

# CSE 317: Design and Analysis of Algorithms

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# Computational Complexity

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# Decision Problems vs Optimization Problems

## Decision Problem

A computational problem is called *decision problem* if the answer is either **yes** or **no**. For example:

- Is there a path of length  $k$  in a graph  $G$ ?
- Is a given array  $A$  sorted in non-decreasing order?

## Optimization Problem

A computational problem is called *optimization problem* if the answer is a value that needs to be maximized or minimized. For example:

- Find the shortest path between two nodes in a graph  $G$ .
- Find the minimum spanning tree of a graph  $G$ .

# Computational Complexity

- **Computational complexity** is a field that deals with measuring the resources required to solve a computational problem and classifying problems based on their resource requirements
- The most common resource is **time** required to solve a problem
- Often we are only interested in the decision problems and we are interested in the time required to solve the problem

## Model of Computation

- We need a model of computation to measure the time required to solve a problem
- The most common model of computation is the **Turing machine**
- A **Turing machine** is an abstract machine that can simulate any algorithm
- The time complexity of an algorithm is measured in terms of the number of steps required to solve a problem

# Languages and Decision Problems

## Language

A *language* is a set of strings over a finite alphabet ( $\Sigma$ ) For example, the set of all binary strings that represent prime numbers is a language

- We can now redefine the decision problem as follows:
- **Decision Problem:** Given a string  $x \in \Sigma^*$ , is  $x$  in the language  $L$ ?
- That is., a decision problem is a kind of membership query for a string in a language

# Algorithms and Machines

- Let  $\Sigma$  be a finite alphabet and  $\mathcal{A}$  be an algorithm (machine) that implements a mapping

$$\varphi : \Sigma^* \rightarrow \{0, 1\}$$

- We say  $\mathcal{A}$  is a polynomial time algorithm if there exists a polynomial  $p(n)$  such that for all  $x \in \Sigma^*$ , the algorithm  $\mathcal{A}$  halts in at most  $p(|x|)$  steps
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# The Class P

## The class P

The class **P** contains all languages (decision problems) which can be decided by some polynomial time algorithm i.e., the membership queries can be computed in polynomial time by some algorithm

## The Class NP

- Let  $\Sigma$  be a finite alphabet
- We say  $Q : \Sigma^* \rightarrow \{0, 1\}$  is a polynomial time predicate if there exists a polynomial time algorithm that for given words  $\alpha, \beta \in \Sigma^*$ , compute  $Q(\alpha, \beta)$

### The class NP

We say that a language  $L \subseteq \Sigma^*$  is in the class **NP** if and only if there exists a polynomial  $q$  and a polynomial time predicate  $Q$  such that for any word  $\alpha \in \Sigma^*$ , we have  $\alpha \in L$  if and only if there exists a word  $\beta \in \Sigma^*$  such that  $Q(\alpha, \beta) = 1$  and  $|\beta| \leq q(|\alpha|)$  in other words:

$$L \subseteq \mathbf{NP} \iff \exists q \ \exists Q \ \forall \alpha \in \Sigma^* \ \exists \beta \in \Sigma^* \ Q(\alpha, \beta) = 1, \quad \text{s.t.} \quad |\beta| \leq q(|\alpha|)$$

The word  $\beta$  is called the *certificate* and the algorithm which computes the predicate  $Q$  is called the *verifier*.

# The Classes P and NP

## Theorem

$P \subseteq NP$ .

## Proof.

- Let  $L \in P$ ,  $L \subseteq \Sigma^*$ , and  $q(n) = 1$
- Let us consider a predicate  $Q(\alpha, \beta)$  such that  $Q(\alpha, \beta) = 1$  if and only if  $\alpha \in L$
- Since  $L \in P$ , the predicate  $Q$  is a polynomial time predicate
- It is clear that  $\alpha \in L$  if and only if there exists  $\beta \in \Sigma^*$  such that  $|\beta| \leq 1$  and  $Q(\alpha, \beta) = 1$  (we can take  $\beta = \epsilon$ )



# Reduction

## Polynomial time reduction

We say that a language  $L_1 \subseteq \Sigma^*$  is a *polynomial-time reducible* to a language  $L_2 \in \Sigma^*$  (written as  $L_1 \leq_{\mathbf{P}} L_2$ ) if there exists a polynomial time computable function  $\varphi : \Sigma^* \rightarrow \Sigma^*$  such that, for all  $\alpha$  we have  $\alpha \in L_1$  if and only if  $\varphi(\alpha) \in L_2$ .

