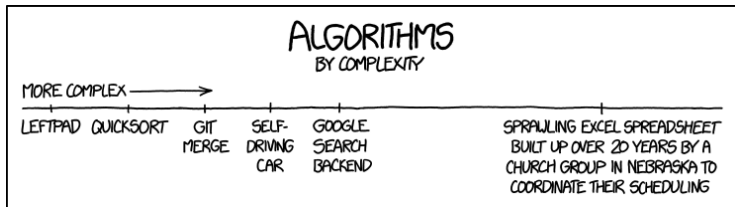


# NP and Computational Intractability

T. M. Murali

November 7, 12, 2018

# Algorithm Design



## • Patterns

- ▶ Greed.
- ▶ Divide-and-conquer.
- ▶ Dynamic programming.
- ▶ Duality.

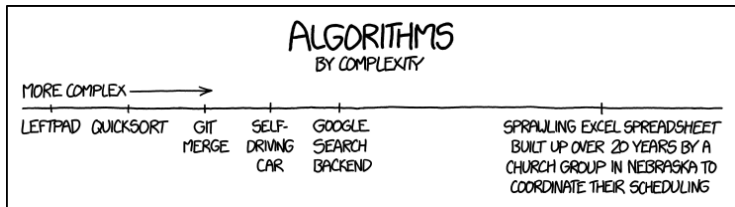
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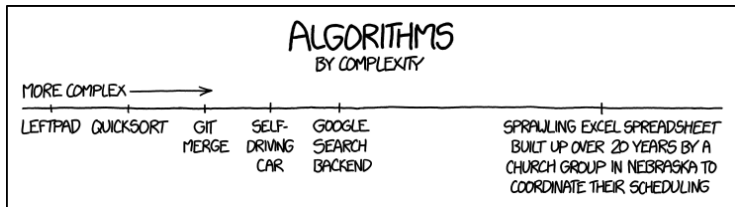
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## • “Anti-patterns”

- ▶ NP-completeness.
- ▶ PSPACE-completeness.
- ▶ Undecidability.

$O(n^k)$  algorithm unlikely.

$O(n^k)$  certification algorithm unlikely.

No algorithm possible.

# Computational Tractability

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## Polynomial time

Shortest path

Matching

Minimum cut

2-SAT

Planar four-colour

Bipartite vertex cover

Primality testing

## Probably not

Longest path

3-D matching

Maximum cut

3-SAT

Planar three-colour

Vertex cover

Factoring



# Problem Classification

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- Some extremely hard problems cannot be solved efficiently (e.g., chess on an  $n$ -by- $n$  board).
- However, classification is unclear for a very large number of discrete computational problems.
- We can prove that these problems are fundamentally equivalent and are manifestations of the same problem!

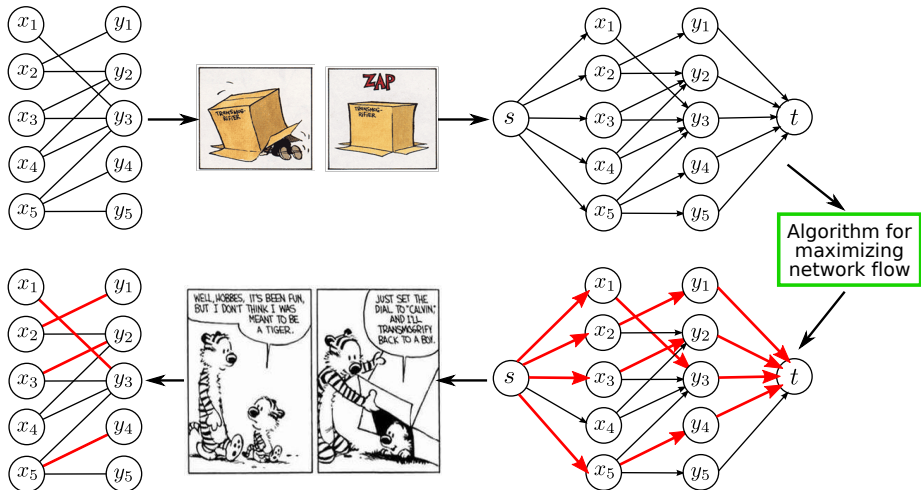
# Polynomial-Time Reduction

- Goal is to express statements of the type “Problem  $X$  is at least as hard as problem  $Y$ .”
- Use the notion of *reductions*.
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- $Y \leq_P X$  implies that “ $X$  is at least as hard as  $Y$ .”
- Such reductions are *Karp reductions*. *Cook reductions* allow a polynomial number of calls to the black box that solves  $X$ .

# Usefulness of Reductions

- Claim: If  $Y \leq_P X$  and  $X$  can be solved in polynomial time, then  $Y$  can be solved in polynomial time.

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- Contrapositive: If  $Y \leq_P X$  and  $Y$  cannot be solved in polynomial time, then  $X$  cannot be solved in polynomial time.
- Informally: If  $Y$  is hard, and we can show that  $Y$  reduces to  $X$ , then the hardness “spreads” to  $X$ .

# Reduction Strategies

- Simple equivalence.
- Special case to general case.
- Encoding with gadgets.

