

SECTION 8.1

### 8. INTRACTABILITY I

- poly-time reductions
- packing and covering problems
- constraint satisfaction problems
- sequencing problems
- partitioning problems
- graph coloring
- numerical problems

### Algorithm design patterns and antipatterns

### Algorithm design patterns.

- Greedy.
- Divide and conquer.
- Dynamic programming.
- Duality.
- Reductions.
- Local search.
- Randomization.

### Algorithm design antipatterns.

- NP-completeness.  $O(n^k)$  algorithm unlikely.
- **PSPACE**-completeness.  $O(n^k)$  certification algorithm unlikely.
- Undecidability.
   No algorithm possible.

## Classify problems according to computational requirements

Q. Which problems will we be able to solve in practice?

A working definition. Those with poly-time algorithms.



von Neumann (1953)



Nash (1955)



Gödel (1956)



**Cobham** (1964)



Edmonds (1965)



Rabin (1966)

Turing machine, word RAM, uniform circuits, ...

Theory. Definition is broad and robust.

constants tend to be small, e.g.,  $3n^2$ 

Practice. Poly-time algorithms scale to huge problems.

# Classify problems according to computational requirements

Q. Which problems will we be able to solve in practice?

A working definition. Those with poly-time algorithms.

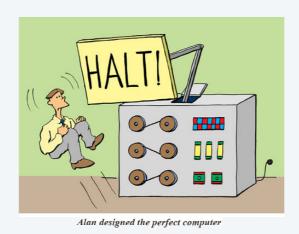
yes	probably no
shortest path	longest path
min cut	max cut
2-satisfiability	3-satisfiability
planar 4-colorability	planar 3-colorability
bipartite vertex cover	vertex cover
matching	3d-matching
primality testing	factoring
linear programming	integer linear programming

# Classify problems

Desiderata. Classify problems according to those that can be solved in polynomial time and those that cannot.

### Provably requires exponential time.

- Given a constant-size program, does it halt in at most k steps?
- Given a board position in an *n*-by-*n* generalization of checkers, can black guarantee a win?





Frustrating news. Huge number of fundamental problems have defied classification for decades.

input size =  $c + \log k$ 

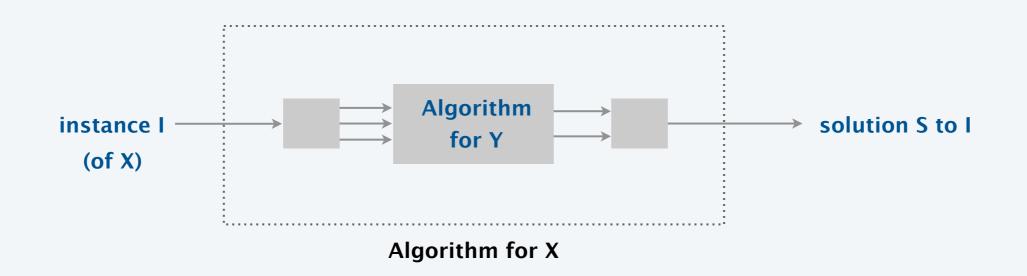
### Poly-time reductions

Desiderata'. Suppose we could solve problem Y in polynomial time. What else could we solve in polynomial time?

Reduction. Problem X polynomial-time (Cook) reduces to problem Y if arbitrary instances of problem X can be solved using:

- Polynomial number of standard computational steps, plus
- Polynomial number of calls to oracle that solves problem Y.

computational model supplemented by special piece of hardware that solves instances of Y in a single step



### Poly-time reductions

Desiderata'. Suppose we could solve problem Y in polynomial time. What else could we solve in polynomial time?

Reduction. Problem X polynomial-time (Cook) reduces to problem Y if arbitrary instances of problem X can be solved using:

- · Polynomial number of standard computational steps, plus
- Polynomial number of calls to oracle that solves problem Y.

Notation.  $X \leq_{P} Y$ .

Note. We pay for time to write down instances of Y sent to oracle  $\Rightarrow$  instances of Y must be of polynomial size.

Novice mistake. Confusing  $X \leq_P Y$  with  $Y \leq_P X$ .

