

# 6.5 REDUCTIONS

- introduction
- designing algorithms
- establishing lower bounds
- classifying problems
- intractability

# Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

http://algs4.cs.princeton.edu

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- intractability-

## Bird's-eye view

Goal. Classify problems according to computational requirements.

| complexity   | order of growth | examples  |
|--------------|-----------------|---|
| linear       | N               | min, max, median, Burrows-Wheeler transform,                |
| linearithmic | $N \log N$      | sorting, element distinctness, closest pair, Euclidean MST, |
| quadratic    | N <sup>2</sup>  | ?   |
| ÷            | <b>:</b>        | <b>:</b>  |
| exponential  | c <sup>N</sup>  | ?   |

Frustrating news. Huge number of problems have defied classification.

### Bird's-eye view

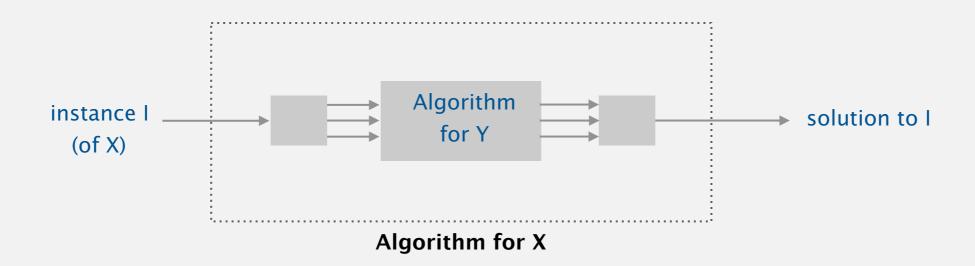
Goal. Classify problems according to computational requirements.

Goal 2. Suppose we could (could not) solve problem *X* efficiently. What else could (could not) we solve efficiently?



<sup>&</sup>quot;Give me a lever long enough and a fulcrum on which to place it, and I shall move the world." — Archimedes

Def. Problem *X* reduces to problem *Y* if you can use an algorithm that solves *Y* to help solve *X*.



Cost of solving X = total cost of solving Y + cost of reduction.perhaps many calls to Y

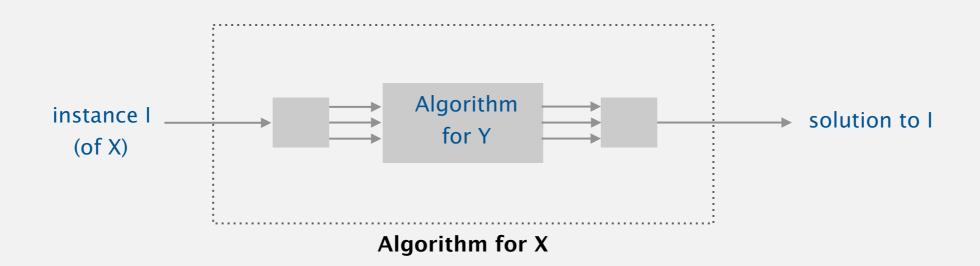
on problems of different sizes

(typically only 1 call)

preprocessing and postprocessing

(typically less than cost of solving Y)

Def. Problem *X* reduces to problem *Y* if you can use an algorithm that solves *Y* to help solve *X*.



Ex 1. [finding the median reduces to sorting]

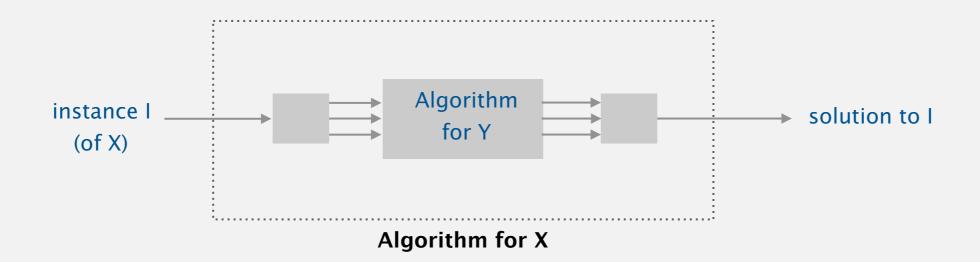
To find the median of *N* items:

- Sort *N* items.
- Return item in the middle.

cost of sorting cost of reduction cost of reduction

Cost of finding the median.  $N \log N + 1$ .

Def. Problem *X* reduces to problem *Y* if you can use an algorithm that solves *Y* to help solve *X*.



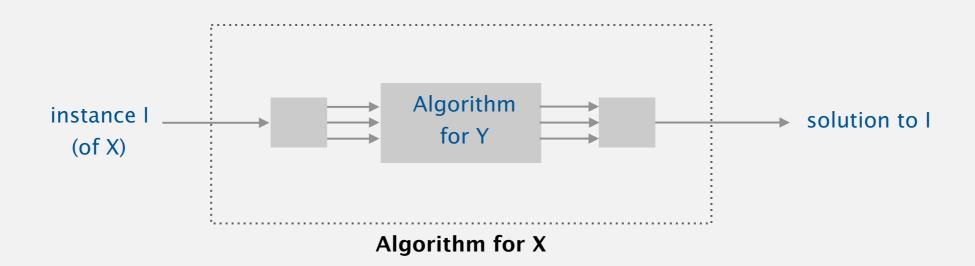
Ex 2. [element distinctness reduces to sorting]

