Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	${\mathcal N}{\mathcal P}$ vs. co- ${\mathcal N}{\mathcal P}$

#### NP and Computational Intractability

#### T. M. Murali

April 23, 28, 2014

# **Algorithm Design**

#### Patterns

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

 $O(n \log n)$  interval scheduling.  $O(n \log n)$  closest pair of points.  $O(n^2)$  edit distance.  $O(n^3)$  maximum flow and minimum cuts.

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- ► Greed.
- Divide-and-conquer.
- Dynamic programming.
- Duality.
- Reductions.
- Local search.
- Randomization.
- "Anti-patterns"
  - NP-completeness.
  - PSPACE-completeness.
  - Undecidability.

 $O(n \log n)$  interval scheduling.  $O(n \log n)$  closest pair of points.  $O(n^2)$  edit distance.  $O(n^3)$  maximum flow and minimum cuts.

 $O(n^k)$  algorithm unlikely.  $O(n^k)$  certification algorithm unlikely. No algorithm possible.

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Polynomial time	Probably not	
Shortest path	Longest path	
Matching	3-D matching	
Minimum cut	Maximum cut	
2-SAT	3-SAT	
Planar four-colour	Planar three-colour	
Bipartite vertex cover	Vertex cover	
Primality testing	Factoring	

### **Problem Classification**

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# **Problem Classification**

- Classify problems based on whether they admit efficient solutions or not.
- 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!

- ▶ Goal is to express statements of the type "Problem X is at least as hard as problem Y."
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- $Y \leq_P X$  implies that "X is at least as hard as Y."
- ► Such reductions are *Cook reductions*. *Karp reductions* allow only one call to the black box that solves *X*.

## **Usefulness of Reductions**

Claim: If Y ≤<sub>P</sub> X and X can be solved in polynomial time, then Y can be solved in polynomial time.

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- ► Contrapositive: If Y ≤<sub>P</sub> 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.

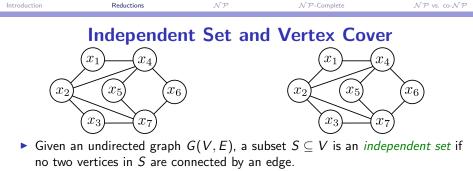
# **Optimisation versus Decision Problems**

- ► So far, we have developed algorithms that solve optimisation problems.
  - Compute the *largest* flow.
  - Find the *closest* pair of points.
  - Find the schedule with the *least* completion time.

# **Optimisation versus Decision Problems**

- ▶ So far, we have developed algorithms that solve optimisation problems.
  - 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.

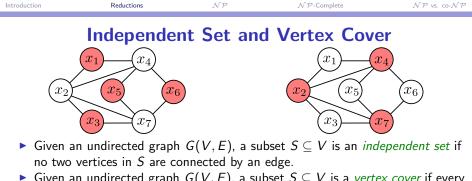
PRIMES INSTANCE: A natural number *n* QUESTION: ls *n* prime?



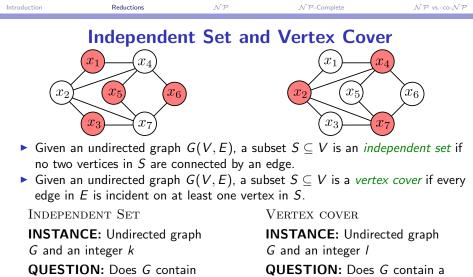
NP

Reductions

• 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.

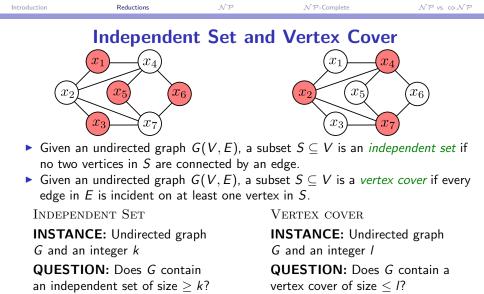


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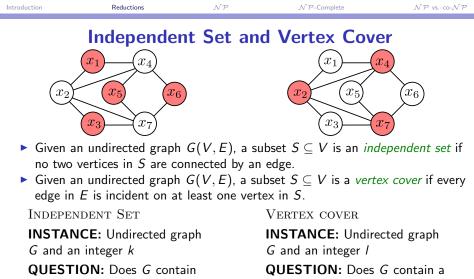


an independent set of size  $\geq k$ ?

vertex cover of size  $\leq I$ ?