## Proposition

1. If  $L_1 \leq_{\mathbf{P}} L_2$  and  $L_2 \in \mathbf{P}$  then  $L_1 \in \mathbf{P}$ .
2. If  $L_1 \leq_{\mathbf{P}} L_2$  and  $L_1 \notin \mathbf{P}$  then  $L_2 \notin \mathbf{P}$ .
3. If  $L_1 \leq_{\mathbf{P}} L_2$  and  $L_2 \leq_{\mathbf{P}} L_3$  then  $L_1 \leq_{\mathbf{P}} L_3$ . (Transitivity property.)

# NP-hard and NP-complete Problems

## NP-hard problems

A language  $L$  is **NP-hard** if for all  $L' \in \text{NP}$ , we have  $L' \leq_P L$ .

## NP-complete problems

A language  $L$  is **NP-complete** if  $L \in \text{NP}$  and  $L$  is **NP-hard**.

- If  $L$  is **NP-hard** and  $L \in \text{P}$ , then  $\text{P} = \text{NP}$ .
- If  $L$  is **NP-hard** and  $\text{P} \neq \text{NP}$ , then  $L \notin \text{P}$ .

# Boolean Functions and CNF

## Boolean Function/Formula

A function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  is called a *boolean function* of  $n$  variables.

- A boolean formula of the following kind is called a *conjunctive normal form* (CNF):

$$f(x_1, x_2, \dots, x_n) = C_1 \wedge C_2 \wedge \dots \wedge C_m$$

where each  $C_j$  is a *clause* of the form  $l_{j1} \vee l_{j2} \vee \dots \vee l_{jk}$  and each  $l_{ji}$  is a *literal* which is either a variable  $x_i$  or its negation  $\neg x_i$

- We assume that each variable occurs in each clause  $C_j$  at most once

# Satisfiability Problem

## Satisfiability Problem

**Decision Problem:** SAT

**Setup:** A boolean formula  $f$  of  $n$  variables in CNF

**Question:** Is there an assignment of values to the variables such that  $f$  evaluates to TRUE?

- We show that SAT is NP-complete
- We need to show that SAT is in NP and it is NP-hard

## Satisfiability Problem is in NP

- Given a boolean formula  $f$  in CNF, we can verify in polynomial time whether a given assignment of values to the variables makes  $f$  evaluate to TRUE
- We can verify this by checking each clause  $C_j$  and checking if at least one literal in the clause evaluates to TRUE
- Thus, SAT is in NP

# Satisfiability Problem is **NP-hard**

## Theorem (Cook-Levin 1971)

**SAT** is **NP-complete**.

- This proof was independently discovered by Stephen Cook (in United States) and Leonid Levin (in, then, Soviet Union) in 1971
- This proof laid down the theoretical/mathematical foundation of the theory of **NP-complete** problems
- In the subsequent we present some **NP-complete** problems all of which can proved to be **NP-hard** using the polynomial time reduction (due to transitivity property of reduction)

# The 3 SAT problem

## The 3 SAT problem

**Decision Problem:**  $\text{3SAT}$

**Setup:** A boolean formula  $f$  of  $n$  variables in CNF where each clause has at most 3 literals

**Question:** Is there an assignment of values to the variables such that  $f$  evaluates to TRUE?

- $\text{3SAT} \in \mathbf{NP}$  (trivial)
- $\text{SAT} \leq_{\mathbf{P}} \text{3SAT}$  (left as an exercise)

# Independent Set Problem

## Independent Set Problem

Let  $G = (V, E)$  be an undirected graph. We say a subset of vertices  $S \subseteq V$  is *independent set* if no two vertices in  $S$  are adjacent. There is no edge that joins any two vertices in  $S$ .

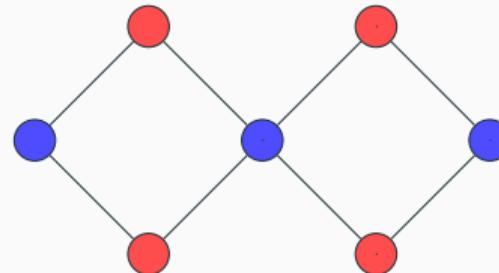
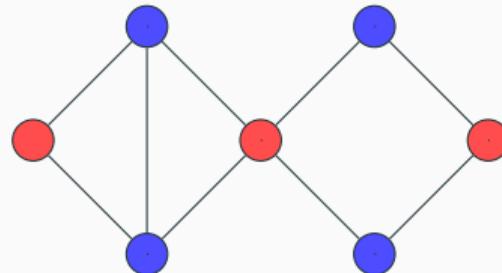
The optimization version of the *independent set problem* is to find an independent set of maximum cardinality in  $G$ .

