

# 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

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**NP** is the class of all search problems.

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- In other words, the class **NP** contains all problems whose solutions can be efficiently verified



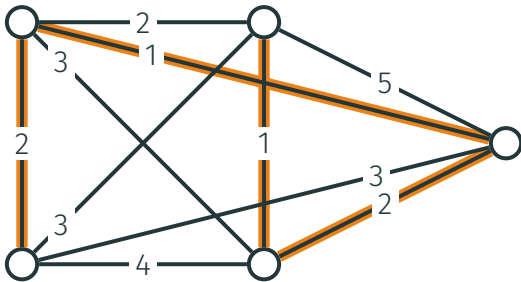
# CLASS P

## Definition

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

# TRAVELING SALESMAN PROBLEM

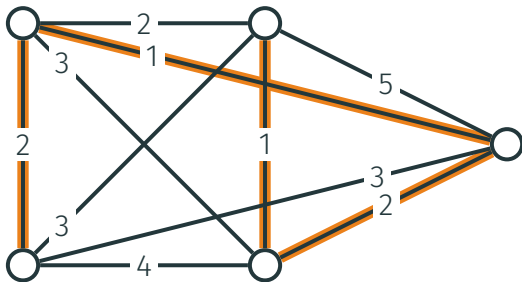
Given a complete weighted graph, find a path of minimum total weight (length) visiting each node exactly once



length: 6

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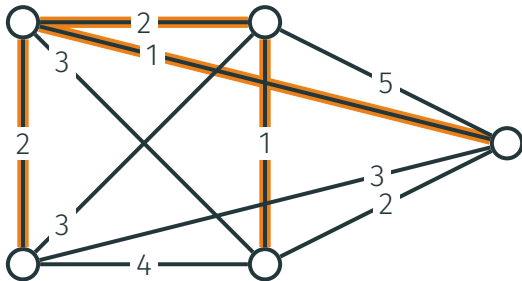
Given a complete weighted graph and a budget  $b$ , find a path of total weight (length)  $\leq b$  visiting each node exactly once



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

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

No polynomial algorithm known!



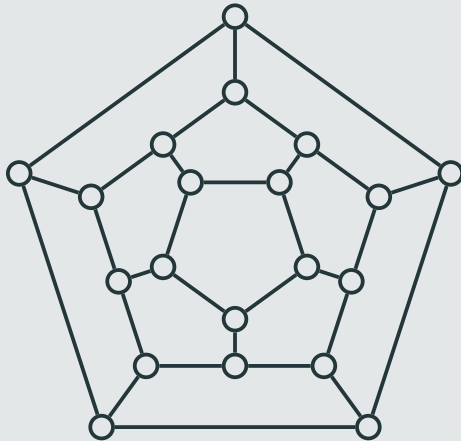
# LONGEST PATH

## Longest path

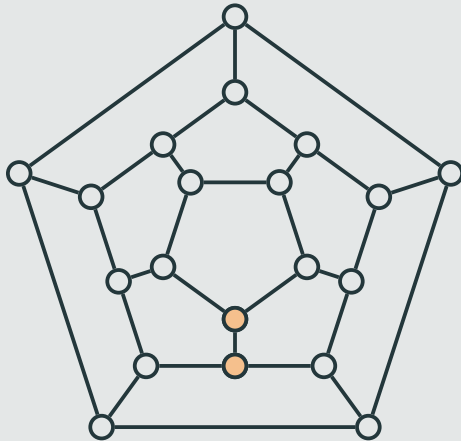
**Input:** A weighted graph, two vertices  $s, t$ , and a budget  $b$ .