# Optimisation versus Decision Problems

- So far, we have developed algorithms that solve optimisation problems.
  - ▶ Compute the *largest* flow.
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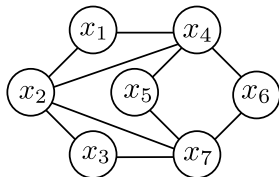
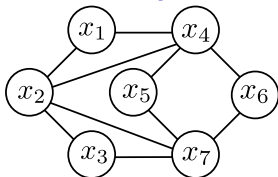
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  - ▶ Compute the *largest* flow.
  - ▶ Find the *closest* pair of points.
  - ▶ Find the schedule with the *least* completion time.
- Now, we will focus on *decision versions* of problems, e.g., is there a flow with value at least  $k$ , for a given value of  $k$ ?
- Decision problem: answer to every input is yes or no.

PRIMES

**INSTANCE:** A natural number  $n$

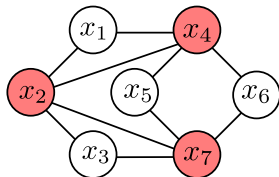
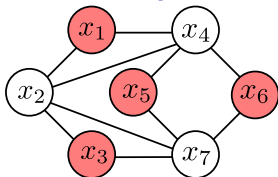
**QUESTION:** Is  $n$  prime?

## Independent Set and Vertex Cover



- Given an undirected graph  $G(V, E)$ , a subset  $S \subseteq V$  is an *independent set* if no two vertices in  $S$  are connected by an edge.
- Given an undirected graph  $G(V, E)$ , a subset  $S \subseteq V$  is a *vertex cover* if every edge in  $E$  is incident on at least one vertex in  $S$ .

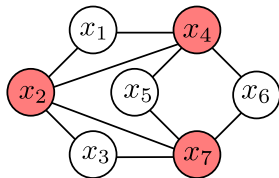
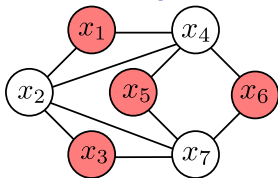
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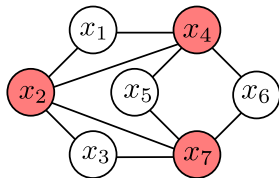
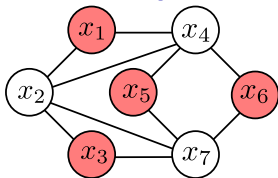
**QUESTION:** Does  $G$  contain an independent set of size  $\geq k$ ?

## VERTEX COVER

**INSTANCE:** Undirected graph  $G$  and an integer  $l$

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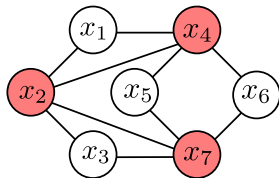
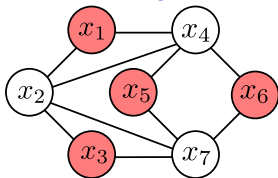
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- Demonstrate simple equivalence between these two problems.
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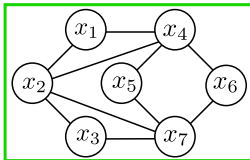
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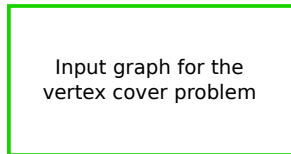
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# Strategy for Proving Indep. Set $\leq_P$ Vertex Cover

$k = 3$



$l = ?$



Yes, there is an independent set of size at least 3

No, every independent set is of size 3 or less



Yes

No

**Black box algorithm for solving vertex cover**

# Strategy for Proving Indep. Set $\leq_P$ Vertex Cover

- ① Start with an arbitrary instance of INDEPENDENT SET: an undirected graph  $G(V, E)$  and an integer  $k$ .
- ② From  $G(V, E)$  and  $k$ , create an instance of VERTEX COVER: an undirected graph  $G'(V', E')$  and an integer  $l$ .
  - ▶  $G'$  related to  $G$  in some way.
  - ▶  $l$  can depend upon  $k$  and size of  $G$ .
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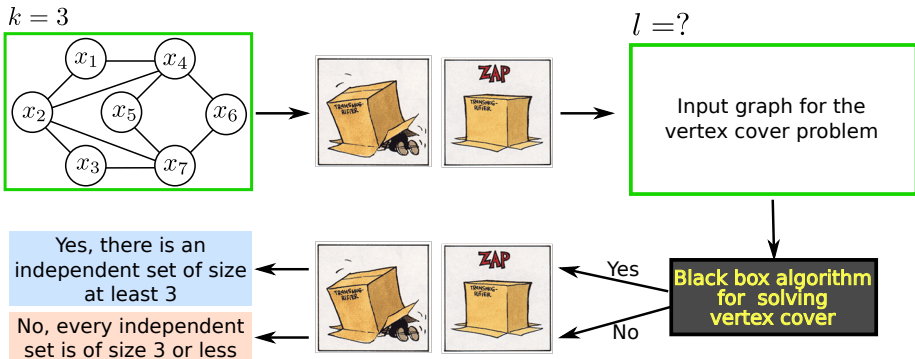


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  - Transformation and proof must be correct for all possible graphs  $G(V, E)$  and all possible values of  $k$ .
  - Why is the proof an iff statement?

