

# Chapter 8

# NP and Computational Intractability



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## Algorithm Design Patterns and Anti-Patterns

## Algorithm design patterns.

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

### Ex.

O(n log n) interval scheduling. O(n log n) FFT. O(n<sup>2</sup>) edit distance. O(n<sup>3</sup>) bipartite matching.

## Algorithm design anti-patterns.

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

O(n<sup>k</sup>) algorithm unlikely. O(n<sup>k</sup>) certification algorithm unlikely. No algorithm possible.

# 8.1 Polynomial-Time Reductions

Classify Problems According to Computational Requirements

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

A working definition. [von Neumann 1953, Godel 1956, Cobham 1964, Edmonds 1965, Rabin 1966] Those with polynomial-time algorithms.

Yes	Probably no
Shortest path	Longest path
Matching	3D-matching
Min cut	Max cut
2-SAT	3-SAT
Planar 4-color	Planar 3-color
Bipartite vertex cover	Vertex cover
Primality testing	Factoring

## 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 Turing machine, does it halt in at most k steps?
- Given a board position in an n-by-n generalization of chess, can black guarantee a win?

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

This chapter. Show that these fundamental problems are "computationally equivalent" and appear to be different manifestations of one really hard problem.

## Polynomial-Time Reduction

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

don't confuse with reduces from

Reduction. Problem X polynomial 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. 
$$X \leq_P Y$$
.

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

## Remarks.

- We pay for time to write down instances sent to black box  $\Rightarrow$  instances of Y must be of *polynomial size in the instance of* X.
- Note: Cook reducibility.

in contrast to Karp reductions

## Polynomial-Time Reduction

Purpose. Classify problems according to relative difficulty.

The use of poly-time reduction.

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

Establish equivalence. If  $X \leq_{P} Y$  and  $Y \leq_{P} X$ , we use notation  $X \equiv_{P} Y$ .

# Reduction By Simple Equivalence

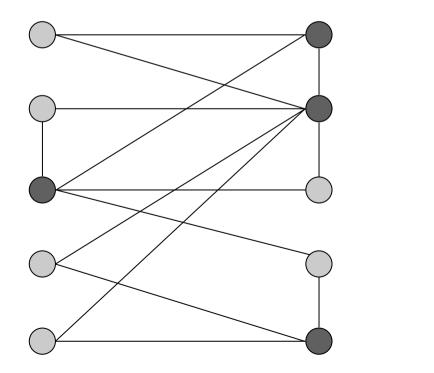
Basic reduction strategies.

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

## Independent Set

**INDEPENDENT SET:** Given a graph G = (V, E) and an integer k, is there a subset of vertices  $S \subseteq V$  such that  $|S| \ge k$ , and for each edge at most one of its endpoints is in S?

Ex. Is there an independent set of size  $\ge 6$ ? Yes. Ex. Is there an independent set of size  $\ge 7$ ? No.

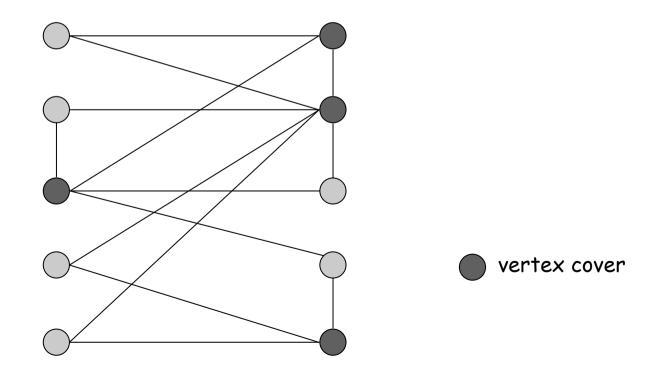




## Vertex Cover

VERTEX COVER: Given a graph G = (V, E) and an integer k, is there a subset of vertices  $S \subseteq V$  such that  $|S| \le k$ , and for each edge, at least one of its endpoints is in S?

Ex. Is there a vertex cover of size  $\leq$  4? Yes. Ex. Is there a vertex cover of size  $\leq$  3? No.



## Hardness of Vertex Cover

VERTEX COVER: Given a graph G = (V, E) and an integer k, is there a subset of vertices  $S \subseteq V$  such that  $|S| \le k$ , and for each edge, at least one of its endpoints is in S?