• Demonstrate simple equivalence between these two problems.



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- Demonstrate simple equivalence between these two problems.
- ► Claim: INDEPENDENT SET ≤<sub>P</sub> VERTEX COVER and VERTEX COVER ≤<sub>P</sub> INDEPENDENT SET.

#### 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 *I*.
  - G' related to G in some way.
  - I can depend upon k and size of G.
- 3. Prove that G(V, E) has an independent set of size  $\geq k$  iff G'(V', E') has a vertex cover of size  $\leq l$ .

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- Transformation and proof must be correct for all possible graphs G(V, E) and all possible values of k.
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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? In the reduction, we are using black box for VERTEX COVER to solve INDEPENDENT SET.
  - (i) 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.
  - (ii) If the black box finds a vertex cover of size  $\leq I$ , we must be sure we can construct an independent set of size  $\geq k$  from this vertex cover.

Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	$\mathcal{NP}$ vs. co- $\mathcal{NP}$
Proc	of that Indep	endent Set	$t \leq_P $ Vertex C	over
	$x_1$ $x_4$ $x_5$ $x_6$ $x_3$ $x_7$		$\begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_7 \end{array}$	x <sub>6</sub>
1. Arbitrar an integ	•	ENDENT SET: a	n undirected graph G	$\mathcal{L}(V, E)$ and
2. Let   <i>V</i>	= <i>n</i> .			
3. Create a integer		EX COVER: sam	ne undirected graph (	$\widehat{\mathcal{G}}(V,E)$ and

Introduc	tion Reductions $\mathcal{NP}$ $\mathcal{NP}$ -Complete $\mathcal{NP}$ vs. co- $\mathcal{NP}$
	<b>Proof that Independent Set</b> $\leq_P$ <b>Vertex Cover</b>
	$x_1$ $x_4$ $x_1$ $x_4$ $x_1$ $x_4$ $x_2$ $x_5$ $x_6$ $x_3$ $x_7$ $x_7$
1.	Arbitrary instance of INDEPENDENT SET: an undirected graph $G(V, E)$ and an integer $k$ .
2.	Let $ V  = n$ .
3.	Create an instance of VERTEX COVER: same undirected graph $G(V, E)$ and integer $n - k$ .
4.	Claim: $G(V, E)$ has an independent set of size $\geq k$ iff $G(V, E)$ has a vertex cover of size $\leq n - k$ .
	Proof: S is an independent set in G iff $V - S$ is a vertex cover in G.

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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.

Set Cover

#### **INSTANCE:** A set U of n

elements, a collection  $S_1, S_2, \ldots, S_m$  of subsets of U, and an integer k.

#### QUESTION: Is there a

collection of  $\leq k$  sets in the collection whose union is *U*?

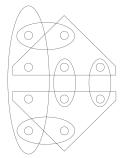
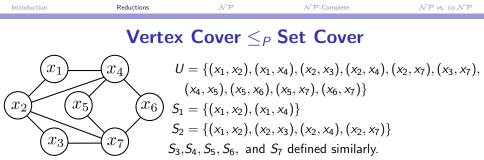


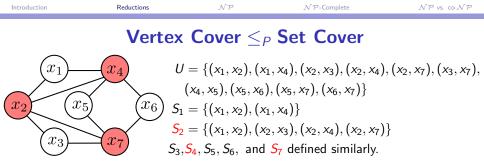
Figure 8.2 An instance of the Set Cover Problem.



