

GEMS OF TCS

P VS NP

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Search Problems

SEARCH PROBLEM

Definition

A **search problem** is defined by an algorithm \mathcal{C} that takes an instance I and a candidate solution S , and runs in time polynomial in the length of I . We say that S is a solution to I iff $\mathcal{C}(S, I) = \text{true}$.

SAT

Example

For SAT, I is a Boolean formula, S is an assignment of Boolean constants to its variables. The corresponding algorithm \mathcal{C} checks whether S satisfies all clauses of I .

CLASS NP

Definition

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Definition

NP is the class of all search problems.

- **NP** stands for “non-deterministic polynomial time”: one can guess a solution, and then verify its correctness in polynomial time

- **NP** stands for “non-deterministic polynomial time”: one can guess a solution, and then verify its correctness in polynomial time
- In other words, the class **NP** contains all problems whose solutions can be efficiently verified

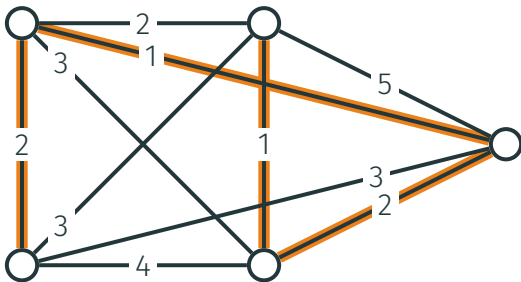
CLASS P

Definition

P is the class of all search problems that can be solved in polynomial time.

TRAVELING SALESMAN PROBLEM

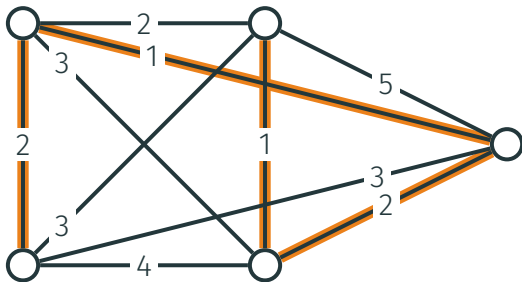
Given a complete weighted graph, find a path of minimum total weight (length) visiting each node exactly once



length: 6

TRAVELING SALESMAN PROBLEM

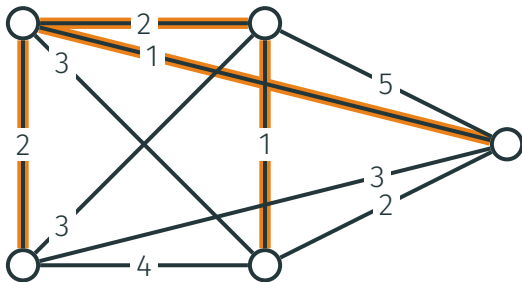
Given a complete weighted graph and a budget b , find a path of total weight (length) $\leq b$ visiting each node exactly once



length: $6 \leq b$

MINIMUM SPANNING TREE

Given a complete weighted graph and a budget b , connect all vertices by $n - 1$ edges of minimum total weight (length)



length: 6

TSP AND MST

MST

Given n cities, connect them by $(n - 1)$ roads of minimal total length

TSP AND MST

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Can be solved efficiently

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TSP

Given n cities, connect them **in a path** of minimal total length

No polynomial algorithm known!

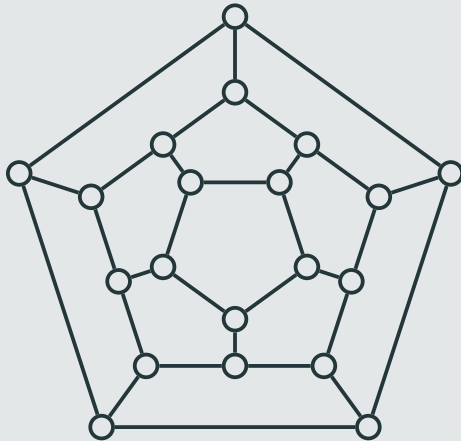
LONGEST PATH

Longest path

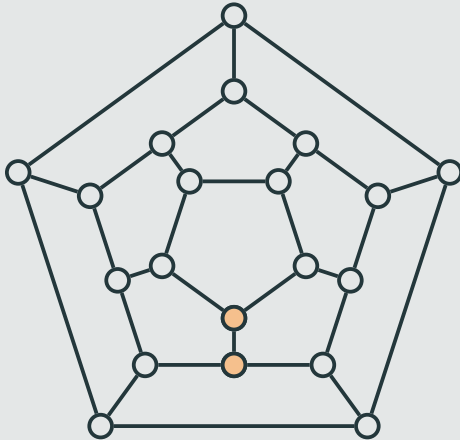
Input: A weighted graph, two vertices s, t , and a budget b .

Output: A simple path (containing no repeated vertices) of total length at least b .

