

## 8. INTRACTABILITY I

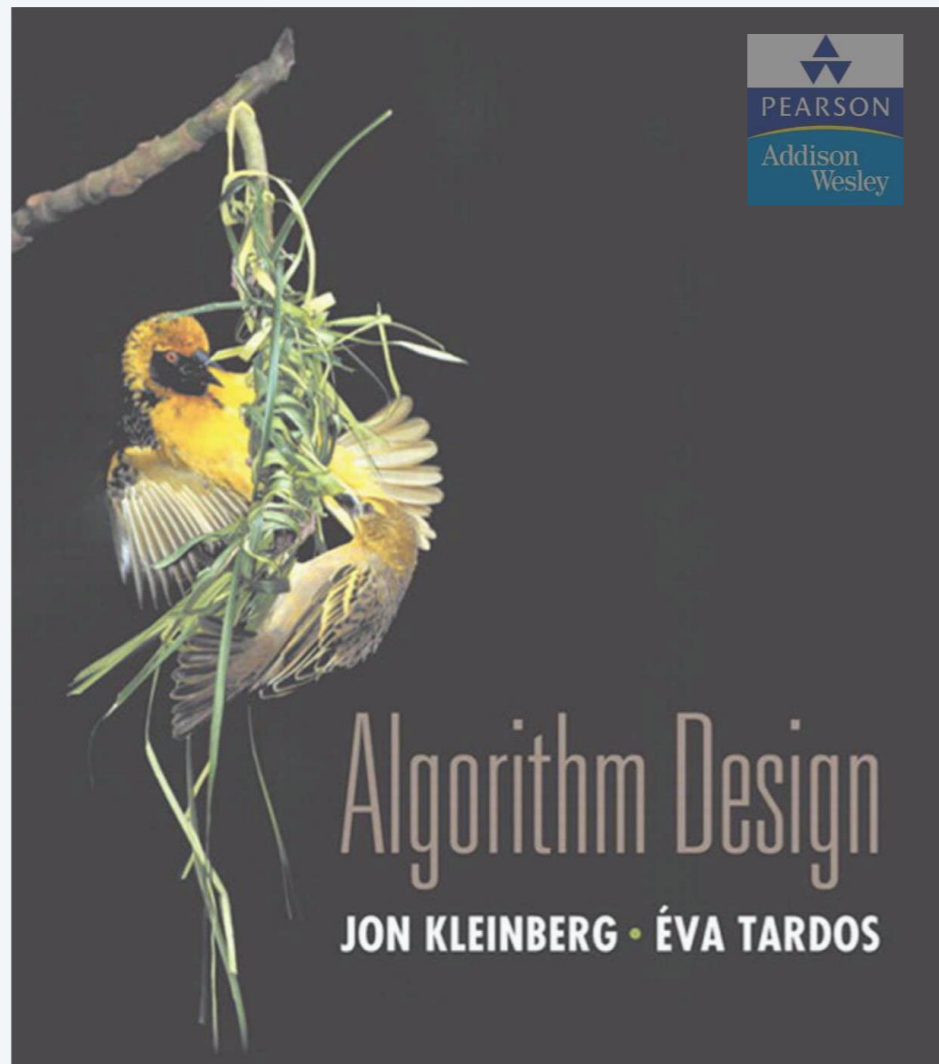
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- ▶ *poly-time reductions*

Lecture slides by Kevin Wayne

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<http://www.cs.princeton.edu/~wayne/kleinberg-tardos>



## SECTION 8.1

# 8. INTRACTABILITY I

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- ▶ *poly-time reductions*

# Algorithm design patterns and antipatterns

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## 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

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**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

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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

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**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?

input size =  $c + \log k$

using forced capture rule



*Alan designed the perfect computer*



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

# Poly-time reductions

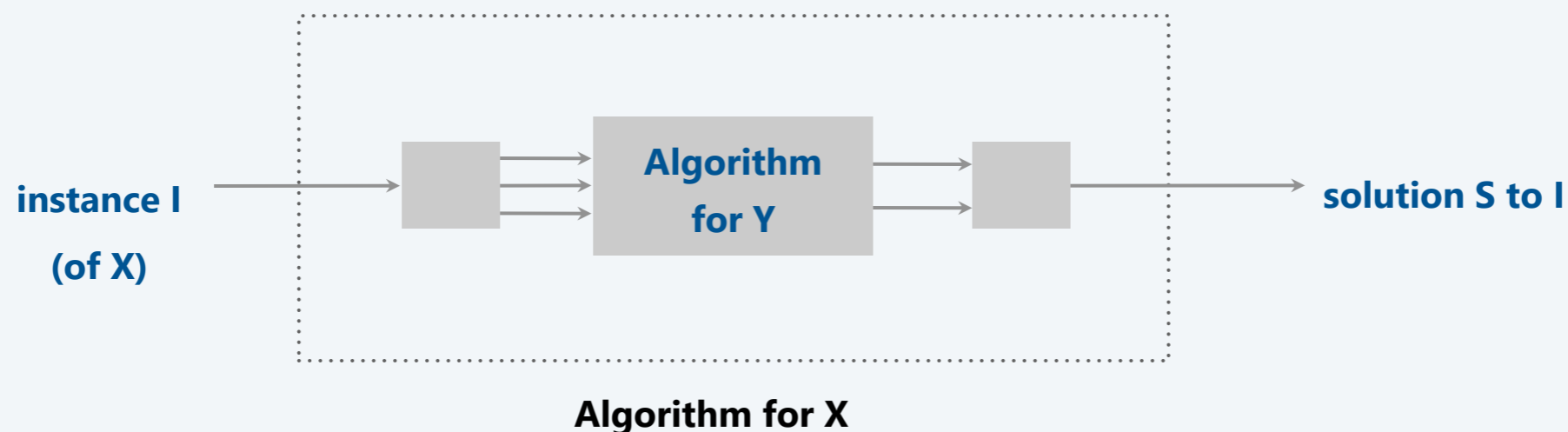
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**Desiderata'.** Suppose we could solve problem  $Y$  in polynomial time. What else could we solve in polynomial time?