To solve element distinctness on N items:

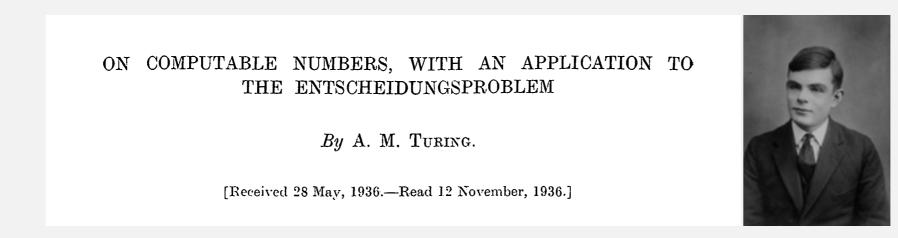
- Sort *N* items.
- Check adjacent pairs for equality.



Def. Problem *X* reduces to problem *Y* if you can use an algorithm that solves *Y* to help solve *X*.



Novice error. Confusing *X reduces to Y* with *Y reduces to X*.



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## Reduction: design algorithms

Def. Problem *X* reduces to problem *Y* if you can use an algorithm that solves *Y* to help solve *X*.

Design algorithm. Given an algorithm for *Y*, can also solve *X*.

#### More familiar reductions.

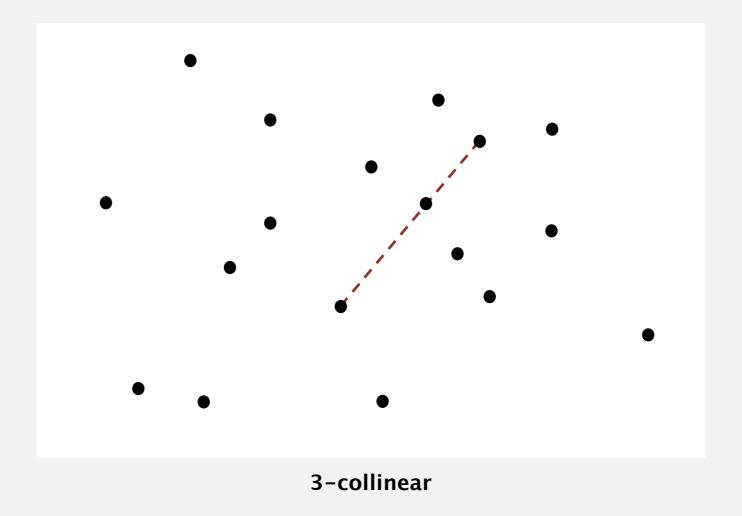
- Mincut reduces to maxflow.
- Arbitrage reduces to negative cycles.
- Bipartite matching reduces to maxflow.
- Seam carving reduces to shortest paths in a DAG.
- · Burrows-Wheeler transform reduces to suffix sort.

. . .

Mentality. Since I know how to solve *Y*, can I use that algorithm to solve *X*?

### 3-collinear

**3-COLLINEAR.** Given *N* distinct points in the plane, are there 3 (or more) that all lie on the same line?

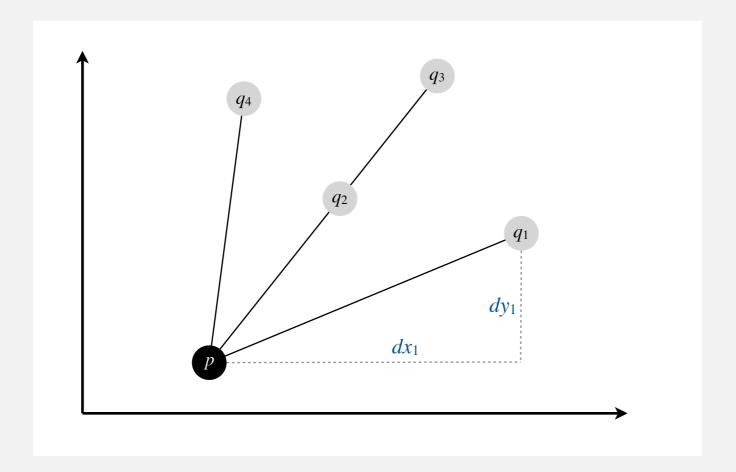


Brute force N<sup>3</sup>. For all triples of points (p, q, r), check if they are collinear.

# 3-collinear reduces to sorting

Sorting-based algorithm. For each point p,

- Compute the slope that each other point q makes with p.
- Sort the N-1 points by slope.
- Collinear points are adjacent.

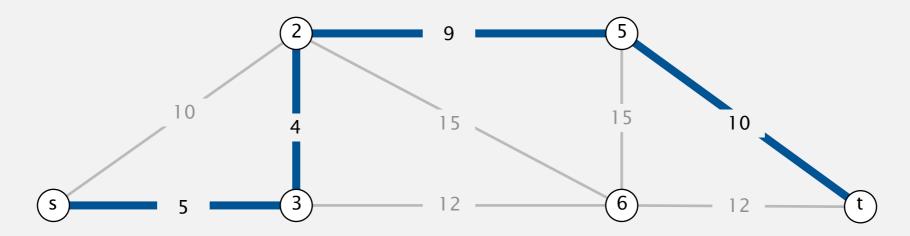


cost of sorting (N times)

