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INTRACTABILITY

- ▶ *introduction*
- ▶ *P vs. NP*
- ▶ *poly-time reductions*
- ▶ *NP-completeness*
- ▶ *Dealing with intractability*

Overview: introduction to advanced topics

Main topics. [final two lectures]

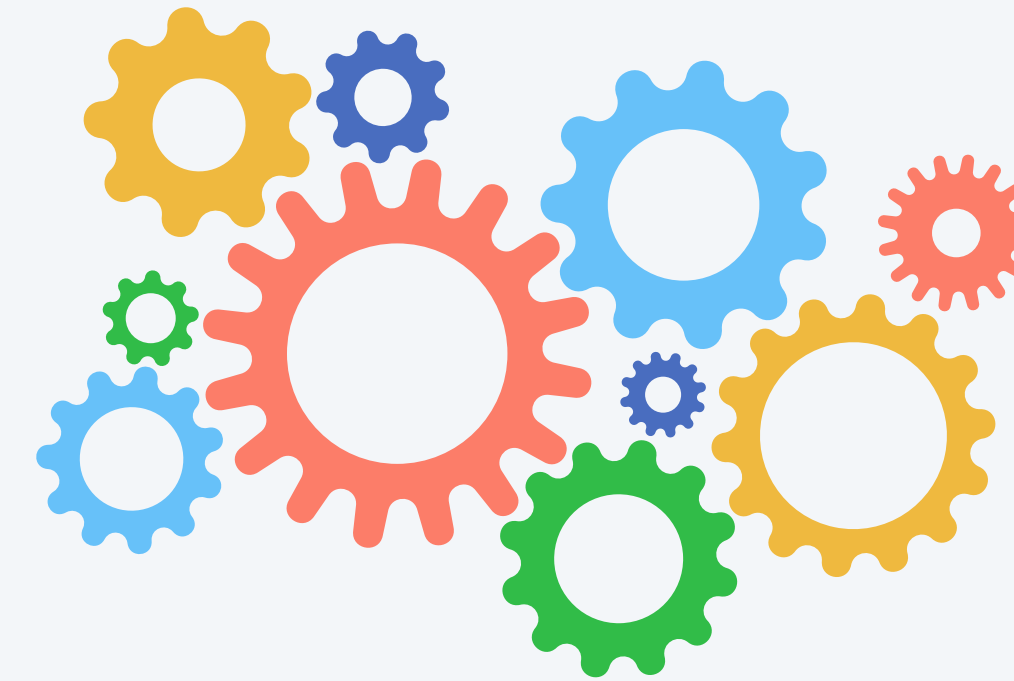
- **Intractability:** barriers to designing efficient algorithms.
- **Algorithm design:** paradigms for solving problems.

Shifting gears.

- From individual problems to problem-solving models/classes.
- From linear/quadratic to poly-time/exponential scale.
- From implementation details to conceptual frameworks.

Goals.

- Introduce you to essential ideas.
- Place algorithms and techniques we've studied in a larger context.





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Fundamental questions

Q1. What is an **algorithm**?

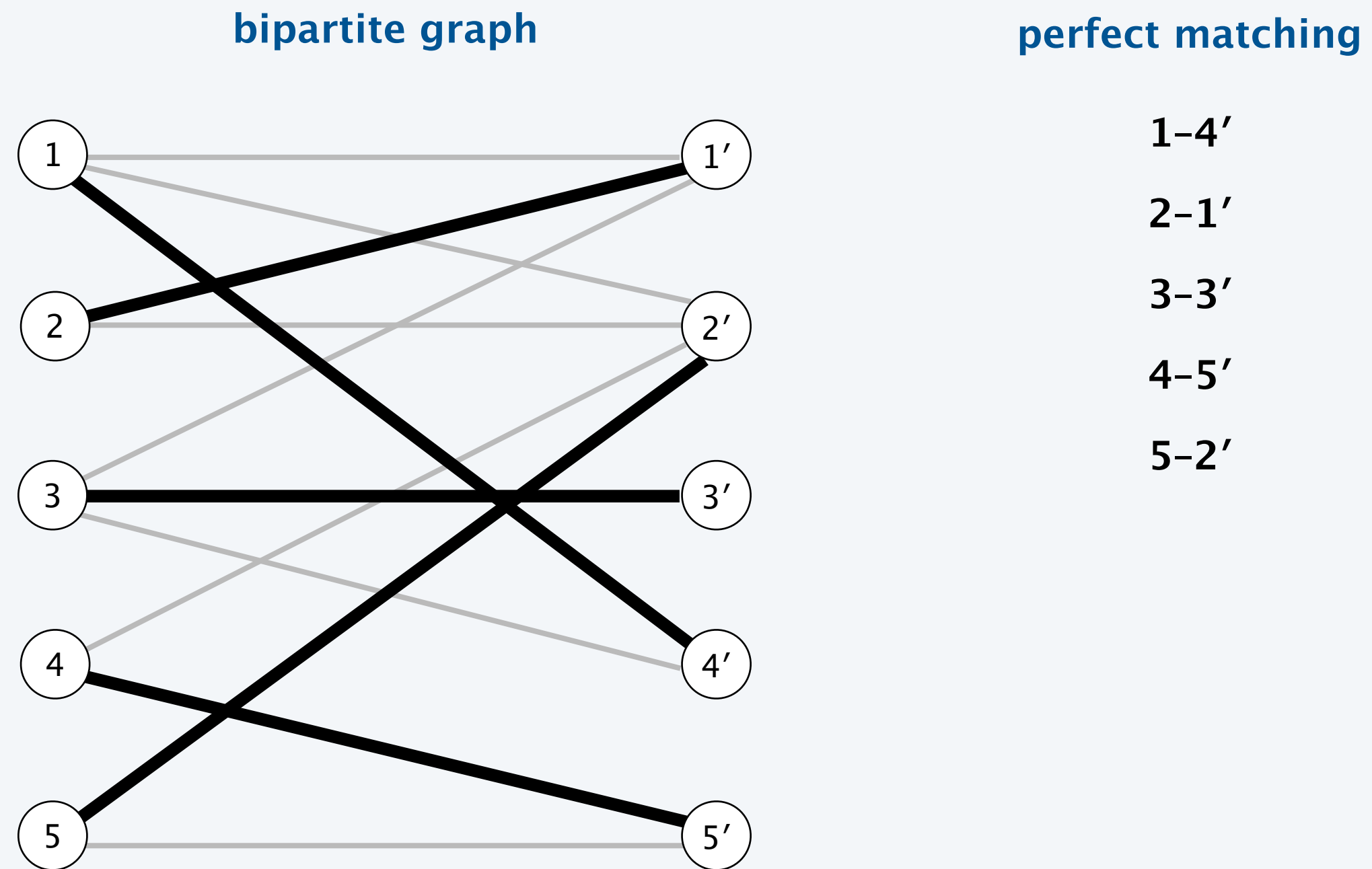
Q2. What is an **efficient** algorithm?

Q3. **Which** problems can be solved **efficiently** and which are **intractable**?


Q4. How can we **prove** that a problem is **intractable**?

A computationally easy problem: perfect matching

Perfect matching (search). Given a **bipartite graph**, find a **perfect matching** ← or report that no such matching exists (set of edges such that every vertex is an endpoint of exactly one edge in the set).



A difficult problem: integer factorization

Integer factorization (search). Given an integer x , find a nontrivial factor.  *or report that no such factor exists*

neither 1 nor x

Ex.

147573952589676412927

a FACTOR instance

193707721

a factor

1350664108659952233496032162788059699388814756056670
2752448514385152651060485953383394028715057190944179
8207282164471551373680419703964191743046496589274256
2393410208643832021103729587257623585096431105640735
0150818751067659462920556368552947521350085287941637
7328533906109750544334999811150056977236890927563

Core application area. Cryptography.

Brute-force search. Try all possible divisors between 2 and \sqrt{x} .

a very challenging FACTOR instance
(factor to earn an A+ in COS 226)

Q. Can we do anything substantially more clever?

*if there's a nontrivial factor larger
than \sqrt{x} , there is one smaller than \sqrt{x}*

Another difficult problem: boolean satisfiability

Boolean satisfiability (search). Given a system of boolean equations, find a satisfying truth assignment.

Ex.

$$\begin{aligned} \neg x_1 \text{ or } x_2 \text{ or } x_3 &= \text{true} \\ x_1 \text{ or } \neg x_2 \text{ or } x_3 &= \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } \neg x_3 &= \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } &\text{or } x_4 = \text{true} \\ &\neg x_2 \text{ or } x_3 \text{ or } x_4 = \text{true} \end{aligned}$$

a SAT instance

CNF, conjunctive normal form (AND of ORs)

or report that no such assignment is possible

$$\begin{aligned} x_1 &= \text{false} \\ x_2 &= \text{false} \\ x_3 &= \text{true} \\ x_4 &= \text{true} \end{aligned}$$

a satisfying truth assignment

Applications.

- Automatic verification systems for software.
- Mean field diluted spin glass model in physics.
- Electronic design automation (EDA) for hardware.
- ...

Another difficult problem: boolean satisfiability

Boolean satisfiability (search). Given a system of boolean equations, find a satisfying truth assignment.

Ex.

$$\begin{array}{l} \neg x_1 \text{ or } x_2 \text{ or } x_3 = \text{true} \\ x_1 \text{ or } \neg x_2 \text{ or } x_3 = \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } \neg x_3 = \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } \text{ or } x_4 = \text{true} \\ \neg x_2 \text{ or } x_3 \text{ or } x_4 = \text{true} \end{array}$$

a SAT instance



needle in a haystack

Brute-force search. Try all 2^n possible truth assignments, where $n = \#$ variables.

Q. Can we do anything substantially more clever?

A. Probably no. [stay tuned]

How difficult can it be?

Imagine a galactic computer...

- With as many processors as electrons in the universe.
- Each processor having the power of today's supercomputers.
- Each processor working for the lifetime of the universe.

quantity	estimate
<i>electrons in universe</i>	10^{79}
<i>instructions per second</i>	10^{13}
<i>age of universe in seconds</i>	10^{17}



Q. Could galactic computer solve satisfiability instance with 1,000 variables using brute-force search?

A. Not even close: $2^{1000} > 10^{300} \gg 10^{79} \cdot 10^{13} \cdot 10^{17} = 10^{109}$.

Lesson. Exponential growth dwarfs technological change.

Polynomial time

Q2. What is an **efficient algorithm**?

A2. Algorithm whose running time is at most polynomial in input size n .