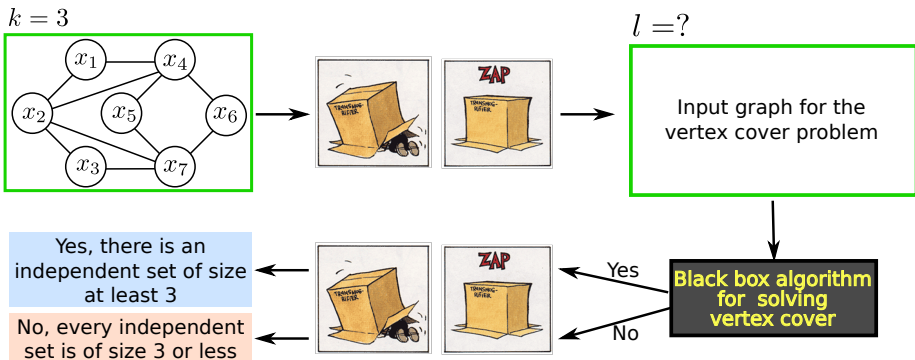


# Reason for Two-Way Proof



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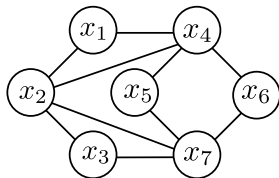
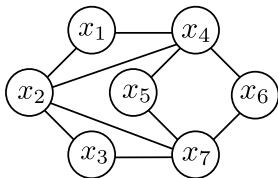
# Reason for Two-Way Proof



- Why is the proof an **iff** statement? In the reduction, we are using black box for VERTEX COVER to solve INDEPENDENT SET.
  - ⊕ If there is an independent set size  $\geq k$ , we must be sure that there is a vertex cover of size  $\leq l$ , so that we know that the black box will find this vertex cover.
  - ⊕ If the black box finds a vertex cover of size  $\leq l$ , we must be sure we can construct an independent set of size  $\geq k$  from this vertex cover.

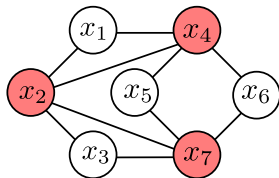
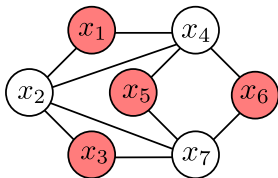


# Proof that Independent Set $\leq_P$ Vertex Cover



- 1 Arbitrary instance of INDEPENDENT SET: an undirected graph  $G(V, E)$  and an integer  $k$ .
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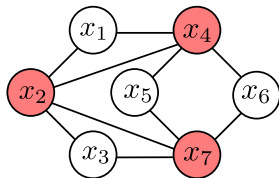
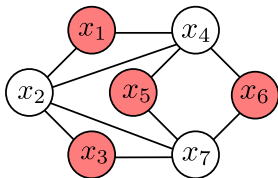
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- Same idea proves that VERTEX COVER  $\leq_P$  INDEPENDENT SET

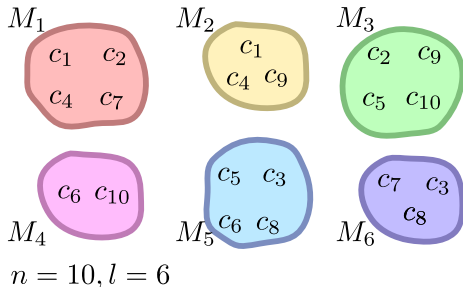
# Vertex Cover and Set Cover

- INDEPENDENT SET is a “packing” problem: pack as many vertices as possible, subject to constraints (the edges).
- VERTEX COVER is a “covering” problem: cover all edges in the graph with as few vertices as possible.
- There are more general covering problems.

## MICROBE COVER

**INSTANCE:** A set  $U$  of  $n$  compounds, a collection  $M_1, M_2, \dots, M_l$  of microbes, where each microbe can make a subset of compounds in  $U$ , and an integer  $k$ .

**QUESTION:** Is there a subset of  $\leq k$  microbes that can together make all the compounds in  $U$ ?



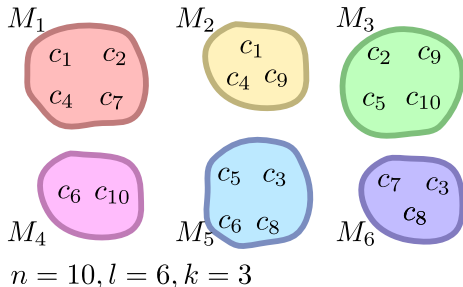
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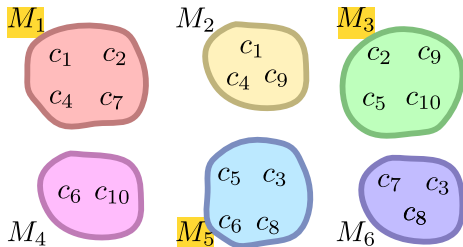
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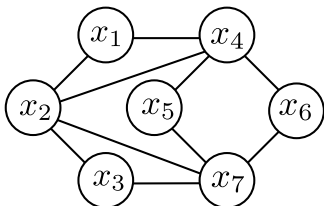
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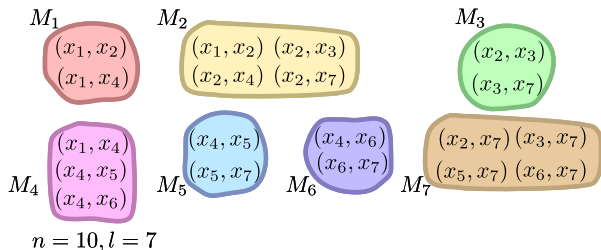
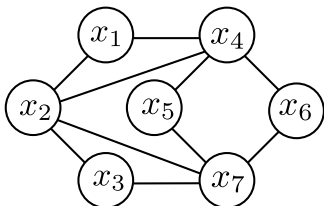
$$n = 10, l = 6, k = 3$$

