

CS 580: Algorithm Design and Analysis

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Homework 4: Due tomorrow (March 9) at 11:59 PM

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.
- $O(n^2)$ edit distance.
- $O(n^2)$ bipartite matching.

Algorithm design anti-patterns.

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

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Recap

Linear Programming

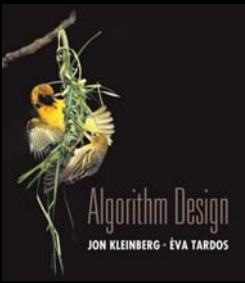
- Very Powerful Technique (Subject of Entire Courses)
- Our Focus: Using Linear Programming as a Tool
 - Solving Network Flow using Linear Programming
 - Finding Minimax Optimal Strategy in 2-Player Zero Sum Game
 - Operations Research (Brewery Example)

Solving Linear Programs

- Simplex Intuition:
 - Optimal point is an "extreme point"
 - No "local optimum"
- Simplex Runs in Exponential Time in Worst Case
 - But other algorithms (e.g., Ellipsoid) run in polynomial time

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8.1 Polynomial-Time Reductions



Chapter 8

NP and Computational Intractability

Algorithm Design
JON KLEINBERG · ÉVA TARDOS

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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

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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.

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Reduction By Simple Equivalence

Basic reduction strategies.

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

Polynomial-Time Reduction

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

don't confuse with reduces from

Reduction. Problem X **polynomially 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.
- Note: Cook reducibility. in contrast to Karp reductions

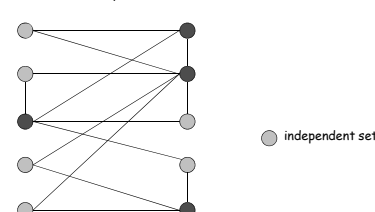
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Independent Set

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

Ex. Is there an independent set of size ≥ 6 ? Yes.

Ex. Is there an independent set of size ≥ 7 ? No.



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Polynomial-Time Reduction

Purpose. Classify problems according to **relative** difficulty.

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.

Establish equivalence. If $X \leq_p Y$ and $Y \leq_p X$, we use notation $X \equiv_p Y$. up to cost of reduction

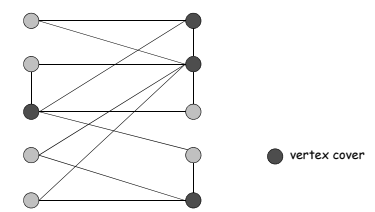
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Vertex Cover

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

Ex. Is there a vertex cover of size ≤ 4 ? Yes.

Ex. Is there a vertex cover of size ≤ 3 ? No.



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Vertex Cover and Independent Set

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

independent set
 vertex cover

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Set Cover

SET COVER: Given a set U of elements, a collection S_1, S_2, \dots, 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.
- 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.

Ex:

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

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Vertex Cover and Independent Set

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

\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) .

\Leftarrow

- Let $V - S$ be any vertex cover.
- Consider two nodes $u \in S$ and $v \in S$.
- 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.

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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.

- Create SET-COVER instance:
 - $k = k$, $U = E$, $S_v = \{e \in E : e \text{ incident to } v\}$
- Set-cover of size $\leq k$ iff vertex cover of size $\leq k$.

VERTEX COVER

$k = 2$

SET COVER

$U = \{1, 2, 3, 4, 5, 6, 7\}$
 $k = 2$
 $S_1 = \{3, 7\}$ $S_5 = \{2, 4\}$
 $S_2 = \{3, 4, 5, 6\}$ $S_6 = \{5\}$
 $S_3 = \{1\}$ $S_7 = \{1, 2, 6, 7\}$

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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.

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Polynomial-Time Reduction

Basic strategies.

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

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8.2 Reductions via "Gadgets"

Basic reduction strategies.

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

3 Satisfiability Reduces to Independent Set

Claim. G contains independent set of size $k = |\Phi|$ iff Φ is satisfiable.

Pf. \Rightarrow Let S be independent set of size k .

- S must contain exactly one vertex in each triangle.
- Set these literals to true. — and any other variables in a consistent way
- Truth assignment is consistent and all clauses are satisfied.

Pf \Leftarrow Given satisfying assignment, select one true literal from each triangle. This is an independent set of size k .