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

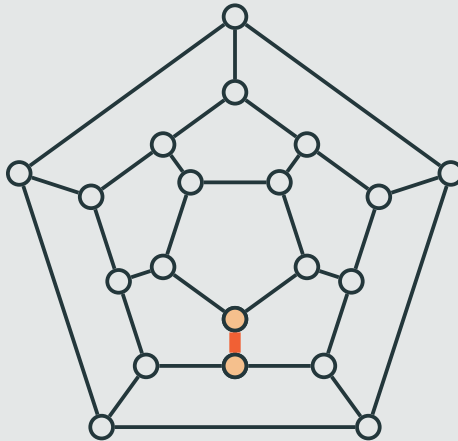
# Example



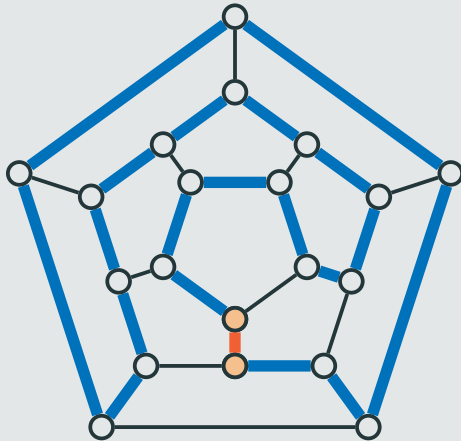
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# INTEGER LINEAR PROGRAMMING PROBLEM

## Integer linear programming

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

**Output:** Integer solution.

## Example

$$x_1 \geq 0.5$$

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

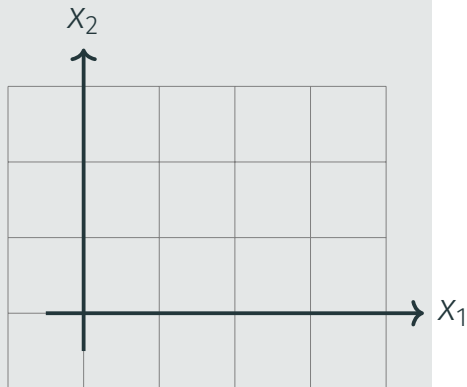
$$-x_1 - 8x_2 \geq -8$$

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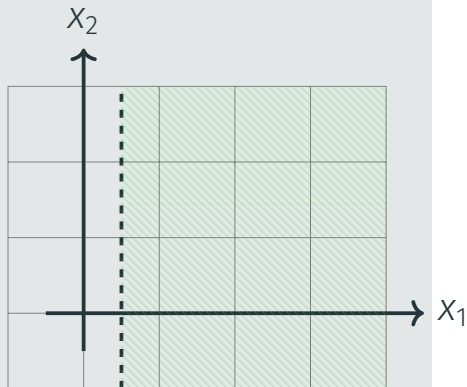


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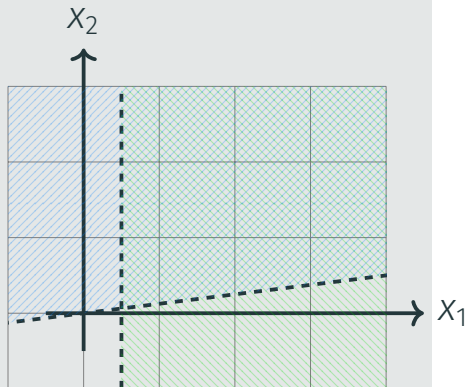


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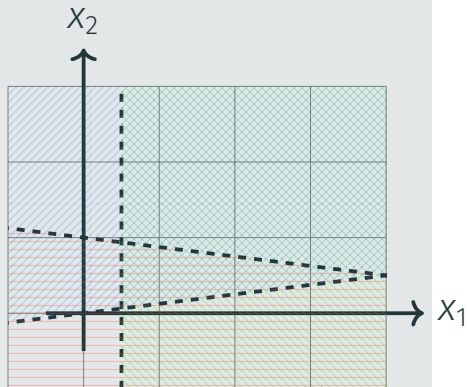


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# INTEGER LINEAR PROGRAMMING

