

Design and Analysis of Algorithms

CS218M

NP Completeness

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Complexity Class \mathbb{P} : Polynomial time Solvable Problems

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Theorem

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- Example:
 $PRIMES = \{x \in \{0,1\}^* \mid x \text{ represents a prime in binary}\}.$

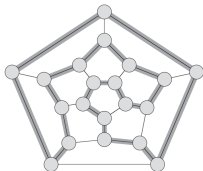
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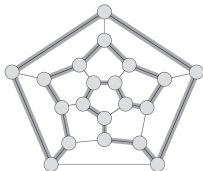
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- *PATH* all graphs having a path from specified s to t of length at most k .

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- $EULER \in \mathbb{P}$ (why?)
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- **Examples:**
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- Above examples belong to an important class of problems called NP problems.

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- **Example:** For boolean formula $((x_1 \rightarrow x_2) \vee \neg((\neg x_1 \leftrightarrow x_3) \vee x_4)) \wedge \neg x_2$
certificate $x_1 = 0, x_2 = 0, x_3 = 1, x_4 = 1$

Definition of Class NP

Definition

$L \in \mathsf{NP}$ if and only if there exists a two input algorithm $A(x, y) \in \mathbb{P}$ and a constant c such that

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Algorithms to Solve L : Given such A and c , for any input x

- Nondeterministically **guess** y . Output $A(x, y)$.
- Systematically **enumerate every** y and compute $A(x, y)$. If any instance $A(X, Y)$ has answer "yes" then we answer "yes" and return. Otherwise try next y . Finally. answer "no". The time of this Deterministic algorithm is $O(2^{\text{Poly}(|X|)})$.

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Natural Question: Is $SAT \in \mathbb{P}$

A Major Open Problem in CS [Cook71-Levin73]

Is $\mathbb{P} = \mathbb{NP}$?

Clay Millenium Problems



The Millennium Prize Problems

In order to celebrate mathematics in the new millennium, The Clay Mathematics Institute of Cambridge, Massachusetts (CMI) established seven *Prize Problems*. The Prizes were conceived to record some of the most difficult problems with which mathematicians were grappling at the turn of the second millennium; to elevate in the consciousness of the general public the fact that in mathematics, the frontier is still open and abounds in important unsolved problems; to emphasize the importance of working towards a solution of the deepest, most difficult problems; and to recognize achievement in mathematics of historical magnitude.

Clay Millenium Problems (List)

Millennium Problems

Yang–Mills and Mass Gap

Experiment and computer simulations suggest the existence of a "mass gap" in the solution to the quantum versions of the Yang-Mills equations. But no proof of this property is known.

Riemann Hypothesis

The prime number theorem determines the average distribution of the primes. The Riemann hypothesis tells us about the deviation from the average. Formulated in Riemann's 1859 paper, it asserts that all the 'non-obvious' zeros of the zeta function are complex numbers with real part $1/2$.

P vs NP Problem

If it is easy to check that a solution to a problem is correct, is it also easy to solve the problem? This is the essence of the P vs NP question. Typical of the NP problems is that of the Hamiltonian Path Problem: given N cities to visit, how can one do this without visiting a city twice? If you give me a solution, I can easily check that it is correct. But I cannot so easily find a solution.

Navier–Stokes Equation

This is the equation which governs the flow of fluids such as water and air. However, there is no proof for the most basic questions one can ask: do solutions exist, and are they unique? Why ask for a proof? Because a proof gives not only certitude, but also understanding.

Hodge Conjecture

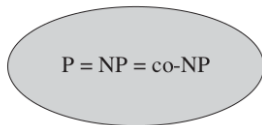
The answer to this conjecture determines how much of the topology of the solution set of a system of algebraic equations can be defined in terms of further algebraic equations. The Hodge conjecture is known in certain special cases, e.g., when the solution set has dimension less than four. But in dimension four it is unknown.

$\overline{\text{Prime}} = \{ \langle n \rangle \mid n \text{ is not a prime} \}$

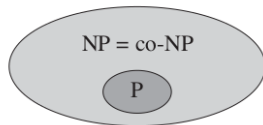
$\overline{\text{Prime}} \in \text{NP}$

$\text{Prime} \in \text{Co-NP}$

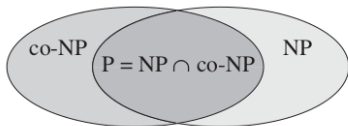
Potential Scenarios



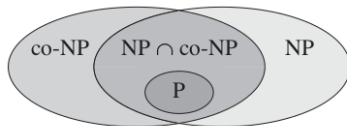
(a)



(b)



(c)



(d)

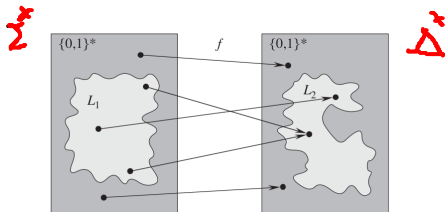
There is strong belief (but no proof!) that $P \subset NP$.

Polynomial Time Reduction

Let $L_1 \subseteq \Sigma^*$ and $L_2 \subseteq \Delta^*$ be two decision problems. A function $f : \Sigma^* \rightarrow \Delta^*$ is called a polynomial time reduction from L_1 to L_2 (denoted $f : L_1 \leq_P L_2$) iff

- f is total.
- There exists a c such that $f(x)$ is computable in time $O(|x|^c)$ for all x .
- $x \in L_1$ iff $f(x) \in L_2$.