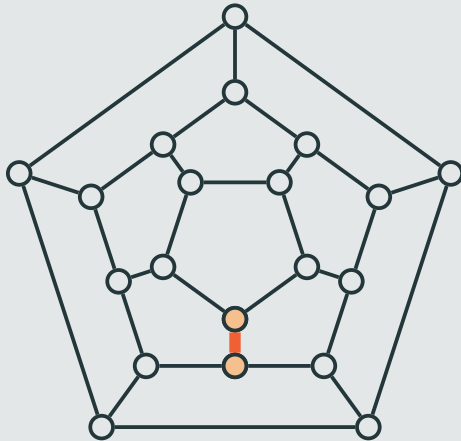
Example



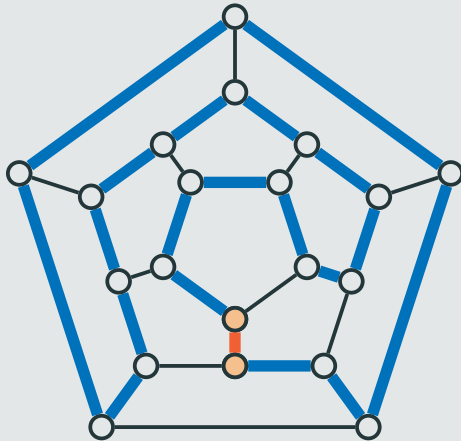
Example



Example



Example



Shortest path

Find a simple path from s to t of total length at most b

Shortest path

Find a simple path from s to t of total length **at most** b

Can be solved
efficiently

Shortest path

Find a simple path from s to t of total length **at most** b

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Longest path

Find a simple path from s to t of total length **at least** b

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Can be solved efficiently

Longest path

Find a simple path from s to t of total length **at least** b

No polynomial algorithm known!

INTEGER LINEAR PROGRAMMING PROBLEM

Integer linear programming

Input: A set of linear inequalities $\mathbf{Ax} \leq \mathbf{b}$.

Output: Integer solution.

Example

$$x_1 \geq 0.5$$

$$-x_1 + 8x_2 \geq 0$$

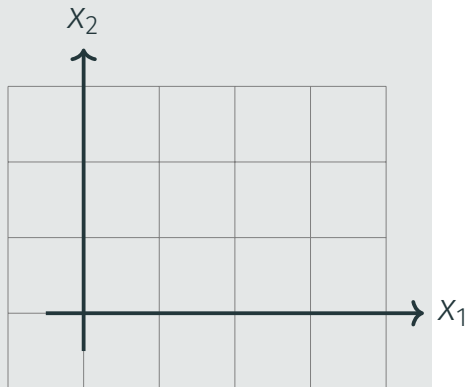
$$-x_1 - 8x_2 \geq -8$$

Example

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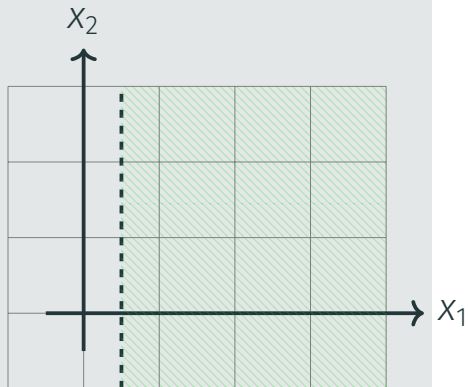


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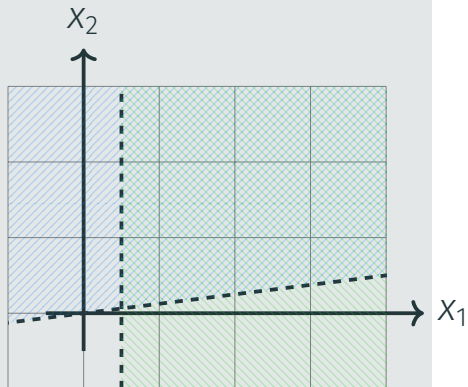


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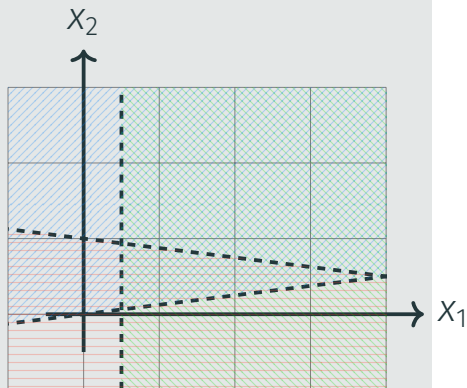


Example

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INTEGER LINEAR PROGRAMMING

LP

Find a **real**
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INTEGER LINEAR PROGRAMMING

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ILP

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ILP

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No polynomial algorithm known!