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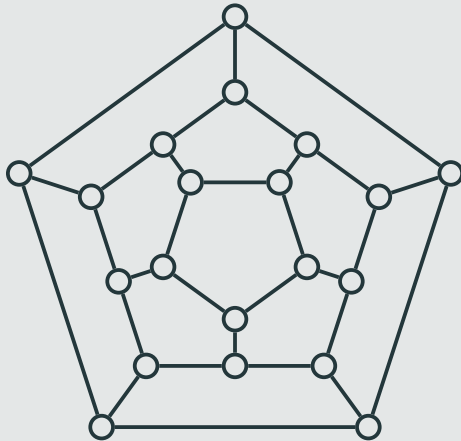
# INDEPENDENT SET PROBLEM

## Independent set

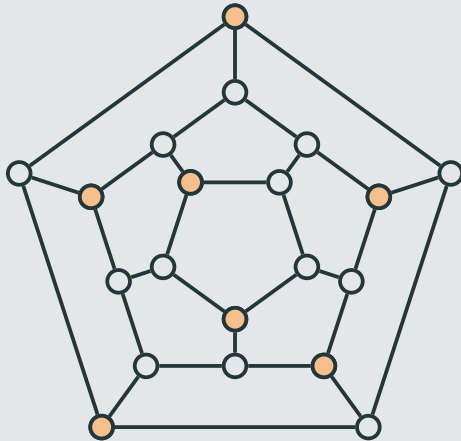
**Input:** A graph and a budget  $b$ .

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

# Example

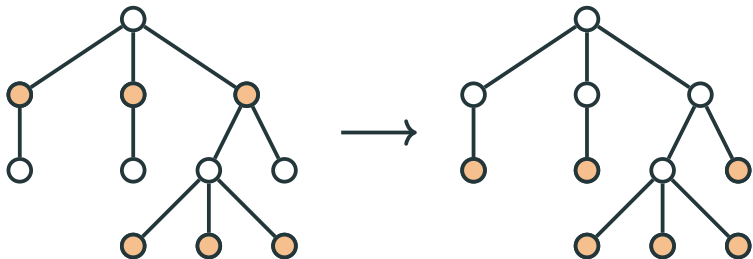


# Example



# 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



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# 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!

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Problems whose solution can be **verified** efficiently

- TSP
- Longest path
- ILP
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The main open problem in Computer Science

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

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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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Algorithm for  $A$

Algorithm for  $B$



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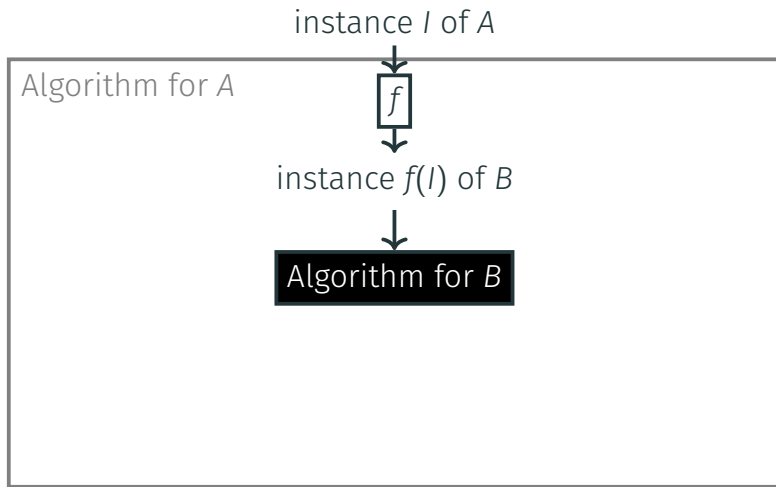
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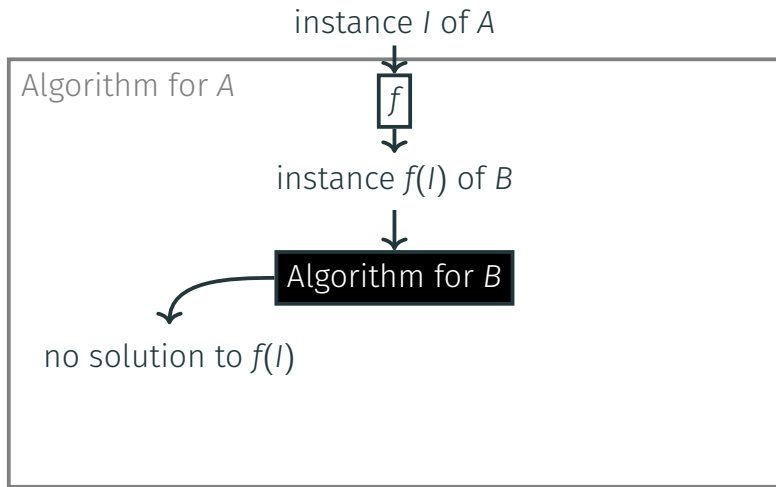