**Decision Problem:** [IS](#)

**Setup:** An undirected graph  $G = (V, E)$  and an integer  $k$

**Question:** Is there an independent set of size at least  $k$  in  $G$ ?

## Independent Set Problem



## Proving Independent Set to **NP-complete**

1. We need to show that **IS**  $\in \text{NP}$
2. We show that for some **NP**-complete problem  $\Pi$ ,  $\Pi \leq_P \text{IS}$

## Showing Independent Set is in NP

- Given a graph  $G = (V, E)$  and an integer  $k$ , we can verify in polynomial time whether a given subset of vertices  $S \subseteq V$  is an independent set of size at least  $k$
- We can verify this by checking if no two vertices in  $S$  are adjacent and if  $|S| \geq k$
- Thus, **IS** is in **NP**

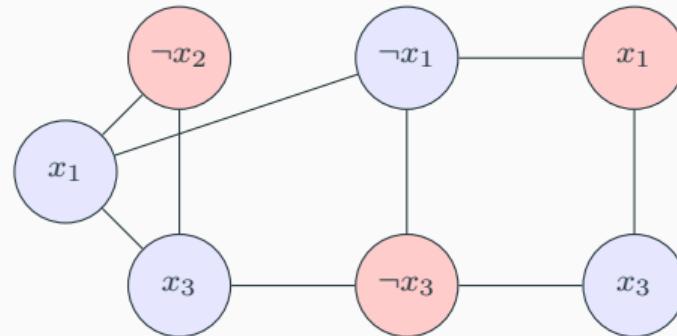
## Showing Independent Set is NP-hard

- We show that IS is NP-hard by showing that  $\text{3SAT} \leq_{\text{P}} \text{IS}$
- Given a boolean formula  $f = C_1 \wedge C_2 \wedge \dots \wedge C_m$  in 3CNF, where each  $C_j = l_{j1} \vee l_{j2} \vee l_{j3}$ , we construct a graph  $G = (V, E)$  such that  $f$  is satisfiable if and only if  $G$  has an independent set of size at least  $k$
- We construct the graph  $G$  as follows:
  - Each literal in each clause has a corresponding vertex in  $G$
  - Each clause  $C_j$  has a corresponding triangle in  $G$
  - Each vertex  $v$  from each clause is connected to  $\neg v$  from other clauses and vice versa
  - We set  $k = m$
  - We show that  $f$  is satisfiable if and only if  $G$  has an independent set of size at least  $k$

## Example Reduction

- Polynomial-time reduction of **3SAT** to **IS**

$$f = (x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_3) \wedge (x_1 \vee x_3)$$



- Above formula is satisfied:  $x_1 = 1, x_2 = 0, x_3 = 0$

# Vertex Cover Problem

## Vertex Cover Problem

Let  $G = (V, E)$  be an undirected graph. We say a subset of vertices  $S \subseteq V$  is a *vertex cover* if every edge in  $E$  has one endpoint in  $S$ .

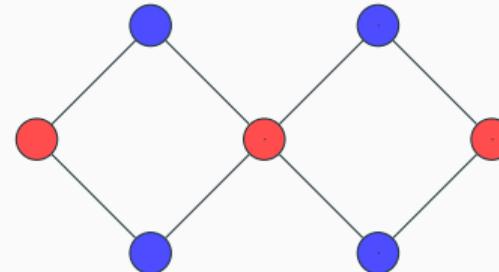
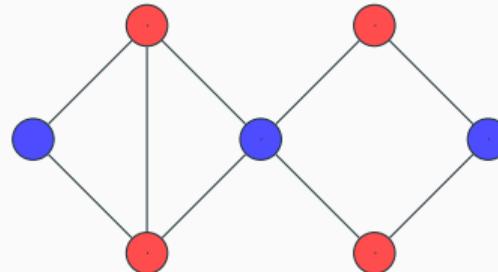
The optimization version of the *vertex cover problem* is to find a vertex cover of minimum cardinality in  $G$ .

**Decision Problem:** [VC](#)

**Setup:** An undirected graph  $G = (V, E)$  and an integer  $k$

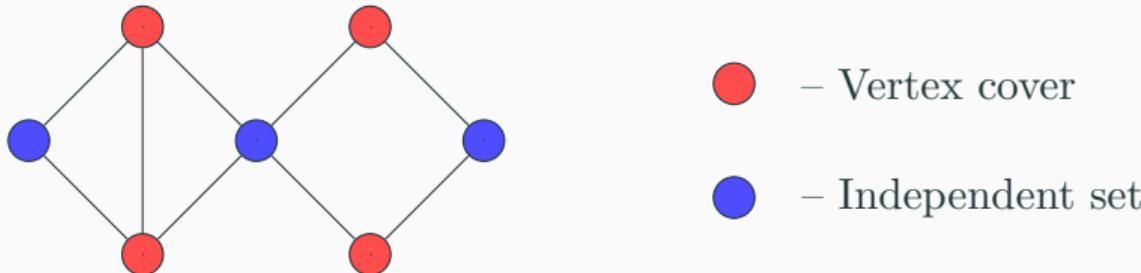
**Question:** Is there a vertex cover of size at most  $k$  in  $G$ ?