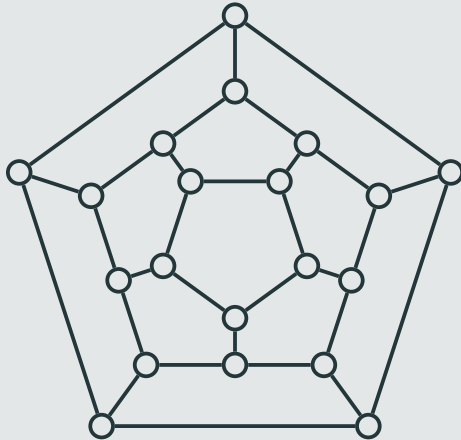
INDEPENDENT SET PROBLEM

Independent set

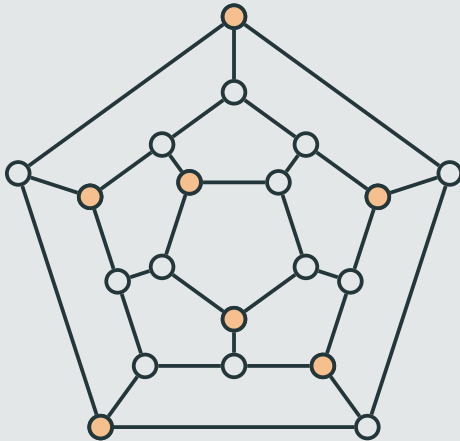
Input: A graph and a budget b .

Output: A subset of vertices of size at least b such that no two of them are adjacent.

Example

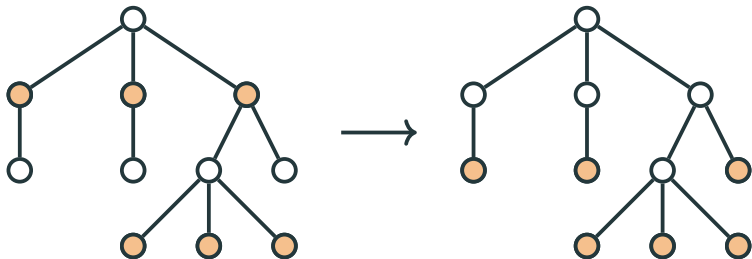


Example



INDEPENDENT SETS IN A TREE

A maximum independent set in a tree can be found by a simple greedy algorithm: it is safe to take into a solution all the leaves.



Independent set in a tree

Find an independent
set of size at least b in
a given **tree**

Independent set in a tree

Find an independent
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Can be solved
efficiently

Independent set in a tree

Find an independent set of size at least b in a given **tree**

Can be solved efficiently

Independent set in a graph

Find an independent set of size at least b in a given **graph**

Independent set in a tree

Find an independent set of size at least b in a given **tree**

Can be solved efficiently

Independent set in a graph

Find an independent set of size at least b in a given **graph**

No polynomial algorithm known!

NP

It turns out that all these hard problems are in a sense a single hard problem: a polynomial time algorithm for any of these problems can be used to solve all of them in polynomial time!

Class P

Problems whose
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Problems whose solution can be found efficiently

- MST
- Shortest path
- LP
- IS on trees

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Problems whose solution can be **found** efficiently

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Class NP

Problems whose solution can be **verified** efficiently

Class P

Problems whose solution can be **found** efficiently

- MST
- Shortest path
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Class NP

Problems whose solution can be **verified** efficiently

- TSP
- Longest path
- ILP
- IS on graphs

The main open problem in Computer Science

Is **P** equal to **NP**?

The main open problem in Computer Science

Is **P** equal to **NP**?

Millenium Prize Problem

Clay Mathematics Institute: \$1M prize for solving the problem

- If $P=NP$, then all search problems can be solved in polynomial time.

- If $P=NP$, then all search problems can be solved in polynomial time.
- If $P \neq NP$, then there exist search problems that cannot be solved in polynomial time.

Reductions

INFORMALLY

We say that a search problem A is reduced to a search problem B and write $A \rightarrow B$, if a polynomial time algorithm for B can be used (as a black box) to solve A in polynomial time.