Algorithm for  $B$

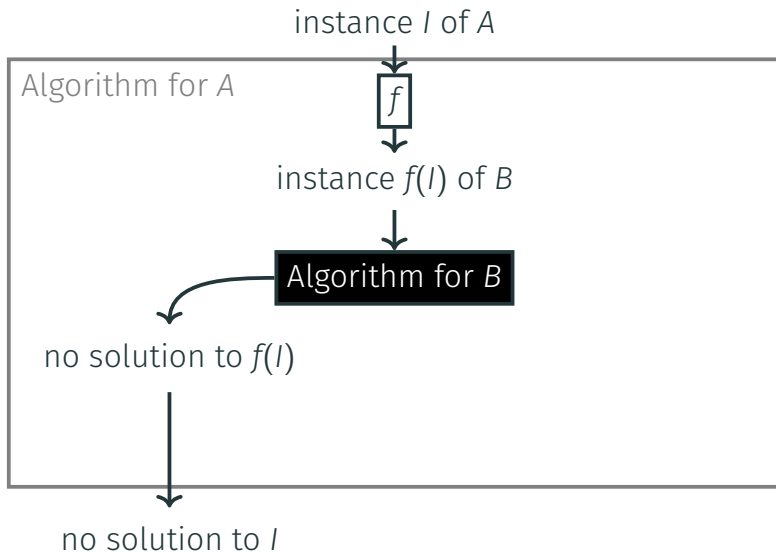
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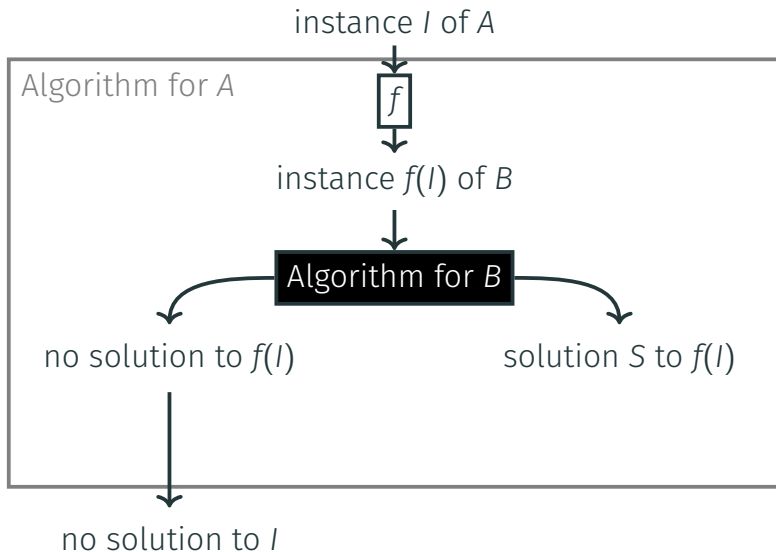
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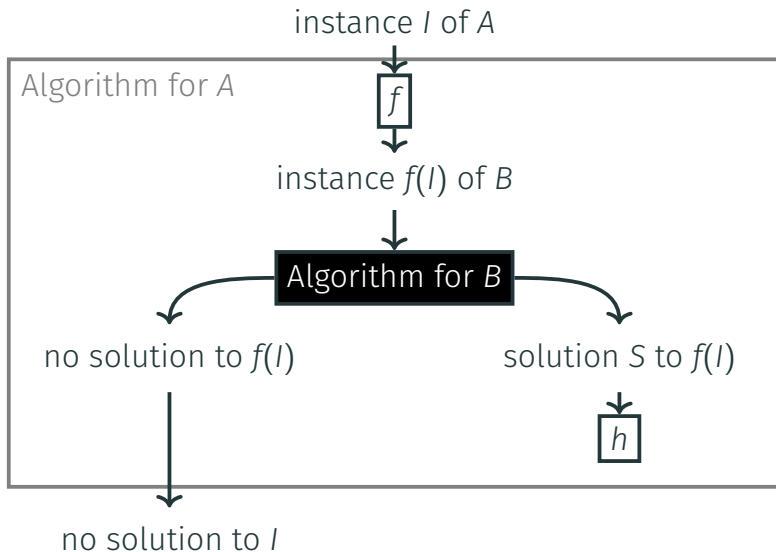