## Vertex Cover $\leq_P$ Microbe Cover



- Input to VERTEX COVER: an undirected graph  $G(V, E)$  and an integer  $k$ .
- Let  $|V| = l$ .
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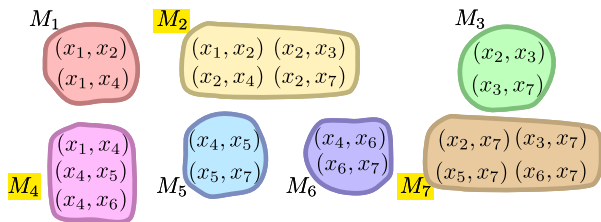
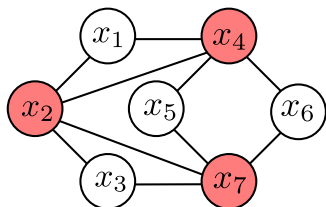
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# Vertex Cover $\leq_P$ Microbe Cover



$$n = 10, l = 7$$

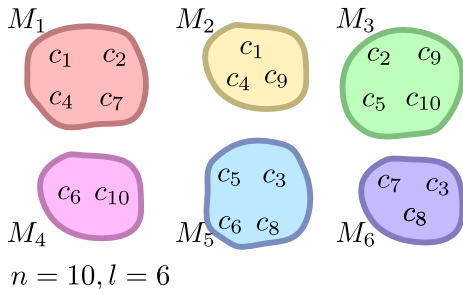
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  - ▶  $U = E$ , i.e., each element of  $U$  is an edge of  $G$ , and
  - ▶ for each node  $i \in V$ , create a microbe  $M_i$  whose compounds are the set of edges incident on  $i$ .
- Claim:  $U$  can be covered with  $\leq k$  microbes iff  $G$  has a vertex cover with at  $\leq k$  nodes.
- Proof strategy:
  - 1 If  $G$  has a vertex cover of size  $\leq k$ , then  $U$  can be covered with  $\leq k$  microbes.
  - 2 If  $U$  can be covered with  $\leq k$  microbes, then  $G$  has a vertex cover of size  $\leq k$ .

# Microbe Cover and Set Cover

## MICROBE COVER

**INSTANCE:** A set  $U$  of  $n$  compounds, a collection  $M_1, M_2, \dots, M_l$  of microbes, where each microbe can make a subset of compounds in  $U$ , and an integer  $k$ .

**QUESTION:** Is there a subset of  $\leq k$  microbes that can together make all the compounds in  $U$ ?



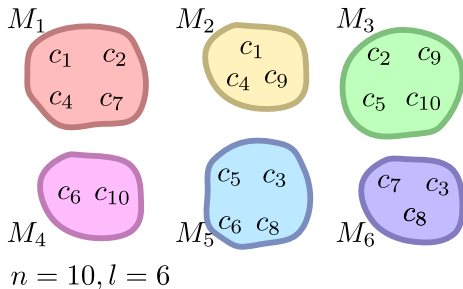
- Purely combinatorial problem: a “microbe” is just a set of “compounds.”

# Microbe Cover and Set Cover

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**QUESTION:** Is there a subset of  $\leq k$  microbes that can together make all the compounds in  $U$ ?



- Purely combinatorial problem: a “microbe” is just a set of “compounds.”

## SET COVER

**INSTANCE:** A set  $U$  of  $n$  elements, a collection  $S_1, S_2, \dots, S_m$  of subsets of  $U$ , and an integer  $k$ .

**QUESTION:** Is there a collection of  $\leq k$  sets in the collection whose union is  $U$ ?

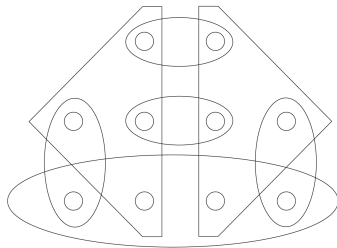


Figure 8.2 An instance of the Set Cover Problem.

# Boolean Satisfiability

- Abstract problems formulated in Boolean notation.