- Input to VERTEX COVER: an undirected graph G(V, E) and an integer k.
- Let |V| = n.
- ▶ Create an instance  $\{U, \{S_1, S_2, ..., S_n\}\}$  of SET COVER where



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- Create an instance  $\{U, \{S_1, S_2, \dots, S_n\}\}$  of SET COVER where
  - ► U = E,
  - ▶ for each vertex  $i \in V$ , create a set  $S_i \subseteq U$  of the edges incident on i.



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  - ► U = E,
  - ▶ for each vertex  $i \in V$ , create a set  $S_i \subseteq U$  of the edges incident on i.
- Claim: U can be covered with fewer than k subsets iff G has a vertex cover with at most k nodes.
- Proof strategy:
  - 1. If G(V, E) has a vertex cover of size at most k, then U can be covered with at most k subsets.
  - If U can be covered with at most k subsets, then G(V, E) has a vertex cover of size at most k.

# **Boolean Satisfiability**

- Abstract problems formulated in Boolean notation.
- Often used to specify problems, e.g., in Al.

Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP}$ -Complete	$\mathcal{NP}$ vs. co- $\mathcal{NP}$		
Boolean Satisfiability						
<ul> <li>Abstract problems formulated in Boolean notation.</li> </ul>						

- Often used to specify problems, e.g., in AI.
- We are given a set  $X = \{x_1, x_2, \dots, x_n\}$  of *n* Boolean variables.
- Each variable can take the value 0 or 1.
- A *term* is a variable  $x_i$  or its negation  $\overline{x_i}$ .
- A clause of length I is a disjunction (or) of I distinct terms  $t_1 \vee t_2 \vee \cdots \mid t_l$ .
- A truth assignment for X is a function  $\nu : X \to \{0, 1\}$ .
- ► An assignment 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<sub>1</sub>, C<sub>2</sub>,... C<sub>k</sub> if it causes all clauses to evaluate to 1, i.e., C<sub>1</sub> ∧ C<sub>2</sub> ∧ · · · C<sub>k</sub> = 1.
  - $\nu$  is a satisfying assignment with respect to  $C_1, C_2, \ldots C_k$ .
  - set of clauses  $C_1, C_2, \ldots C_k$  is *satisfiable*.

Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	${\cal NP}$ vs. co- ${\cal NP}$
	9	SAT and 3	SAT	
SATI	SFIABILITY PROB	lem (SAT)		
INSTA	NCE: A set of cla	uses $C_1, C_2, \ldots C_n$	- k	over a
set $X =$	$\{x_1, x_2, \dots x_n\}$ of	n variables.		
OUEST	<b>FION</b> . Is there a st	atisfying truth a	signment for X with	n respect to

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

		C.A.T.	
	SAT and 3	-SAT	

3-Satisfiability Problem (SAT)

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

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**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.

- $C_1 = x_1 \lor 0 \lor 0$
- $C_2 = x_2 \lor 0 \lor 0$
- $\bullet \quad C_3 = \overline{x_1} \vee \overline{x_2} \vee 0$

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- $C_2 = x_2 \lor 0 \lor 0$
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- 1. Is  $C_1 \wedge C_2$  satisfiable?

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- 3. Is  $C_2 \wedge C_3$  satisfiable?

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- 3. Is  $C_2 \wedge C_3$  satisfiable? Yes, by  $x_1 = 0, x_2 = 1$ .
- 4. Is  $C_1 \wedge C_2 \wedge C_3$  satisfiable? No.

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

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

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

▶ We want to prove 3-SAT  $\leq_P$  INDEPENDENT SET.

Introduction	Reductions	5 ¢ 7	JV / -Complete	JV / V3. CO-JV /
	3-SAT	and Inde	pendent Set	
$C_1 = \mathbf{x_1} \lor C_2 = \overline{x_1} \lor$	1. 366	$ect\ x_1 = 1, x_2 =$	$1, x_3 = 1, x_4 = 1.$	

NP

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

- ▶ We want to prove 3-SAT  $\leq_P$  INDEPENDENT SET.
- ▶ Two ways to think about 3-SAT:

Reductions

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.