## Intractability: quiz 1



### Suppose that $X \leq_P Y$ . Which of the following can we infer?

- A. If X can be solved in polynomial time, then so can Y.
- **B.** X can be solved in poly time iff Y can be solved in poly time.
- **C.** If *X* cannot be solved in polynomial time, then neither can *Y*.
- **D.** If *Y* cannot be solved in polynomial time, then neither can *X*.

## Intractability: quiz 2



### Which of the following poly-time reductions are known?

- A. FIND-MAX-FLOW  $\leq_P$  FIND-MIN-CUT.
- **B.** FIND-MIN-CUT  $\leq_P$  FIND-MAX-FLOW.
- C. Both A and B.
- **D.** Neither A nor B.

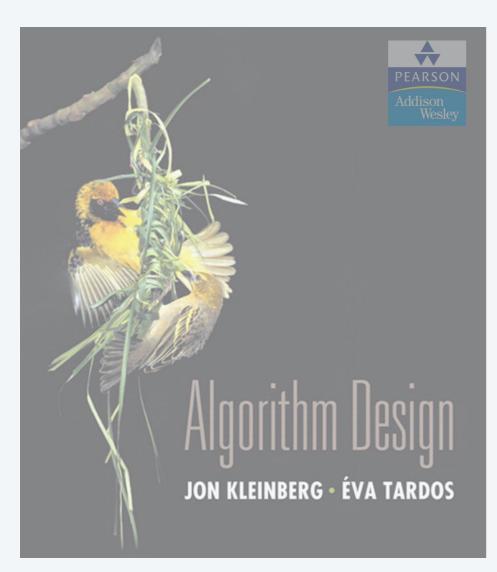
### Poly-time reductions

Design algorithms. If  $X \leq_P Y$  and Y can be solved in polynomial time, then X can be solved in polynomial time.

Establish intractability. If  $X \leq_P Y$  and X cannot be solved in polynomial time, then Y cannot be solved in polynomial time.

Establish equivalence. If both  $X \leq_P Y$  and  $Y \leq_P X$ , we use notation  $X \equiv_P Y$ . In this case, X can be solved in polynomial time iff Y can be.

Bottom line. Reductions classify problems according to relative difficulty.



SECTION 8.1

### 8. INTRACTABILITY I

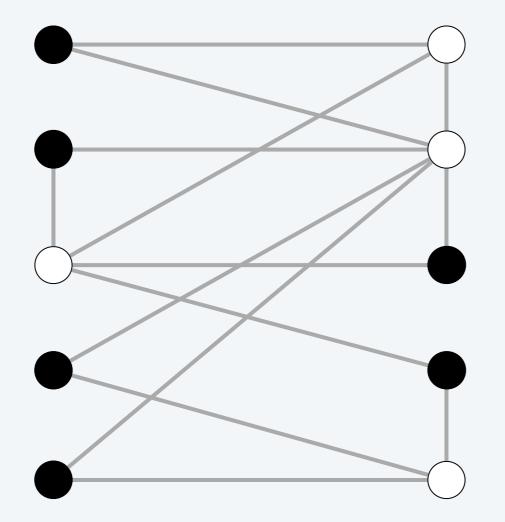
- poly-time reductions
- packing and covering problems
- constraint satisfaction problems
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### Independent set

INDEPENDENT-SET. Given a graph G = (V, E) and an integer k, is there a subset of k (or more) vertices such that no two are adjacent?

Ex. Is there an independent set of size  $\geq 6$ ?

Ex. Is there an independent set of size  $\geq 7$ ?



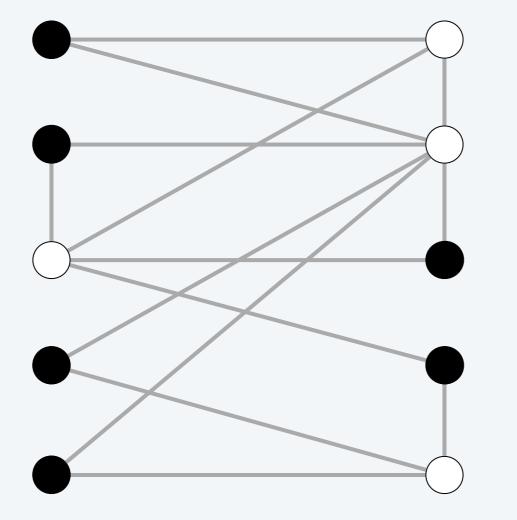
independent set of size 6

#### Vertex cover

VERTEX-COVER. Given a graph G = (V, E) and an integer k, is there a subset of k (or fewer) vertices such that each edge is incident to at least one vertex in the subset?