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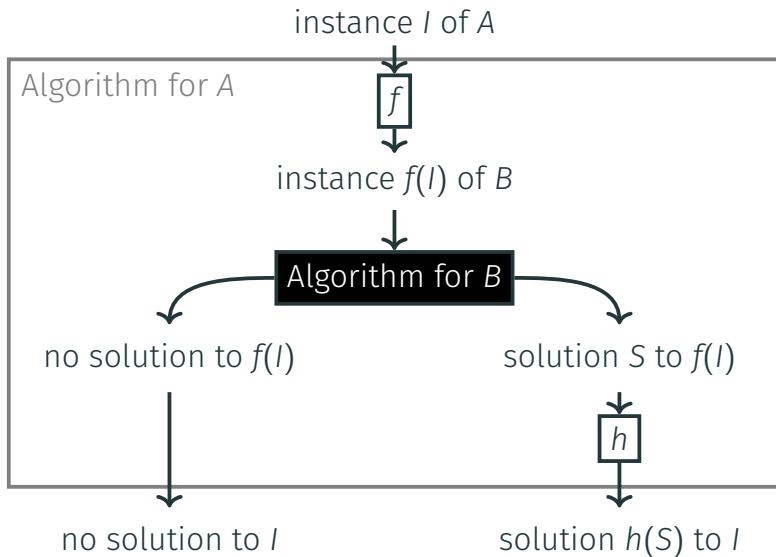
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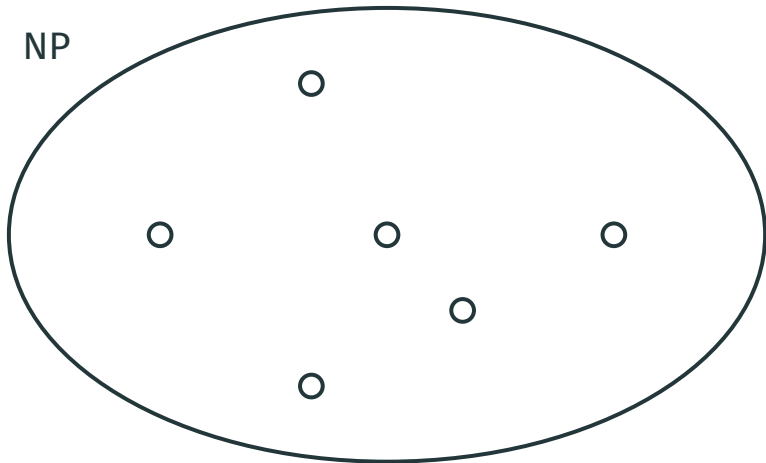
# FORMALLY