# Boolean Satisfiability

- Abstract problems formulated in Boolean notation.
- Given a set  $X = \{x_1, x_2, \dots, x_n\}$  of  $n$  Boolean variables.
- Each variable can take the value 0 or 1.
- *Term*: a variable  $x_i$  or its negation  $\overline{x_i}$ .
- *Clause* of *length*  $l$ : (or) of  $l$  distinct terms  $t_1 \vee t_2 \vee \dots \vee t_l$ .
- *Truth assignment* for  $X$ : is a function  $\nu : X \rightarrow \{0, 1\}$ .
- An assignment  $\nu$  *satisfies* a clause  $C$  if it causes at least one term in  $C$  to evaluate to 1 (since  $C$  is an or of terms).
- An assignment *satisfies* a collection of clauses  $C_1, C_2, \dots, C_k$  if it causes all clauses to evaluate to 1, i.e.,  $C_1 \wedge C_2 \wedge \dots \wedge C_k = 1$ .
  - ▶  $\nu$  is a *satisfying assignment* with respect to  $C_1, C_2, \dots, C_k$ .
  - ▶ set of clauses  $C_1, C_2, \dots, C_k$  is *satisfiable*.

# Example

- $X = \{x_1, x_2, x_3, x_4\}$
- Terms:  $x_1, \overline{x_1}, x_2, \overline{x_2}, x_3, \overline{x_3}, x_4, \overline{x_4}$

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- Terms:  $x_1, \overline{x_1}, x_2, \overline{x_2}, x_3, \overline{x_3}, x_4, \overline{x_4}$
- Clauses:

$$x_1 \vee \overline{x_2} \vee \overline{x_3}$$

$$x_2 \vee \overline{x_3} \vee x_4$$

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- Clauses:
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  - $x_2 \vee \overline{x_3} \vee x_4$
  - $x_3 \vee \overline{x_4}$
- Assignment:  $x_1 = 1, x_2 = 0, x_3 = 1, x_4 = 0$



# Example

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- Terms:  $x_1, \overline{x_1}, x_2, \overline{x_2}, x_3, \overline{x_3}, x_4, \overline{x_4}$
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  - $x_2 \vee \overline{x_3} \vee x_4$
  - $x_3 \vee \overline{x_4}$
- Assignment:  $x_1 = 1, x_2 = 0, x_3 = 1, x_4 = 0$ 
  - $x_1 \vee \overline{x_2} \vee \overline{x_3}$
  - $x_2 \vee \overline{x_3} \vee x_4$
  - $x_3 \vee \overline{x_4}$
  - ▶ Not a satisfying assignment

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- ▶ Is a satisfying assignment

# SAT and 3-SAT

## SATISFIABILITY PROBLEM (SAT)

**INSTANCE:** A set of clauses  $C_1, C_2, \dots, C_k$  over a set  $X = \{x_1, x_2, \dots, x_n\}$  of  $n$  variables.

**QUESTION:** Is there a satisfying truth assignment for  $X$  with respect to  $C$ ?

# SAT and 3-SAT

## 3-SATISFIABILITY PROBLEM (3-SAT)

**INSTANCE:** A set of clauses  $C_1, C_2, \dots, C_k$ , each of length three, over a set  $X = \{x_1, x_2, \dots, x_n\}$  of  $n$  variables.

**QUESTION:** Is there a satisfying truth assignment for  $X$  with respect to  $C$ ?

# SAT and 3-SAT

## 3-SATISFIABILITY PROBLEM (SAT)

**INSTANCE:** A set of clauses  $C_1, C_2, \dots, C_k$ , each of length three, over a set  $X = \{x_1, x_2, \dots, x_n\}$  of  $n$  variables.

**QUESTION:** Is there a satisfying truth assignment for  $X$  with respect to  $C$ ?

- SAT and 3-SAT are fundamental combinatorial search problems.
- We have to make  $n$  independent decisions (the assignments for each variable) while satisfying a set of constraints.
- Satisfying each constraint in isolation is easy, but we have to make our decisions so that all constraints are satisfied simultaneously.

# Examples of 3-SAT

Example:

- ▶  $C_1 = x_1 \vee 0 \vee 0$
- ▶  $C_2 = x_2 \vee 0 \vee 0$
- ▶  $C_3 = \overline{x_1} \vee \overline{x_2} \vee 0$

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❶ Is  $C_1 \wedge C_2$  satisfiable?



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- 3 Is  $C_2 \wedge C_3$  satisfiable?

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- 4 Is  $C_1 \wedge C_2 \wedge C_3$  satisfiable? No.

## 3-SAT and Independent Set

$$C_1 = x_1 \vee \overline{x_2} \vee \overline{x_3}$$

$$C_2 = \overline{x_1} \vee x_2 \vee x_4$$

$$C_3 = \overline{x_1} \vee x_3 \vee \overline{x_4}$$

- We want to prove  $3\text{-SAT} \leq_P \text{INDEPENDENT SET}$ .



## 3-SAT and Independent Set

$$C_1 = x_1 \vee \overline{x_2} \vee \overline{x_3} \quad \textcircled{1} \text{ Select } x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1.$$

$$C_2 = \overline{x_1} \vee x_2 \vee x_4$$

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- We want to prove  $3\text{-SAT} \leq_P \text{INDEPENDENT SET}$ .
- Two ways to think about 3-SAT:
  - ① Make an independent 0/1 decision on each variable and succeed if we achieve one of three ways in which to satisfy each clause.