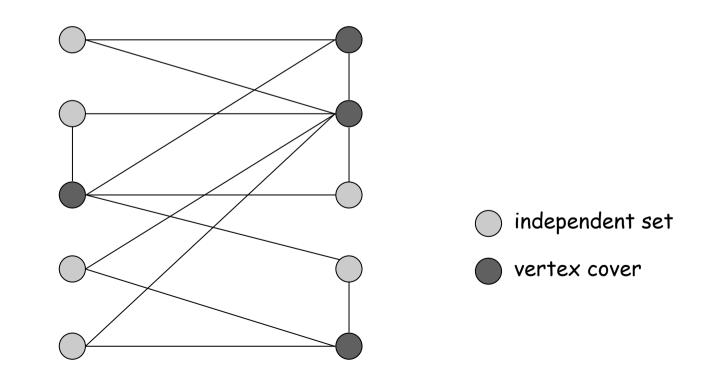
Excercise: Greedy does not work!

- 1) Select the node v with maximal degree (break ties arbitraly)
- -2) remove **v** and its edges
- -3) apply step 1 to the new graph until all edges are covered.

-Show the algorithm does not find (always) the optimum.

## Vertex Cover and Independent Set

Claim. VERTEX-COVER  $\equiv_P$  INDEPENDENT-SET. Pf. We show S is an independent set iff V - S is a vertex cover.



## Vertex Cover and Independent Set

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Claim 1. S is an independent set iff V - S is a vertex cover Pf.
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#### $\Rightarrow$

- Let S be any independent set.
- Consider an arbitrary edge (u, v).
- S independent  $\Rightarrow$  u  $\notin$  S or v  $\notin$  S  $\Rightarrow$  u  $\in$  V S or v  $\in$  V S.
- Thus, V S covers (u, v).

#### (

- Let V S be any vertex cover.
- Consider two nodes  $\mathbf{u} \in \mathbf{S}$  and  $\mathbf{v} \in \mathbf{S}$ .
- Observe that  $(u, v) \notin E$  since V S is a vertex cover.
- Thus, no two nodes in S are joined by an edge  $\rightarrow$  S independent set. •

Vertex Cover and Independent Set

Thm. VERTEX-COVER  $\equiv_P$  INDEPENDENT-SET. Pr. VC  $\leq_P$  IS; Hypothesis: Poly-time Alg. A(\*\*) for IS

Given instance X = <G(V,E) ; k>0 > of VC;

- Transform it into instance Y=<G(V,E) , k'=n-k>;

-- Apply Algo A(Y) (thanks to Claim 1)

 $IS \leq_{P} VC;$ 

-Similar.

# Reduction from Special Case to General Case

Basic reduction strategies.

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

## Set Cover

SET COVER: Given a set of elements, a collection  $S_1, S_2, \ldots, S_m$  of subsets of U, and an integer k, does there exist a collection of  $\leq k$  of these sets whose union is equal to U?

## Sample application.

- m available pieces of software W1,...,Wm.
- Set U of n capabilities that we would like our system to have.
- The ith piece of software provides the set  $S_i \subseteq U$  of capabilities.
- Goal: achieve all n capabilities using fewest pieces of software.

Ex:

 $U = \{ 1, 2, 3, 4, 5, 6, 7 \}$ k = 2  $S_1 = \{ 3, 7 \}$   $S_4 = \{ 2, 4 \}$   $S_2 = \{ 3, 4, 5, 6 \}$   $S_5 = \{ 5 \}$  $S_3 = \{ 1 \}$   $S_6 = \{ 1, 2, 6, 7 \}$ 

## Vertex Cover Reduces to Set Cover

Claim. VERTEX-COVER  $\leq_{P}$  SET-COVER. Pf. Given a VERTEX-COVER instance G = (V, E), k, we construct a set cover instance whose size equals the size of the vertex cover instance. Construction.

-GOAL of reductions: **Programming with the language of the second problem** 

VC Language: Q. What means when we select a node v in V? A. We cover all its edges

SC Language: 1 edge  $\rightarrow$  1 element of the Universe U 1 vertex  $\mathbf{v} \rightarrow$  1 subset of U = edges covered by  $\mathbf{v}$ 

• Create SET-COVER instance:

- k = k, U = E, for any  $v \in V$ :  $S_v = \{e \in E : e \text{ incident to } v\}$ 

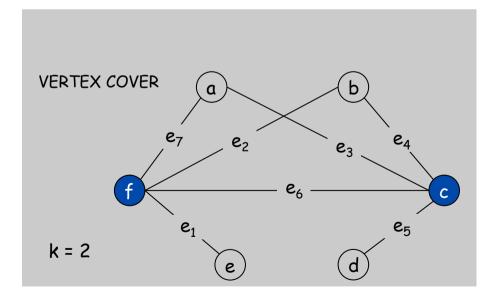
## Vertex Cover Reduces to Set Cover

Claim. VERTEX-COVER  $\leq_{P}$  SET-COVER.

