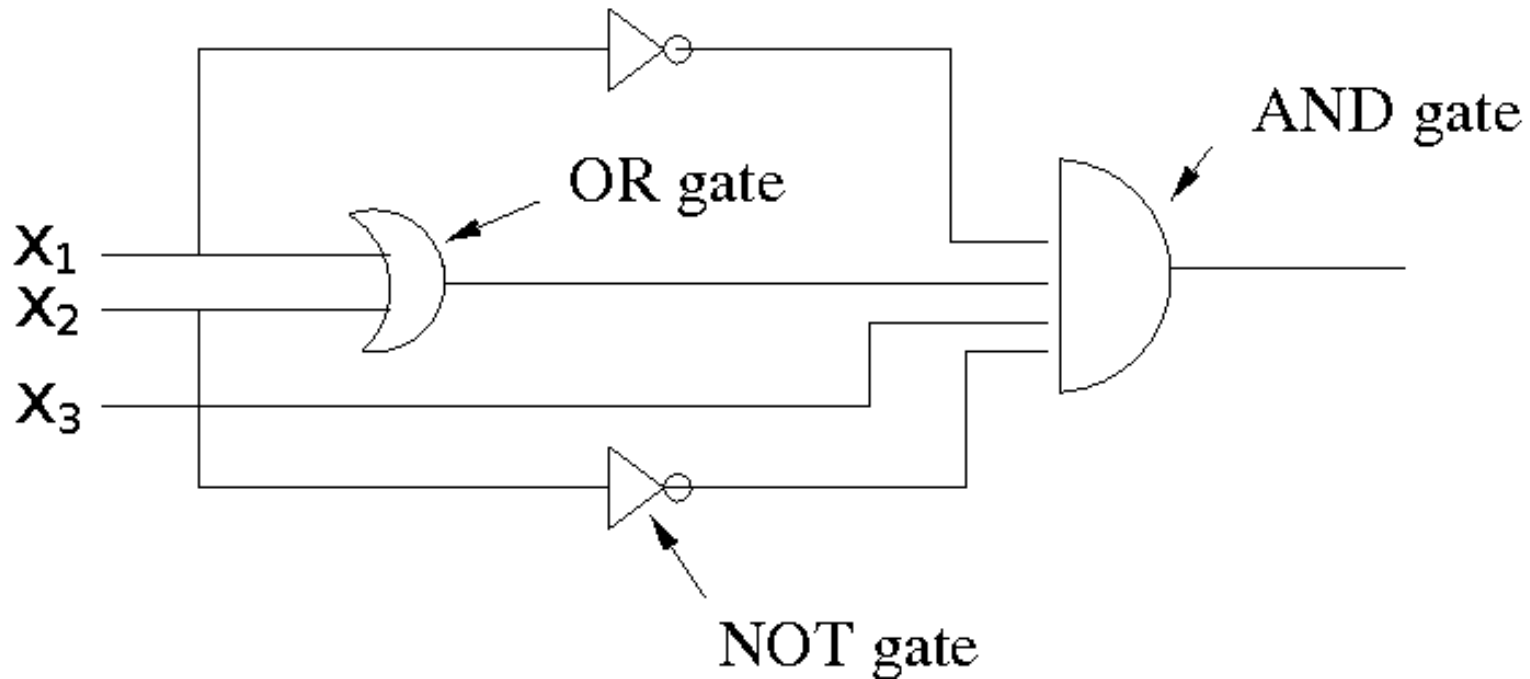
- **Definition:** L is NP-hard if $\forall L' \in \text{NP}, L' \leq_p L$.
- **Definition:** L is NP-Complete if:
 - 1) $L \in \text{NP}$.
 - 2) $L \in \text{NP-hard}$.

A very likely possibility:



Circuit Satisfiability problem

A Boolean combinational circuit



CKT-SAT problem

Decision problem

- Is there an assignment to the input that makes the circuit evaluate to TRUE?

$\text{CKT-SAT} = \{\langle \text{CKT} \rangle : \text{CKT has a satisfying assignment}\}.$

- What is the running time of a brute force algorithm?

CKT-SAT is NP-complete

- Lemma 34.5: CKT-SAT \in NP.
 - We can take the number of gates + wires as the size k of the circuit.
 - We can create a binary encoding $\langle \text{CKT} \rangle$ that is polynomial in k .
 - Certificate = an assignment of boolean values to the wires.
 - Checking whether the certificate corresponds to a satisfying assignment takes $O(k)$ time.
- Lemma 34.6: CKT-SAT \in NP-hard (pages 1074–1077).

Alternative definition of NP-completeness

Lemma 34.8: If L is a language such that $L' \leq_p L$ for some $L' \in \text{NPC}$, then L is NP-hard. Moreover, if $L \in \text{NP}$, then $L \in \text{NPC}$.