$n = \# \text{ of bits in input}$



Polynomial time. Number of *elementary operations* is at most an^b

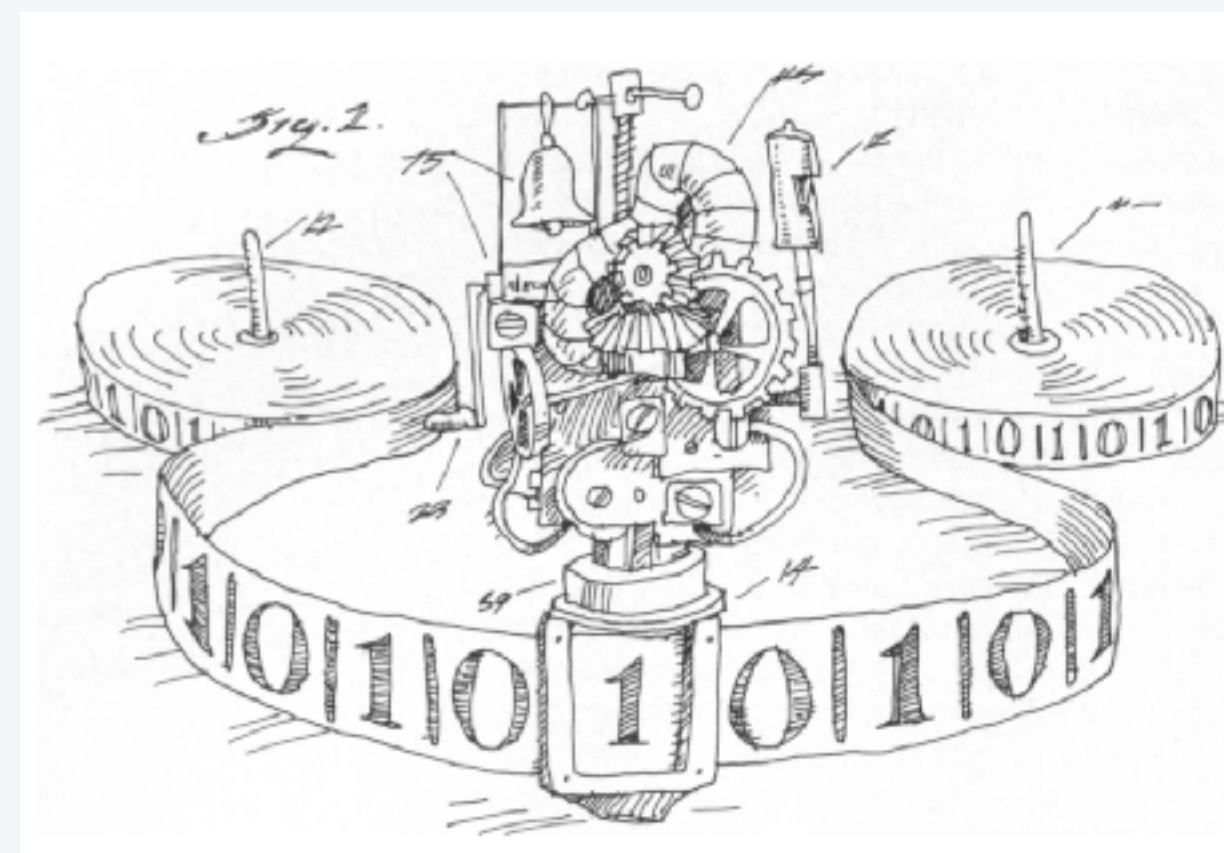
for some constant a and b . ← *must hold for all inputs of size n*

Q1. What is an **algorithm**?

A1. A **Turing Machine**! Equivalently, a program in Java/Python/C++/...

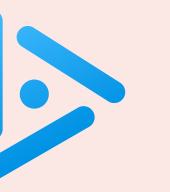


the extended Church-Turing thesis



A Turing machine

order	emoji	name	today
$\Theta(1)$	😍	<i>constant</i>	😊
$\Theta(\log n)$	😎	<i>logarithmic</i>	😊
$\Theta(n)$	😁	<i>linear</i>	😊
$\Theta(n \log n)$	😁	<i>linearithmic</i>	😊
$\Theta(n^2)$	😞	<i>quadratic</i>	😊
$\Theta(n^3)$	😞	<i>cubic</i>	😊
$\Theta(n^{\log n})$	😬	<i>quasipolynomial</i>	😡
$\Theta(1.1^n)$	😭	<i>exponential</i>	😡
$\Theta(2^n)$	😡	<i>exponential</i>	😡
$\Theta(n!)$	😡	<i>factorial</i>	😡



Which of the following are poly-time algorithms?

- A. Brute-force search for boolean satisfiability.
- B. Brute-force search for integer factorization.
- C. Both A and B.
- D. Neither A nor B.

Intractable problems

Q3. Which **problems** can be solved efficiently?

A3. Those for which poly-time algorithms exist.

Why do we define poly-time as efficient?

Def. A problem is **intractable** if no poly-time algorithm solves it.

Q4. How can we prove that a problem is intractable?

A4. Generally no easy way. Focus of today's lecture!

Often times, efficient algorithms require deep math insights.

*require math
insights*

tractable

intractable?

primality

integer factorization

shortest path

longest path

min cut

max cut

Euler cycle

Hamiltonian cycle

2-SAT

3-SAT

⋮

⋮

*3 boolean variables
per equation*

Intractable problems





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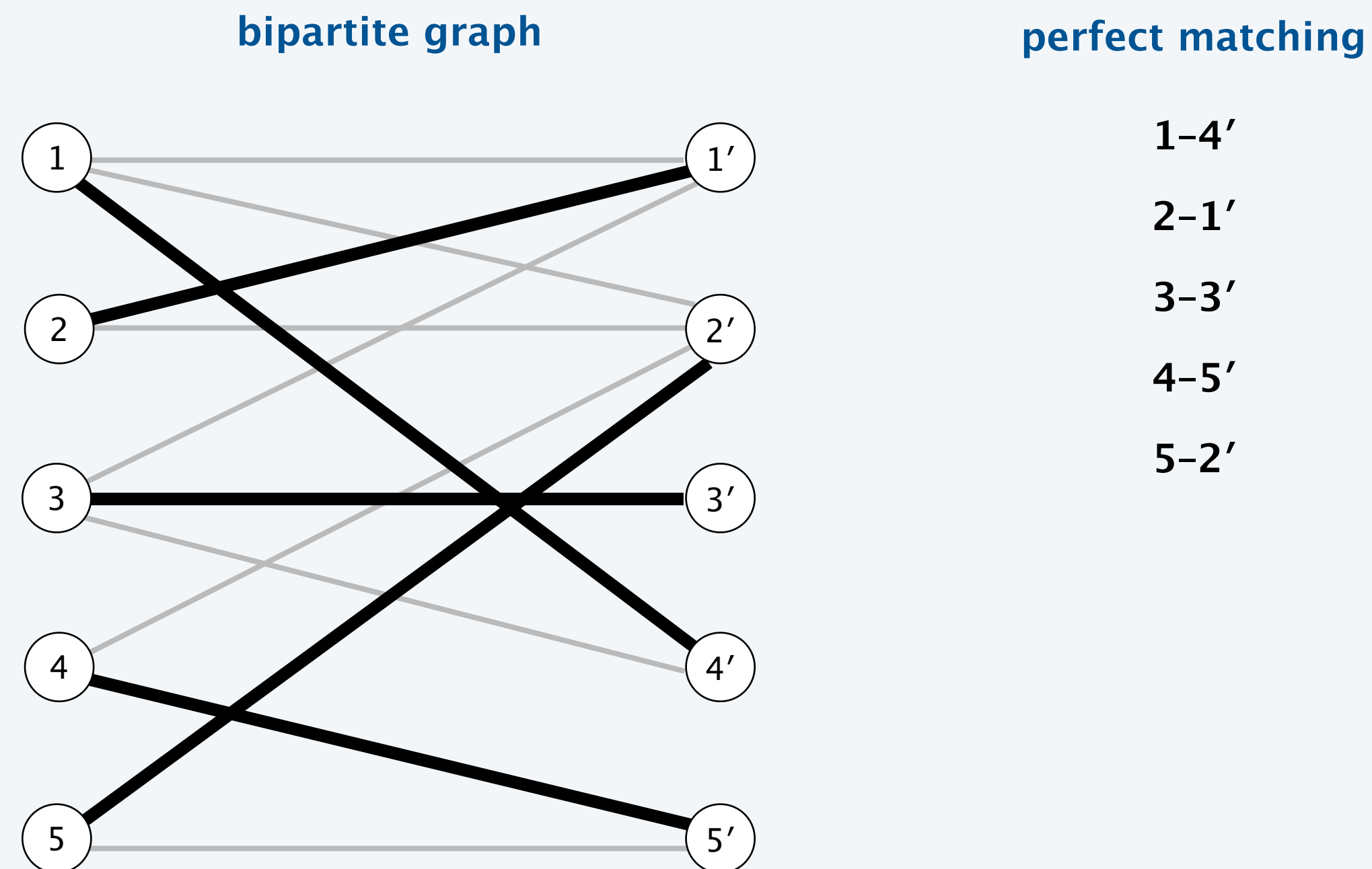
The P complexity class

A **decision problem** is Boolean function (given an input answer YES/NO).

Def. **P** is the set of all decision problems that can be **solved in poly-time**.

Ex 1 – Perfect matching (decision): Given a **bipartite graph**, *is there* a **perfect matching**?

trick – max flow!



Are all “interesting” problems in **P**? Maybe there is always a clever trick for solving...

The NP complexity class

Def. NP is the set of all decision problems for which you can **verify** a **YES** answer in poly-time given a “**witness**” (a.k.a “proof”, “certificate”). ← **NP** = *Nondeterministic Poly-time*

Ex 1 – boolean satisfiability (*decision*): Given a system of m boolean equations in n variables, *is there* an assignment that satisfies all equations?

$$\begin{array}{l} \neg x_1 \quad \text{or} \quad x_2 \quad \text{or} \quad x_3 \quad = \quad \text{true} \\ x_1 \quad \text{or} \quad \neg x_2 \quad \text{or} \quad x_3 \quad = \quad \text{true} \\ \neg x_1 \quad \text{or} \quad \neg x_2 \quad \text{or} \quad \neg x_3 \quad = \quad \text{true} \\ \neg x_1 \quad \text{or} \quad \neg x_2 \quad \text{or} \quad \quad \quad \text{or} \quad x_4 \quad = \quad \text{true} \\ \quad \quad \quad \neg x_2 \quad \text{or} \quad x_3 \quad \text{or} \quad x_4 \quad = \quad \text{true} \end{array}$$

A SAT instance

$$\begin{array}{l} x_1 = \text{false} \\ x_2 = \text{false} \\ x_3 = \text{true} \\ x_4 = \text{true} \end{array}$$

witness

Witness. A satisfying assignment.

Poly-time verification algorithm. Plug values of assignment into the equations and check. ← *verify in $O(mn)$ time*

The NP complexity class

Def. **NP** is the set of all decision problems for which you can **verify** a **YES** answer in poly-time given a “**witness**” (a.k.a “proof”, “certificate”).