$k = 3$

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

Satisfiability

Literal: A Boolean variable or its negation. x_i or \bar{x}_i

Clause: A disjunction of literals. $C_j = x_1 \vee \bar{x}_2 \vee x_3$

Conjunctive normal form: A propositional formula Φ that is the conjunction of clauses. $\Phi = C_1 \wedge C_2 \wedge C_3 \wedge C_4$

SAT: Given CNF formula Φ , does it have a satisfying truth assignment?

3-SAT: SAT where each clause contains exactly 3 literals.
each corresponds to a different variable

Ex: $(\bar{x}_1 \vee x_2 \vee x_3) \wedge (\bar{x}_1 \vee \bar{x}_2 \vee x_3) \wedge (x_2 \vee x_3) \wedge (\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_3)$

Yes: $x_1 = \text{true}, x_2 = \text{true}, x_3 = \text{false}$.

Review

Basic reduction strategies.

- Simple equivalence: INDEPENDENT-SET \equiv_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.

3 Satisfiability Reduces to Independent Set

Claim. 3-SAT \leq_p INDEPENDENT-SET.

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

Construction.

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

$k = 3$

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

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

8.3 Definition of NP

NP

Certification algorithm intuition.

- Certifier views things from "managerial" viewpoint.
- Certifier doesn't determine whether $s \in X$ on its own; rather, it checks a proposed proof t that $s \in X$.

Def. Algorithm $C(s, t)$ is a **certifier** for problem X if for every string s , $s \in X$ iff there exists a string t such that $C(s, t) = \text{yes}$.

"certificate" or "witness"

NP. Decision problems for which there exists a **poly-time** certifier.

$C(s, t)$ is a poly-time algorithm and $|t| \leq p(|s|)$ for some polynomial $p(\cdot)$.

Remark. NP stands for **nondeterministic** polynomial-time.

Decision Problems

Decision problem.

- X is a set of strings.
- Instance: string s .
- Algorithm A solves problem X : $A(s) = \text{yes}$ iff $s \in X$.

Polynomial time. Algorithm A runs in poly-time if for every string s , $A(s)$ terminates in at most $p(|s|)$ "steps", where $p(\cdot)$ is some polynomial.

length of s

PRIMES: $X = \{ 2, 3, 5, 7, 11, 13, 17, 23, 29, 31, 37, \dots \}$

Algorithm. [Agrawal-Kayal-Saxena, 2002] $p(|s|) = |s|^8$.

Certifiers and Certificates: Composite

COMPOSITES. Given an integer s , is s composite?

Certificate. A nontrivial factor t of s . Note that such a certificate exists iff s is composite. Moreover $|t| \leq |s|$.

Certifier.

```
boolean C(s, t) {
  if (t <= 1 or t >= s)
    return false
  else if (s is a multiple of t)
    return true
  else
    return false
}
```

Instance. $s = 437,669$.

Certificate. $t = 541$ or 809 . — $437,669 = 541 \times 809$

Conclusion. COMPOSITES is in NP.

Definition of P

P. Decision problems for which there is a poly-time algorithm.

Problem	Description	Algorithm	Yes	No
MULTIPLE	Is x a multiple of y ?	Grade school division	51, 17	51, 16
RELPRIME	Are x and y relatively prime?	Euclid (300 BCE)	34, 39	34, 51
PRIMES	Is x prime?	AKS (2002)	53	51
EDIT-DISTANCE	Is the edit distance between x and y less than 5?	Dynamic programming	niether	acgggt ttttta
LSOLVE	Is there a vector x that satisfies $Ax = b$?	Gauss-Edmonds elimination	$\begin{bmatrix} 0 & 1 & 1 & 4 \\ 2 & 4 & -1 & 2 \\ 0 & 3 & 15 & 36 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix}$

Certifiers and Certificates: 3-Satisfiability

SAT. Given a CNF formula Φ , is there a satisfying assignment?

Certificate. An assignment of truth values to the n boolean variables.

Certifier. Check that each clause in Φ has at least one true literal.

Ex.

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

instance s

$$x_1 = 1, x_2 = 1, x_3 = 0, x_4 = 1$$

certificate t

Conclusion. SAT is in NP.