**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

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**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  $Y \leq_p X$ .



# Poly-time reductions

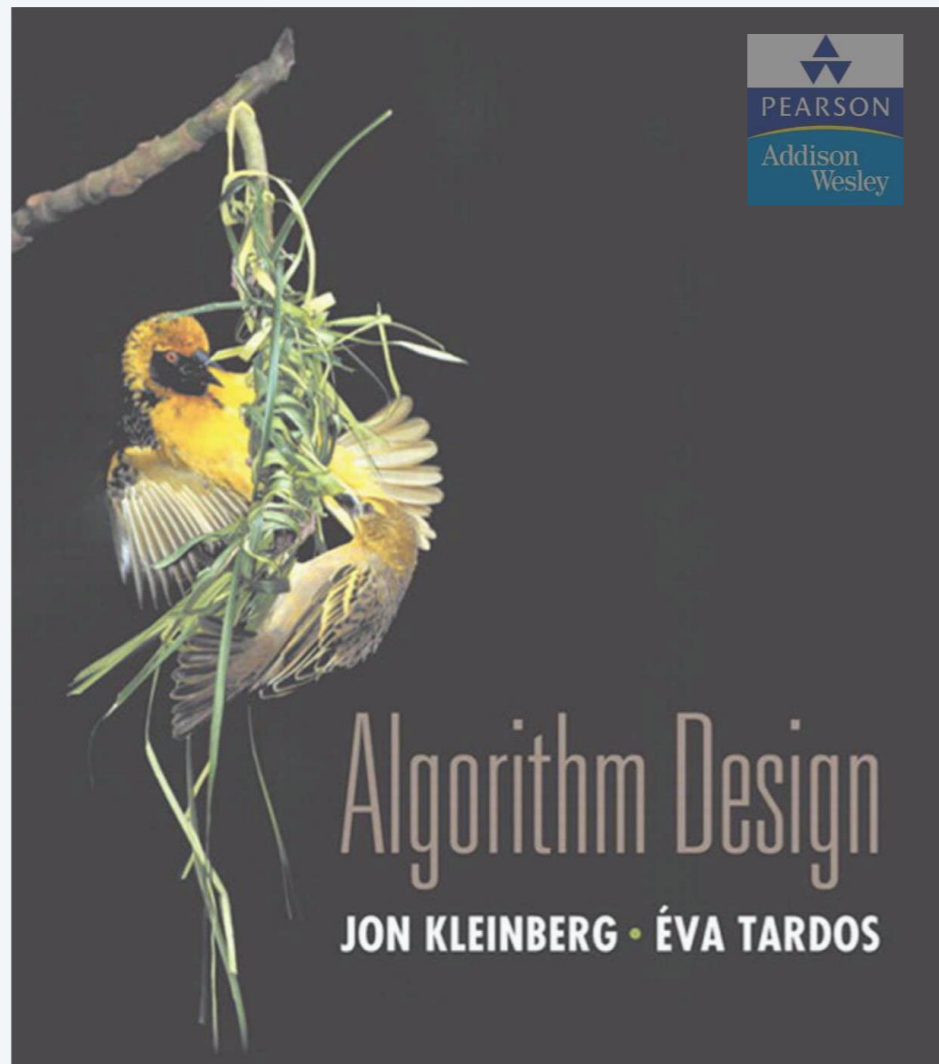
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**Design algorithms.** If  $X \leq_p Y$  and  $Y$  can be solved in polynomial time, then  $X$  can be solved in polynomial time.

**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

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- ▶ *packing and covering problems*

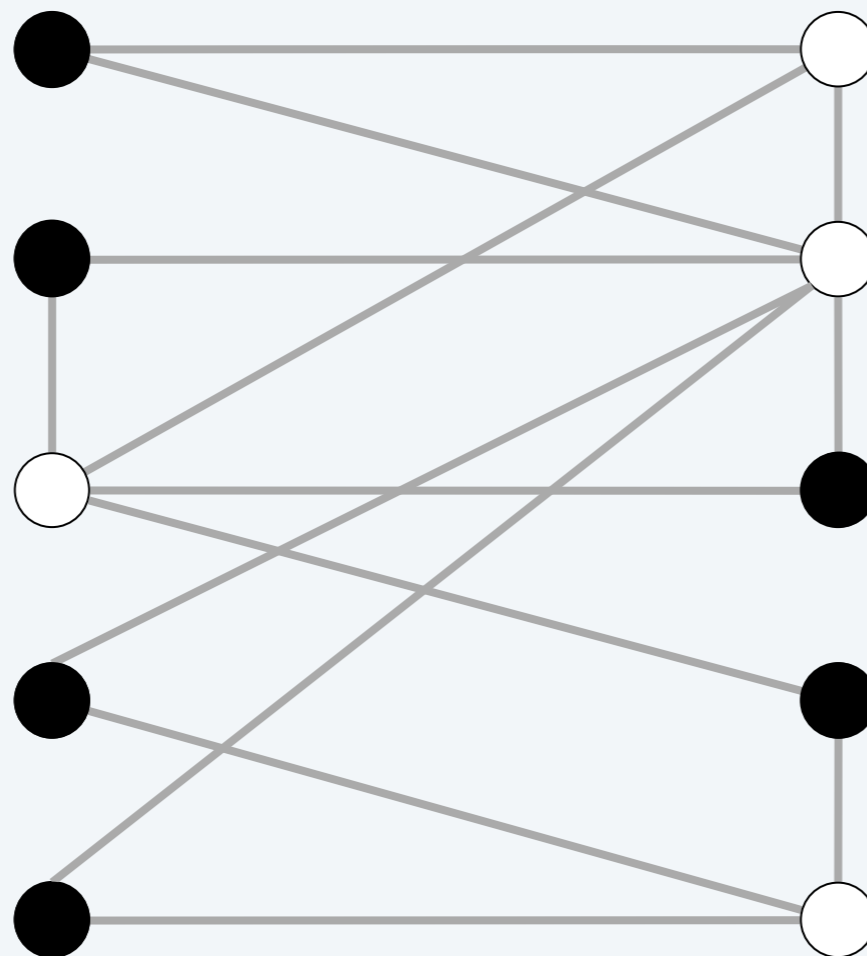
# Independent set

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**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