Ex 2 – integer factorization (*decision*): Given two integers x and k , does x have a nontrivial factor greater than k ?

$x = 147573952589676412927$

$k = 100,000,000$

193,707,721

a FACTOR instance

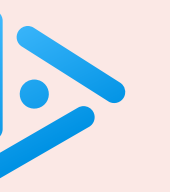
witness

Witness. A nontrivial factor of x greater than k .

Poly-time verification algorithm. Check that the witness is greater than k and that it's a divisor of x . $\longleftarrow O(n^2)$ time via long division

Note: For a problem to be in **NP**, it suffices to verify a *purported* witness for a **YES** answer.

- Doesn't need to *find* the witness (e.g., a candidate factor is given).
- Doesn't need verify a **NO** answer (e.g., no factor greater than k).



Which decision version of longest path is in NP?

- A. Given a graph G and an integer k , is the longest simple path in G of length *at most* k edges.
- B. Given a graph G and an integer k , is the longest simple path in G of length *at least* k edges.
- C. Both A and B.
- D. Neither A nor B.

P vs. NP

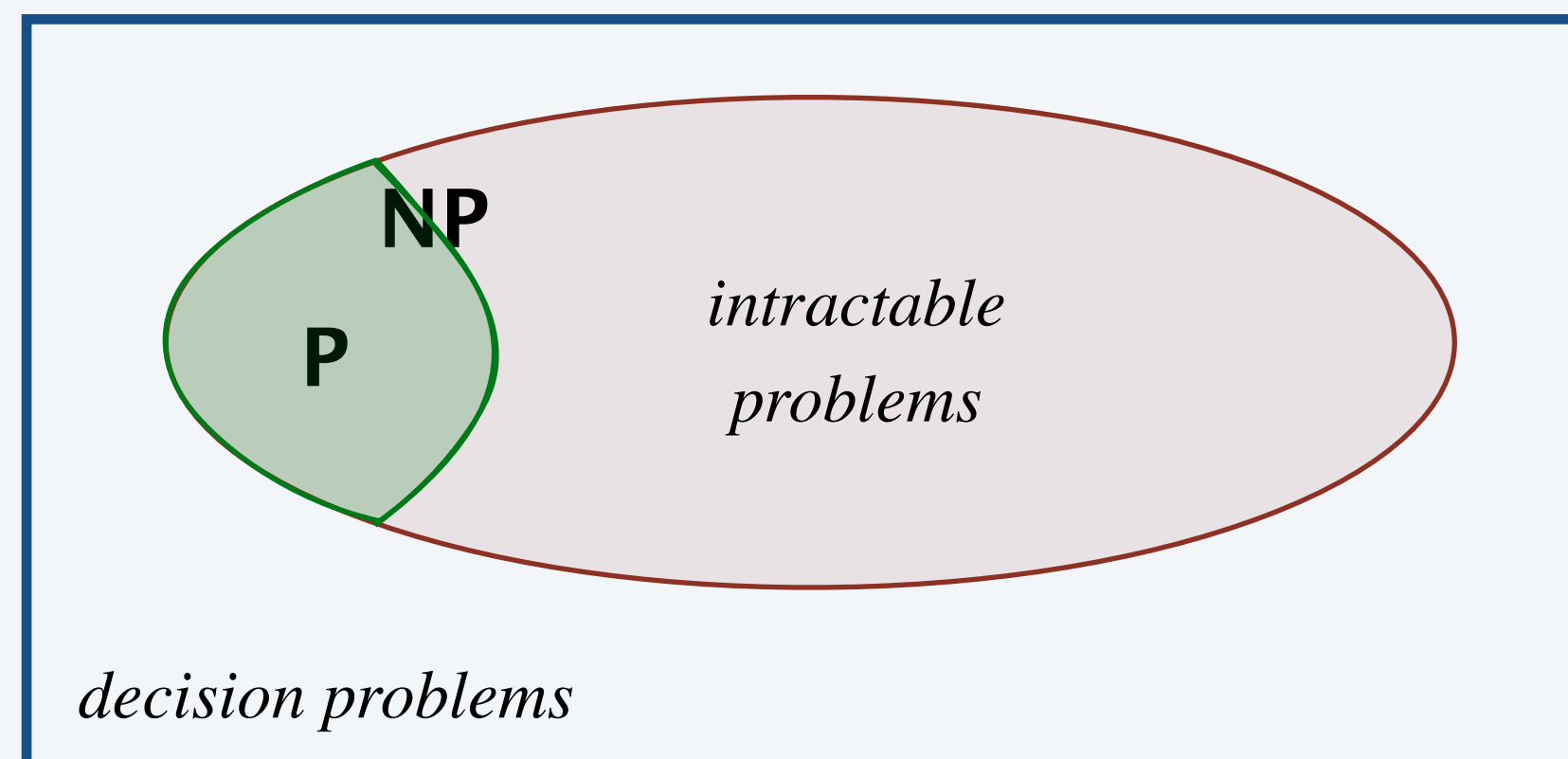
P = set of problems whose solution can be *computed* efficiently (in poly-time).

NP = set of problems whose solution can be *verified* efficiently (in poly-time).

Observation. **NP** contains **P** ← *any string serves as witness*
← *e.g., perfect matching is in NP*

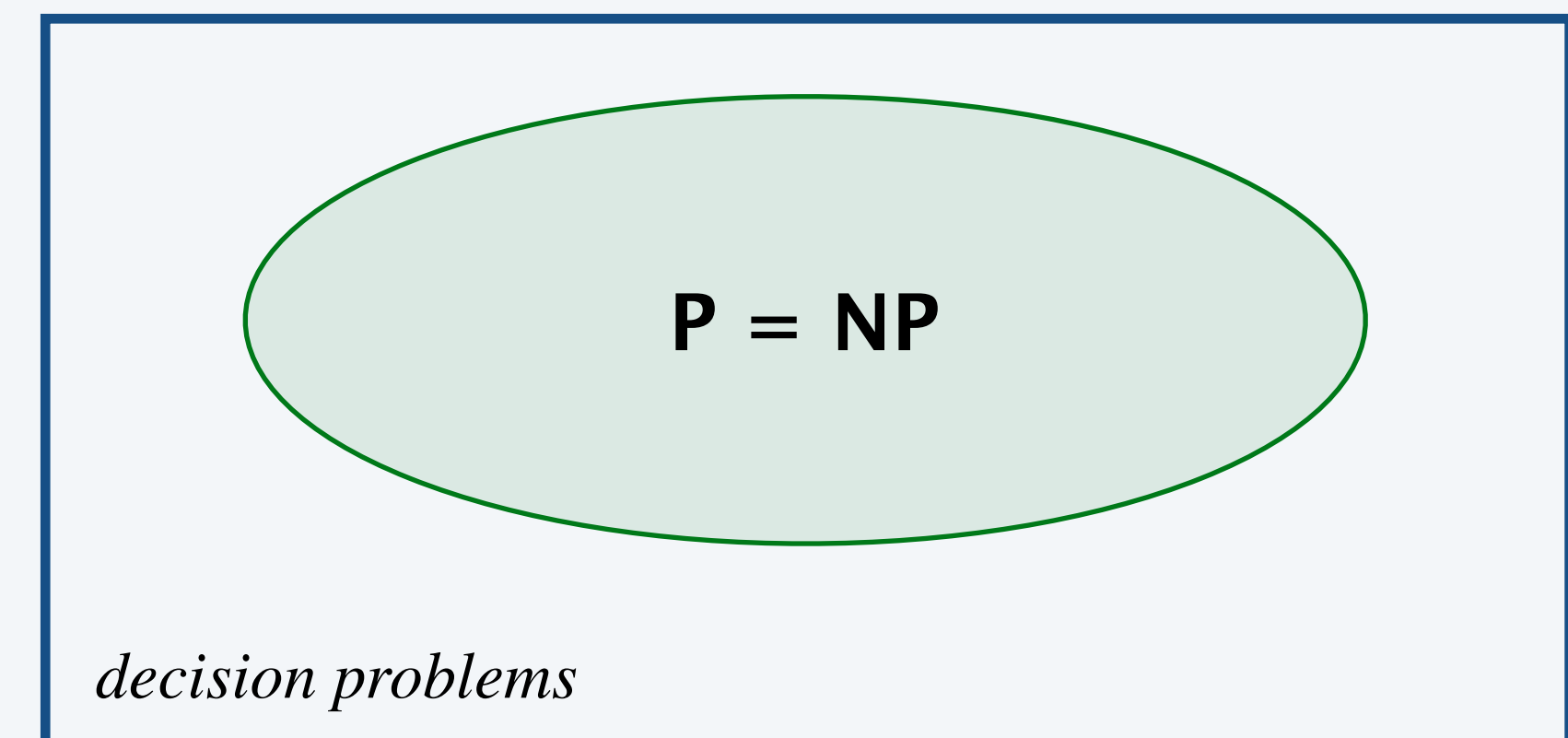
THE question. Does **P = NP**? ← *\$1M*

Two possible worlds.



P ≠ NP

*brute-force search may be
the best we can do*



P = NP

*poly-time algorithms for
FACTOR, SAT, LONGEST-PATH, ...*

Conjecture. **P ≠ NP**. ← *why?*

Why is P vs NP so central?

P vs NP is central in math, science, technology and beyond.