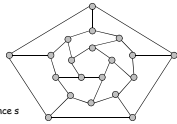
Certifiers and Certificates: Hamiltonian Cycle

HAM-CYCLE. Given an undirected graph $G = (V, E)$, does there exist a simple cycle C that visits every node?

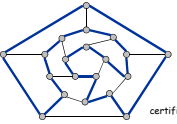
Certificate. A permutation of the n nodes.

Certifier. Check that the permutation contains each node in V exactly once, and that there is an edge between each pair of adjacent nodes in the permutation.

Conclusion. HAM-CYCLE is in NP.

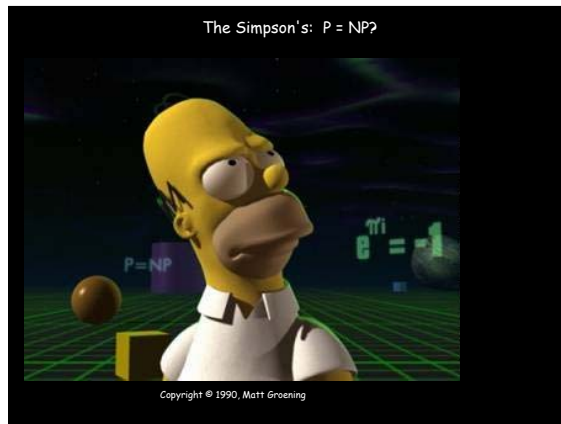


instance s



certificate t

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P, NP, EXP

P. Decision problems for which there is a **poly-time algorithm**.

EXP. Decision problems for which there is an **exponential-time algorithm**.

NP. Decision problems for which there is a **poly-time certifier**.

Claim. $P \subseteq NP$.

Pf. Consider any problem X in P .

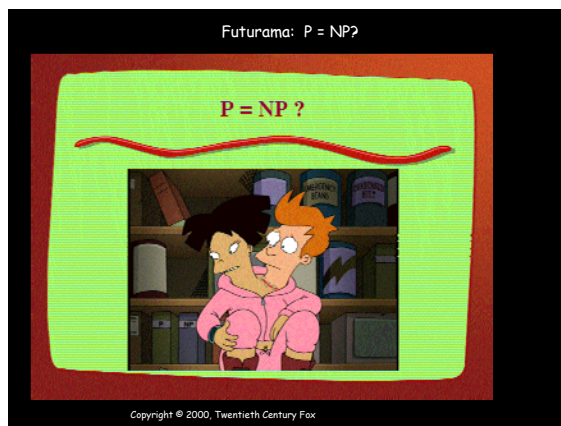
- By definition, there exists a poly-time algorithm $A(s)$ that solves X .
- Certificate: $t = \epsilon$, certifier $C(s, t) = A(s)$.

Claim. $NP \subseteq EXP$.

Pf. Consider any problem X in NP .

- By definition, there exists a poly-time certifier $C(s, t)$ for X .
- To solve input s , run $C(s, t)$ on all strings t with $|t| \leq p(|s|)$.
- Return **yes**, if $C(s, t)$ returns **yes** for any of these.

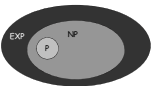
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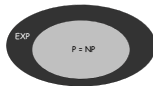
The Main Question: P Versus NP

Does $P = NP?$ [Cook 1971, Edmonds, Levin, Yablonski, Gödel]

- Is the decision problem as easy as the certification problem?
- Clay \$1 million prize.



If $P = NP$



If $P = NP$

would break RSA cryptography
(and potentially collapse economy)

If yes: Efficient algorithms for 3-COLOR, TSP, FACTOR, SAT, ...

If no: No efficient algorithms possible for 3-COLOR, TSP, SAT, ...

Consensus opinion on $P = NP?$ Probably no.

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Looking for a Job?

Some writers for the Simpsons and Futurama.

- J. Steward Burns. M.S. in mathematics, Berkeley, 1993.
- David X. Cohen. M.S. in computer science, Berkeley, 1992.
- Al Jean. B.S. in mathematics, Harvard, 1981.
- Ken Keeler. Ph.D. in applied mathematics, Harvard, 1990.
- Jeff Westbrook. Ph.D. in computer science, Princeton, 1989.

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