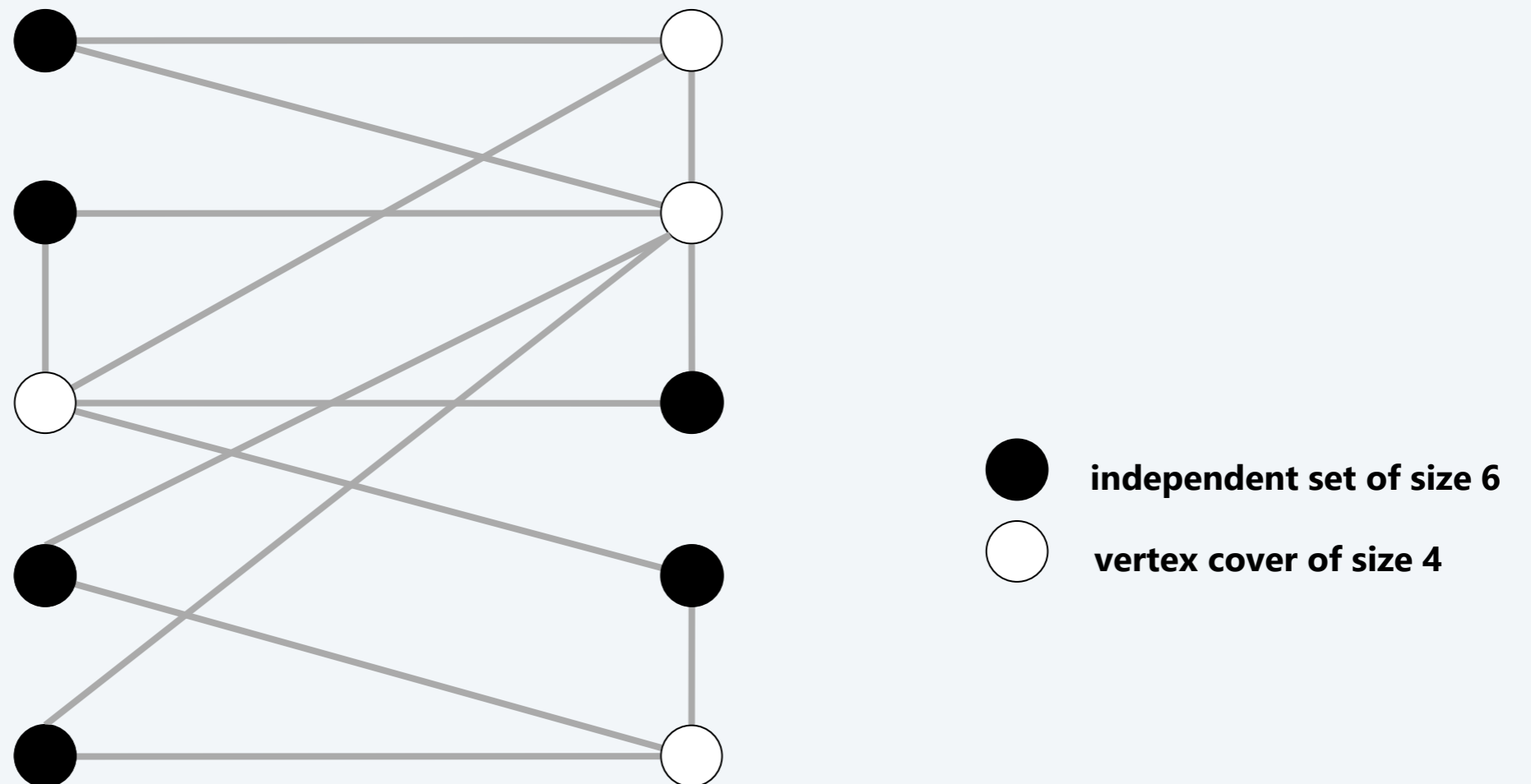
# Vertex cover

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**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  $\leq 3$ ?