NP models many intellectual challenges humanity faces: *Why would you attempt to solve a problem if you cannot even tell if a solution is good?*



creative genius

domain	problem	witness/solution
<i>mathematics</i>	is a conjecture correct?	mathematical proof
<i>engineering</i>	given constraints (size, weight, energy), find a design (bridge, medicine, computer)	blueprint
<i>science</i>	given data on a phenomenon, find a theory explaining it	a scientific theory
<i>the arts</i>	write a beautiful poem / novel / pop song, draw a beautiful picture	a poem, novel, pop song, drawing



ordinary appreciation

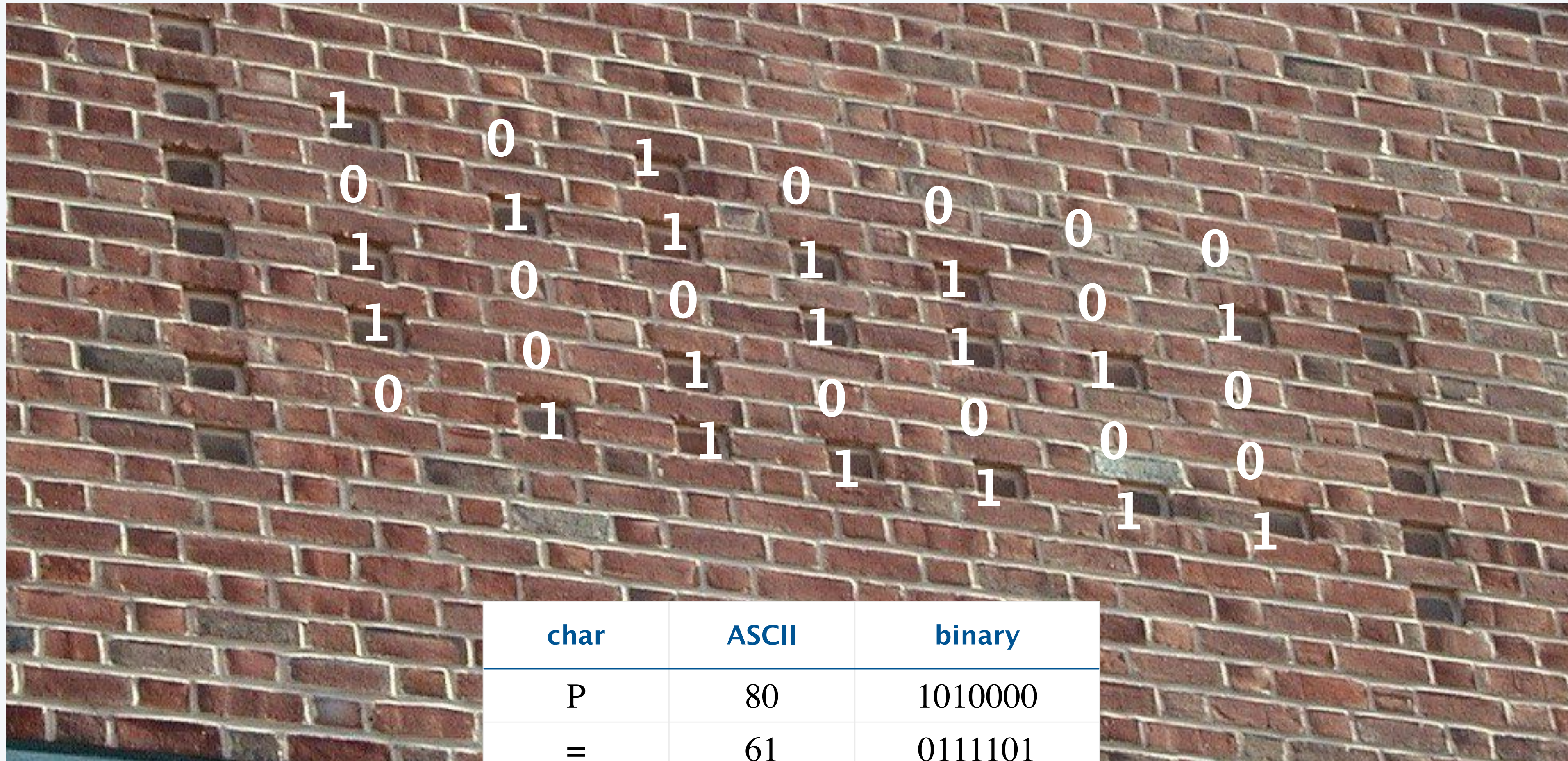
Verifying a solution seems like it should be way easier than finding it! This suggests **P ≠ NP**.

Analogy for P vs NP. Creative genius vs. ordinary appreciation of creativity.

Princeton computer science building



Princeton computer science building (closeup)



char	ASCII	binary
P	80	1010000
=	61	0111101
N	78	1001110
P	80	1010000
?	63	0111111



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Bird's-eye view

Goal. Classify **problems** according to computational requirements.

Goal'. Suppose we could (not) solve problem X efficiently.

What else could we (not) solve efficiently?

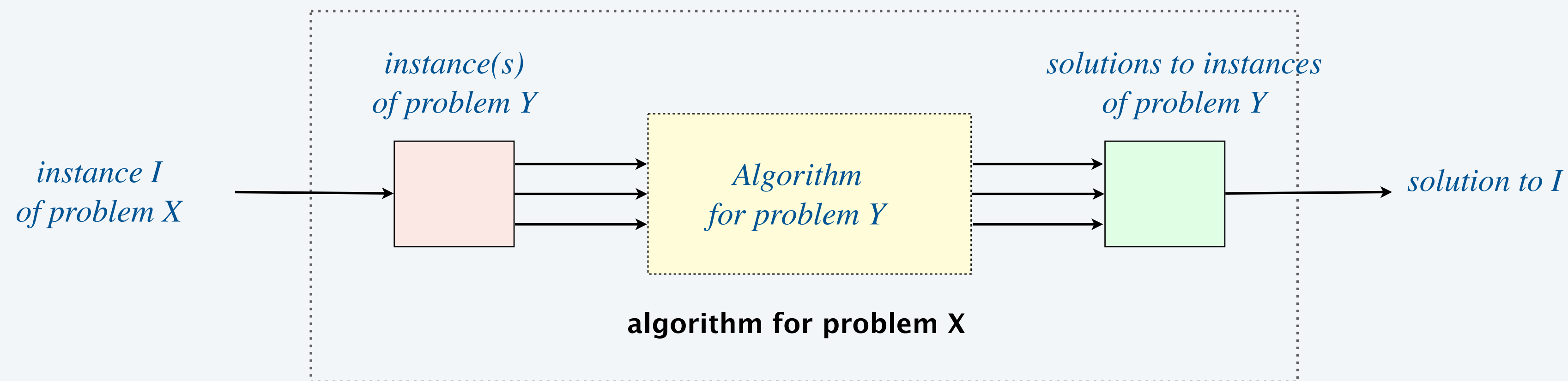
“ Give me a lever long enough and a fulcrum on which to place it, and I shall move the world.” — Archimedes



Poly-time reduction

Def. Problem X **poly-time reduces to** problem Y if X can be solved with: $\longleftarrow X \leq Y$

- Polynomial number of elementary operations.
- Polynomial number of calls to Y . \longleftarrow Cook reduction



Ex 1. MEDIAN poly-time reduces to SORT.

Ex 2. BIPARTITE-MATCHING poly-time reduces to MAX-FLOW.

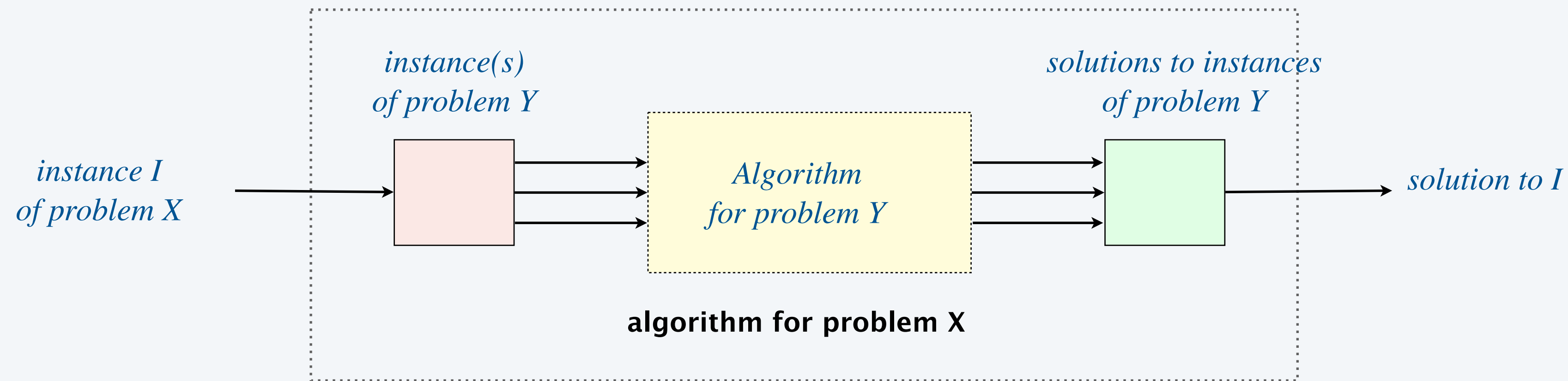
Design algorithms. If X poly-time reduces to Y and Y can be solved efficiently, then X can be solved efficiently.

Establish intractability. If SAT is intractable and SAT poly-time reduces to Y , then Y is intractable.

Poly-time reduction

Def. Problem X **poly-time reduces to** problem Y if X can be solved with:

- Polynomial number of elementary operations.
- Polynomial number of calls to Y .



Common mistake. Confuse X poly-time reduces to Y with Y poly-time reduces to Y .



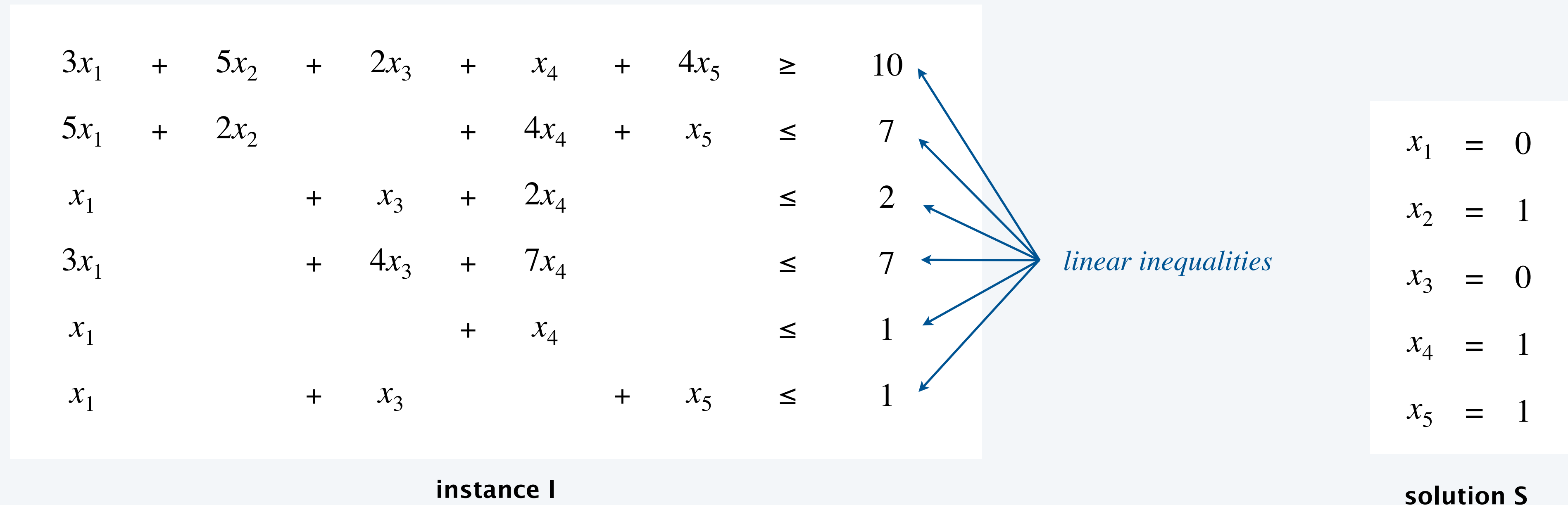
X reduces to SAT: X is no harder than SAT (A solution to SAT implies a solution to X)