Ex. Is there a vertex cover of size  $\leq 4$ ?

Ex. Is there a vertex cover of size  $\leq 3$ ?

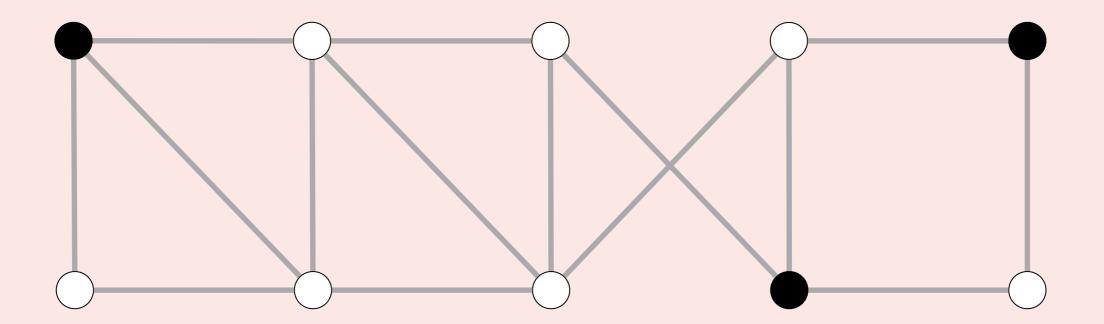


- independent set of size 6
- vertex cover of size 4



### Consider the following graph G. Which are true?

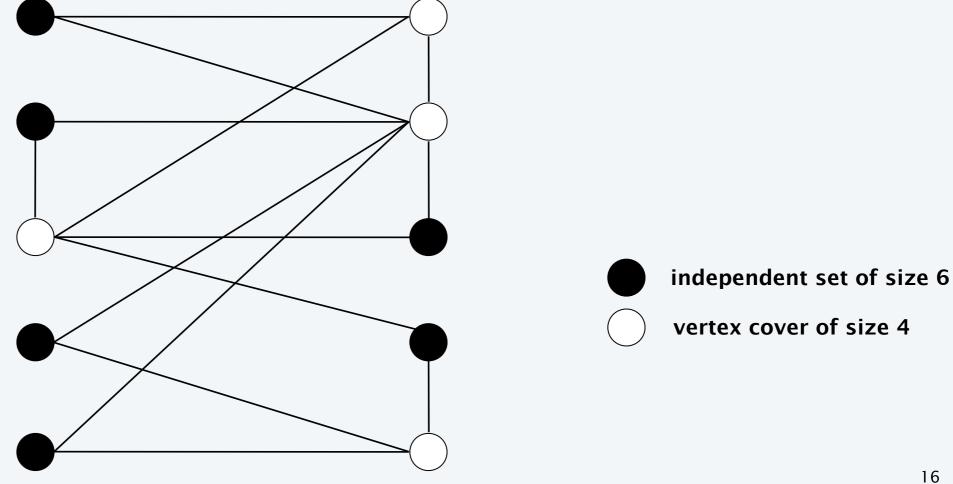
- **A.** The white vertices are a vertex cover of size 7.
- **B.** The black vertices are an independent set of size 3.
- C. Both A and B.
- **D.** Neither A nor B.



## Vertex cover and independent set reduce to one another

Theorem. Independent-Set  $\equiv_P$  Vertex-Cover.

Pf. We show S is an independent set of size k iff V - S is a vertex cover of size n-k.



### Vertex cover and independent set reduce to one another

Theorem. INDEPENDENT-SET  $\equiv_P$  VERTEX-COVER.