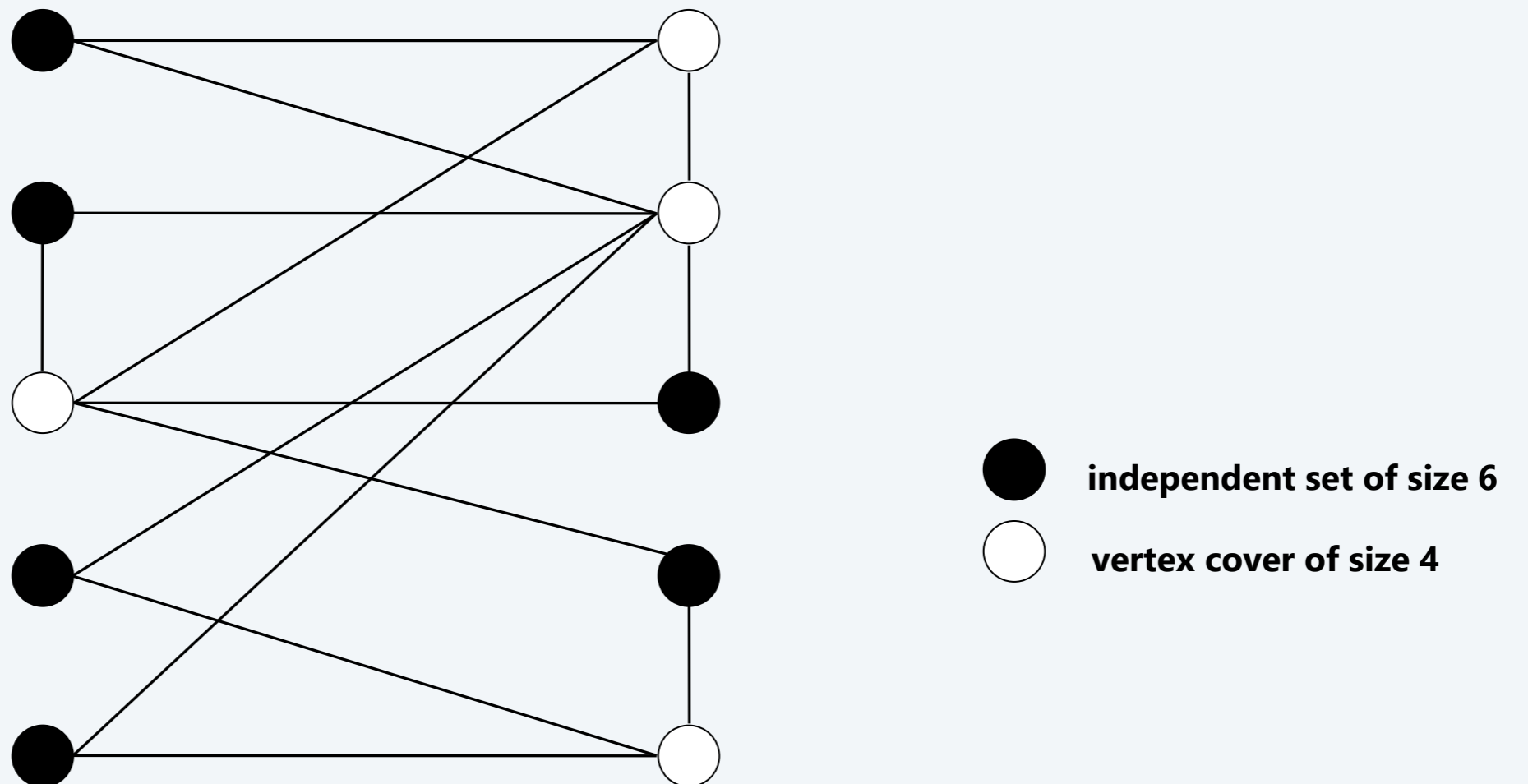


# Vertex cover and independent set reduce to one another

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**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

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**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  $n - k$ .

$\Rightarrow$

- 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  $(u, v)$ . ▪

# Vertex cover and independent set reduce to one another

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**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  $V - S$  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

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**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  $i^{\text{th}}$  piece of software provides the set  $S_i \subseteq U$  of capabilities.
- Goal: achieve all  $n$  capabilities using fewest pieces of software.

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

$$S_a = \{ 3, 7 \}$$

$$S_b = \{ 2, 4 \}$$

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

$$S_d = \{ 5 \}$$

$$S_e = \{ 1 \}$$

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

$$k = 2$$

**a set cover instance**



One more example

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

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

$$S_a = \{ 1, 4, 6 \}$$

$$S_b = \{ 1, 6, 7 \}$$

$$S_c = \{ 1, 2, 3, 6 \}$$

$$S_d = \{ 1, 3, 5, 7 \}$$

$$S_e = \{ 2, 6, 7 \}$$

$$S_f = \{ 3, 4, 5 \}$$

minimum size: 3

# Vertex cover reduces to set cover

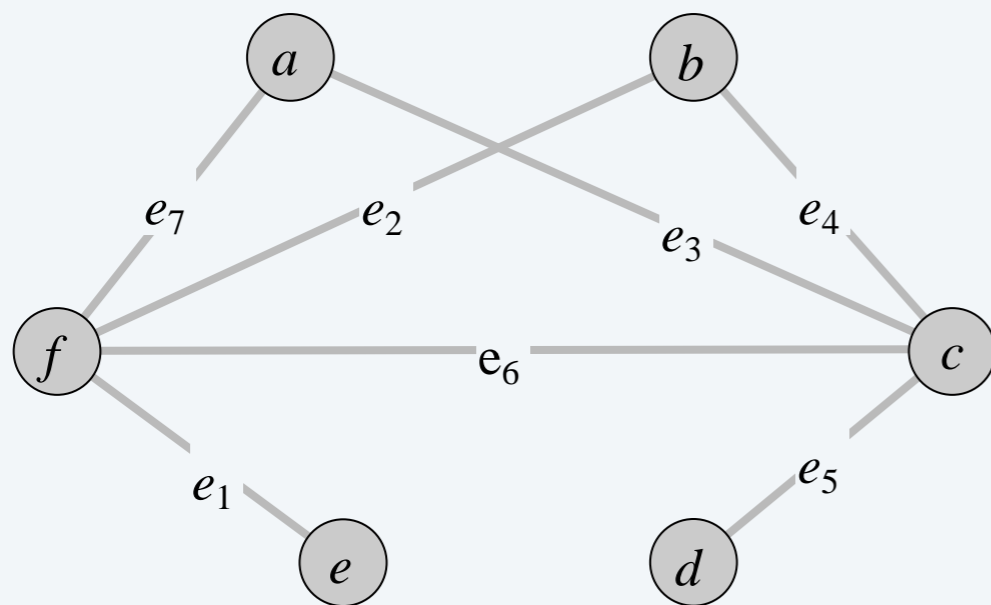
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**Theorem.** VERTEX-COVER  $\leq_P$  SET-COVER.