• Create SET-COVER instance:

- k = k, U = E, for any  $v \in V$ :  $S_v = \{e \in E : e \text{ incident to } v\}$ 

• Set Cover of size  $\leq k$  iff Vertex Cover of size  $\leq k$ .



SET COVER	
S <sub>c</sub> = {3, 4, 5, 6}	} S <sub>b</sub> = {2, 4} S <sub>d</sub> = {5} S <sub>f</sub> = {1, 2, 6, 7}

## Polynomial-Time Reduction

### Basic strategies.

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction by encoding with gadgets.

# 8.2 Reductions via "Gadgets"

#### Basic reduction strategies.

- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction via "gadgets."

## Satisfiability

Literal: A Boolean variable or its negation.  $x_i$  or  $x_i$ Clause: A disjunction of literals.  $C_j = x_1 \vee \overline{x_2} \vee x_3$ 

Conjunctive normal form: A propositional formula  $\Phi$  that is the conjunction of clauses.

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

**Def.** The size  $|\Phi|$  is the number of clauses. SAT: Given CNF formula  $\Phi$ , does it have a satisfying truth assignment?

3-SAT: SAT where each clause contains exactly 3 literals.

Ex: 
$$(\overline{X_1} \lor X_2 \lor X_3) \land (X_1 \lor \overline{X_2} \lor X_3) \land (X_2 \lor X_3) \land (\overline{X_1} \lor \overline{X_2} \lor \overline{X_3})$$
  
Yes:  $x_1 = \text{true}, x_2 = \text{true} x_3 = \text{false}.$ 

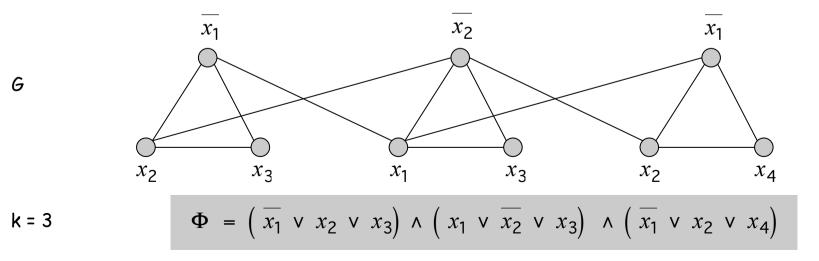
3 Satisfiability Reduces to Independent Set

Claim.  $3-SAT \leq_{P} INDEPENDENT-SET.$ 

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 iff  $\Phi$  is satisfiable.

Construction.

- G contains 3 vertices for each clause, one for each literal.
- Connect 3 literals in a clause in a triangle = GADGET
- Connect positive literal to each of its negations.
- set k = |Φ|

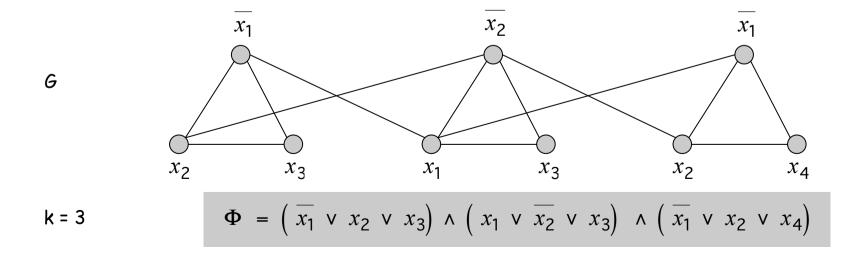


## 3 Satisfiability Reduces to Independent Set

Claim. G contains independent set of size  $\mathbf{k} = |\Phi|$  iff  $\Phi$  is satisfiable. Pf.  $\Rightarrow$  Let S be independent set of size  $\mathbf{k}$ .

- S must contain exactly one vertex (= literal) in each triangle.
- Set these literals (= vertices) to **true**.
- No adjacent vertices in S → nonconstructive field in the consistent way
- Truth assignment is consistent and all clauses are satisfied.

 $Pf \leftarrow Given satisfying assignment, select one true literal from each triangle. No contradictions <math>\rightarrow$  This is an independent set of size k.



## Review

Basic reduction strategies.

- Simple equivalence: INDEPENDENT-SET =  $_{P}$  VERTEX-COVER.
- Special case to general case: VERTEX-COVER  $\leq_{P}$  SET-COVER.
- Encoding with gadgets:  $3-SAT \leq P$  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:  $3-SAT \leq_{P} INDEPENDENT-SET \leq_{P} VERTEX-COVER \leq_{P} SET-COVER$ .

## Self-Reducibility

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

Self-reducibility. Search 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