REDUCTION: $A \rightarrow B$

instance I of A

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instance I of A

Algorithm for A

Algorithm for B

REDUCTION: $A \rightarrow B$

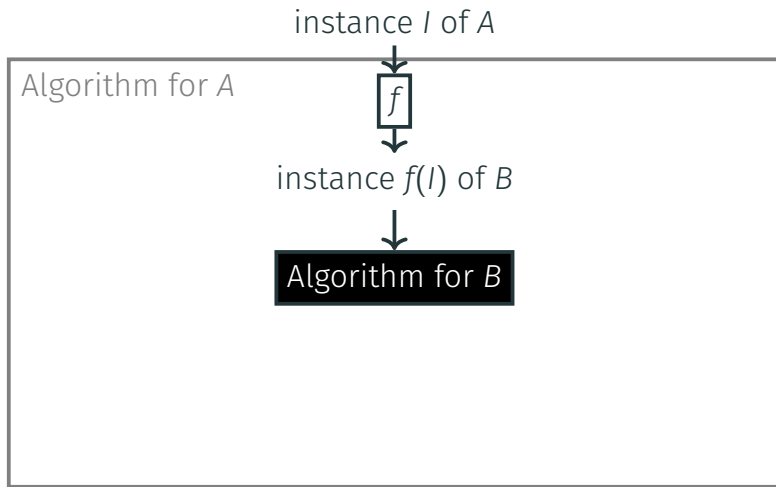
instance I of A

Algorithm for A

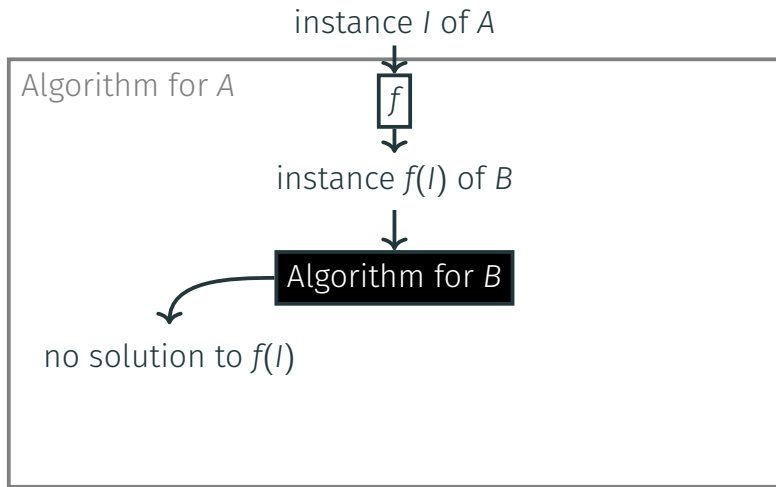


Algorithm for B

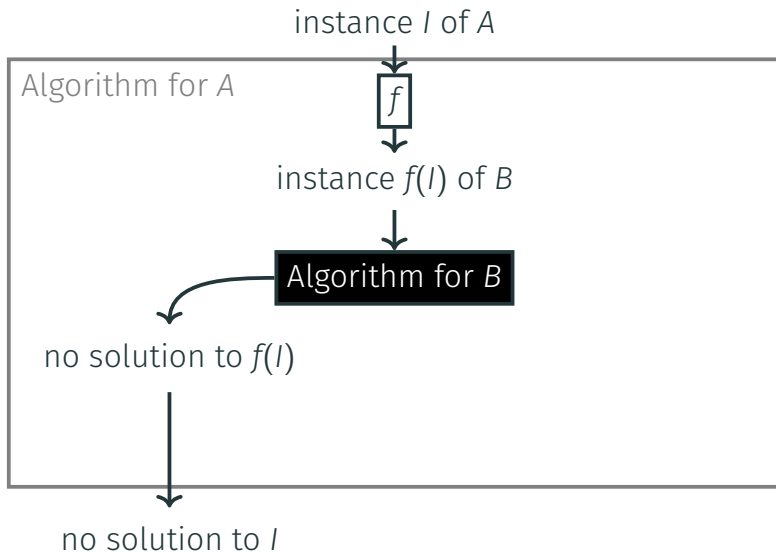
REDUCTION: $A \rightarrow B$



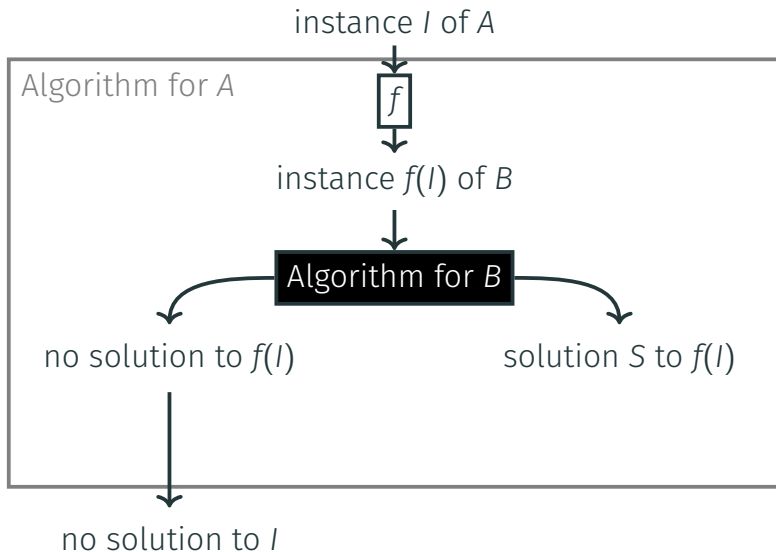
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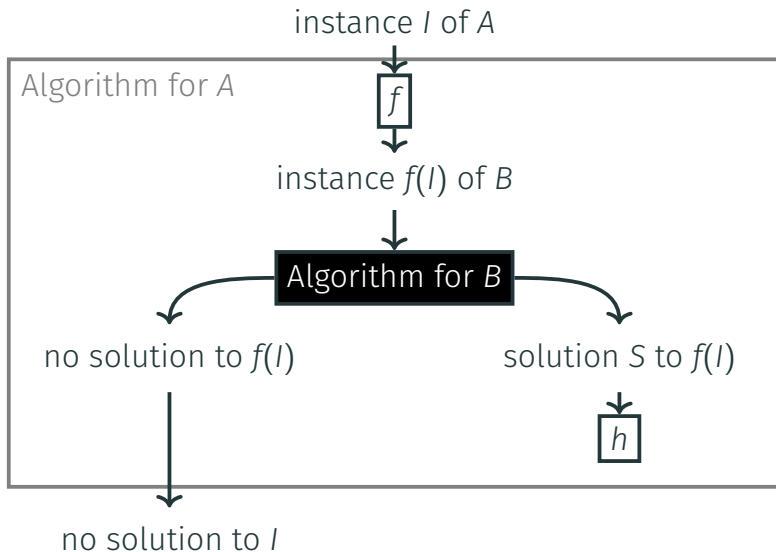
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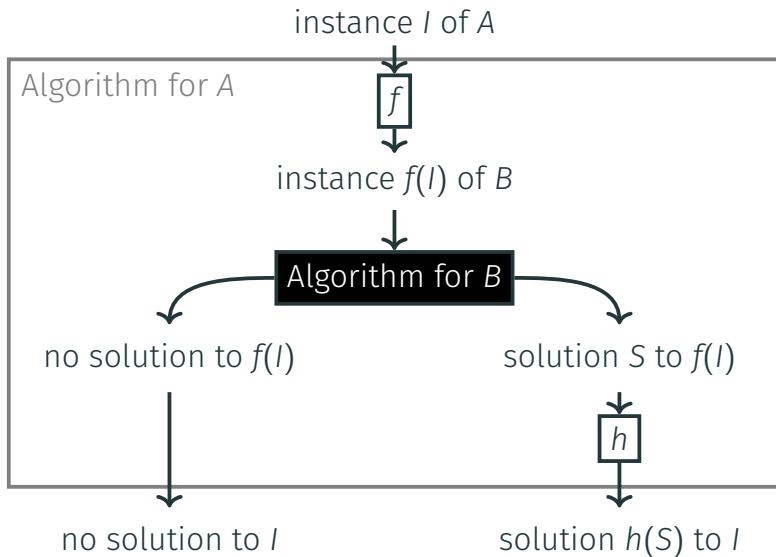