Introduction

NP VS. CO-NP

Introduction	Reductions	NP	N P-Complete	NP vs. co-NP
	3-SAT	and Inde	pendent Set	
$C_1 = x_1^{-1}$	$\sqrt{\overline{x_2}}\sqrt{\overline{x_3}}$ 1 Selec	$r_{1} = 1 r_{2} = 1$	$1 x_{0} - 1 x_{1} - 1$	

1. Select  $x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1$ .

 $C_2 = \overline{x_1} \lor x_2 \lor \mathbf{x_4}$ 2. Choose one literal from each clause to evaluate to true.  $C_3 = \overline{X_1} \vee X_3 \vee \overline{X_4}$ 

- ▶ We want to prove 3-SAT <<sub>P</sub> 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.
  - 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  $C_2$ .

Introduction	Reductions	NP	N P-Complete	NP vs. co-NP
	3-SAT	and Inde	pendent Set	
$C_1 = x_1$	$\sqrt{\overline{x_2}}\sqrt{\overline{x_3}}$ 1 Soler	* v. – 1 v. –	1 - x - 1 - 1	

- $\begin{array}{ll} \mathcal{C}_1 = x_1 \sqrt{x_2} \sqrt{x_3} & 1. \text{ Select } x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1. \\ \mathcal{C}_2 = \overline{x_1} \sqrt{x_2} \sqrt{x_4} & 2. \end{array}$ 
  - $2. \ \mbox{Choose}$  one literal from each clause to evaluate to true.

• Choices of selected literals imply  $x_1 = 0, x_2 = 0, x_4 = 1$ .

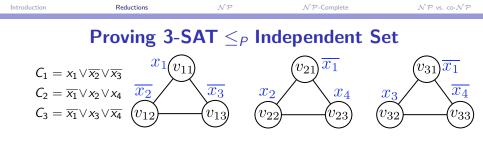
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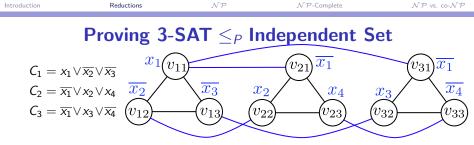
- 1. Make an independent 0/1 decision on each variable and succeed if we achieve one of three ways in which to satisfy each clause.
- 2. Choose (at least) one term from each clause. Find a truth assignment that causes each chosen term to evaluate to 1. Ensure that no two terms selected *conflict*, e.g., select  $\overline{x_2}$  in  $C_1$  and  $x_2$  in  $C_2$ .

Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-}Complete$	$\mathcal{NP}$ vs. co- $\mathcal{NP}$
	Proving 3-9	SAT ≤ <sub>P</sub> Iı	ndependent S	et
$C_1 = x_1$ $C_2 = \overline{x_1}$ $C_3 = \overline{x_1}$	$\forall x_2 \lor x_4$			
5 1	given an instance of	of 3-SAT with	k clauses of length th	ree over n

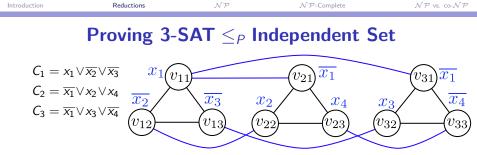
• Construct an instance of independent set: graph G(V, E) with 3k nodes.



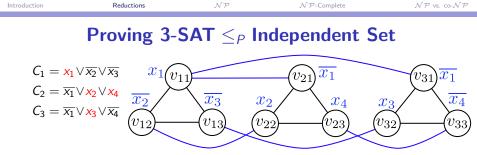
- We are given an instance of 3-SAT with k clauses of length three over n variables.
- Construct an instance of independent set: graph G(V, E) with 3k nodes.
  - For each clause C<sub>i</sub>, 1 ≤ i ≤ k, add a triangle of three nodes v<sub>i1</sub>, v<sub>i2</sub>, v<sub>i3</sub> and three edges to G.
  - Label each node  $v_{ij}$ ,  $1 \le j \le 3$  with the *j*th term in  $C_i$ .