## Vertex Cover Problem



## Proving Vertex Cover is NP-complete

- We show that **VC** is **NP-hard** by showing that **IS**  $\leq_P$  **VC**
- Clearly, if  $S$  is an independent set in  $G$ , then  $V \setminus S$  is a vertex cover in  $G$



- If  $S$  is an independent set, each  $e \in E$  has at least one endpoint in  $V \setminus S$
- Therefore,  $V \setminus S$  is a vertex cover in  $G$
- It implies that  $G$  has an independent set with at least  $k$  vertices if and only if  $G$  has a vertex cover with at most  $n - k$  vertices
- Therefore, **VC** is **NP-complete**

# Set Cover Problem

## Set Cover Problem

Let  $\mathcal{U}$  be a set and  $\mathcal{F}$  a family of subsets of  $\mathcal{U}$  s.t.,  $\mathcal{U} = \bigcup_{S \in \mathcal{F}} S$ .

The optimization version of the *set cover problem* is to find a set cover of minimum cardinality in  $\mathcal{F}$ .

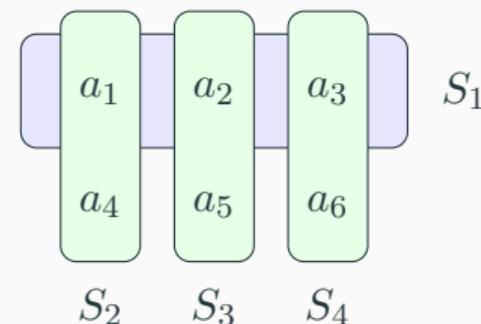
**Decision Problem:** [SC](#)

**Setup:** A finite set  $\mathcal{U}$ , a collection of subsets  $\mathcal{F}$  of  $\mathcal{U}$ , and an integer  $k$

**Question:** Is there a set cover of size at most  $k$  in  $\mathcal{F}$ ?

## Set Cover Problem

- Let  $\mathcal{U} = \{a_1, a_2, a_3, a_4, a_5, a_6\}$  and  $\mathcal{F} = \{S_1, S_2, S_3, S_4\}$
- $S_1 = \{a_1, a_2, a_3\}$ ,  $S_2 = \{a_1, a_4\}$ ,  $S_3 = \{a_2, a_5\}$ ,  $S_4 = \{a_3, a_6\}$



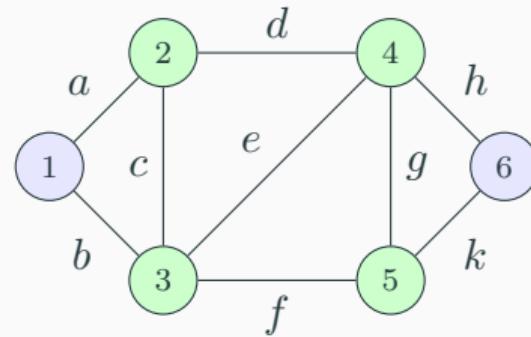
## Set Cover Problem

- It is easy to show that **SC** is in **NP**
- We show that **VC**  $\leq_{\text{P}}$  **SC**
- Let  $G = (V, E)$  be an undirected graph and  $k$  be an integer
- We construct a set cover problem  $(\mathcal{U}_G, \mathcal{F}_G)$  as follows:
  - Let  $\mathcal{U}_G = E$  and  $\mathcal{F}_G = \{S_v \mid v \in V\}$
  - Where  $S_v$  is the set of edges from  $E$  that have  $v$  as an endpoint
  - We set  $k = |V|$
  - We show that  $\{v_{i_1}, \dots, v_{i_k}\}$  is a vertex cover for  $G$  if and only if:

$$\bigcup_{S_v \in \{S_{v_{i_1}}, \dots, S_{v_{i_k}}\}} S_v = \mathcal{U}_G$$

- Thus, **SC** is **NP**-complete

## Set Cover Problem



- The set cover problem  $\mathcal{U}_G, \mathcal{F}_G$  for the above graph  $G = (V, E)$  is:
- $\mathcal{U}_G = \{a, b, c, d, e, f, g, h, k\}$
- $\mathcal{F}_G = \{S_1, S_2, S_3, S_4, S_5, S_6\}$
- $S_1 = \{a, d\}, S_2 = \{a, c\}, S_3 = \{d, e\}, S_4 = \{d, h\}, S_5 = \{e, g\}, S_6 = \{h, k\}$