Pf. We show S is an independent set of size k iff V - S is a vertex cover of size n - k.



- Let S be any independent set of size k.
- V-S is of size n-k.
- Consider an arbitrary edge  $(u, v) \in E$ .
- *S* independent  $\Rightarrow$  either  $u \notin S$ , or  $v \notin S$ , or both.  $\Rightarrow$  either  $u \in V - S$ , or  $v \in V - S$ , or both.
- Thus, V S covers (u, v).  $\blacksquare$

### Vertex cover and independent set reduce to one another

Theorem. INDEPENDENT-SET  $\equiv_P$  VERTEX-COVER.

Pf. We show S is an independent set of size k iff V - S is a vertex cover of size n - k.

 $\Leftarrow$ 

- Let V S be any vertex cover of size n k.
- S is of size k.
- Consider an arbitrary edge  $(u, v) \in E$ .
- V S is a vertex cover  $\Rightarrow$  either  $u \in V S$ , or  $v \in V S$ , or both.  $\Rightarrow$  either  $u \notin S$ , or  $v \notin S$ , or both.
- Thus, *S* is an independent set. •

#### Set cover

**SET-COVER.** Given a set U of elements, a collection S of subsets of U, and an integer k, are there  $\leq k$  of these subsets whose union is equal to U?

### Sample application.

- *m* available pieces of software.
- Set *U* of *n* capabilities that we would like our system to have.
- The  $i^{th}$  piece of software provides the set  $S_i \subseteq U$  of capabilities.
- Goal: achieve all n capabilities using fewest pieces of software.

$$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$$
 $S_a = \{ 3, 7 \}$ 
 $S_b = \{ 2, 4 \}$ 
 $S_c = \{ 3, 4, 5, 6 \}$ 
 $S_d = \{ 5 \}$ 
 $S_e = \{ 1 \}$ 
 $S_f = \{ 1, 2, 6, 7 \}$ 
 $k = 2$ 

### Intractability: quiz 4



Given the universe  $U = \{ 1, 2, 3, 4, 5, 6, 7 \}$  and the following sets, which is the minimum size of a set cover?

- **A.** 1
- **B.** 2
- **C.** 3
- **D.** None of the above.

$$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$$
 $S_a = \{ 1, 4, 6 \}$ 
 $S_b = \{ 1, 6, 7 \}$ 
 $S_c = \{ 1, 2, 3, 6 \}$ 
 $S_d = \{ 1, 3, 5, 7 \}$ 
 $S_e = \{ 2, 6, 7 \}$ 
 $S_f = \{ 3, 4, 5 \}$ 

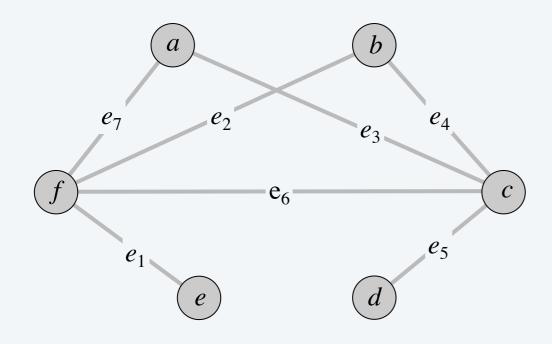
### Vertex cover reduces to set cover

Theorem. Vertex-Cover  $\leq_P$  Set-Cover.

Pf. Given a Vertex-Cover instance G = (V, E) and k, we construct a Set-Cover instance (U, S, k) that has a set cover of size k iff G has a vertex cover of size k.

#### Construction.

- Universe U = E.
- Include one subset for each node  $v \in V$ :  $S_v = \{e \in E : e \text{ incident to } v \}$ .



$$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$$
  
 $S_a = \{ 3, 7 \}$   $S_b = \{ 2, 4 \}$   
 $S_c = \{ 3, 4, 5, 6 \}$   $S_d = \{ 5 \}$   
 $S_e = \{ 1 \}$   $S_f = \{ 1, 2, 6, 7 \}$ 

vertex cover instance (k = 2)

set cover instance (k = 2)