Notation $L_1 \leq_P L_2$ denotes $\exists f$ such that $f : L_1 \leq_P L_2$.

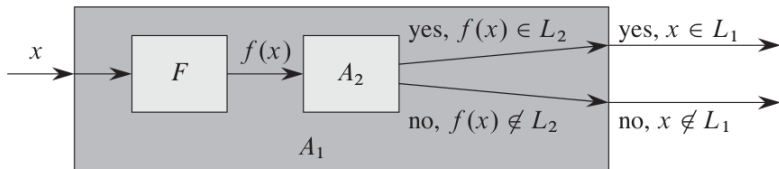


Using Poly Time Reduction

Theorem

If $f : L_1 \leq_P L_2$ and $L_2 \in \mathbb{P}$ then $L_1 \in \mathbb{P}$.

Proof:



Time taken for computing A_1 is polynomial in x (why?)

$$|f(x)| = |x|^c$$

NP-Complete Problems

A problem $L \subseteq \Sigma^*$ is called NP-complete (denoted NPC) if

- $L \in \text{NP}$.
- For every $L' \in \text{NP}$, we have $L' \leq_P L$.
(This shows that L is at least as hard as L' .)

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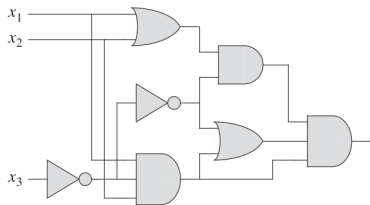
Proof: Let L, L' as above. Then for all $L'' \in \text{NP}$, we have $L'' \leq_P L'$. By transitivity, then $L'' \leq_P L$. Hence, $L \in \text{NPC}$.

We can prove a wide class of problems to be NPC.

- (Initial Problem) **Circuit Satisfaction Problem:** Given a combinational circuit made out of *AND*, *OR*, *NOT* gates, decide whether any input makes the circuit output 1.
- (More problems by Reduction) **Indepandat Set, Vertex Cover, k-Clique, set Packing, set Cover, sat, 3-sat, Hamiltonian Cycle, Travelling Salesman, 3-dimnesional-matching, graph-coloring, subsetsum**

Circuit Satisfiability Problem

- Boolean Combinational Circuit made of *AND*, *OR* *NOT* gates, connected by wires.
- Input wires, single output wire, internal wire
- gates with n inputs, one output, fan out.
- **Circuit satisfiability problem** Give a circuit C is there an assignment of values to input wires which makes the output wire 1? (Similar to SAT).
 $CIRCUIT_SAT = \{ \langle C \rangle \mid C \text{ has a satisfying assignment} \}.$



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- Hence, $CIRCUIT_SAT \in NP$.

CIRCUIT_SAT is NP-hard

(Proof Schema) Let $L \in \text{NP}$. Hence, there is $A(x, y) \in \mathbb{P}$ s.t.
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- A is like a machine program. Execution of A on given input x, y goes through a sequence of machine configurations:

$$c_1 \rightarrow c_2 \rightarrow \dots \rightarrow c_{T(n)}$$

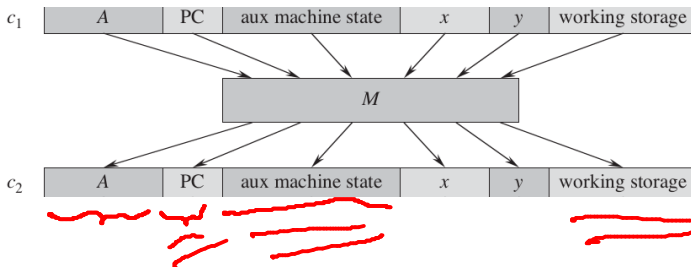
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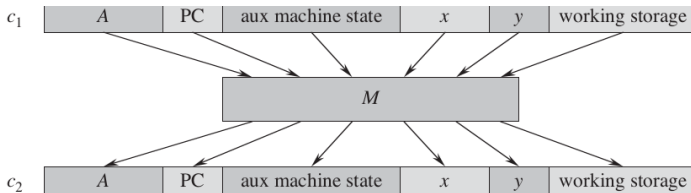
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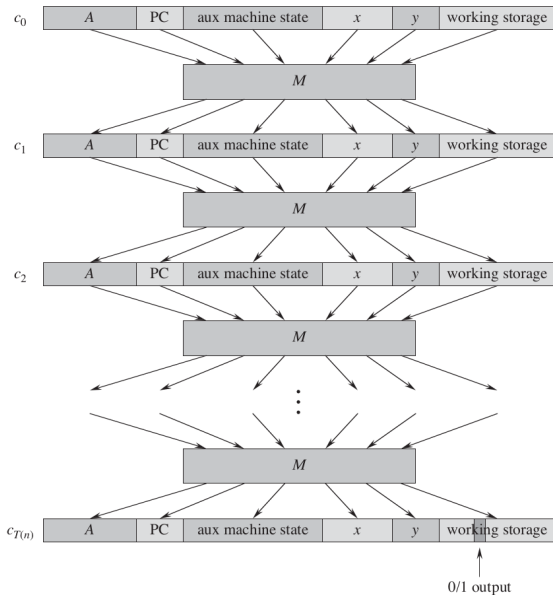
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- We have a **circuit** M which inputs previous configuration and outputs next configuration.

Circuit C_A for $A(x, y)$ taking time $T(n)$



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- It outputs the result bit in last configuration as output wire.