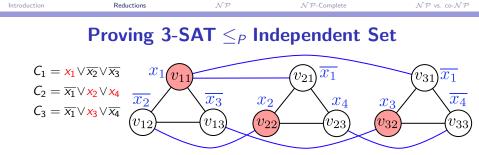
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  - Add an edge between each pair of nodes whose labels correspond to terms that conflict.



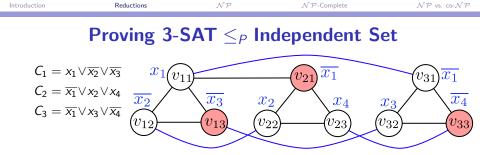
▶ Claim: 3-SAT instance is satisfiable iff *G* has an independent set of size *k*.



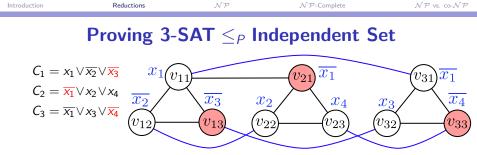
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- Satisfiable assignment  $\rightarrow$  independent set of size k:



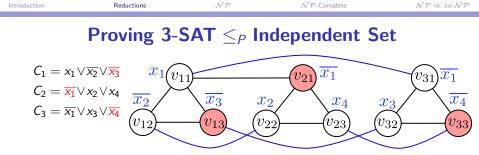
- ▶ Claim: 3-SAT instance is satisfiable iff *G* has an independent set of size *k*.
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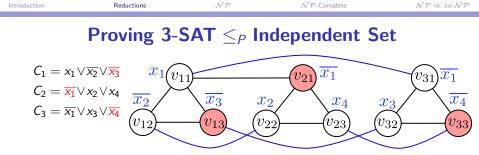
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  - If  $x_i$  is the label of a node in S, set  $x_i = 1$ ; else set  $x_i = 0$ .
  - Why is each clause satisfied?

## **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-SAT  $\leq_P$  Independent Set  $\leq_P$  Vertex Cover  $\leq_P$  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?

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- 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.

#### PRIMES **INSTANCE:** A natural number *n* **QUESTION:** Is *n* prime?

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- ► A has a polynomial running time if there is a polynomial function p(·) such that for every input s, A terminates on s in at most O(p(|s|)) steps.
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- $\mathcal{P}$ : set of problems X for which there is a polynomial time algorithm.
- ► A decision problem X is in P iff there is an algorithm A with polynomial running time that solves X.

- ► A "checking" algorithm for a decision problem X has a different structure from an algorithm that solves X.
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- ► An algorithm *B* is an *efficient certifier* for a problem *X* if
  - 1. B is a polynomial time algorithm that takes two inputs s and t and
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### $\blacktriangleright$ $\mathcal{NP}$ is the set of all problems for which there exists an efficient certifier.

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▶ 3-SAT  $\in \mathcal{NP}$ :

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### ▶ Set Cover $\in \mathcal{NP}$ :

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Introduction Reductions NP NP-Complete NP vs. co-NP

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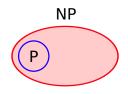
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  - Certifier *B*: checks if their union of these sets is *U*.

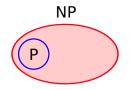
Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	${\mathcal N}{\mathcal P}$ vs. co- ${\mathcal N}{\mathcal P}$
		${\cal P}$ vs. ${\cal N}$	P	

▶ Claim:  $\mathcal{P} \subseteq \mathcal{NP}$ .

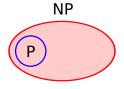


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• If  $X \in P$ , then there is a polynomial time algorithm A that solves X. B ignores t and returns A(s). Why is B an efficient certifier?