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  - ② Choose (at least) one term from each clause. Find a truth assignment that causes each chosen term to evaluate to 1. Ensure that no two terms selected *conflict*, e.g., select  $\overline{x_2}$  in  $C_1$  and  $x_2$  in  $C_2$ .

## 3-SAT and Independent Set

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$$C_3 = \overline{x_1} \vee x_3 \vee \overline{x_4} \quad \triangleright \text{ Choices of selected literals imply } x_1 = 0, x_2 = 0, x_4 = 1.$$

- We want to prove  $3\text{-SAT} \leq_P \text{INDEPENDENT SET}$ .
- Two ways to think about 3-SAT:
  - ① Make an independent 0/1 decision on each variable and succeed if we achieve one of three ways in which to satisfy each clause.
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## Proving $3\text{-SAT} \leq_P \text{Independent Set}$

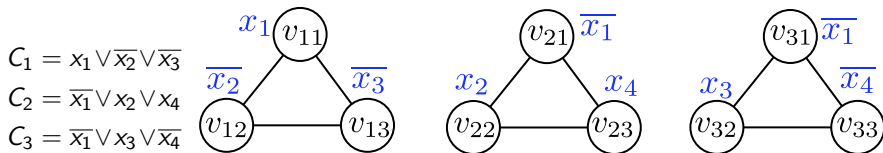
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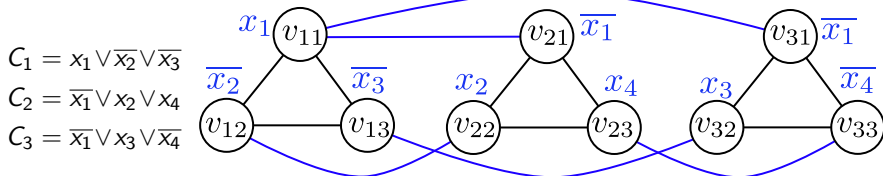
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- Construct an instance of independent set: graph  $G(V, E)$  with  $3k$  nodes.

# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



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# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



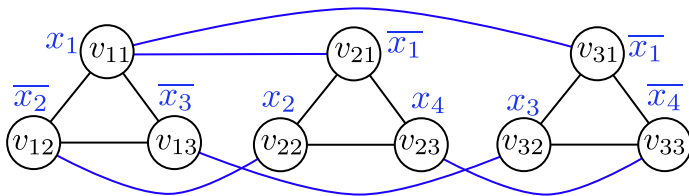
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# Proving $3\text{-SAT} \leq_P \text{Independent Set}$

$$C_1 = x_1 \vee \overline{x_2} \vee \overline{x_3}$$

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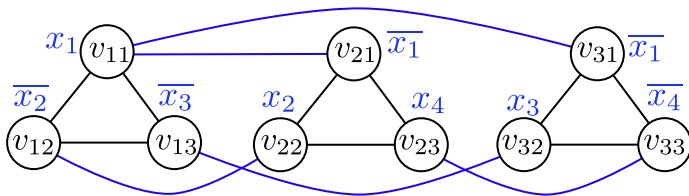
- Claim: 3-SAT instance is satisfiable iff  $G$  has an independent set of size  $k$ .

# Proving $3\text{-SAT} \leq_P \text{Independent Set}$

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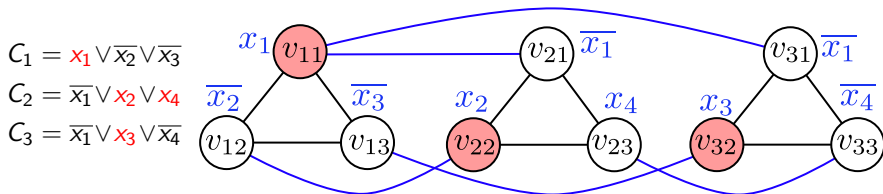
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- Claim: 3-SAT instance is satisfiable iff  $G$  has an independent set of size  $k$ .
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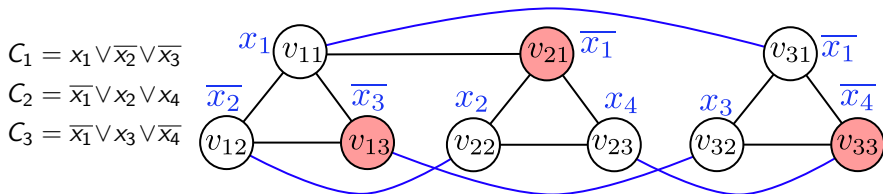


# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



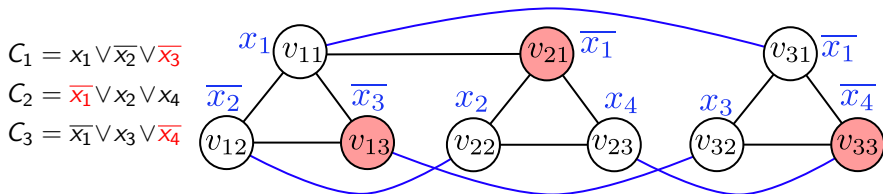
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- Satisfiable assignment  $\rightarrow$  independent set of size  $k$ : Each triangle in  $G$  has at least one node whose label evaluates to 1. Set  $S$  of nodes consisting of one such node from each triangle forms an independent set of size  $= k$ . Why?

# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



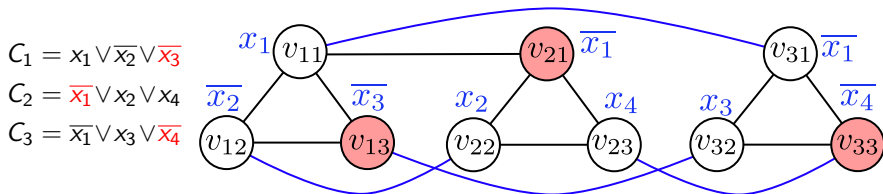
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- Independent set  $S$  of size  $k \rightarrow$  satisfiable assignment:

# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



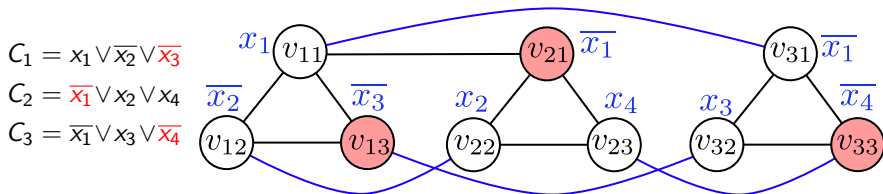
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# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



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- Independent set  $S$  of size  $k \rightarrow$  satisfiable assignment: the size of this set is  $k$ . How do we construct a satisfying truth assignment from the nodes in the independent set?
  - For each variable  $x_i$ , only  $x_i$  or  $\overline{x_i}$  is the label of a node in  $S$ . Why?

# Proving $3\text{-SAT} \leq_P \text{Independent Set}$



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- Independent set  $S$  of size  $k \rightarrow$  satisfiable assignment: the size of this set is  $k$ . How do we construct a satisfying truth assignment from the nodes in the independent set?
  - ▶ For each variable  $x_i$ , only  $x_i$  or  $\overline{x_i}$  is the label of a node in  $S$ . Why?
  - ▶ If  $x_i$  is the label of a node in  $S$ , set  $x_i = 1$ ; else set  $x_i = 0$ .
  - ▶ Why is each clause satisfied?

# Transitivity of Reductions

- Claim: If  $Z \leq_P Y$  and  $Y \leq_P X$ , then  $Z \leq_P X$ .

# Transitivity of Reductions

- Claim: If  $Z \leq_P Y$  and  $Y \leq_P X$ , then  $Z \leq_P X$ .
- We have shown

$3\text{-SAT} \leq_P \text{INDEPENDENT SET} \leq_P \text{VERTEX COVER} \leq_P \text{SET COVER}$

## Finding vs. Certifying

- Is it easy to check if a given set of vertices in an undirected graph forms an independent set of size at least  $k$ ?
- Is it easy to check if a particular truth assignment satisfies a set of clauses?



## Finding vs. Certifying

- Is it easy to check if a given set of vertices in an undirected graph forms an independent set of size at least  $k$ ?
- Is it easy to check if a particular truth assignment satisfies a set of clauses?
- We draw a contrast between *finding* a solution and *checking* a solution (in polynomial time).
- Since we have not been able to develop efficient algorithms to solve many decision problems, let us turn our attention to whether we can check if a proposed solution is correct.

# Problems and Algorithms

PRIMES

**INSTANCE:** A natural number  $n$

**QUESTION:** Is  $n$  prime?

- Decision problem  $X$ : for every input  $s$ , answer  $X(s)$  is yes or no.

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A decision problem  $X$  is in  $\mathcal{P}$  iff there is an algorithm  $A$  with polynomial running time that solves  $X$ .

# Efficient Certification

- A “checking” algorithm for a decision problem  $X$  has a different structure from an algorithm that solves  $X$ .
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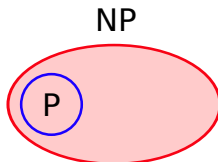
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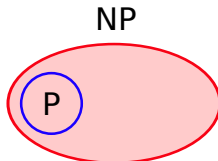
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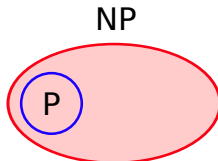
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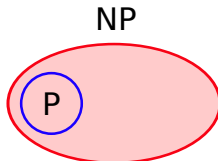
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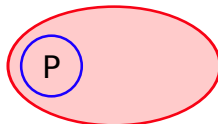
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# Summary

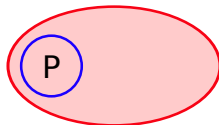
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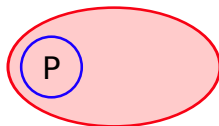
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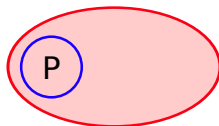
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  - 2 Are there two problems  $X_1$  and  $X_2$  in  $\mathcal{NP}$  such that there is no problem  $X \in \mathcal{NP}$  where  $X_1 \leq_P X$  and  $X_2 \leq_P X$ ?