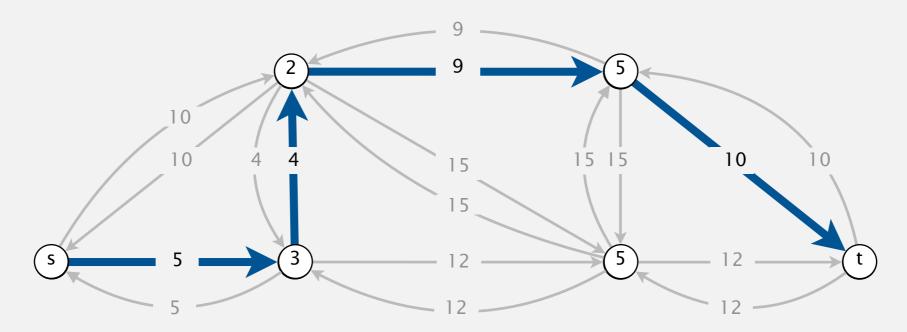
Cost of solving 3-COLLINEAR.  $N^2 \log N + N^2$ .

# Shortest paths on edge-weighted graphs and digraphs

Proposition. Undirected shortest paths (with nonnegative weights) reduces to directed shortest path.

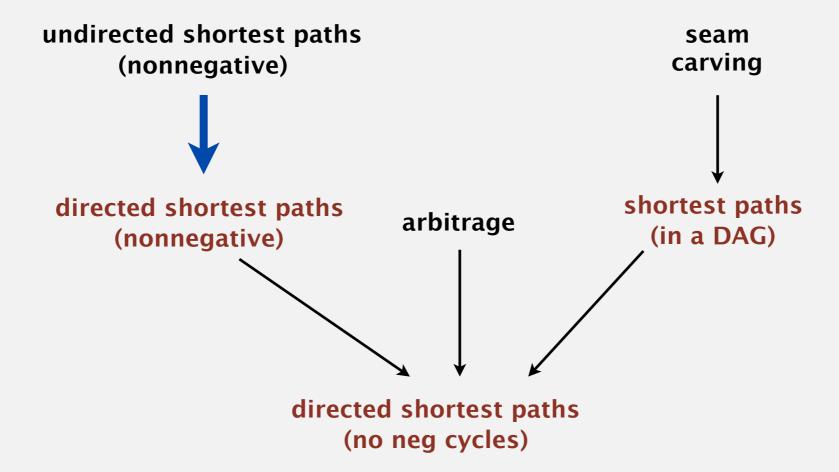


Pf. Replace each undirected edge by two directed edges.



Cost of solving undirected shortest paths.  $E \log V + (E + V)$ .

# Some reductions in combinatorial optimization



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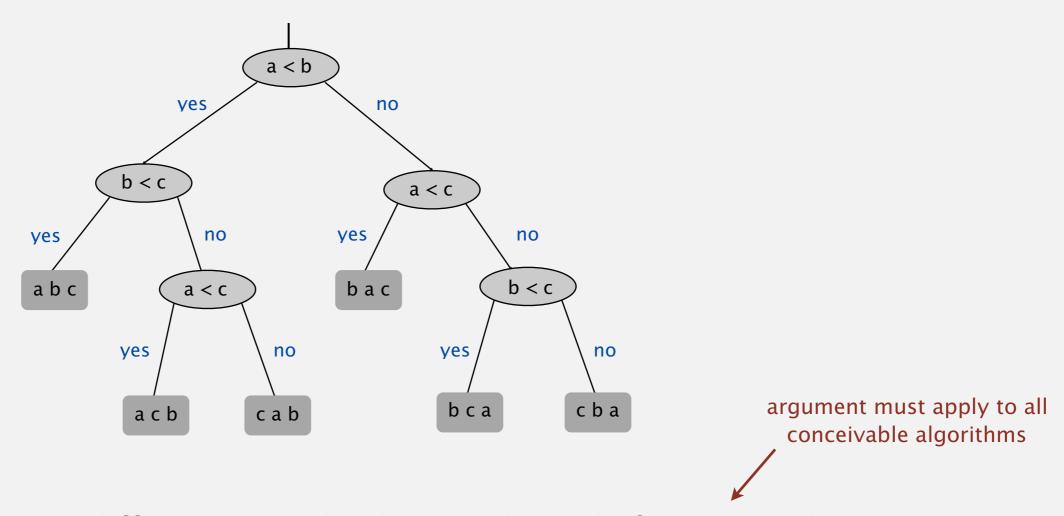
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### Bird's-eye view

Goal. Prove that a problem requires a certain number of steps.

Ex. In decision tree model, any compare-based sorting algorithm requires  $\Omega(N \log N)$  compares in the worst case.



Bad news. Very difficult to establish lower bounds from scratch. Good news. Spread  $\Omega(N \log N)$  lower bound to Y by reducing sorting to Y.

#### Linear-time reductions

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

- Linear number of standard computational steps.
- Constant number of calls to Y.

#### Establish lower bound:

- If *X* takes  $\Omega(N \log N)$  steps, then so does *Y*.
- If X takes  $\Omega(N^2)$  steps, then so does Y.

