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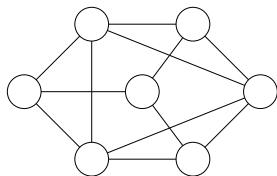
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- Far away from polynomial time
- HC is **NP-hard**: it is **unlikely** that it can be solved in polynomial time.

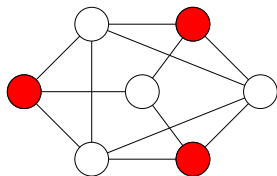
Maximum Independent Set Problem

Def. An **independent set** of $G = (V, E)$ is a subset $I \subseteq V$ such that no two vertices in I are adjacent in G .



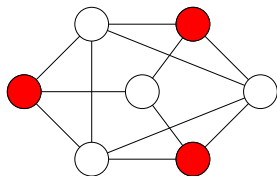
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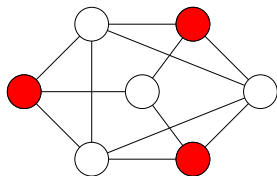
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- Maximum Independent Set is NP-hard

Formula Satisfiability

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Input: boolean formula with n variables, with \vee, \wedge, \neg operators.

Output: whether the boolean formula is satisfiable

- Example: $\neg((\neg x_1 \wedge x_2) \vee (\neg x_1 \wedge \neg x_3) \vee x_1 \vee (\neg x_2 \wedge x_3))$ is not satisfiable
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Outline

- 1 Some Hard Problems
- 2 P, NP and Co-NP**
- 3 Polynomial Time Reductions and NP-Completeness
- 4 NP-Complete Problems
- 5 Dealing with NP-Hard Problems
- 6 Summary

Decision Problem Vs Optimization Problem

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Fact For each optimization problem X , there is a decision version X' of the problem. If we have a polynomial time algorithm for the decision version X' , we can solve the original problem X in polynomial time.

Shortest Path

Input: graph $G = (V, E)$, weight w, s, t and a bound L

Output: whether there is a path from s to t of length at most L

Optimization to Decision

Shortest Path

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Maximum Independent Set

Input: a graph G and a bound k

Output: whether there is an independent set of size at least k

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- Input: (3, 6, 100, 9, 60)
- Binary: (11, 110, 1100100, 1001, 111100)

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- String:

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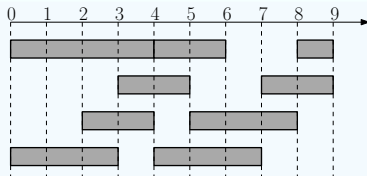
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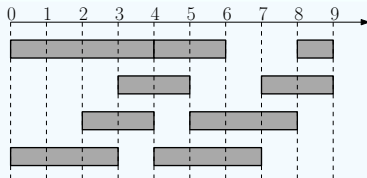
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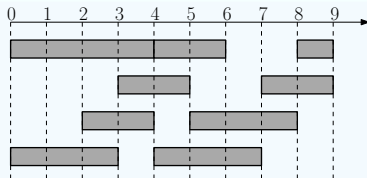


- $(0, 3, 0, 4, 2, 4, 3, 5, 4, 6, 4, 7, 5, 8, 7, 9, 8, 9)$

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Example: Interval Scheduling Problem



- $(0, 3, 0, 4, 2, 4, 3, 5, 4, 6, 4, 7, 5, 8, 7, 9, 8, 9)$
- Encode the sequence into a binary string as before

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A: No! As long as we are using a “natural” encoding. We only care whether the running time is polynomial or not

Define Problem as a Function

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

Def. A **decision problem** X is a function mapping $\{0, 1\}^*$ to $\{0, 1\}$ such that for any $s \in \{0, 1\}^*$, $X(s)$ is the correct output for input s .

- $\{0, 1\}^*$: the set of all binary strings of any length.

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Def. A has a **polynomial running time** if there is a polynomial function $p(\cdot)$ so that for every string s , the algorithm A terminates on s in at most $p(|s|)$ steps.

Complexity Class P

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- The decision versions of interval scheduling, shortest path and minimum spanning tree all in P.

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Def. The message Alice sends to Bob is called a **certificate**, and the algorithm Bob runs is called a **certifier**.

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- Certificate: a set of size k
- Certifier: check if the given set is really an independent set

The Complexity Class NP

Def. B is an **efficient certifier** for a problem X if

- B is a polynomial-time algorithm that takes two input strings s and t , and outputs 0 or 1.
- there is a polynomial function p such that, $X(s) = 1$ if and only if there is string t such that $|t| \leq p(|s|)$ and $B(s, t) = 1$.

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Def. The complexity class NP is the set of all problems for which there exists an efficient certifier.