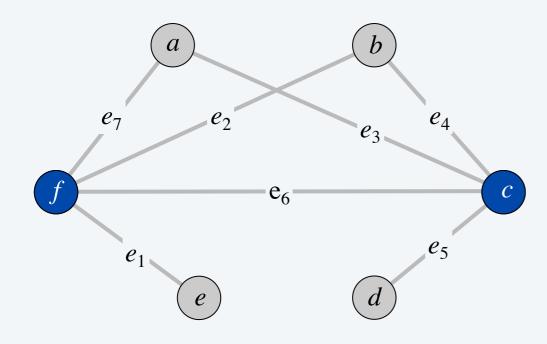
### Vertex cover reduces to set cover

Lemma. G = (V, E) contains a vertex cover of size k iff (U, S, k) contains a set cover of size k.

Pf.  $\Rightarrow$  Let  $X \subseteq V$  be a vertex cover of size k in G.

• Then  $Y = \{ S_v : v \in X \}$  is a set cover of size k. •

"yes" instances of VERTEX-COVER are solved correctly



vertex cover instance 
$$(k = 2)$$

$$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$$

$$S_a = \{ 3, 7 \} \qquad S_b = \{ 2, 4 \}$$

$$S_c = \{ 3, 4, 5, 6 \} \qquad S_d = \{ 5 \}$$

$$S_e = \{ 1 \} \qquad S_f = \{ 1, 2, 6, 7 \}$$

set cover instance (k = 2)

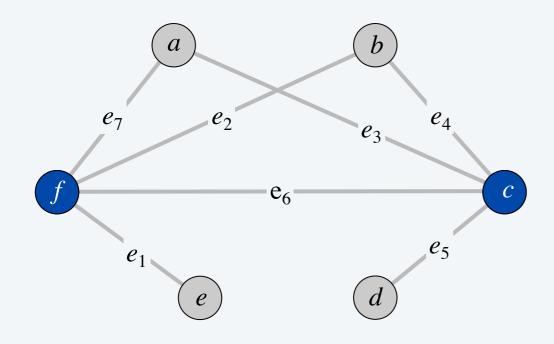
### Vertex cover reduces to set cover

Lemma. G = (V, E) contains a vertex cover of size k iff (U, S, k) contains a set cover of size k.

Pf.  $\leftarrow$  Let  $Y \subseteq S$  be a set cover of size k in (U, S, k).

• Then  $X = \{ v : S_v \in Y \}$  is a vertex cover of size k in G. •

"no" instances of VERTEX-COVER are solved correctly



vertex cover instance

(k = 2)

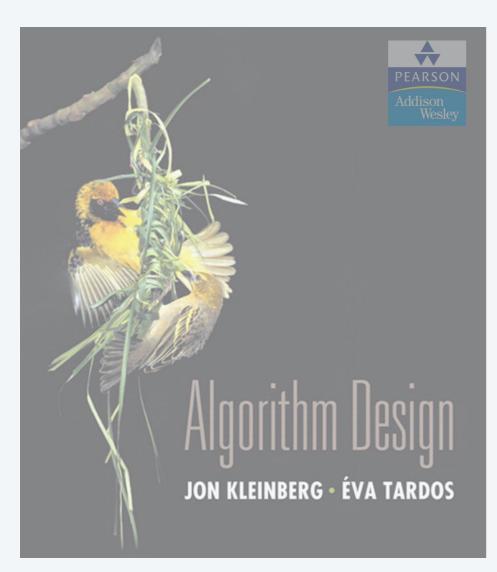
$$U = \{ 1, 2, 3, 4, 5, 6, 7 \}$$

$$S_a = \{ 3, 7 \} \qquad S_b = \{ 2, 4 \}$$

$$S_c = \{ 3, 4, 5, 6 \} \qquad S_d = \{ 5 \}$$

$$S_e = \{ 1 \} \qquad S_f = \{ 1, 2, 6, 7 \}$$

set cover instance (k = 2)



SECTION 8.2

### 8. INTRACTABILITY I

- poly-time reductions
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## Satisfiability

Literal. A Boolean variable or its negation.

$$x_i$$
 or  $\overline{x_i}$ 

Clause. A disjunction of literals.

