

# NP-Completeness

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# Classification of the Problems

- ❑ **Unsolvable:** There is no algorithm to solve them (Halting Problem)
- ❑ **Intractable:** As they grow large, we are unable to solve them in reasonable time (Hamiltonian Cycle)
- ❑ **Tractable:** We are able to solve them in reasonable time (Sorting)

**Q:** What constitutes reasonable time?

**A:** Standard working definition: *polynomial time*

- On an input of size  $n$ , the worst-case running time is  $O(n^k)$  for some constant  $k$
- Polynomial time:  $O(n^2)$ ,  $O(n^3)$ ,  $O(1)$ ,  $O(n \lg n)$
- Not in polynomial time:  $O(2^n)$ ,  $O(n^n)$ ,  $O(n!)$

# P vs. NP

- **P:** The class of problems, for which a Polynomial-time algorithm exists.
  - **Ex:** fractional knapsack, finding maximum subarray, rod cutting, etc.
- **NP:** The class of problems for which a solution can be *verified* in polynomial time
  - **Ex:** Hamiltonian cycle, graph coloring, 3SAT, etc.

# Decision vs. Optimization

- ❑ **Decision problem:** a decision problem is a problem that can be posed as a **yes-no question** of the input values.
- ❑ **Optimization problem:** optimization problems are concerned with finding the **best** answer to a particular input.

# Decision vs. Optimization

## Example (Knapsack Problem)

Suppose that we have  $n$  objects, say  $o_i$  ( $i = 1, 2, \dots, n$ ), each with corresponding weight ( $w_i$ ) and profit ( $p_i$ ), and a weight bound  $b$ .

- **Optimization:** Find an  $X = (x_1, x_2, \dots, x_n)$  that maximize  $\sum_{i=1}^n x_i p_i$  with respect to  $\sum_{i=1}^n x_i w_i \leq b$ .
- **Decision:** For a given  $k$ , is there a feasible solution, say  $X = (x_1, x_2, \dots, x_n)$ , where  $\sum_{i=1}^n x_i p_i \geq k$ ?

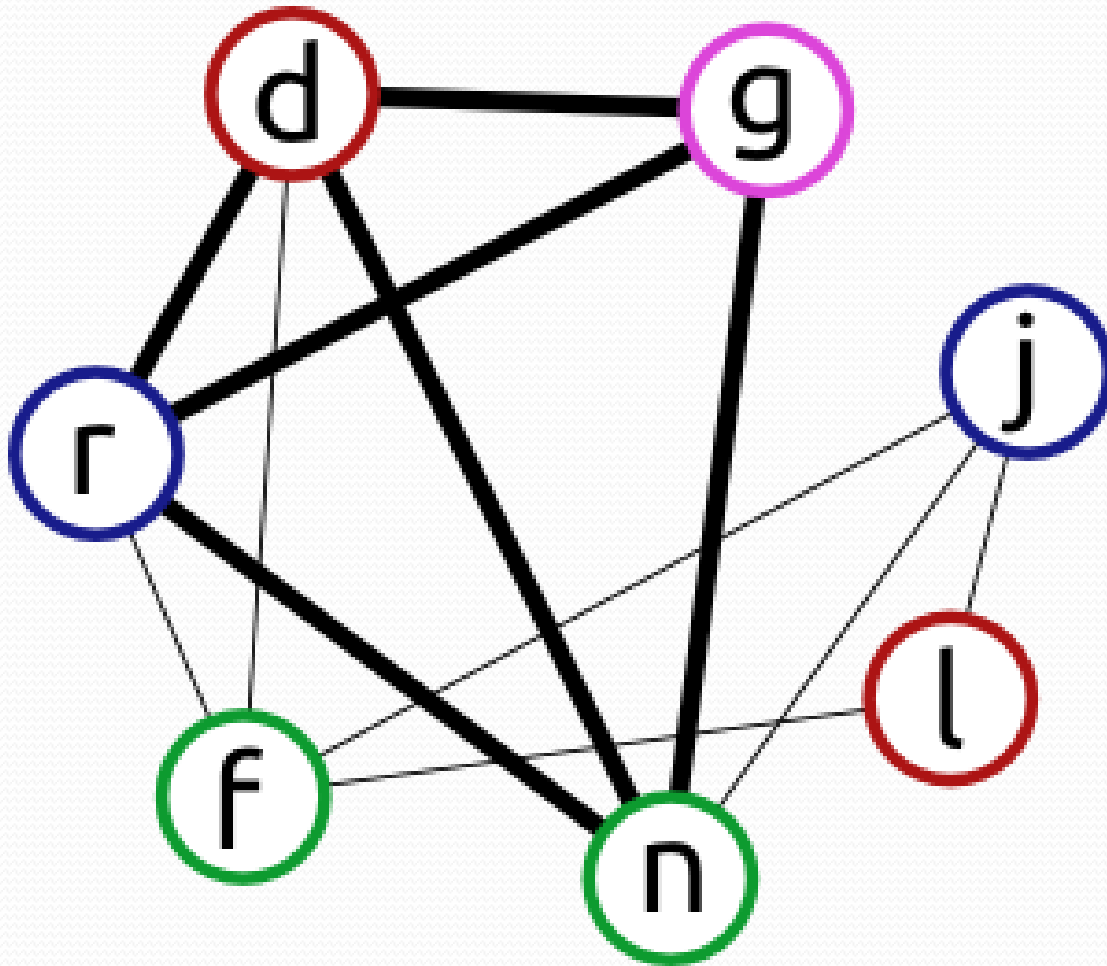
# Decision vs. Optimization

## Example (Max-Clique Problem)

Suppose that a graph  $G = (V, E)$  is given.