HC (Hamiltonian Cycle) \in NP

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- $\text{HC}(G) = 1 \iff \exists S, B(G, S) = 1$

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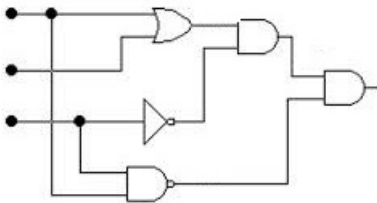
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Circuit Satisfiability (Circuit-Sat) Problem

Input: a circuit with and/or/not gates

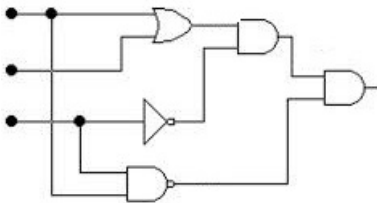
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Input: a circuit with and/or/not gates

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- Is Circuit-Sat \in NP?

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- Unlikely
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- $\overline{\text{HC}} \in \text{Co-NP}$

The Complexity Class Co-NP

Def. For a problem X , the problem \overline{X} is the problem such that $\overline{X}(s) = 1$ if and only if $X(s) = 0$.

Def. **Co-NP** is the set of decision problems X such that $\overline{X} \in \text{NP}$.

Def. A **tautology** is a boolean formula that always evaluates to 1.

Tautology Problem

Input: a boolean formula

Output: whether the formula is a tautology

- e.g. $(\neg x_1 \wedge x_2) \vee (\neg x_1 \wedge \neg x_3) \vee x_1 \vee (\neg x_2 \wedge x_3)$ is a tautology

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- Thus Tautology \in Co-NP

$P \subseteq NP$

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- Thus, $X \in NP$ and $P \subseteq NP$
- Similarly, $P \subseteq \text{Co-NP}$, thus $P \subseteq NP \cap \text{Co-NP}$

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- We assume $P \neq NP$ and prove that problems do not have polynomial time algorithms.
- We said it is **unlikely** that Hamiltonian Cycle can be solved in polynomial time:
 - if $P \neq NP$, then $HC \notin P$
 - $HC \notin P$, unless $P = NP$

Is $NP = Co-NP$?

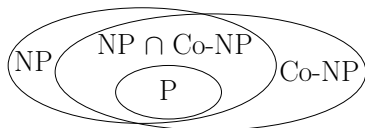
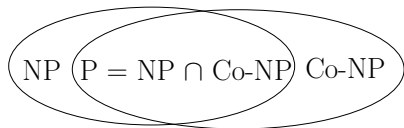
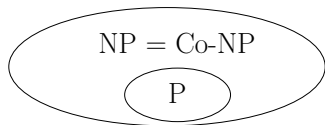
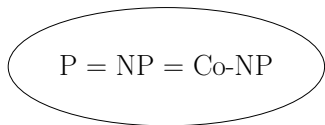
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Is $NP = Co-NP$?

- Again, a big open problem
- Most researchers believe $NP \neq Co-NP$.

4 Possibilities of Relationships

Notice that $X \in \text{NP} \iff \bar{X} \in \text{Co-NP}$ and $P \subseteq \text{NP} \cap \text{Co-NP}$



- People commonly believe we are in the 4th scenario

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Polynomial-Time Reductions

Def. Given a black box algorithm A that solves a problem X , if any instance of a problem Y can be solved using a polynomial number of standard computational steps, plus a polynomial number of calls to A , then we say Y is polynomial-time reducible to X , denoted as $Y \leq_P X$.

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To prove negative results:

Suppose $Y \leq_P X$. If Y cannot be solved in polynomial time, then X cannot be solved in polynomial time.

Polynomial-Time Reduction: Example

Hamiltonian-Path (HP) problem

Input: $G = (V, E)$ and $s, t \in V$

Output: whether there is a Hamiltonian path from s to t in G

Polynomial-Time Reduction: Example

Hamiltonian-Path (HP) problem

Input: $G = (V, E)$ and $s, t \in V$

Output: whether there is a Hamiltonian path from s to t in G

Lemma $HP \leq_P HC$.

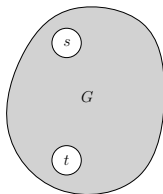
Polynomial-Time Reduction: Example

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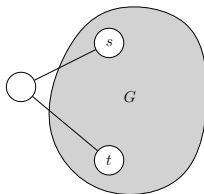
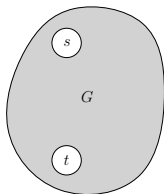
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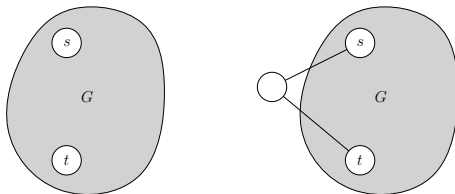
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Obs. G has a HP from s to t if and only if graph on right side has a HC.