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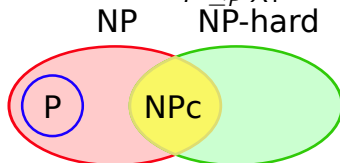
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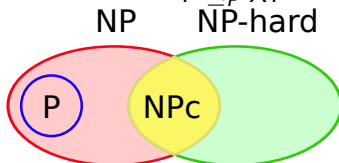
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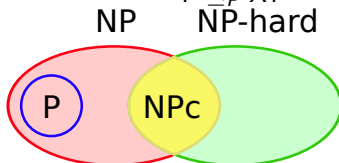
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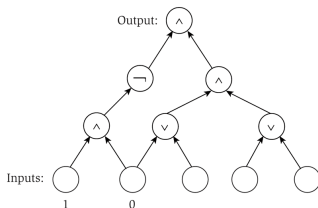
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- Does even one  $\mathcal{NP}$ -Complete problem exist?!** If it does, how can we prove that every problem in  $\mathcal{NP}$  reduces to this problem?

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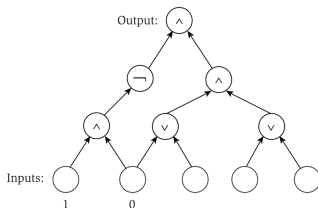
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CIRCUIT SATISFIABILITY

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**QUESTION:** Is there a truth assignment to the inputs that causes the output to have value 1?

► Skip proof; read textbook or Chapter 2.6 of Garey and Johnson.

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- $s \in X$  iff there is an assignment of the input bits of  $K$  that makes  $K$  satisfiable.



# Example of Transformation to Circuit Satisfiability

- Does a graph  $G$  on  $n$  nodes have a two-node independent set?

# Example of Transformation to Circuit Satisfiability

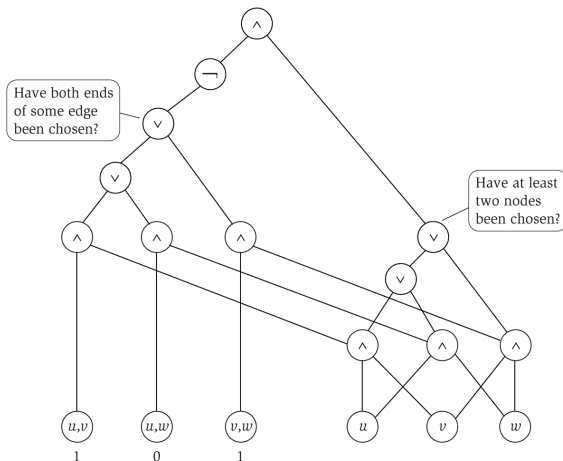
- Does a graph  $G$  on  $n$  nodes have a two-node independent set?
- $s$  encodes the graph  $G$  with  $\binom{n}{2}$  bits.
- $t$  encodes the independent set with  $n$  bits.
- Certifier needs to check if
  - 1 at least two bits in  $t$  are set to 1 and
  - 2 no two bits in  $t$  are set to 1 if they form the ends of an edge (the corresponding bit in  $s$  is set to 1).

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- Suppose  $G$  contains three nodes  $u$ ,  $v$ , and  $w$  with  $v$  connected to  $u$  and  $w$ .

# Example of Transformation to Circuit Satisfiability

- Suppose  $G$  contains three nodes  $u, v$ , and  $w$  with  $v$  connected to  $u$  and  $w$ .



**Figure 8.5** A circuit to verify whether a 3-node graph contains a 2-node independent set.

# Asymmetry of Certification

- Definition of efficient certification and  $\mathcal{NP}$  is fundamentally asymmetric:
  - ▶ An input  $s$  is a “yes” instance iff there exists a short certificate  $t$  such that  $B(s, t) = \text{yes}$ .
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- For a decision problem  $X$ , its *complementary problem*  $\overline{X}$  is the set of inputs  $s$  such that  $s \in \overline{X}$  iff  $s \notin X$ .

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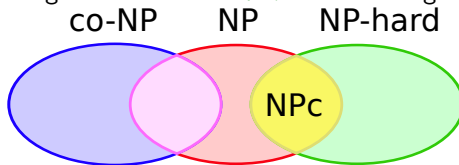
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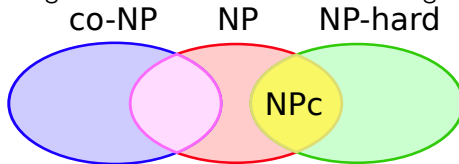
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- Claim: If  $\mathcal{NP} \neq \text{co-}\mathcal{NP}$  then  $\mathcal{P} \neq \mathcal{NP}$ .

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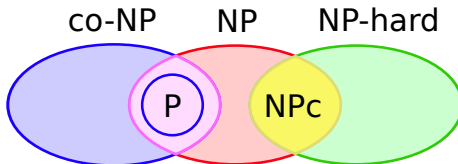
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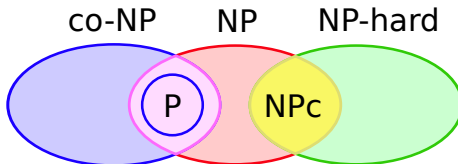
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