## 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  $I$ .

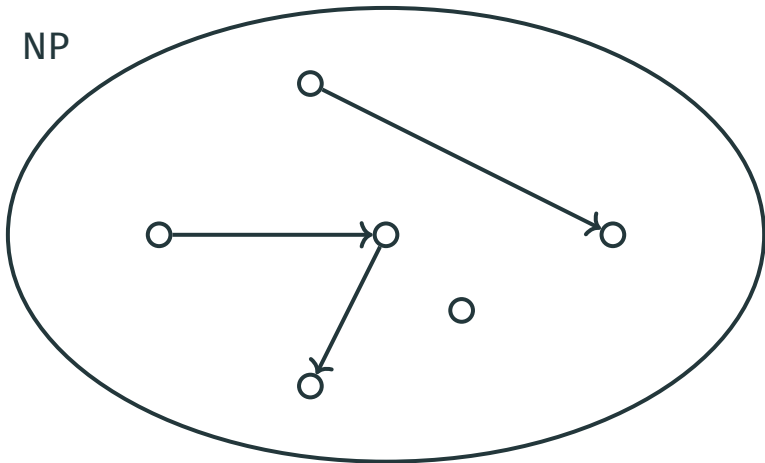


# GRAPH OF SEARCH PROBLEMS



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NP



# NP-COMPLETE PROBLEMS

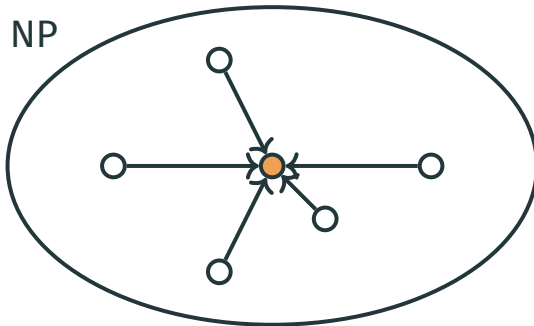
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A search problem is called **NP-complete** if all other search problems reduce to it.

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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  $B$