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

Conjunctive normal form (CNF). A propositional formula  $\Phi$  that is a conjunction of clauses.

$$\Phi = C_1 \wedge C_2 \wedge C_3 \wedge C_4$$

SAT. Given a CNF formula  $\Phi$ , does it have a satisfying truth assignment? 3-SAT. SAT where each clause contains exactly 3 literals (and each literal corresponds to a different variable).

$$\Phi = \left(\overline{x_1} \vee x_2 \vee x_3\right) \wedge \left(x_1 \vee \overline{x_2} \vee x_3\right) \wedge \left(\overline{x_1} \vee x_2 \vee x_4\right)$$

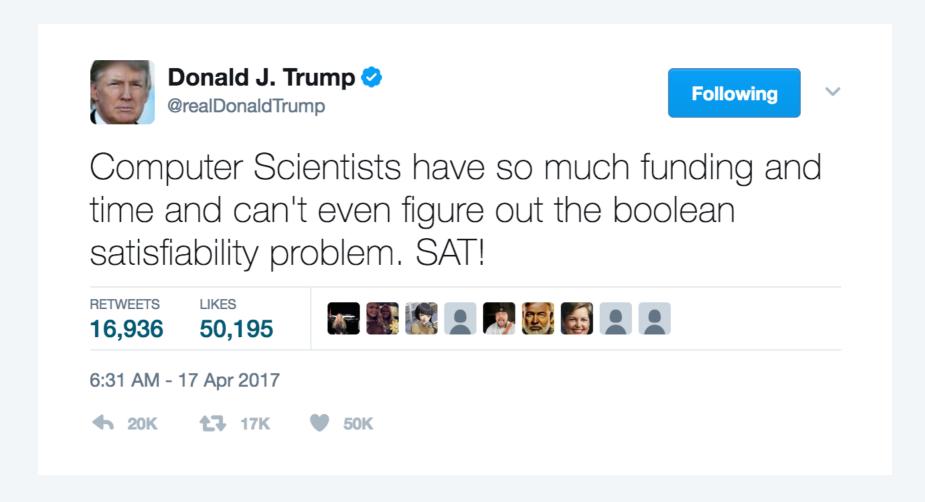
yes instance:  $x_1 = \text{true}, x_2 = \text{true}, x_3 = \text{false}, x_4 = \text{false}$ 

Key application. Electronic design automation (EDA).

### Satisfiability is hard

Scientific hypothesis. There does not exists a poly-time algorithm for 3-SAT.

P vs. NP. This hypothesis is equivalent to  $P \neq NP$  conjecture.



https://www.facebook.com/pg/npcompleteteens

### 3-satisfiability reduces to independent set

Theorem. 3-SAT  $\leq_P$  INDEPENDENT-SET.

Pf. Given an instance  $\Phi$  of 3-SAT, we construct an instance (G, k) of INDEPENDENT-SET that has an independent set of size  $k = |\Phi|$  iff  $\Phi$  is satisfiable.

#### Construction.

G

- G contains 3 nodes for each clause, one for each literal.
- Connect 3 literals in a clause in a triangle.
- Connect literal to each of its negations.

 $\overline{x_1}$   $\overline{x_2}$   $\overline{x_1}$   $\overline{x_2}$   $\overline{x_1}$   $\overline{x_2}$   $\overline$ 

 $\Phi = \left(\overline{x_1} \vee x_2 \vee x_3\right) \wedge \left(x_1 \vee \overline{x_2} \vee x_3\right) \wedge \left(\overline{x_1} \vee x_2 \vee x_4\right)$ 

## 3-satisfiability reduces to independent set

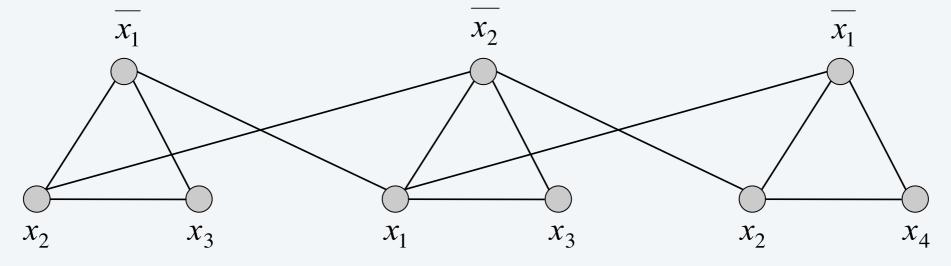
**Lemma.**  $\Phi$  is satisfiable iff G contains an independent set of size  $k = |\Phi|$ .