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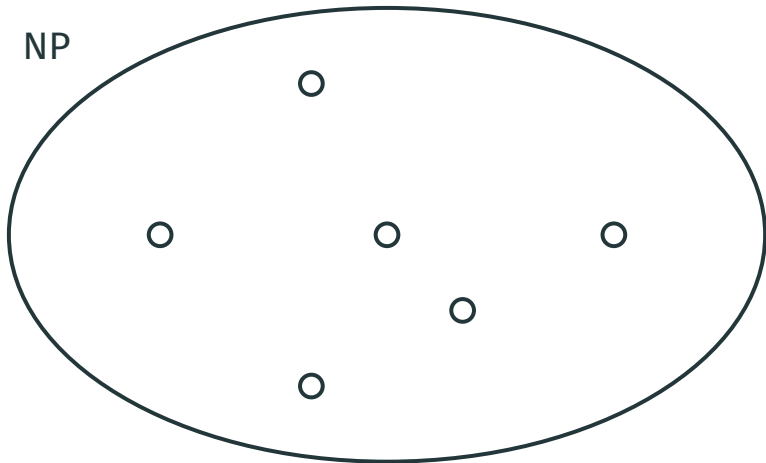


FORMALLY

Definition

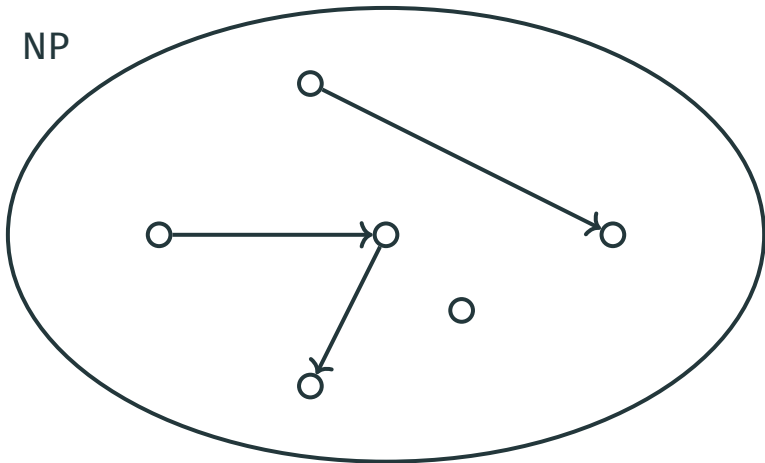
We say that a search problem A is reduced to a search problem B and write $A \rightarrow B$, if there exists a polynomial time algorithm f that converts any instance I of A into an instance $f(I)$ of B , together with a polynomial time algorithm h that converts any solution S to $f(I)$ back to a solution $h(S)$ to A . If there is no solution to $f(I)$, then there is no solution to I .