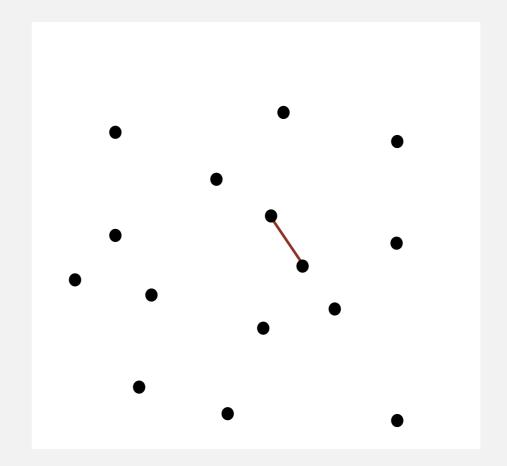
#### Mentality.

- If I could easily solve Y, then I could easily solve X.
- I can't easily solve X.
- Therefore, I can't easily solve *Y*.

#### ELEMENT-DISTINCTNESS linear-time reduces to 2D-CLOSEST-PAIR

ELEMENT-DISTINCTNESS. Given N elements, are any two equal? 2D-CLOSEST-PAIR. Given N points in the plane, find the closest pair.

element distinctness



2d closest pair

#### ELEMENT-DISTINCTNESS linear-time reduces to 2D-CLOSEST-PAIR

ELEMENT-DISTINCTNESS. Given N elements, are any two equal? 2D-CLOSEST-PAIR. Given N points in the plane, find the closest pair.

Proposition. ELEMENT-DISTINCTNESS linear-time reduces to 2D-CLOSEST-PAIR. Pf.

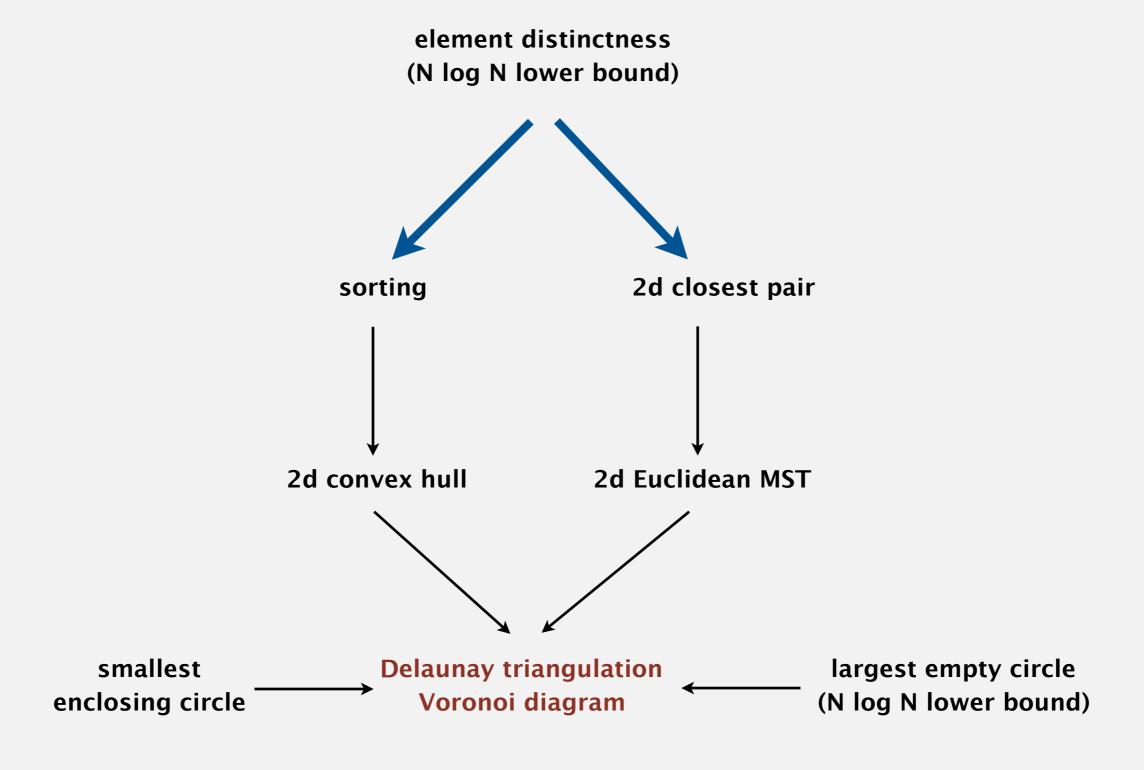
- ELEMENT-DISTINCTNESS instance:  $x_1, x_2, ..., x_N$ .
- 2D-CLOSEST-PAIR instance:  $(x_1, x_1), (x_2, x_2), ..., (x_N, x_N)$ .
- The *N* elements are distinct iff distance of closest pair > 0.

```
allows linear tests like x_i < x_j
and quadratic tests like (x_i - x_k)^2 + (x_j - x_k)^2 > 4
```

ELEMENT-DISTINCTNESS lower bound. In quadratic decision tree model, any algorithm that solves ELEMENT-DISTINCTNESS takes  $\Omega(N \log N)$  steps.

Implication. In quadratic decision tree model, any algorithm for 2D-CLOSEST-PAIR takes  $\Omega(N \log N)$  steps.