Pf.  $\Rightarrow$  Consider any satisfying assignment for  $\Phi$ .

- · Select one true literal from each clause/triangle.
- This is an independent set of size  $k = |\Phi|$ . •

"yes" instances of 3-SAT are solved correctly

G



$$\Phi = (\overline{x_1} \vee x_2 \vee x_3) \wedge (x_1 \vee \overline{x_2} \vee x_3) \wedge (\overline{x_1} \vee x_2 \vee x_4)$$

## 3-satisfiability reduces to independent set

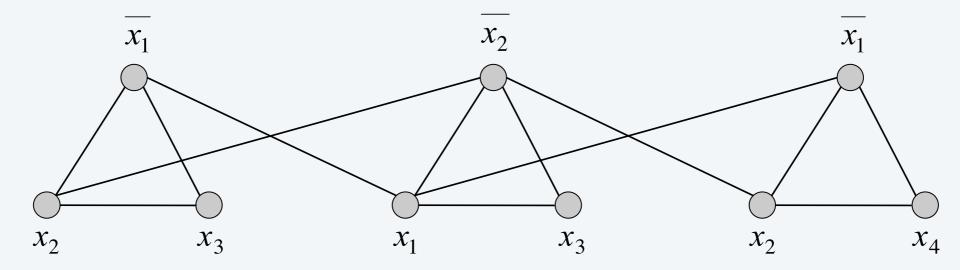
**Lemma.**  $\Phi$  is satisfiable iff G contains an independent set of size  $k = |\Phi|$ .

Pf.  $\leftarrow$  Let S be independent set of size k.

- S must contain exactly one node in each triangle.
- Set these literals to true (and remaining literals consistently).
- All clauses in  $\Phi$  are satisfied.  $\blacksquare$

G

"no" instances of 3-SA are solved correctly



$$\Phi = (\overline{x_1} \vee x_2 \vee x_3) \wedge (x_1 \vee \overline{x_2} \vee x_3) \wedge (\overline{x_1} \vee x_2 \vee x_4)$$

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#### Review

### Basic reduction strategies.

- Simple equivalence: Independent-Set  $\equiv_P$  Vertex-Cover.
- Special case to general case: Vertex-Cover ≤ P Set-Cover.
- Encoding with gadgets:  $3-SAT \leq_P INDEPENDENT-SET$ .

Transitivity. If  $X \le_P Y$  and  $Y \le_P Z$ , then  $X \le_P Z$ . Pf idea. Compose the two algorithms.

Ex. 3-SAT  $\leq_P$  INDEPENDENT-SET  $\leq_P$  VERTEX-COVER  $\leq_P$  SET-COVER.

# DECISION, SEARCH, AND OPTIMIZATION PROBLEMS



Decision problem. Does there exist a vertex cover of size  $\leq k$ ? Search problem. Find a vertex cover of size  $\leq k$ . Optimization problem. Find a vertex cover of minimum size.

Goal. Show that all three problems poly-time reduce to one another.

#### Self-Reducibility

Decision problem. Does there exist a vertex cover of size  $\leq k$ ? Optimization problem. Find vertex cover of minimum cardinality.

Self-reducibility. Optimization problem  $\leq p$  decision version.

- Applies to all (NP-complete) problems in this chapter.
- Justifies our focus on decision problems.

#### Ex: to find min cardinality vertex cover.

- (Binary) search for cardinality k\* of min vertex cover.
- Find a vertex v such that  $G \{v\}$  has a vertex cover of size  $\leq k^* 1$ .
  - any vertex in any min vertex cover will have this property
- Include v in the vertex cover.
- Recursively find a min vertex cover in  $G \{v\}$ .

delete v and all incident edges