GRAPH OF SEARCH PROBLEMS



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NP



NP-COMPLETE PROBLEMS

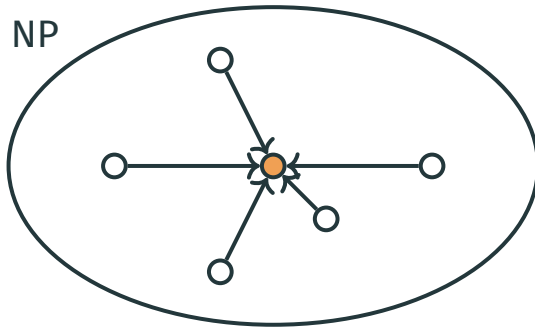
Definition

A search problem is called **NP-complete** if all other search problems reduce to it.

NP-COMPLETE PROBLEMS

Definition

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Do they exist?

It's not at all immediate that **NP**-complete problems even exist. We'll see later that all hard problems that we've seen in the previous part are in fact **NP**-complete!

Two ways of using $A \rightarrow B$:

- if B is easy (can be solved in polynomial time), then so is A
- if A is hard (cannot be solved in polynomial time), then so is B

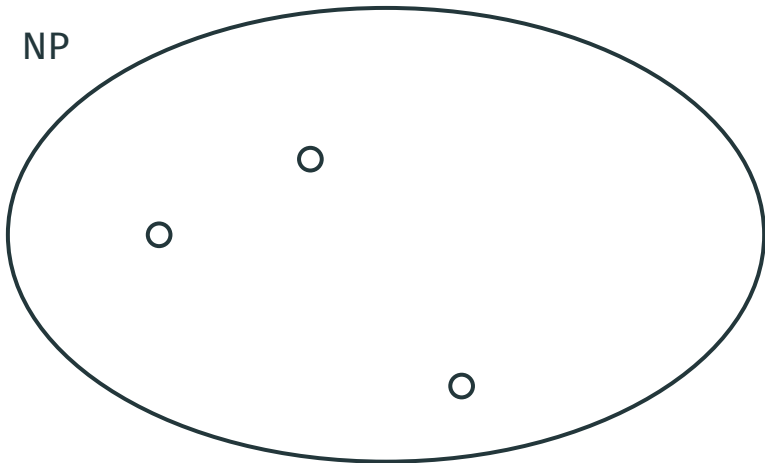
REDUCTIONS COMPOSE

Lemma

If $A \rightarrow B$ and $B \rightarrow C$, then $A \rightarrow C$.