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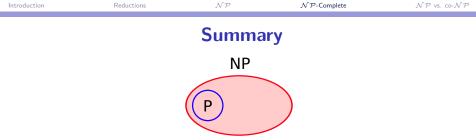


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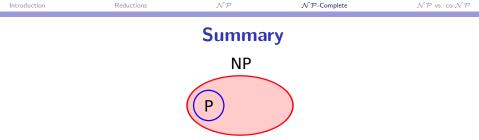
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Is P = NP or is NP − P ≠ Ø? One of the major unsolved problems in computer science. \$1M prize offered by Clay Mathematics Institute.

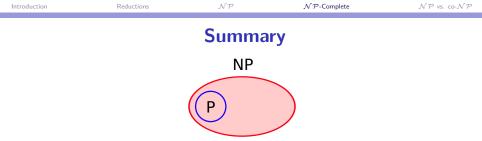




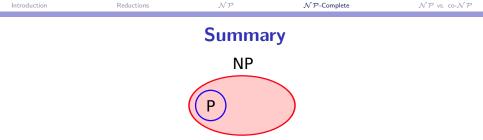
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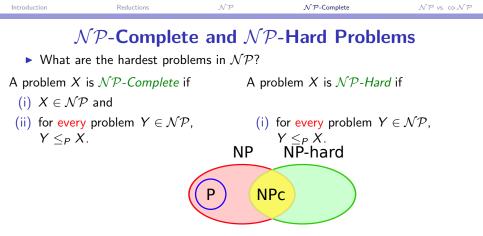


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  - 2. Are there two problems  $X_1$  and  $X_2$  in  $\mathcal{NP}$  such that there is no problem  $X \in \mathcal{NP}$  where  $X_1 \leq_P X$  and  $X_2 \leq_P X$ ?

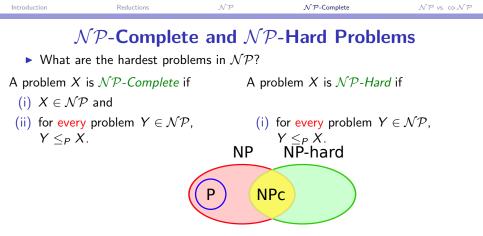
### $\mathcal{NP}$ -Complete and $\mathcal{NP}$ -Hard Problems

• What are the hardest problems in  $\mathcal{NP}$ ?

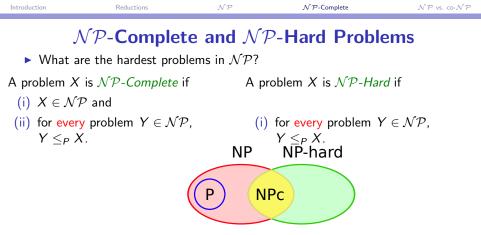
Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	$\mathcal{NP}$ vs. co- $\mathcal{NP}$		
$\mathcal{NP} extsf{-}$ Complete and $\mathcal{NP} extsf{-}$ Hard Problems						
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A problem X (i) $X \in \mathcal{N}\mathcal{N}$	is $\mathcal{NP}$ - $Complete$ if $\mathcal P$ and	A	problem $X$ is $\mathcal{NP}$ -Hard	if		
(ii) for every $Y \leq_P X$	y problem $Y \in \mathcal{NP}$ , K.	(	i) for every problem $Y$ $Y \leq_P X$ .	$\in \mathcal{NP}$ ,		



▶ Claim: Suppose X is  $\mathcal{NP}$ -Complete. Then  $X \in \mathcal{P}$  iff  $\mathcal{P} = \mathcal{NP}$ .



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- Corollary: If there is any problem in NP that cannot be solved in polynomial time, then no NP-Complete problem can be solved in polynomial time.
- Does even one *NP*-Complete problem exist?! If it does, how can we prove that *every* problem in *NP* reduces to this problem?

## **Circuit Satisfiability**

### ► Cook-Levin Theorem: CIRCUIT SATISFIABILITY is *NP*-Complete.

### **Circuit Satisfiability**

- ► Cook-Levin Theorem: CIRCUIT SATISFIABILITY is *NP*-Complete.
- A circuit K is a labelled, directed acyclic graph such that
  - 1. the *sources* in *K* are labelled with constants (0 or 1) or the name of a distinct variable (the *inputs* to the circuit).
  - 2. every other node is labelled with one Boolean operator  $\land$ ,  $\lor$ , or  $\neg$ .
  - 3. a single node with no outgoing edges represents the *output* of K.