**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  $v \in V$ :  $S_v = \{e \in E : e \text{ incident to } v\}$ .



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

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

**set cover instance**  
( $k = 2$ )

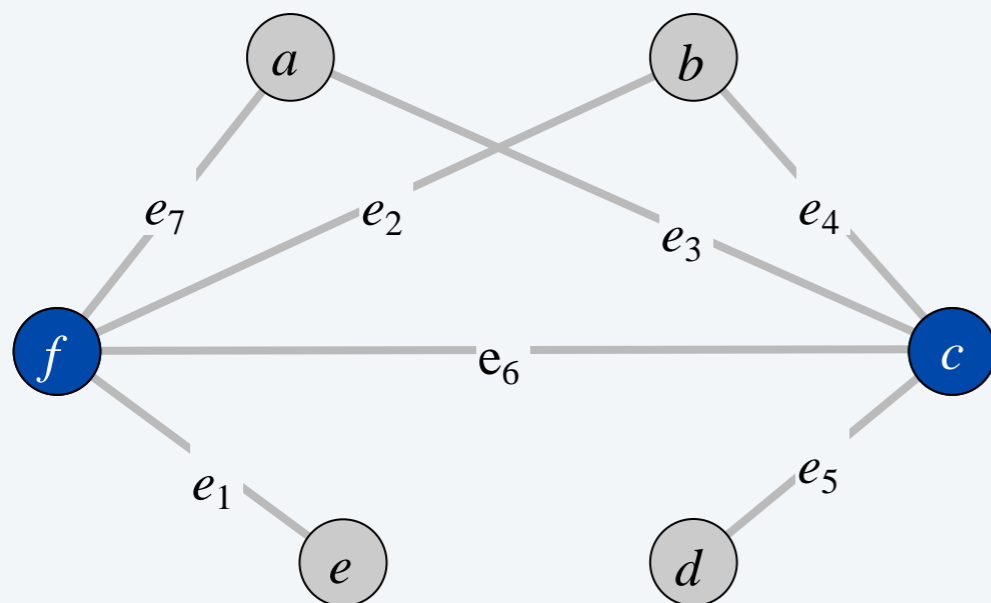
# Vertex cover reduces to set cover

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

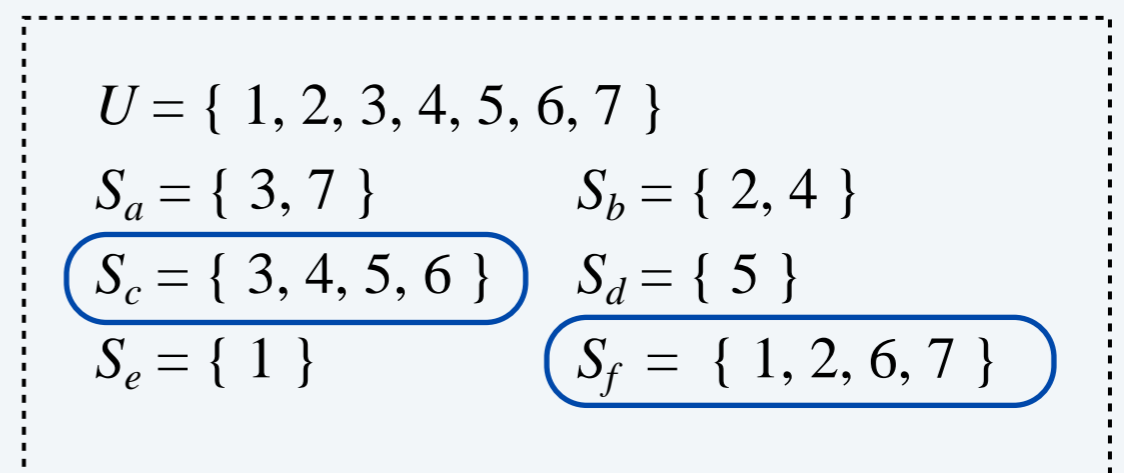
**Pf.**  $\Rightarrow$  Let  $X \subseteq V$  be a vertex cover of size  $k$  in  $G$ .

- 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**  
( $k = 2$ )