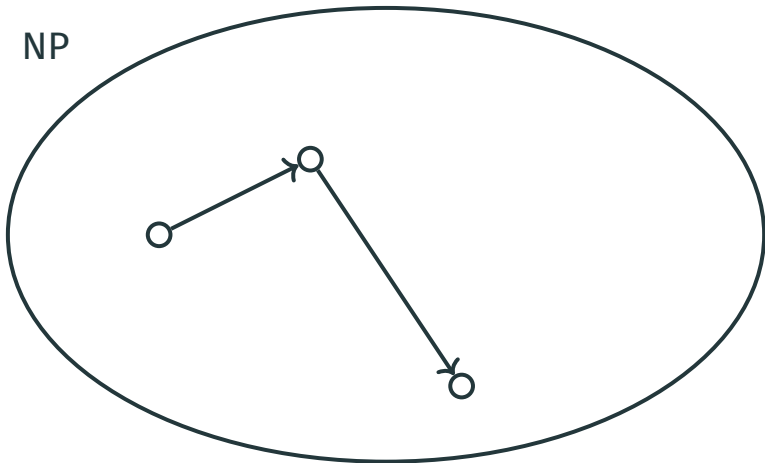
PICTORIALLY

NP



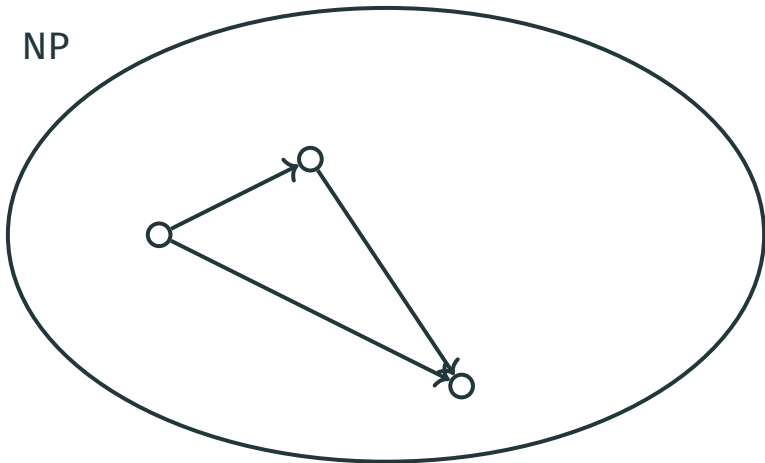
PICTORIALLY

NP



PICTORIALLY

NP



SHOWING NP-COMPLETENESS

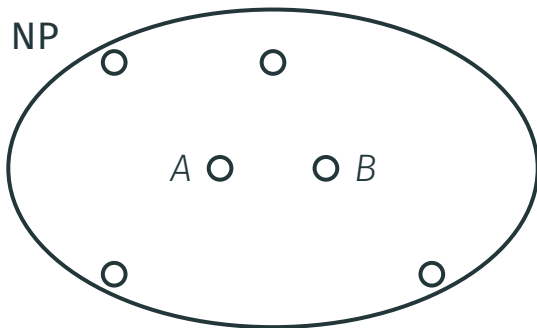
Corollary

If $A \rightarrow B$ and A is **NP**-complete, then so is B .

SHOWING NP-COMPLETENESS

Corollary

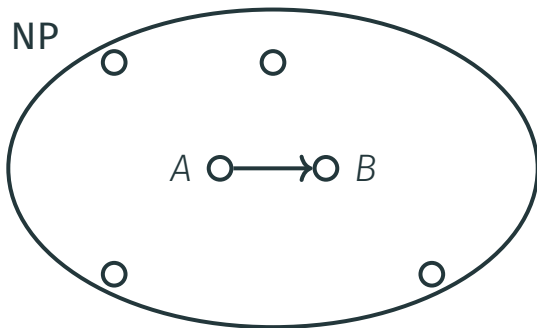
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SHOWING NP-COMPLETENESS

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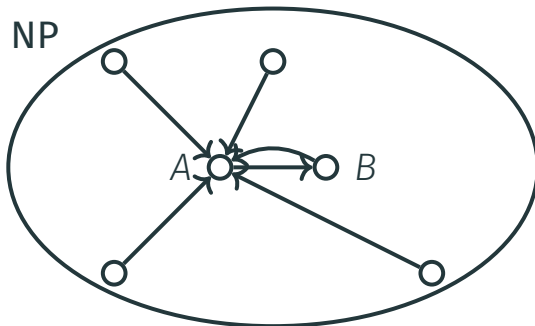
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SHOWING NP-COMPLETENESS

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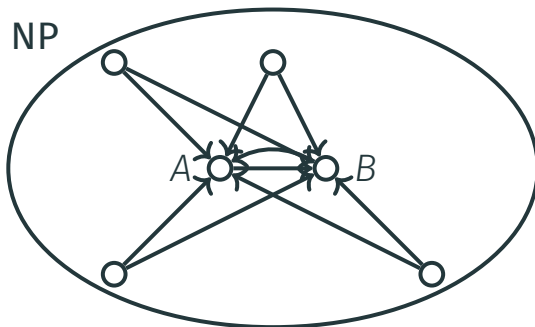
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SHOWING NP-COMPLETENESS

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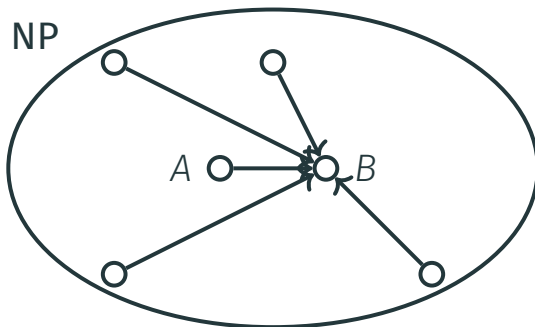
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SHOWING NP-COMPLETENESS

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NP-Completeness of SAT

Goal

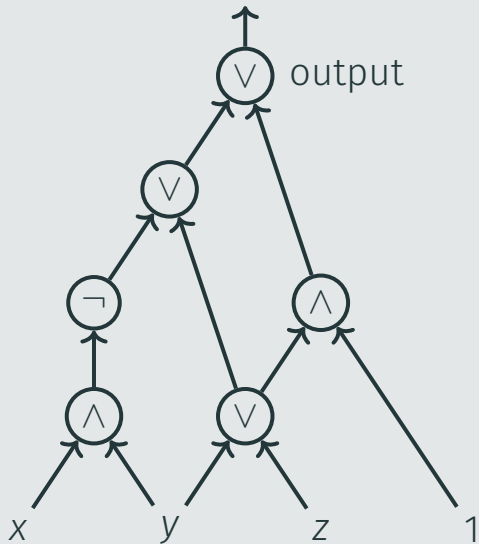
Show that **every** search problem reduces to SAT.