- **Optimization:** Find a maximal subset  $V' \subseteq V$ , such that the induced graph by  $V'$  is complete graph (The **clique** is a complete subgraph).
- **Decision:** For a given  $k$ , is there a subset  $V' \subseteq V$  with  $|V'| \geq k$ , such that the induced graph by  $V'$  is complete graph (a clique of size at least  $k$ )?

# Decision vs. Optimization



# Decision vs. Optimization

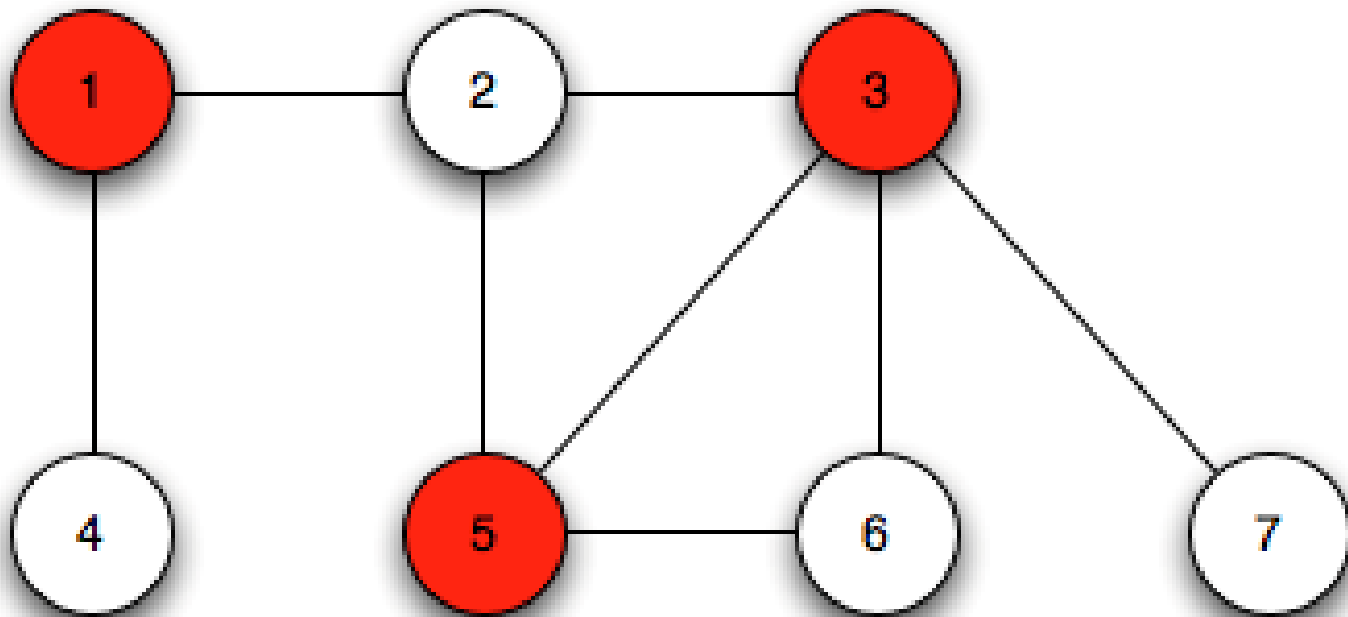
## Example (Min-Vertex Cover Problem)

Suppose that a graph  $G = (V, E)$  is given.

- **Optimization:** Find a minimal subset  $V' \subseteq V$ , such that for each  $e = (v_1, v_2) \in E$ , either  $v_1 \in V'$  or  $v_2 \in V'$  ( $V'$  covers the  $E$ ).
- **Decision:** For a given  $k$ , is there a subset  $V' \subseteq V$  with  $|V'| \leq k$ , such that  $V'$  covers the  $E$ ?