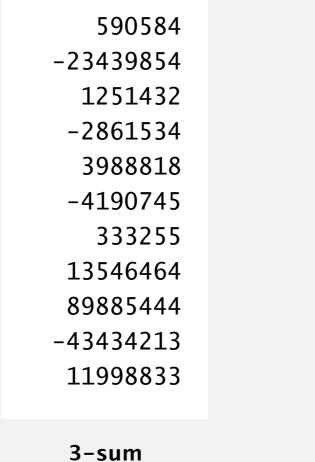
# Some linear-time reductions in computational geometry

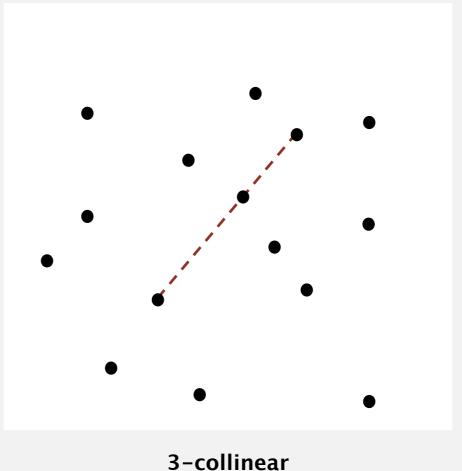


### Lower bound for 3-COLLINEAR

**3-Sum.** Given *N* distinct integers, are there three that sum to 0?

**3-COLLINEAR.** Given *N* distinct points in the plane, are there 3 (or more) that lie on the same line?





#### Lower bound for 3-COLLINEAR

**3-Sum.** Given *N* distinct integers, are there three that sum to 0?

**3-COLLINEAR.** Given *N* distinct points in the plane, are there 3 (or more) that lie on the same line?

Proposition. 3-Sum linear-time reduces to 3-Collinear.

Pf. [next two slides]

lower-bound mentality:

if I can't solve 3-SUM in N<sup>1.99</sup> time,

I can't solve 3-COLLINEAR

in N<sup>1.99</sup> time either

Conjecture. Any algorithm for 3-Sum requires  $\Omega(N^{2-\epsilon})$  steps.

Implication. No sub-quadratic algorithm for 3-Collinear likely.

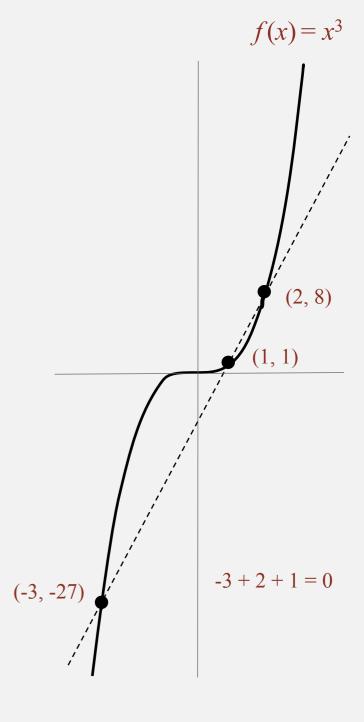
our N<sup>2</sup> log N algorithm was pretty good

#### 3-SUM linear-time reduces to 3-COLLINEAR

Proposition. 3-Sum linear-time reduces to 3-Collinear.

- 3-Sum instance:  $x_1, x_2, ..., x_N$ .
- 3-COLLINEAR instance:  $(x_1, x_1^3), (x_2, x_2^3), ..., (x_N, x_N^3)$ .

Lemma. If a, b, and c are distinct, then a + b + c = 0 if and only if  $(a, a^3)$ ,  $(b, b^3)$ , and  $(c, c^3)$  are collinear.



#### 3-SUM linear-time reduces to 3-COLLINEAR

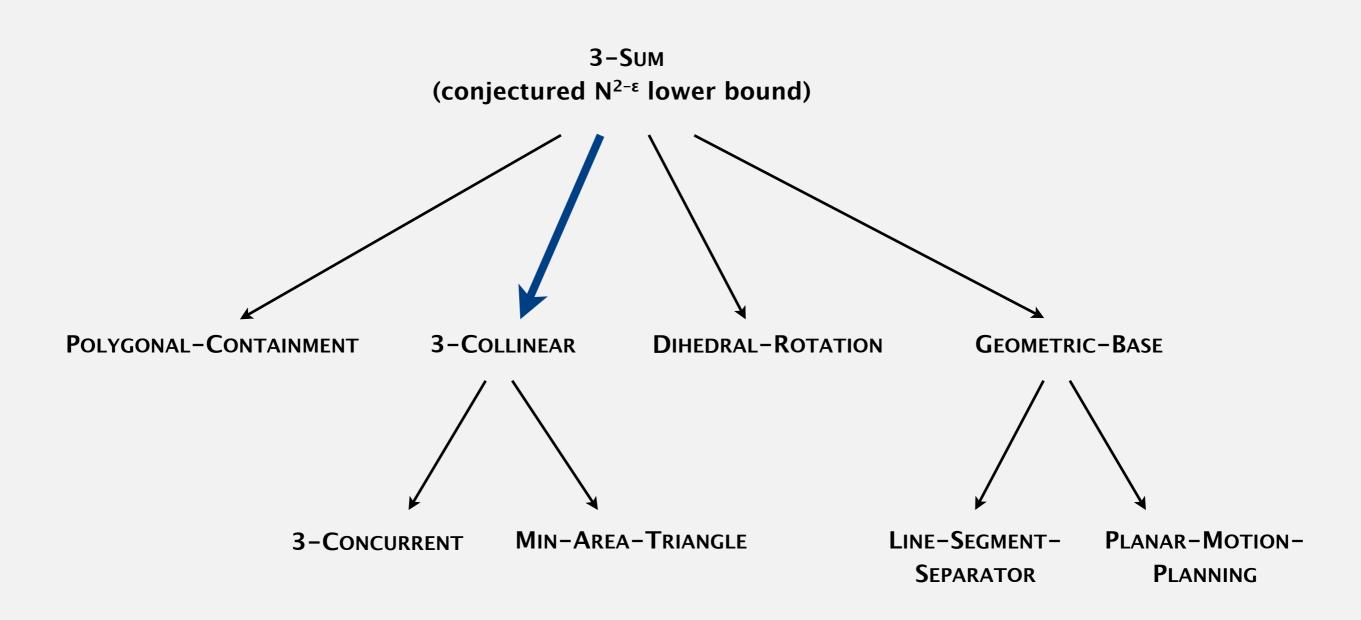
Proposition. 3-Sum linear-time reduces to 3-Collinear.