Goal

Show that **every** search problem reduces to SAT.

Instead, we show that any problem reduces to Circuit SAT problem, which, in turn, reduces to SAT.

Circuit



Definition

A **circuit** is a directed acyclic graph of in-degree at most 2. Nodes of in-degree 0 are called **inputs** and are marked by Boolean variables and constants. Nodes of in-degree 1 and 2 are called **gates**: gates of in-degree 1 are labeled with NOT, gates of in-degree 2 are labeled with AND or OR. One of the sinks is marked as **output**.

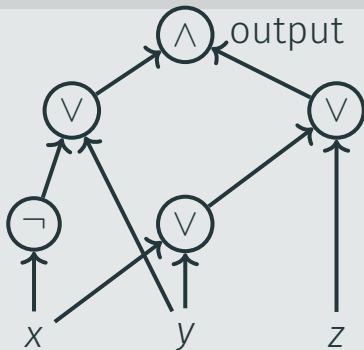
Circuit-SAT

Input: A circuit.

Output: An assignment of Boolean values to the input variables of the circuit that makes the output true.

SAT is a special case of Circuit-SAT as a SAT formula can be represented as a circuit:

Example: $(x \vee y \vee z)(y \vee \bar{x})$



CIRCUIT-SAT \rightarrow SAT

To reduce Circuit-SAT to SAT, we need to design a polynomial time algorithm that for a given circuit outputs a SAT formula which is satisfiable, if and only if the circuit is satisfiable

IDEA

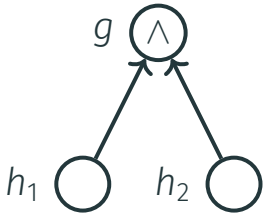
- Introduce a Boolean variable for each gate
- For each gate, write down a few clauses that describe the relationship between this gate and its direct predecessors

NOT GATES



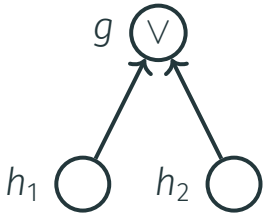
$$(h \vee g)(\bar{h} \vee \bar{g})$$

AND GATES



$$(h_1 \vee \bar{g})(h_2 \vee \bar{g})(\bar{h}_1 \vee \bar{h}_2 \vee g)$$

OR GATES



$$(\bar{h}_1 \vee g)(\bar{h}_2 \vee g)(h_1 \vee h_2 \vee \bar{g})$$

OUTPUT GATE

$g \bigcirc \text{output} \quad (g)$

- The resulting SAT formula is consistent with the initial circuit: in any satisfying assignment of the formula, the value of g is equal to the value of the gate labeled with g in the circuit

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- Therefore, the SAT formula and the circuit are equisatisfiable
- The reduction takes polynomial time

Goal

Reduce every search problem to Circuit-SAT.

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- Let A be a search problem

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Goal

Reduce every search problem to Circuit-SAT.

- Let A be a search problem
- We know that there exists an algorithm \mathcal{C} that takes an instance I of A and a candidate solution S and checks whether S is a solution for I in time polynomial in $|I|$
- In particular, $|S|$ is polynomial in $|I|$

TURN AN ALGORITHM INTO A CIRCUIT

- Note that a computer is in fact a circuit implemented on a chip

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- Note that a computer is in fact a circuit implemented on a chip
- Each step of the algorithm $\mathcal{C}(I, S)$ is performed by this computer's circuit
- This gives a circuit of size polynomial in $|I|$ that has input bits for I and S and outputs whether S is a solution for I (a separate circuit for each input length)

REDUCTION

To solve an instance I of the problem A :

- take a circuit corresponding to $\mathcal{C}(I, \cdot)$

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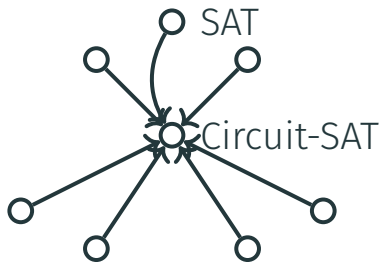
- take a circuit corresponding to $\mathcal{C}(I, \cdot)$
- the inputs to this circuit encode candidate solutions

REDUCTION

To solve an instance I of the problem A :

- take a circuit corresponding to $\mathcal{C}(I, \cdot)$
- the inputs to this circuit encode candidate solutions
- use a Circuit-SAT algorithm for this circuit to find a solution (if exists)

SUMMARY



SUMMARY