SAT reduces to X : X is no easier than SAT (A solution to X implies a solution to SAT)

“up to polynomials”

Integer linear programming

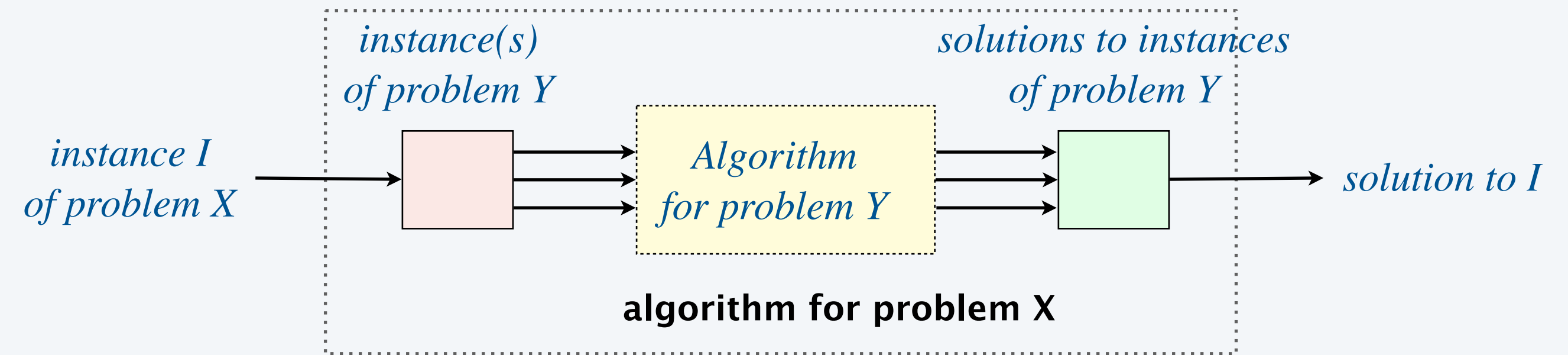
ILP. Given a system of linear inequalities, is there a solution where all variables take **integer** values?



Context. Cornerstone problem in operations research.

Remark. Finding a **real-valued** solution can be solved in poly-time (linear programming).

SAT poly-time reduces to ILP

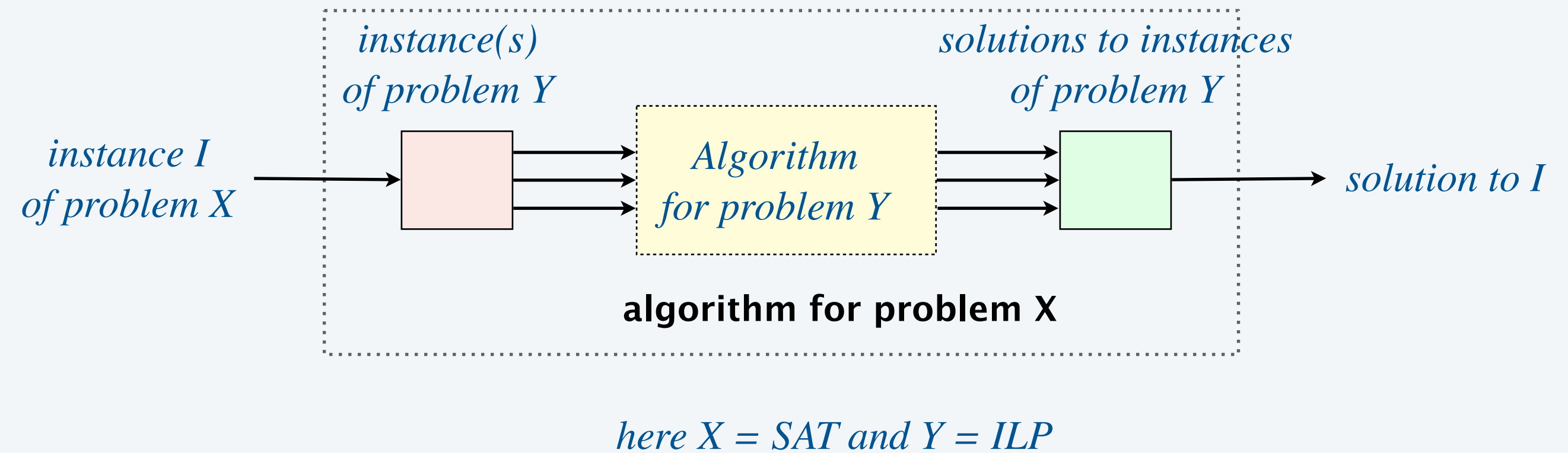


here $X = SAT$ and $Y = ILP$

SAT poly-time reduces to ILP

SAT. Given a system of m boolean equations in n variables, is there an assignment that satisfies all equations?

$$\begin{aligned} \neg x_1 \text{ or } x_2 \text{ or } x_3 &= \text{true} \\ x_1 \text{ or } \neg x_2 \text{ or } x_3 &= \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } \neg x_3 &= \text{true} \\ \neg x_1 \text{ or } \neg x_2 \text{ or } &\text{or } x_4 = \text{true} \\ &\neg x_2 \text{ or } x_3 \text{ or } x_4 = \text{true} \end{aligned}$$



ILP. Given a system of linear inequalities, is there an assignment where all variables take **integer** values?

$$\begin{aligned} (1 - y_1) + y_2 + y_3 &\geq 1 & 0 \leq y_1 \leq 1 \\ y_1 + (1 - y_2) + y_3 &\geq 1 & 0 \leq y_2 \leq 1 \\ (1 - y_1) + (1 - y_2) + (1 - y_3) + y_4 &\geq 1 & 0 \leq y_3 \leq 1 \\ (1 - y_1) + (1 - y_2) + &+ y_4 \geq 1 & 0 \leq y_4 \leq 1 \\ &(1 - y_2) + y_3 + y_4 \geq 1 & \end{aligned}$$

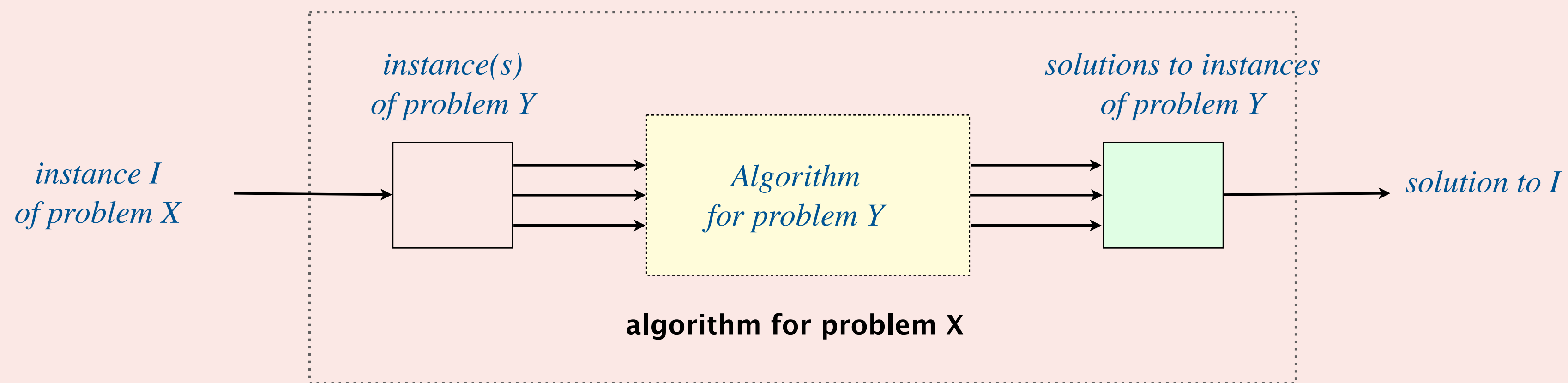
$y_i = 0 \Rightarrow x_i = \text{false}$
 $y_i = 1 \Rightarrow x_i = \text{true}$



Suppose that Problem X poly-time reduces to Problem Y .

Which of the following can we infer?

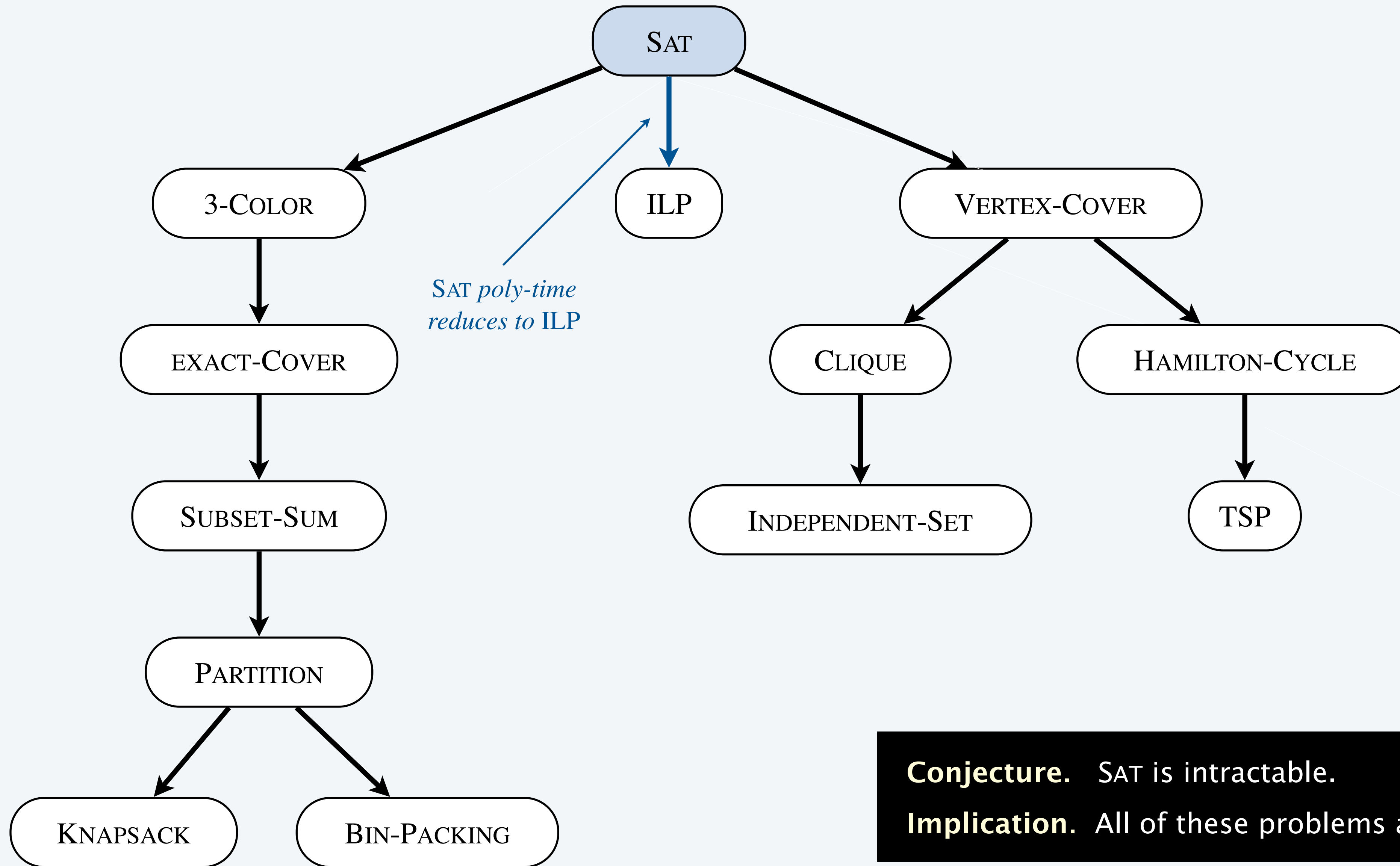
- A. If X can be solved in poly-time, then so can Y .
- B. If Y can be solved in $\Theta(n^3)$ time, then X can be solved in $\Theta(n^3)$ time.
- C. If Y can be solved in $\Theta(n^3)$ time, then X can be solved in poly-time.
- D. If X cannot be solved in $\Theta(n^3)$ time, then Y cannot be solved in poly-time.
- E. If Y cannot be solved in poly-time, then neither can X .