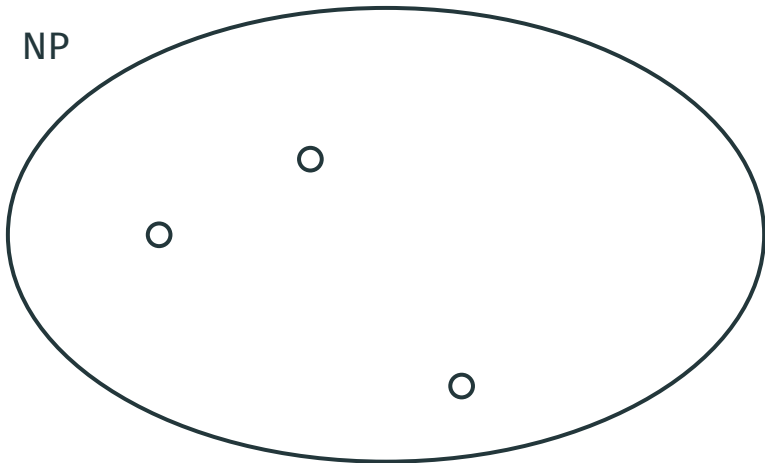
# REDUCTIONS COMPOSE

## Lemma

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

# PICTORIALLY

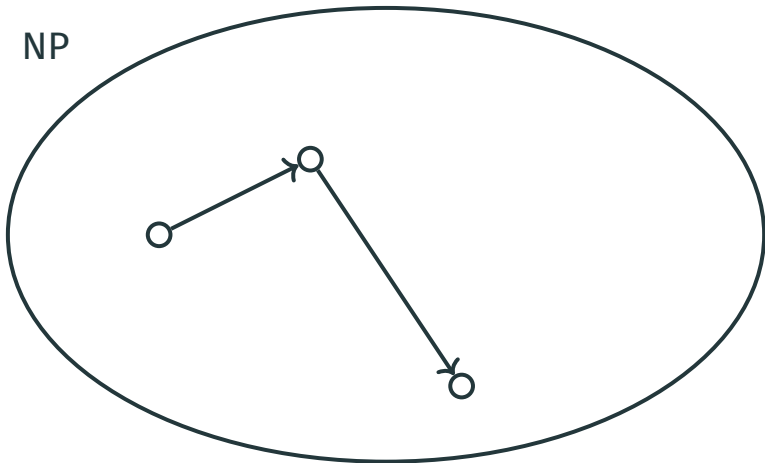
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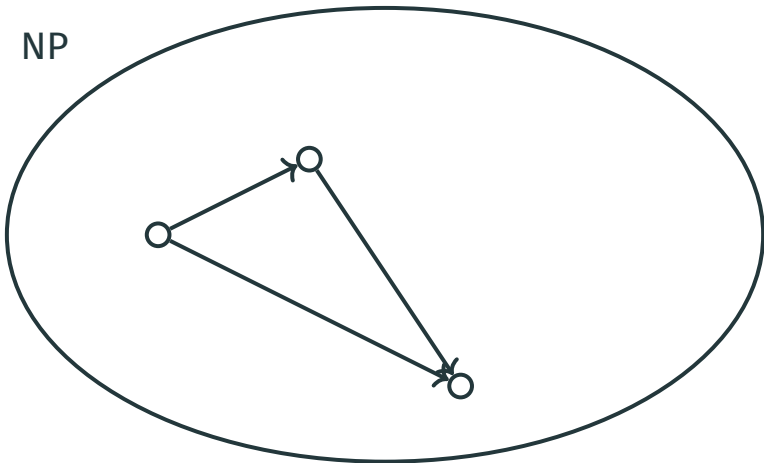
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# SHOWING NP-COMPLETENESS

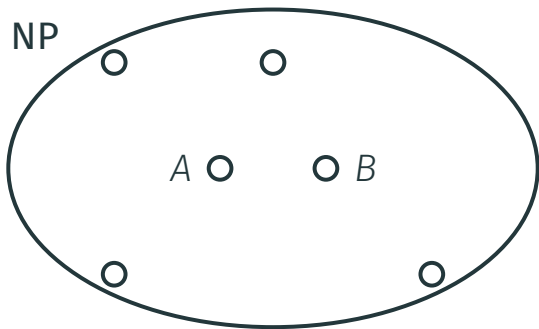
## Corollary

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

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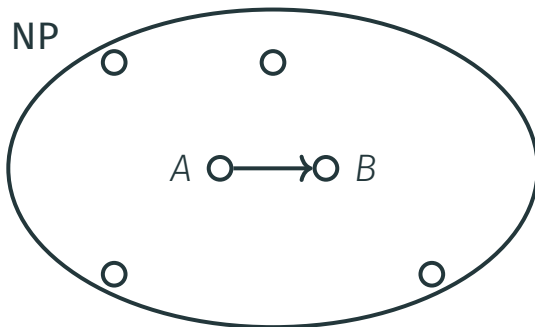
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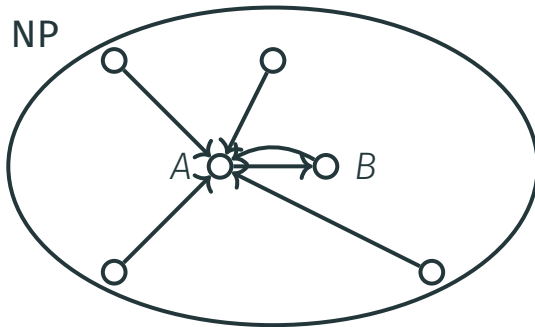
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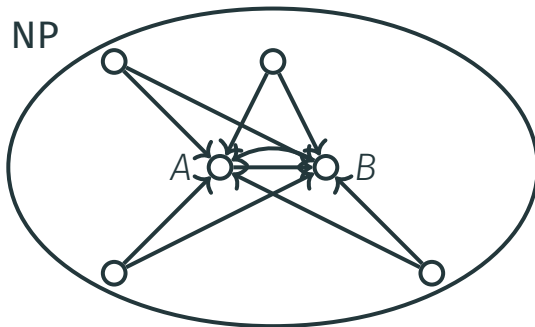
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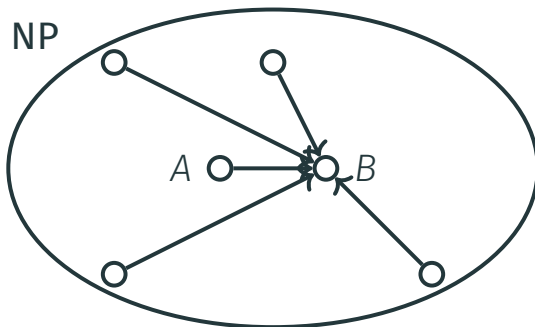
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# SHOWING NP-COMPLETENESS