Proof: Since L' is NP-complete, for all $L'' \in \text{NP}$, we have $L'' \leq_p L'$. By supposition, $L' \leq_p L$, and thus by transitivity, we have $L'' \leq_p L$, which shows that L is NP-hard. If $L \in \text{NP}$, then we also have $L \in \text{NPC}$.

Transitivity: If $L_1 \leq_p L_2$ and $L_2 \leq_p L_3$, then $L_1 \leq_p L_3$ (*Exercise 34.3-2*).

$L \in \text{NPC}$: Generic Proof

- Step 1: prove $L \in \text{NP}$.
- Step 2: prove $L \in \text{NP-hard}$:
 1. Select a known NP-complete language L' .
 2. Find a reduction algorithm F , s.t. $x \in L' \Leftrightarrow F(x) \in L$.
 3. Prove that the algorithm F runs in poly-time.

Up to this point, the only NPC problem we know is CKT-SAT.

Another NPC problem: SAT

Formula satisfiability problem (SAT)

- A instance of SAT is a Boolean formula ϕ composed of
 - 1) n Boolean variables: x_1, x_2, \dots, x_n .
 - 2) m Boolean connectors: \wedge (AND), \vee (OR), \neg (NOT), \rightarrow (implication), \leftrightarrow (if and only if).
 - 3) parentheses.
- For example: $\phi = ((x_1 \rightarrow x_2) \wedge (\neg x_1 \vee x_2 \vee x_3)) \rightarrow (x_1 \wedge \neg x_2)$.
- $\text{SAT} = \{\langle \phi \rangle : \phi \text{ has a satisfying assignment (an assignment causes } \phi \text{ to evaluate to 1)}\}$.
- For example, $x_1 \vee x_2 \in \text{SAT}$, while $x_1 \wedge \neg x_1 \notin \text{SAT}$.

SAT is NP-Complete

Proof:

- Step 1: $\text{SAT} \in \text{NP}$.

Certificate is the “truth assignment”. Algorithm merely has to verify, in polynomial time, that the truth assignment produces TRUE.

- Step 2: $\text{SAT} \in \text{NP-hard}$.

by proving $\text{CKT-SAT} \leq_p \text{SAT}$.

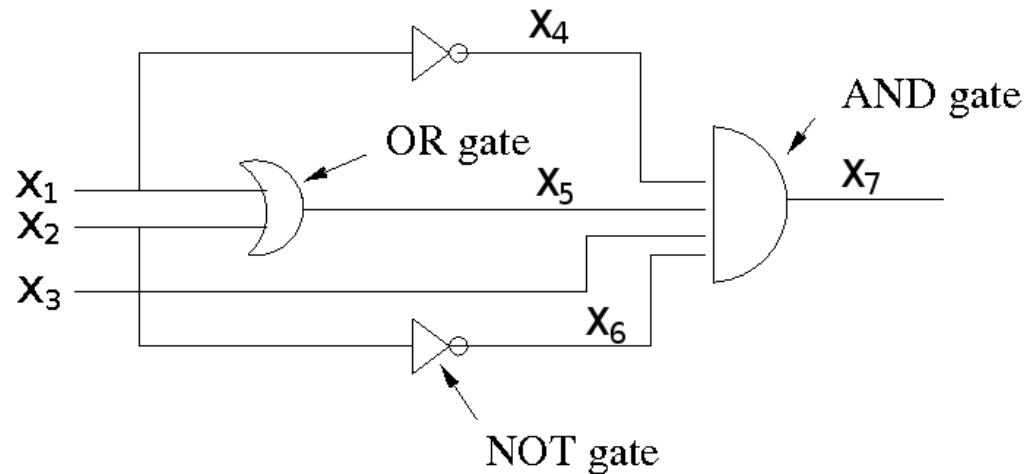
SAT: Poly-time reduction

The reduction is as follows:

- For each wire x_i in the circuit C , the formula ϕ has a variable x_i .
- For each gate in C , make a formula involving the variables of its incident wires that fully describes the behaviour of the gate.
 - For example, the operation of the output OR gate (figure on the next page) is $x_5 \leftrightarrow (x_1 \vee x_2)$.
- The formula ϕ produced by the reduction is the AND of the circuit-output variable with the conjunction of clauses describing the operation of each gate.

SAT: Poly-time reduction

A Boolean combinational circuit



- For the above circuit C , the formula is

$$\begin{aligned}\phi = & x_7 \wedge (\neg x_1 \leftrightarrow x_4) \\ & \wedge (x_5 \leftrightarrow (x_1 \vee x_2)) \\ & \wedge (x_2 \leftrightarrow \neg x_6) \\ & \wedge (x_7 \leftrightarrow (x_4 \wedge x_5 \wedge x_3 \wedge x_6))\end{aligned}$$

SAT: Poly-time reduction

- Easy to see that C is satisfiable $\Leftrightarrow \phi$ is satisfiable.
- The reduction runs in polynomial time.