More poly-time reductions from SAT



Richard Karp
(1972)



Conjecture. SAT is intractable.

Implication. All of these problems are intractable.



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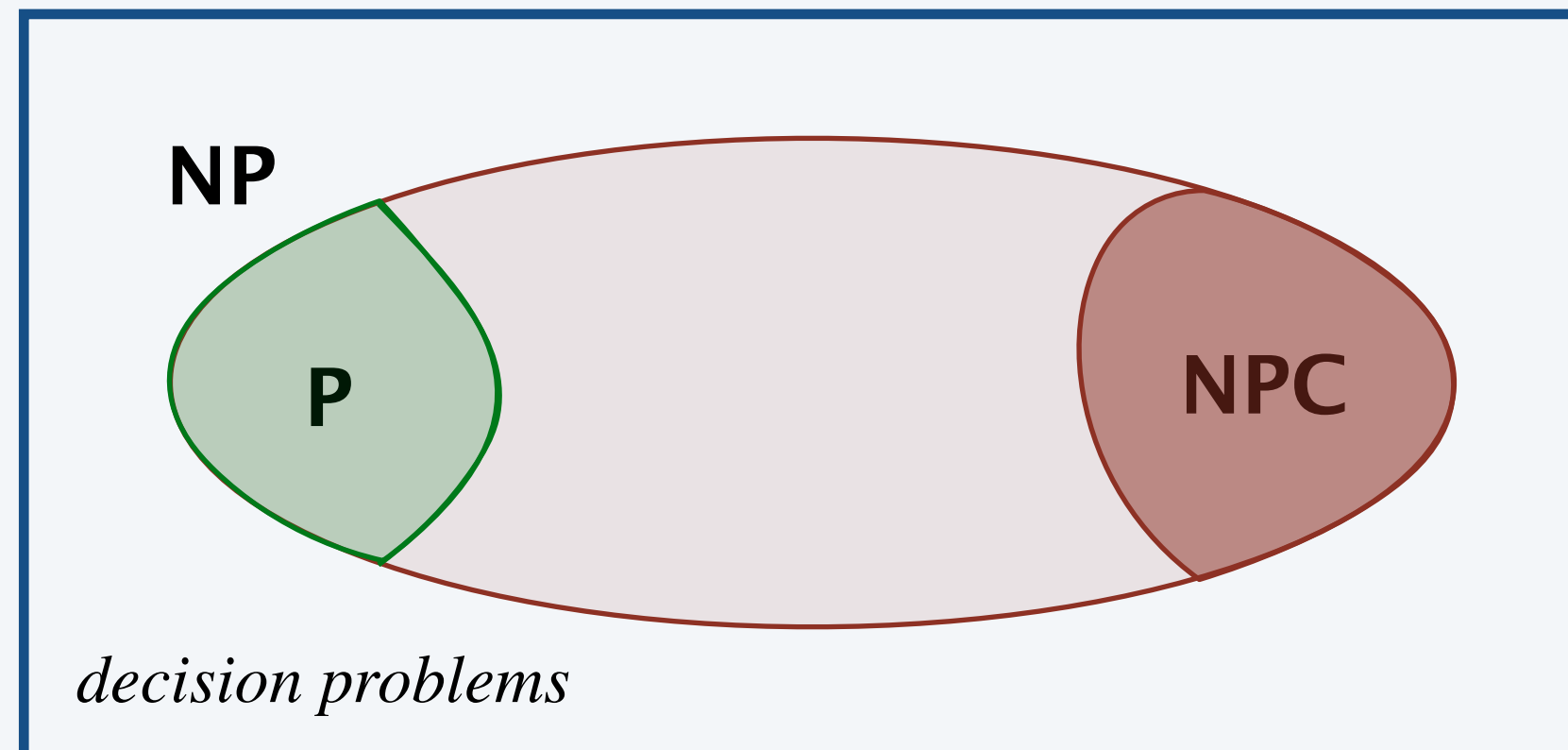
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NP-completeness

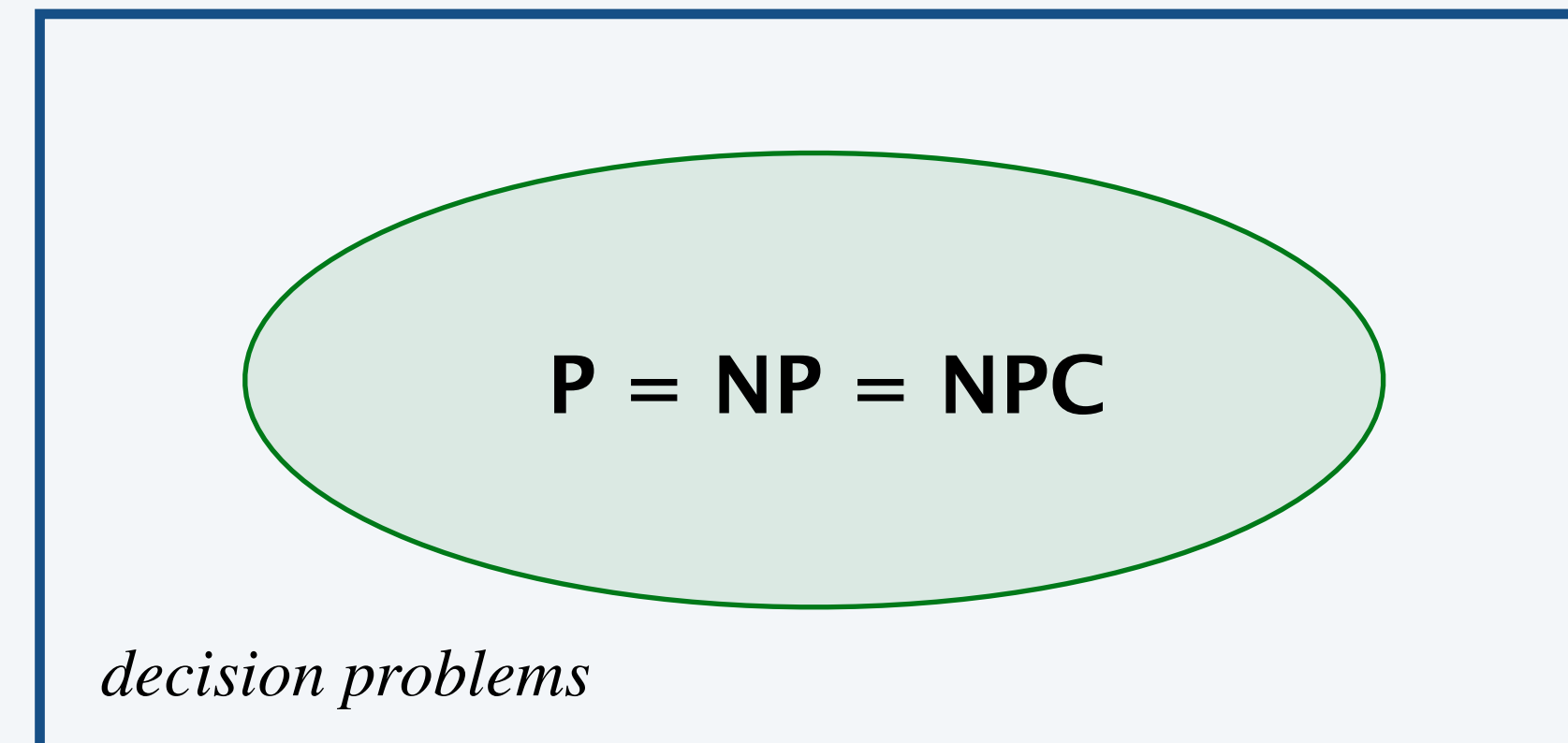
Def. A decision problem is **NP-complete** if

- It is in **NP**.
- All problems in **NP** poly-time to reduce to it. ← *intuitively, the “hardest problems” in NP*

Two worlds.



P ≠ NP



P = NP



Suppose that X is NP-complete. What can you infer?

- I. X is in NP.
- II. If X can be solved in poly-time, then $P = NP$.
- III. If X cannot be solved in poly-time, then $P \neq NP$.

- A. I only.
- B. II only.
- C. I and II only.
- D. I, II, and III.

Cook–Levin theorem

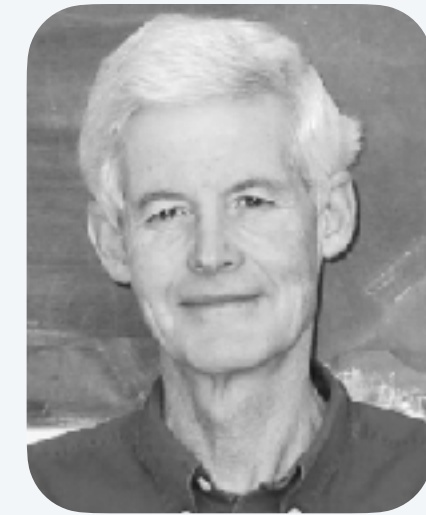
Cook–Levin theorem. SAT is **NP**-complete.

Pioneering result in computer science!

Corollary. SAT can be solved in poly-time if and only if **P = NP**.

Impact. To prove that a new problem Y is **NP**-complete, suffices to show that:

- Y is in **NP**.
- SAT poly-time reduces to Y .