- 3-Sum instance:  $x_1, x_2, ..., x_N$ .
- 3-COLLINEAR instance:  $(x_1, x_1^3), (x_2, x_2^3), ..., (x_N, x_N^3)$ .

Lemma. If a, b, and c are distinct, then a + b + c = 0 if and only if  $(a, a^3)$ ,  $(b, b^3)$ , and  $(c, c^3)$  are collinear.

Pf. Three distinct points  $(a, a^3)$ ,  $(b, b^3)$ , and  $(c, c^3)$  are collinear iff:

$$0 = \begin{vmatrix} a & a^3 & 1 \\ b & b^3 & 1 \\ c & c^3 & 1 \end{vmatrix}$$
$$= a(b^3 - c^3) - b(a^3 - c^3) + c(a^3 - b^3)$$
$$= (a - b)(b - c)(c - a)(a + b + c)$$



## Complexity of 3-SUM

April 2014. Some recent evidence that the complexity might be  $N^{3/2}$ .

Threesomes, Degenerates, and Love Triangles\*

Allan Grønlund MADALGO, Aarhus University

Seth Pettie University of Michigan

April 4, 2014

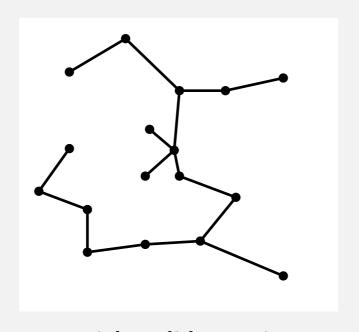
#### Abstract

The 3SUM problem is to decide, given a set of n real numbers, whether any three sum to zero. We prove that the decision tree complexity of 3SUM is  $O(n^{3/2}\sqrt{\log n})$ , that there is a randomized 3SUM algorithm running in  $O(n^2(\log\log n)^2/\log n)$  time, and a deterministic algorithm running in  $O(n^2(\log\log n)^{5/3}/(\log n)^{2/3})$  time. These results refute the strongest version of the 3SUM conjecture, namely that its decision tree (and algorithmic) complexity is  $\Omega(n^2)$ .

## Establishing lower bounds: summary

Establishing lower bounds through reduction is an important tool in guiding algorithm design efforts.

- Q. How to convince yourself no linear-time EUCLIDEAN-MST algorithm exists?
- A1. [hard way] Long futile search for a linear-time algorithm.
- A2. [easy way] Linear-time reduction from element distinctness.







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# Classifying problems: summary

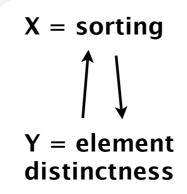
Goal. Problem with algorithm that matches lower bound.

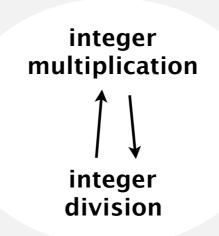
Ex. Sorting and element distinctness have complexity  $N \log N$ .

Goal'. Prove that two problems X and Y have the same complexity. First, show that problem X linear-time reduces to Y.

- Second, show that Y linear-time reduces to X.
- Conclude that *X* has complexity  $N^b$  iff *Y* has complexity  $N^b$  for  $b \ge 1$ .

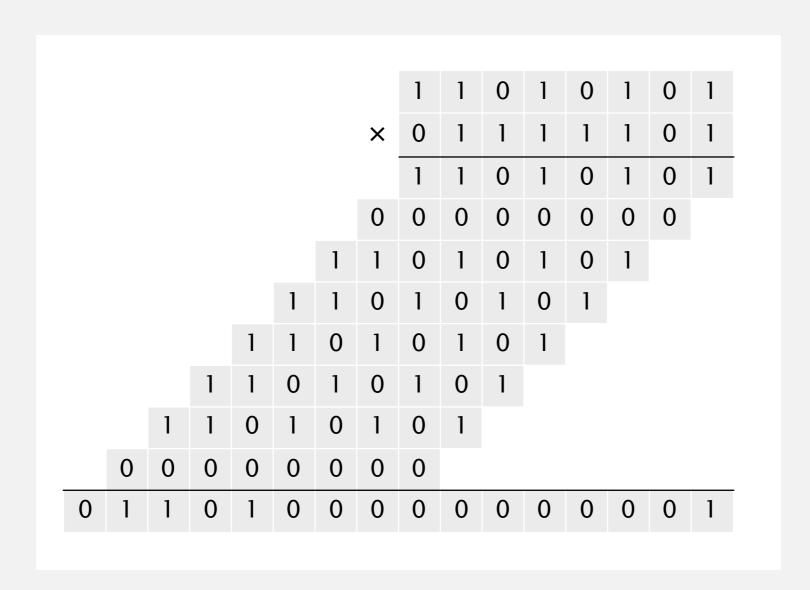
even if we don't know what it is





### Integer arithmetic reductions

Integer multiplication. Given two N-bit integers, compute their product. Brute force.  $N^2$  bit operations.