**set cover instance**  
( $k = 2$ )

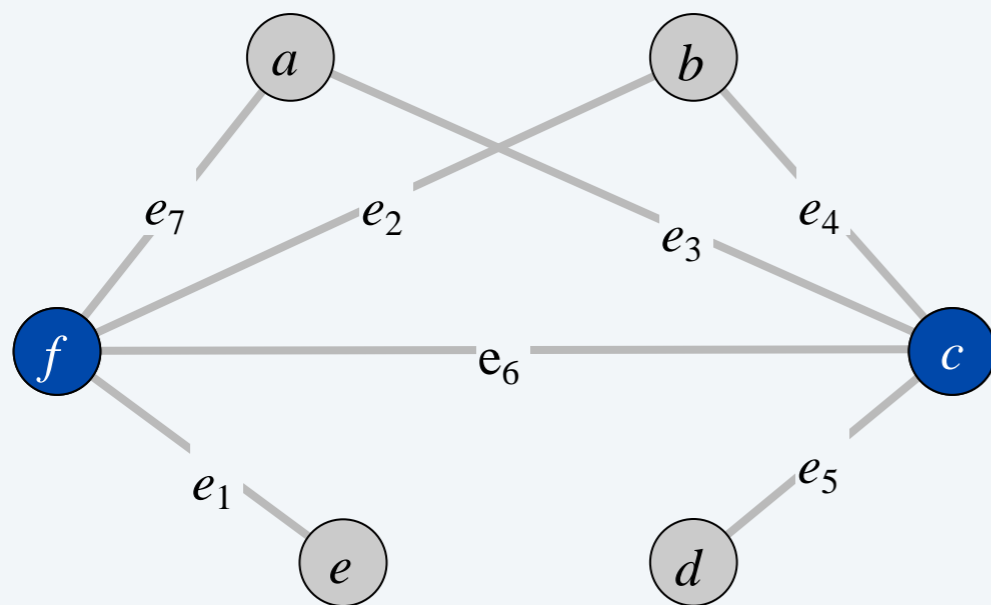
# Vertex cover reduces to set cover

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

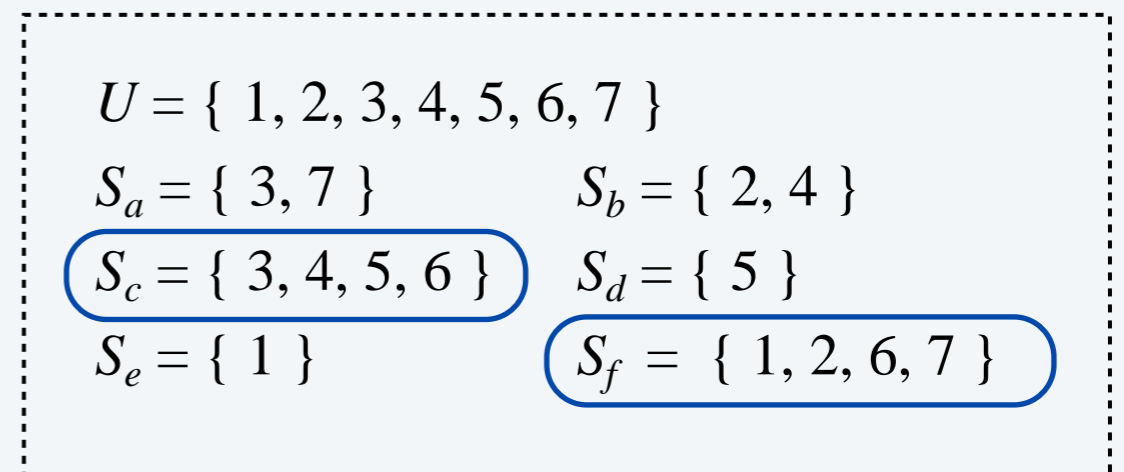
**Pf.**  $\Leftarrow$  Let  $Y \subseteq S$  be a set cover of size  $k$  in  $(U, S, k)$ .

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

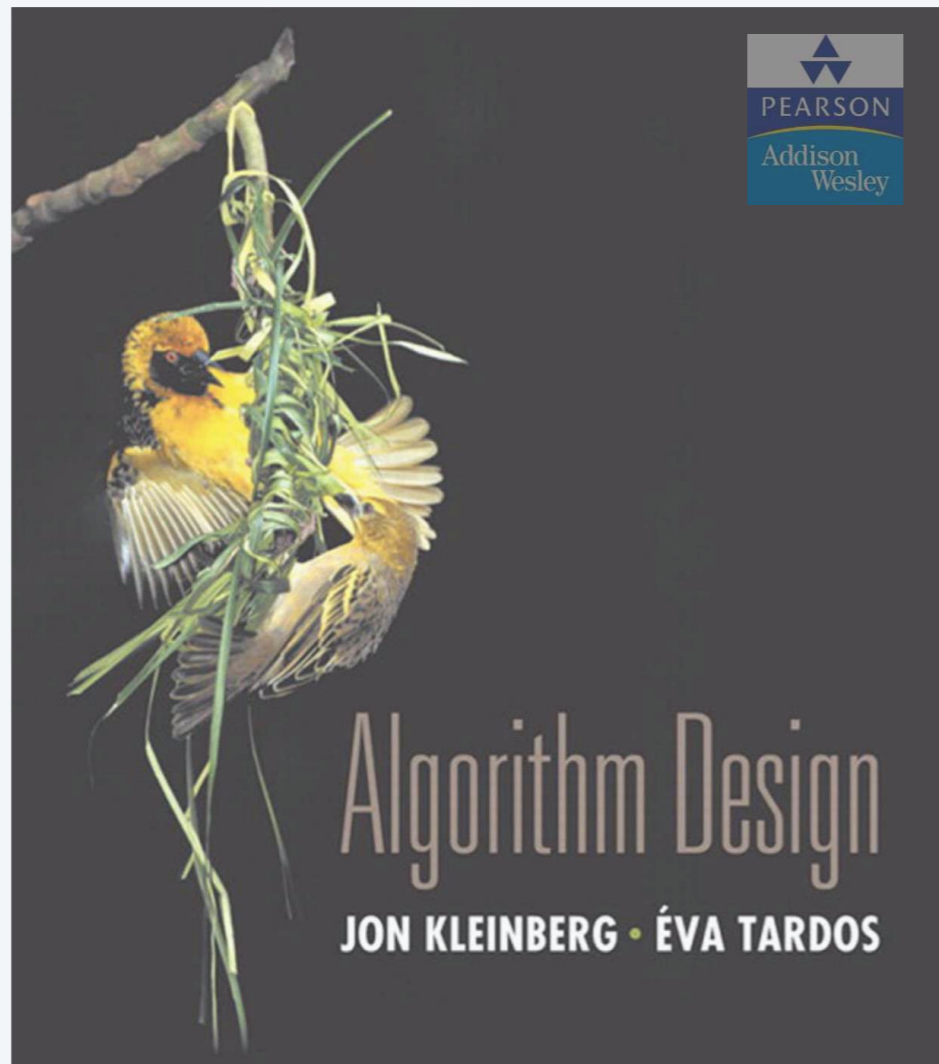
“no” instances of VERTEX-COVER are solved correctly



vertex cover instance ( $k = 2$ )



set cover instance ( $k = 2$ )



## SECTION 8.2

# 8. INTRACTABILITY I

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- ▶ *constraint satisfaction problems*

# Satisfiability

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**Literal.** A Boolean variable or its negation.

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

**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).

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

**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

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

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

# 3-satisfiability reduces to independent set

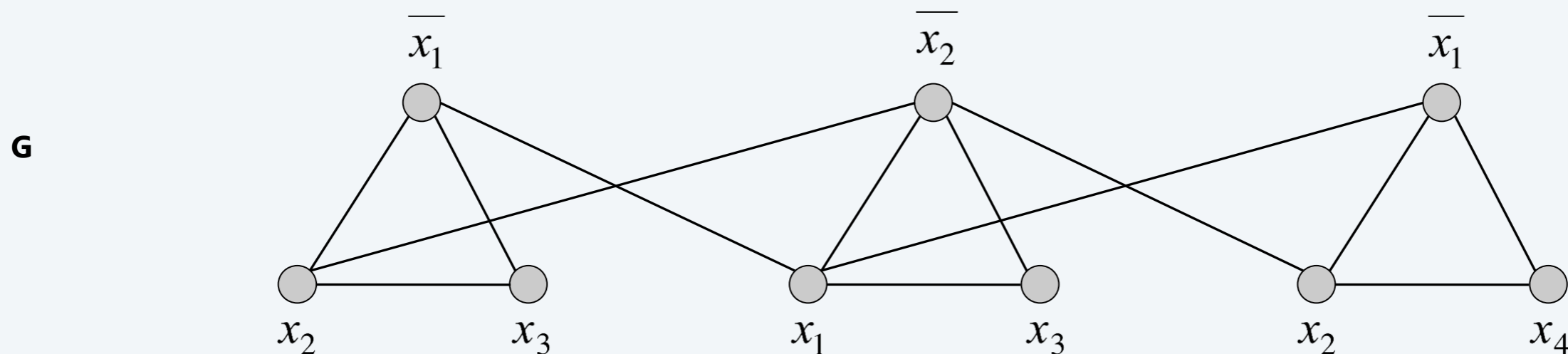
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**Theorem.**  $3\text{-SAT} \leq_P \text{INDEPENDENT-SET}$ .