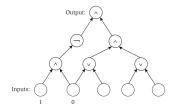
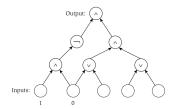


Figure 8.4 A circuit with three inputs, two additional sources that have assigned truth values, and one output.

## **Circuit Satisfiability**

- ► Cook-Levin Theorem: CIRCUIT SATISFIABILITY is *NP*-Complete.
- A circuit K is a labelled, directed acyclic graph such that
  - 1. the *sources* in *K* are labelled with constants (0 or 1) or the name of a distinct variable (the *inputs* to the circuit).
  - 2. every other node is labelled with one Boolean operator  $\wedge,\,\vee,$  or  $\neg.$
  - 3. a single node with no outgoing edges represents the *output* of K.



CIRCUIT SATISFIABILITY

**INSTANCE:** A circuit *K*.

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

Figure 8.4 A circuit with three inputs, two additional sources that have assigned truth values, and one output.

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

## Proving Circuit Satisfiability is $\mathcal{NP}$ -Complete

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▶ Take an arbitrary problem  $X \in \mathcal{NP}$  and show that  $X \leq_P \text{CIRCUIT SATISFIABILITY}.$ 

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  - 1. First *n* sources are hard-coded with the bits of *s*.
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  - 1. First n sources are hard-coded with the bits of s.
  - 2. The remaining p(n) sources labelled with variables representing the bits of t.
- ► s ∈ X iff there is an assignment of the input bits of K that makes K satisfiable.

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

- ▶ 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).

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

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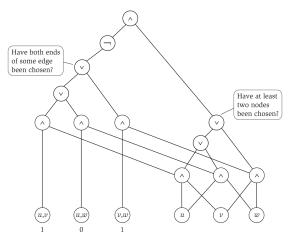


Figure 8.5 A circuit to verify whether a 3-node graph contains a 2-node independent set.

## **Asymmetry of Certification**

- $\blacktriangleright$  Definition of efficient certification and  $\mathcal{NP}$  is fundamentally asymmetric:
  - An input string s is a "yes" instance iff there exists a short string t such that B(s, t) = yes.
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Introduction	Reductions	$\mathcal{NP}$	$\mathcal{NP} ext{-Complete}$	$\mathcal{NP}$ vs. co- $\mathcal{NP}$
		со- $\mathcal{NP}$	2	
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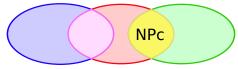
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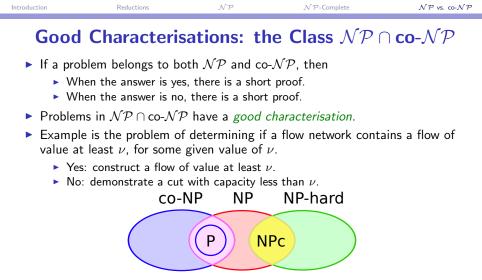
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- Claim: If  $\mathcal{NP} \neq \text{co-}\mathcal{NP}$  then  $\mathcal{P} \neq \mathcal{NP}$ .

#### Good Characterisations: the Class $\mathcal{NP} \cap \text{co-}\mathcal{NP}$

- $\blacktriangleright$  If a problem belongs to both  $\mathcal{NP}$  and co- $\mathcal{NP},$  then
  - When the answer is yes, there is a short proof.
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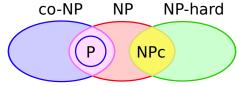
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- Example is the problem of determining if a flow network contains a flow of value at least  $\nu$ , for some given value of  $\nu$ .
  - Yes: construct a flow of value at least v.
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• Claim:  $\mathcal{P} \subseteq \mathcal{NP} \cap \text{co-}\mathcal{NP}$ .

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