# Integer arithmetic reductions

Integer multiplication. Given two N-bit integers, compute their product. Brute force.  $N^2$  bit operations.

| problem                | arithmetic                 | order of growth |
|------------------------|----------------------------|-----------------|
| integer multiplication | $a \times b$               | M(N)            |
| integer division       | $a / b$ , $a \mod b$       | M(N)            |
| integer square         | $a^2$                      | M(N)            |
| integer square root    | $\lfloor \sqrt{a} \rfloor$ | M(N)            |

integer arithmetic problems with the same complexity as integer multiplication

Q. Is brute-force algorithm optimal?

# History of complexity of integer multiplication

| year | algorithm          | order of growth           |
|------|--------------------|---------------------------|
| ?    | brute force        | $N^2$                     |
| 1962 | Karatsuba          | N 1.585                   |
| 1963 | Toom-3, Toom-4     | $N^{1.465}$ , $N^{1.404}$ |
| 1966 | Toom-Cook          | N 1 + ε                   |
| 1971 | Schönhage-Strassen | $N \log N \log \log N$    |
| 2007 | Fürer              | $N \log N 2^{\log^* N}$   |
| ?    | ?                  | N                         |

number of bit operations to multiply two N-bit integers

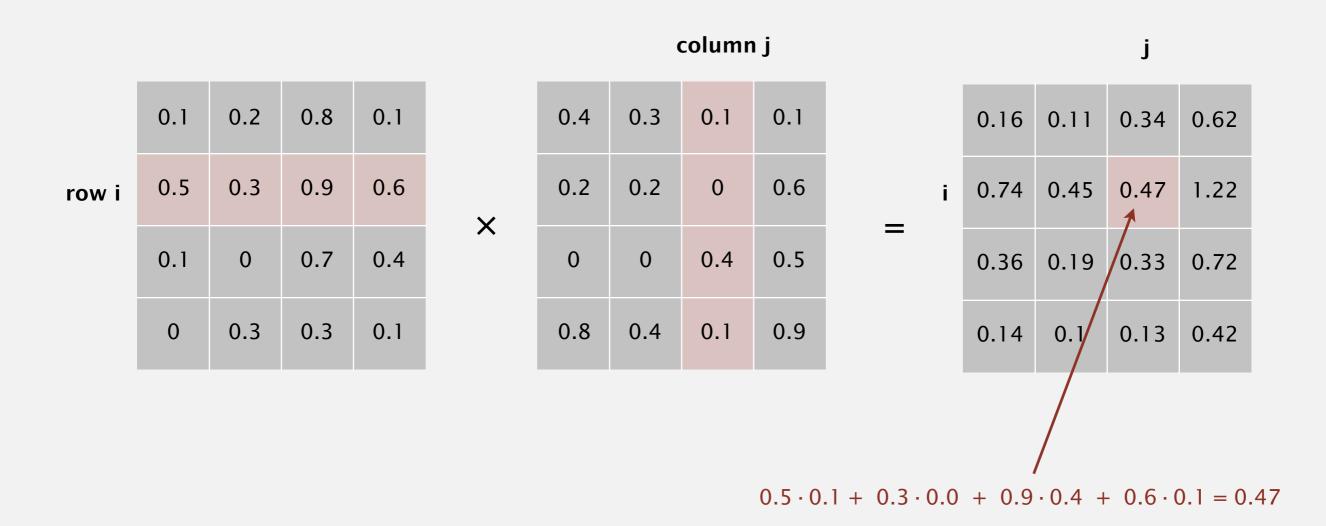
used in Maple, Mathematica, gcc, cryptography, ...

Remark. GNU Multiple Precision Library uses one of five different algorithm depending on size of operands.



## Numerical linear algebra reductions

Matrix multiplication. Given two N-by-N matrices, compute their product. Brute force.  $N^3$  flops.



# Numerical linear algebra reductions

Matrix multiplication. Given two N-by-N matrices, compute their product. Brute force.  $N^3$  flops.

| problem                    | linear algebra      | order of growth |
|----------------------------|---------------------|-----------------|
| matrix multiplication      | $A \times B$        | MM(N)           |
| matrix inversion           | $A^{-1}$            | MM(N)           |
| determinant                | A                   | MM(N)           |
| system of linear equations | Ax = b              | MM(N)           |
| LU decomposition           | A = L U             | MM(N)           |
| least squares              | $\min   Ax - b  _2$ | MM(N)           |

numerical linear algebra problems with the same complexity as matrix multiplication

Q. Is brute-force algorithm optimal?

# History of complexity of matrix multiplication

| year | algorithm            | order of growth  |
|------|----------------------|------------------|
| ?    | brute force          | $N^3$            |
| 1969 | Strassen             | N 2.808          |
| 1978 | Pan                  | N 2.796          |
| 1979 | Bini                 | $N^{2.780}$      |
| 1981 | Schönhage            | N 2.522          |
| 1982 | Romani               | N 2.517          |
| 1982 | Coppersmith-Winograd | N 2.496          |
| 1986 | Strassen             | N 2.479          |
| 1989 | Coppersmith-Winograd | N 2.376          |
| 2010 | Strother             | N 2.3737         |
| 2012 | Williams             | N 2.372873       |
| 2014 | de Gall              | N 2.372864       |
| ?    | ?                    | N <sup>2+ε</sup> |

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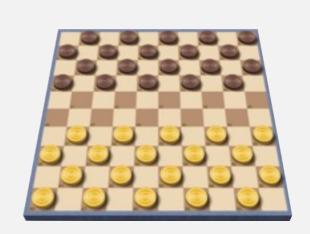
### Bird's-eye view

Def. A problem is intractable if it can't be solved in polynomial time. Goal. Prove that a problem is intractable.