Stephen Cook
(1971)

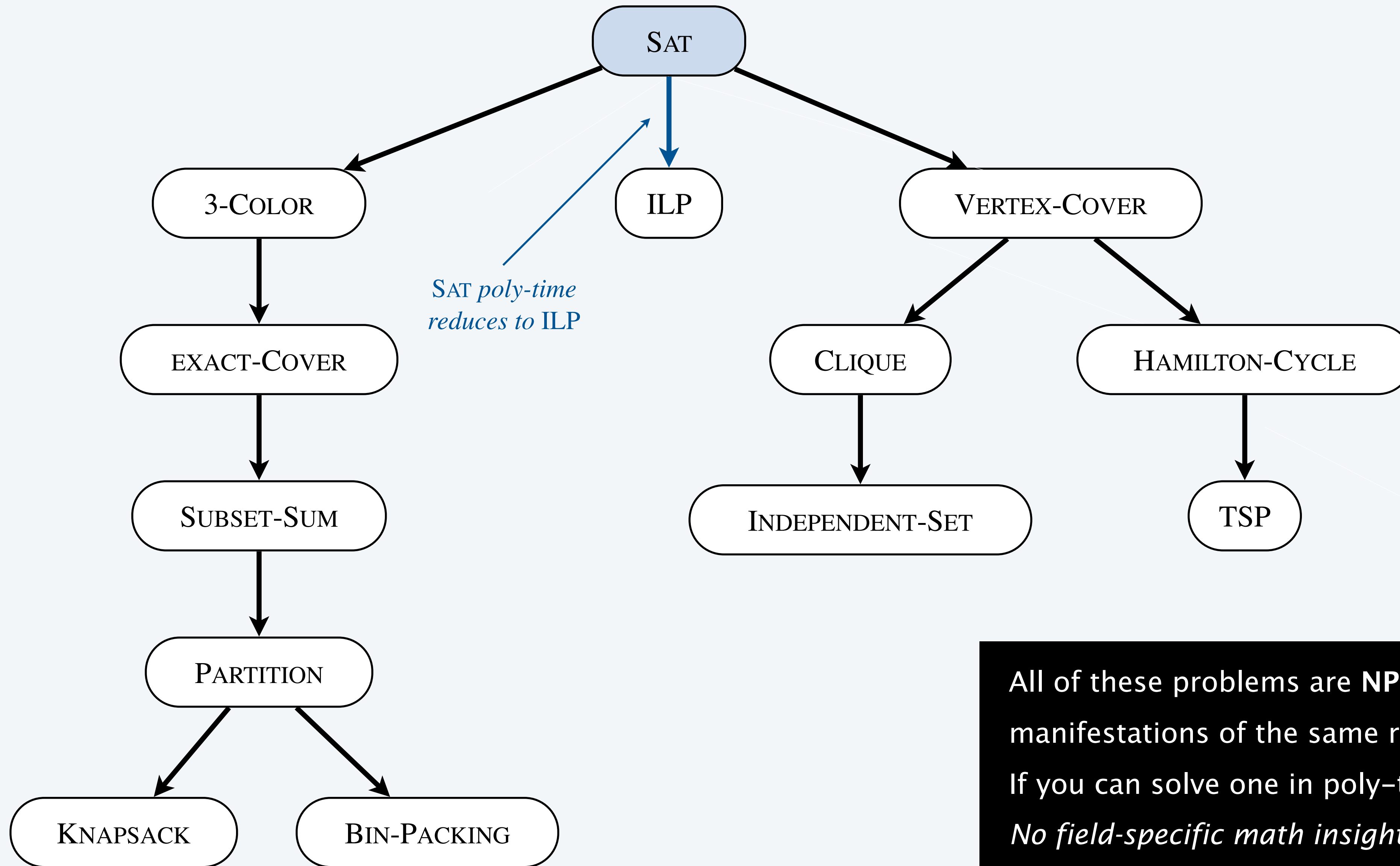


Leonid Levin
(1971)

Implications of Karp + Cook-Levin



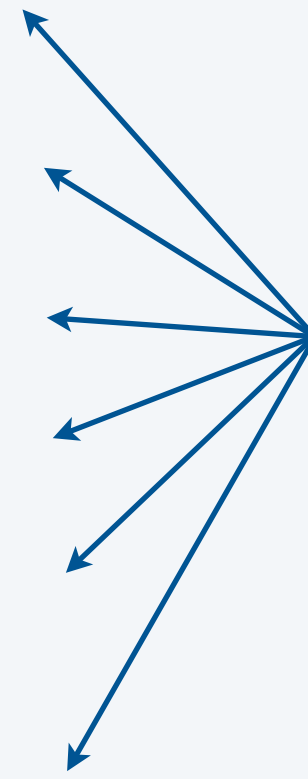
Richard Karp
(1972)



All of these problems are NP-complete; they are manifestations of the same really hard problem. If you can solve one in poly-time, you can solve all! *No field-specific math insights are needed!*

More NP-complete problems

field of study	NP-complete problem
Aerospace engineering	<i>optimal mesh partitioning for finite elements</i>
Biology	<i>phylogeny reconstruction</i>
Chemical engineering	<i>heat exchanger network synthesis</i>
Chemistry	<i>protein folding</i>
Civil engineering	<i>equilibrium of urban traffic flow</i>
Economics	<i>computation of arbitrage in financial markets with friction</i>
Electrical engineering	<i>VLSI layout</i>
Environmental engineering	<i>optimal placement of contaminant sensors</i>
Financial engineering	<i>minimum risk portfolio of given return</i>
Game theory	<i>Nash equilibrium that maximizes social welfare</i>
Mechanical engineering	<i>structure of turbulence in sheared flows</i>
Medicine	<i>reconstructing 3d shape from biplane angiocardialogram</i>
Operations research	<i>traveling salesperson problem, integer programming</i>
Physics	<i>partition function of 3d Ising model</i>
Politics	<i>Shapley–Shubik voting power</i>
Pop culture	<i>versions of Sudoku, Checkers, Minesweeper, Tetris</i>
Statistics	<i>optimal experimental design</i>



6,000+ scientific papers per year.



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Dealing with intractability



Identifying intractable problems

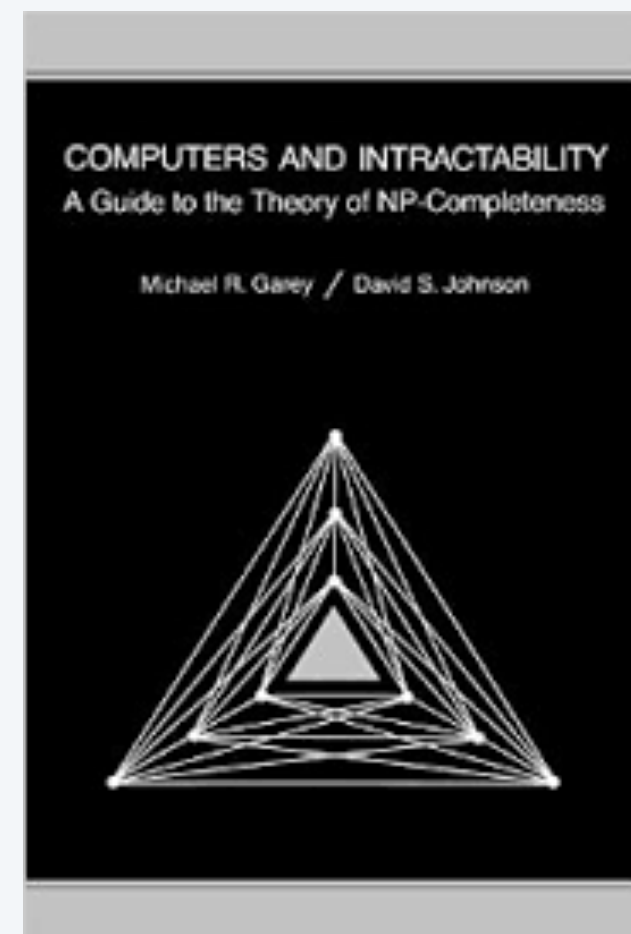
Establishing **NP**-completeness through poly-time reduction is an important tool in guiding algorithm design efforts.

Q4'. How to convince yourself that a problem is (probably) intractable?

A. [hard way] Long futile search for a poly-time algorithm (as for **SAT**).

A. [easy way] Poly-time reduction from **SAT** (or any other **NP**-complete problem).

Caveat. Intricate reductions are common.



Approaches to dealing with intractability

Q. What to do when you identify an NP-complete problem?

A. Safe to assume it is intractable: no *worst-case* poly-time algorithm for *all* problem instances.

Q1. Must your algorithm *always* run fast?

Solve real-world instances. Backtracking, TSP, SAT.

Q2. Do you need the *right/best* solution or a *good* solution?

Approximation algorithms. Look for suboptimal solutions.

Q3. Can you use the problem's hardness in your favor?

Leverage intractability. Cryptography.

Dealing with intractability: find solutions to real-world problem instances

Observations.

- Worst-case inputs may not occur for practical problems.
- Instances that do occur in practice may be easier to solve.
- Reasonable approach: relax the condition of guaranteed poly-time.

Boolean satisfiability.