**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.**

- $G$  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.



**k = 3**

$$\Phi = (\bar{x}_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \bar{x}_2 \vee x_3) \wedge (\bar{x}_1 \vee x_2 \vee x_4)$$



# 3-satisfiability reduces to independent set

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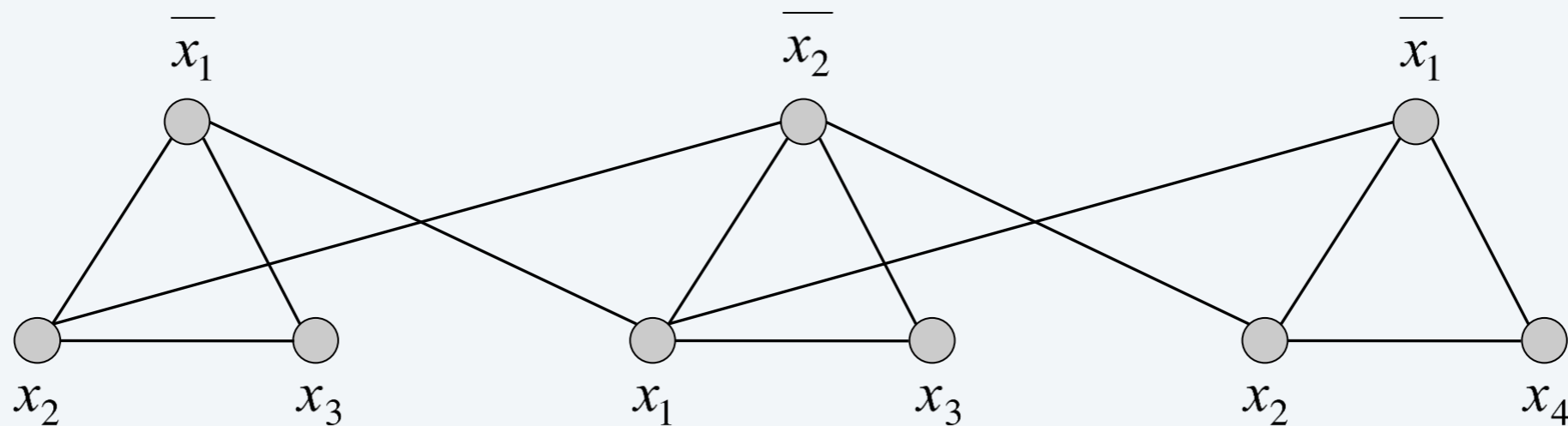
**Lemma.**  $\Phi$  is satisfiable iff  $G$  contains an independent set of size  $k = |\Phi|$ .

**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**



**k = 3**

$$\Phi = (\bar{x}_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \bar{x}_2 \vee x_3) \wedge (\bar{x}_1 \vee x_2 \vee x_4)$$

# 3-satisfiability reduces to independent set

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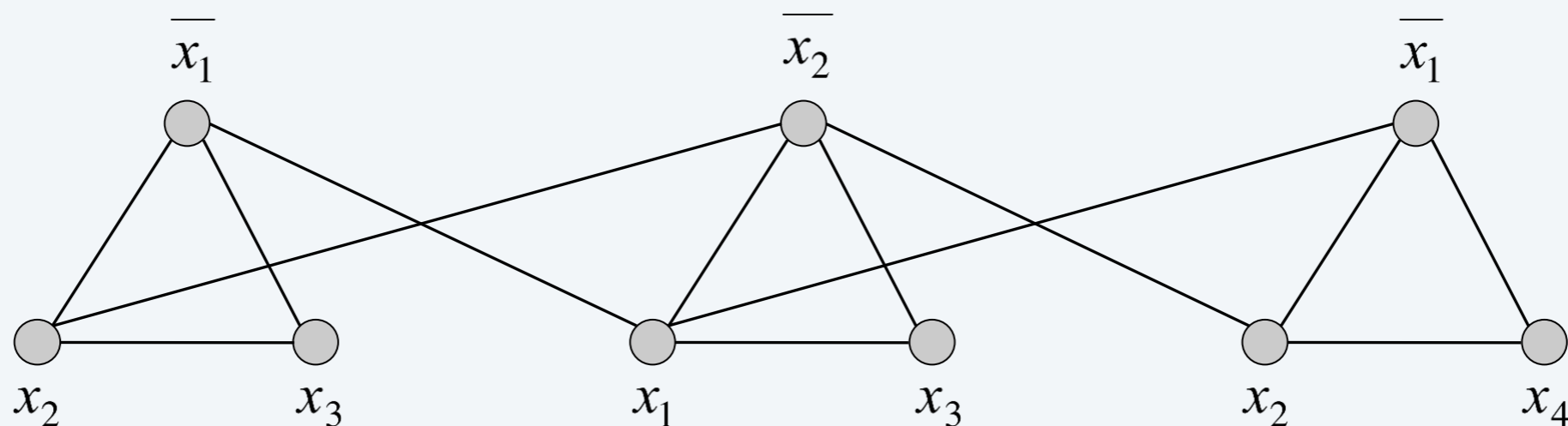
**Lemma.**  $\Phi$  is satisfiable iff  $G$  contains an independent set of size  $k = |\Phi|$ .

**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. ▪

“no” instances of 3-SAT  
are solved correctly

**G**



**k = 3**

$$\Phi = (\bar{x}_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \bar{x}_2 \vee x_3) \wedge (\bar{x}_1 \vee x_2 \vee x_4)$$

# Review

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## Basic reduction strategies.