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# NP-Completeness of SAT

## Goal

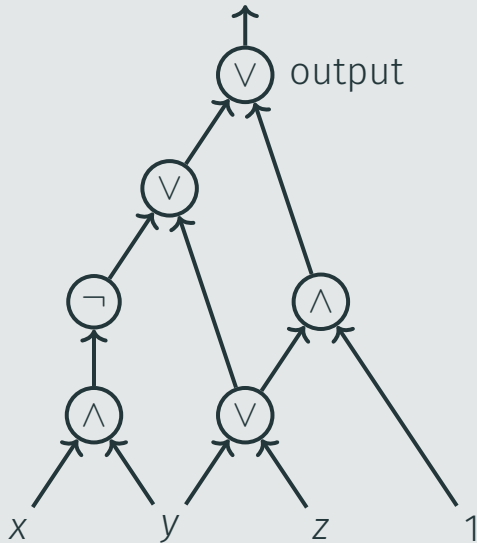
Show that **every** search problem reduces to SAT.

## 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**.

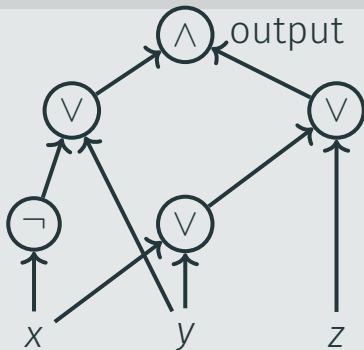
## Circuit-SAT

**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

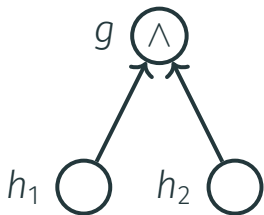
- Introduce a Boolean variable for each gate
- For each gate, write down a few clauses that describe the relationship between this gate and its direct predecessors

# NOT GATES



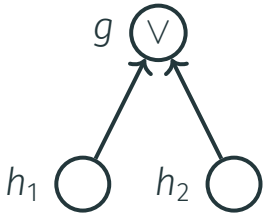
$$(h \vee g)(\bar{h} \vee \bar{g})$$

# 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

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

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



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Reduce every search problem to Circuit-SAT.

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- In particular,  $|S|$  is polynomial in  $|I|$

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- 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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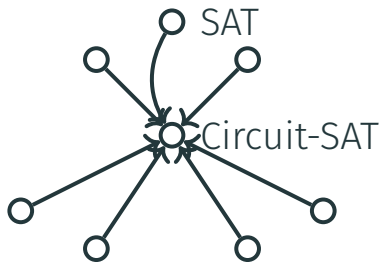
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# REDUCTION

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



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