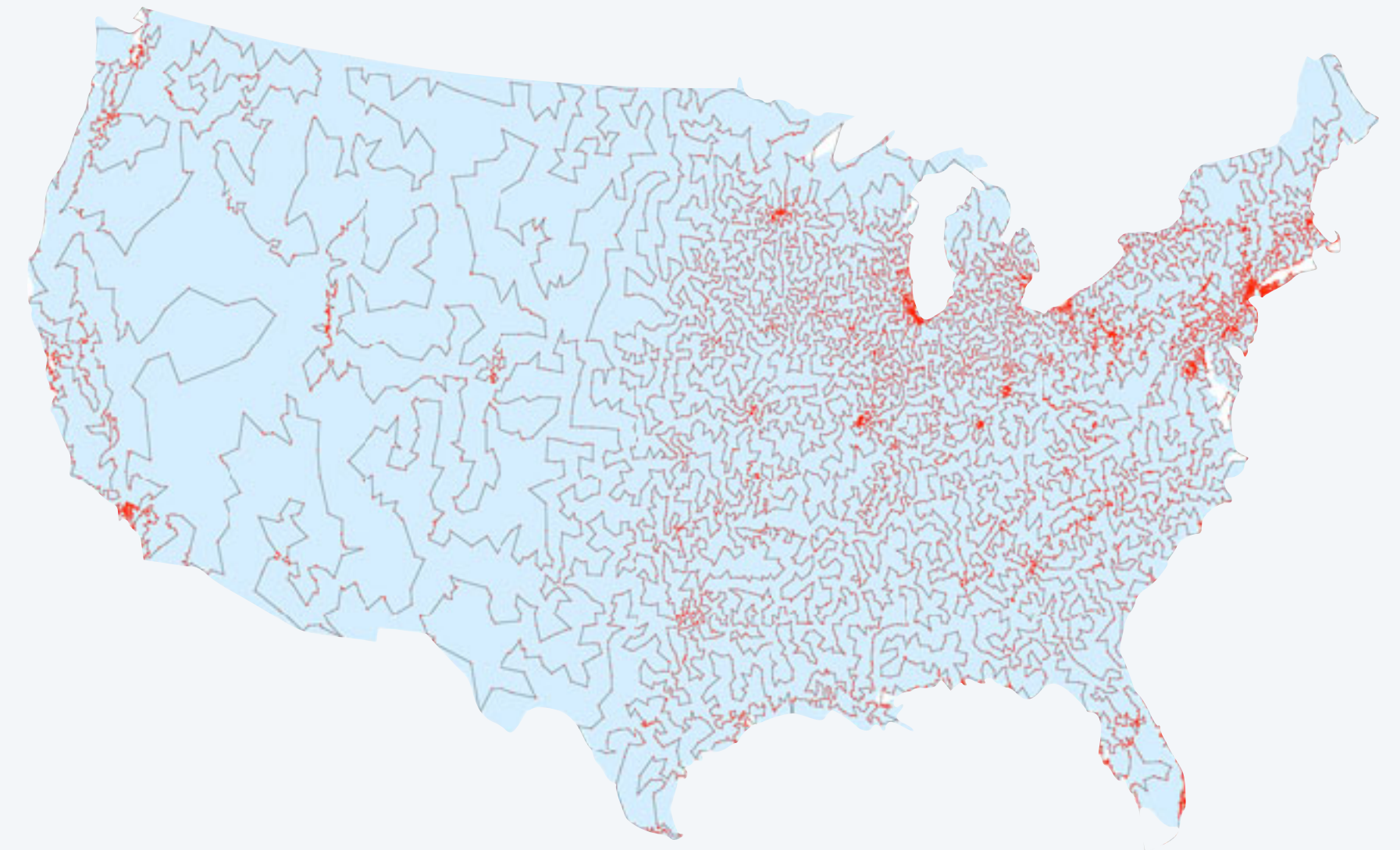
- Chaff solves real-world instances with 10,000+ variables.
- Princeton senior independent work (!) in 2000.

Traveling salesperson problem.

- Concorde routinely solves large real-world instances.
- 85,900-city instance solved in 2006.

Integer linear programming.

- CPLEX routinely solves large real-world instances.
- Routinely used in scientific and commercial applications.

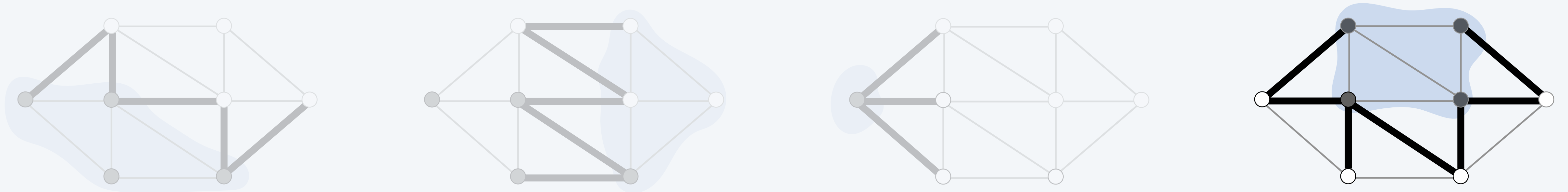


TSP solution for 13,509 US cities

Dealing with intractability: approximation algorithms

MAX-CUT (search): given a graph G , find the cut with maximum number M of crossing edges.

Approximate version: find a large cut.



Algorithm: take a uniformly random cut.

Expected size is $E/2$; random assignment size is $\geq E/2 \geq M/2$ with at least 50% probability.

can improve to $.878M$

Dealing with intractability: approximation algorithms

3-SAT (search): given 3-variable equations on n boolean variables, find satisfying truth assignment.

Approximate version: find assignment that satisfies many equations.

Algorithm: take a uniformly random assignment.

Expected fraction of satisfied equations is $7/8$.

can't be improved (unless $P = NP$)



Remark. Some problems have approximation algorithms with arbitrary precision.

For others, finding better approximations is also **NP**-complete!

Leveraging intractability: RSA cryptosystem

Modern cryptography applications.

- Secure a secret communication.
- Append a digital signature.
- Credit card transactions.
- ...



RSA cryptosystem exploits intractability.

- To use: multiply/divide two n -digit integers (easy).
- To break: factor a $2n$ -digit integer (intractable?).



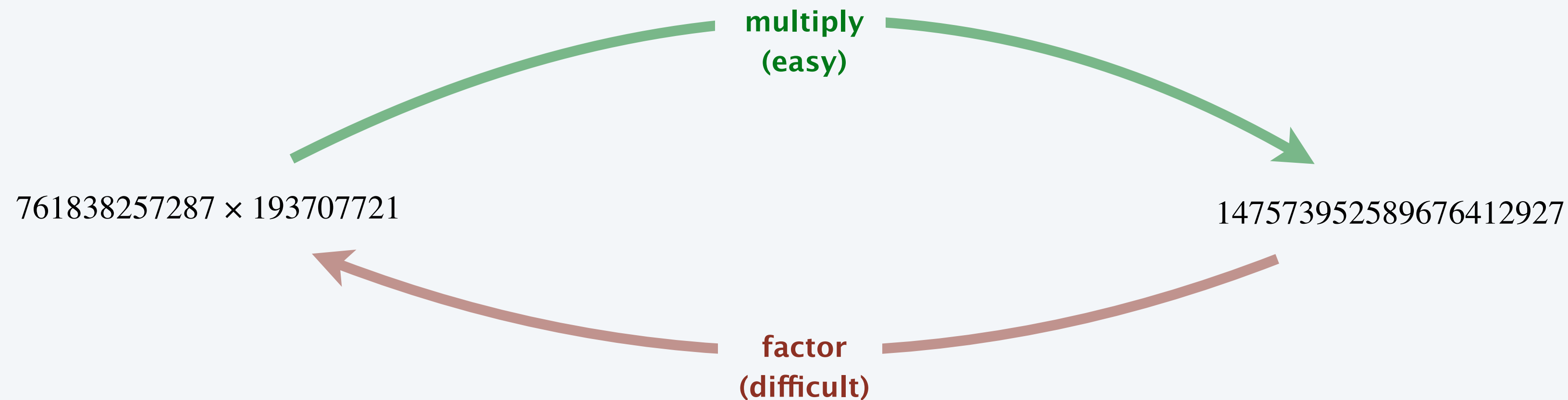
Ron Rivest



Adi Shamir



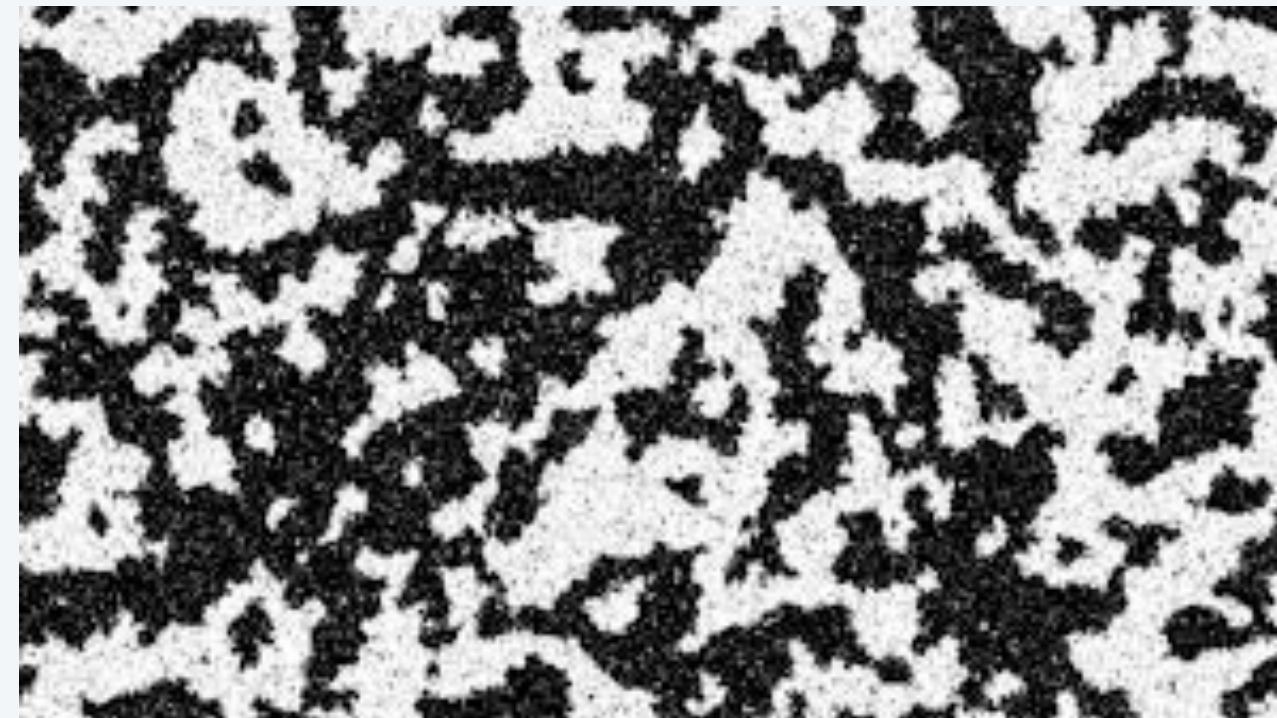
Len Adelman



Leveraging intractability: guiding scientific inquiry

- 1926. Ising introduces a mathematical model for ferromagnetism.
- 1930s. Closed form solution is a holy grail of statistical mechanics.
- 1944. Onsager finds closed form solution to 2D version in tour de force.
- 1950s. Feynman (and others) seek closed form solution to 3D version.
- 2000. Istrail shows that ISING-3D is **NP**-complete.

Bottom line. Search for a closed formula seems futile.



Summary

P. Set of problems solvable in poly-time.

NP. Set of problems checkable in poly-time.

NP-complete. “Hardest” problems in **NP**. ← SAT, LONGEST-PATH, ILP, TSP, ...

Use theory as a guide

- You will confront **NP**-complete problems in your career.
- An poly-time algorithm for an **NP**-complete problem would be a stunning scientific breakthrough (a proof that **P** = **NP**).
- It is safe to assume that **P** ≠ **NP** and that such problems are intractable.
- Identify these situations and proceed accordingly.



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A final thought

“ Now my general conjecture is as follows: for almost all sufficiently complex types of enciphering, [...] the mean key computation length increases exponentially with the length of the key [...].

*The nature of this conjecture is such that I cannot prove it [...].
Nor do I expect it to be proven. ”*

— John Nash