# Decision vs. Optimization

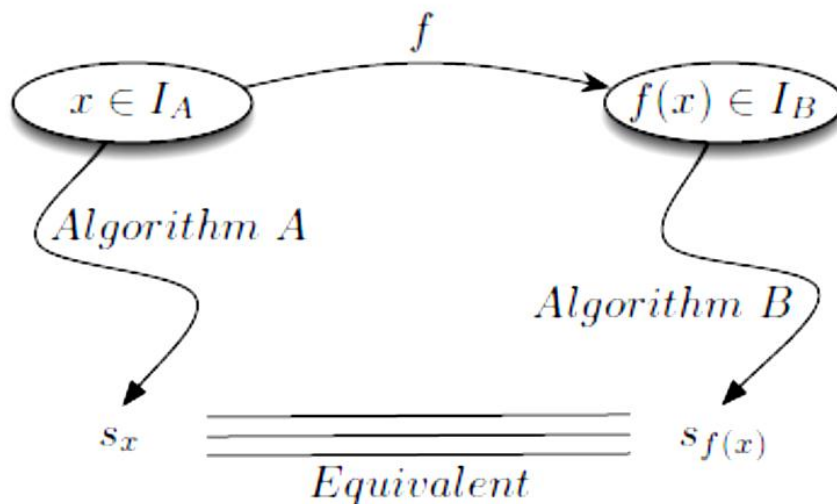


# The principle of Reduction

## Definition

Suppose that  $A$  and  $B$  are two decision problems. We say that  $A$  is reduced to  $B$  (denoted by  $A \leq_P B$ ), if there exists a polynomial algorithm, say  $f$ , such that:

- $x \in \text{Instance}(A) \implies f(x) \in \text{Instance}(B)$ .
- $x$  is a **yes**-instance of  $A \iff f(x)$  is a **yes**-instance of  $B$ .



## Application of reduction

The reduction defines an order over the decision problems with respect to their level of difficulties.

- If  $B$  is easy to solve, then  $A$  is also easy.
- Conversely, If  $A$  is hard to solve, then  $B$  is also hard.

# The theory of NP-Completeness

## Definition

A problem  $L$  is called  $NP$ –Hard if all  $NP$  problems are reduced to  $L$ , i.e.

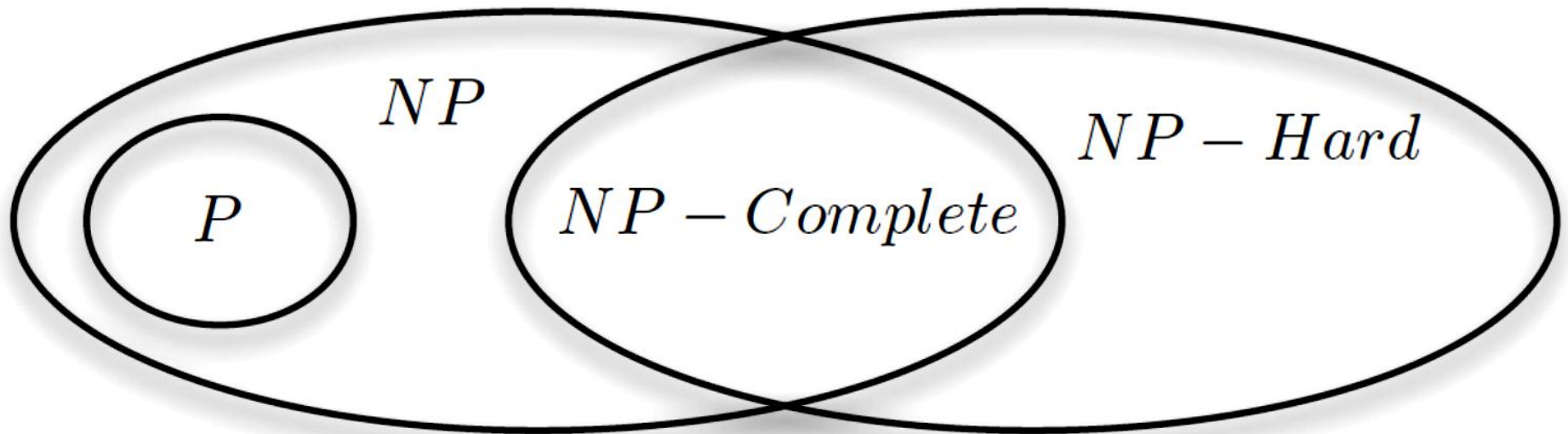
$$L \in NP - Hard \iff \forall L' \in NP : L' \preceq_p L.$$

## Definition

A problem  $L$  is called  $NP$ –Complete if all  $NP$  problems are reduced to  $L$  and  $L \in NP$ , i.e.

$$L \in NP - Complete \iff L \in NP \cap NP - Hard.$$

# The theory of NP-Completeness



# The theory of NP-Completeness

- NP-Complete problems are the “hardest” problems in NP:
  - If any *one* NP-Complete problem can be solved in polynomial time...
  - ...then *every* NP-Complete problem can be solved in polynomial time...
  - ...and in fact *every* problem in **NP** can be solved in polynomial time (which would show **P = NP**)
  - Thus: solve Hamiltonian-cycle in  $O(n^{100})$  time, you’ve proved that **P = NP**. Retire rich & famous.

# Travelling Salesman Problem (TSP)

- ❑ The **travelling salesman problem (TSP)** asks the following question:

Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city and returns to the origin city?

- ❑ **A very important question:**
- ❑ Which complexity class does the TSP belong to? (NP-complete or NP-hard?)