- Simple equivalence:  $\text{INDEPENDENT-SET} \equiv_P \text{VERTEX-COVER}$ .
- Special case to general case:  $\text{VERTEX-COVER} \leq_P \text{SET-COVER}$ .
- Encoding with gadgets:  $3\text{-SAT} \leq_P \text{INDEPENDENT-SET}$ .

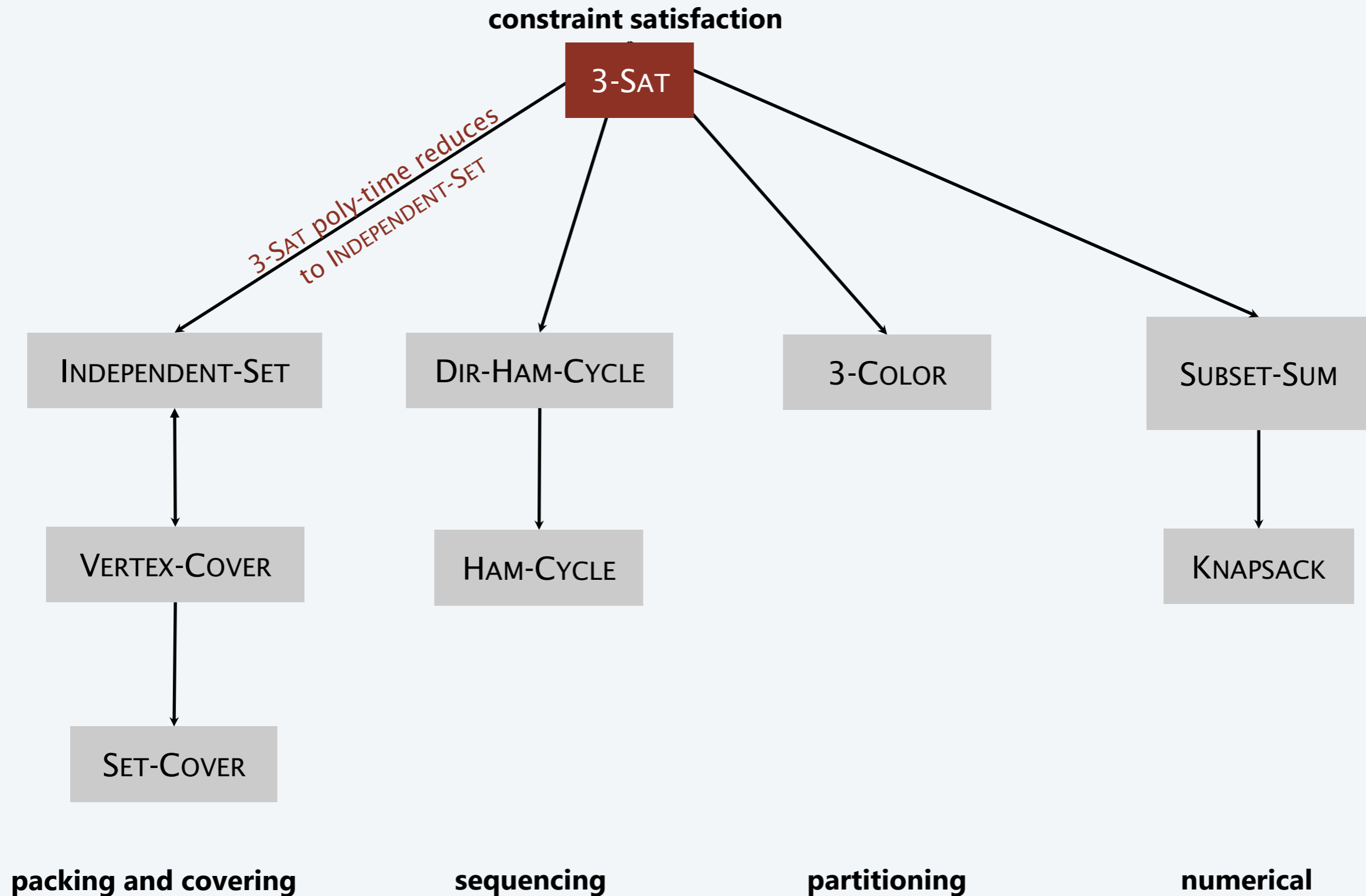
**Transitivity.** If  $X \leq_P Y$  and  $Y \leq_P Z$ , then  $X \leq_P Z$ .

**Pf idea.** Compose the two algorithms.

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

# Poly-time reductions

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# 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.

# SEARCH PROBLEMS VS. DECISION PROBLEMS



**VERTEX-COVER.** Does there exist a vertex cover of size  $\leq k$ ?

**FIND-VERTEX-COVER.** Find a vertex cover of size  $\leq k$ .

**Theorem.**  $\text{VERTEX-COVER} \equiv_p \text{FIND-VERTEX-COVER}$ .

**Pf.  $\leq_p$**  Decision problem is a special case of search problem. ▀

**Pf.  $\geq_p$**

To find a vertex cover of size  $\leq k$  :

- Determine if there exists a vertex cover of size  $\leq k$ .
- Find a vertex  $v$  such that  $G - \{v\}$  has a vertex cover of size  $\leq k - 1$ .  
(any vertex in any vertex cover of size  $\leq k$  will have this property)
- Include  $v$  in the vertex cover.
- Recursively find a vertex cover of size  $\leq k - 1$  in  $G - \{v\}$ . ▀

delete  $v$  and all incident edges

# OPTIMIZATION PROBLEMS VS. SEARCH PROBLEMS

**FIND-VERTEX-COVER.** Find a vertex cover of size  $\leq k$ .

**FIND-MIN-VERTEX-COVER.** Find a vertex cover of minimum size.

**Theorem.**  $\text{FIND-VERTEX-COVER} \equiv_P \text{FIND-MIN-VERTEX-COVER}$ .

**Pf.  $\leq_P$**  Search problem is a special case of optimization problem. ■

**Pf.  $\geq_P$**  To find vertex cover of minimum size:

- Binary search (or linear search) for size  $k^*$  of min vertex cover.
- Solve search problem for given  $k^*$ . ■