Two problems that provably require exponential time. Input size =  $c + \lg K$ 

- Given a constant-size program, does it halt in at most *K* steps?
- Given N-by-N checkers board position, can the first player force a win?





using forced capture rule

Frustrating news. Very few successes.

### A core problem: satisfiability

SAT. Given a system of boolean equations, find a solution.

Ex.

 $x_1$   $x_2$   $x_3$   $x_4$  T F T

instance I

solution S

3-SAT. All equations of this form (with three variables per equation).

#### Key applications.

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

• ...

# Satisfiability is conjectured to be intractable

- Q. How to solve an instance of 3-SAT with *N* variables?
- A. Exhaustive search: try all  $2^N$  truth assignments.



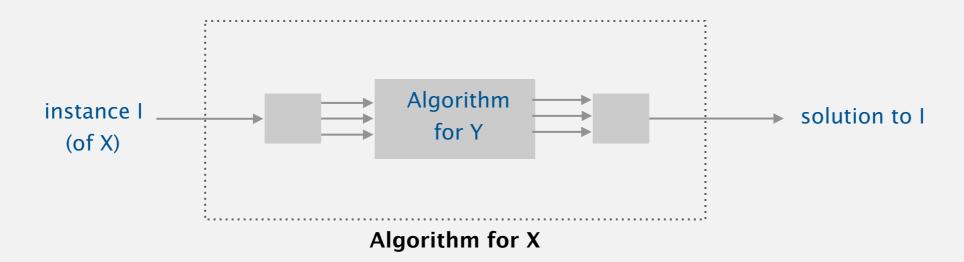
Q. Can we do anything substantially more clever?

Conjecture ( $P \neq NP$ ). 3-SAT is intractable (no poly-time algorithm).

### Polynomial-time reductions

Problem *X* poly-time (Cook) reduces to problem *Y* if *X* can be solved with:

- Polynomial number of standard computational steps.
- Polynomial number of calls to Y.



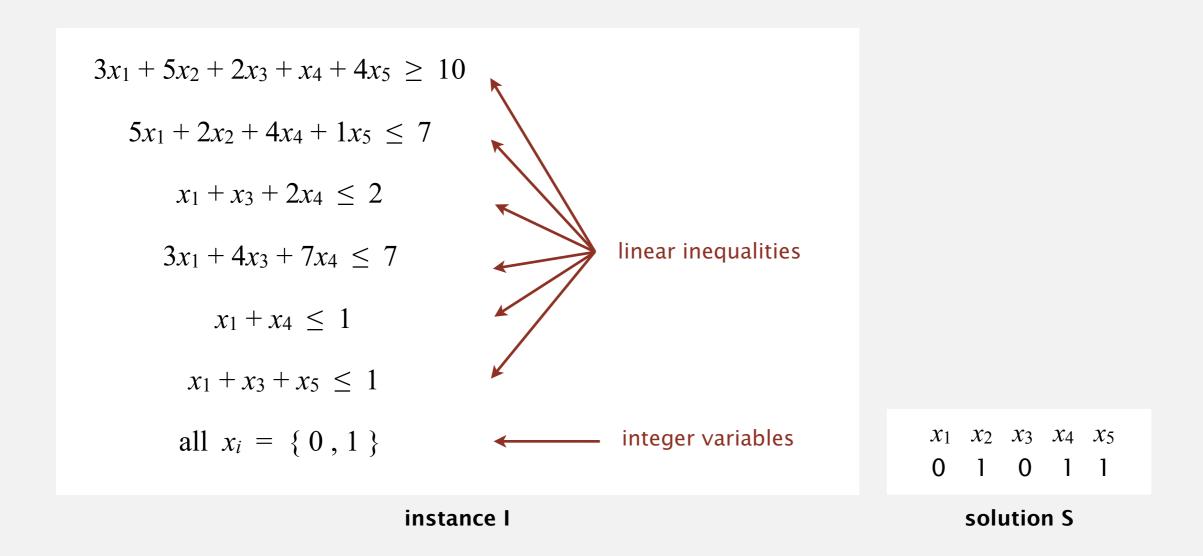
Establish intractability. If 3-SAT poly-time reduces to Y, then Y is intractable. (assuming 3-SAT is intractable)

#### Mentality.

- If I could solve *Y* in poly-time, then I could also solve 3-SAT in poly-time.
- 3-SAT is believed to be intractable.
- Therefore, so is *Y*.

### Integer linear programming

ILP. Given a system of linear inequalities, find an integral solution.



Context. Cornerstone problem in operations research.

Remark. Finding a real-valued solution is tractable (linear programming).

# 3-SAT poly-time reduces to ILP

3-SAT. Given a system of boolean equations, find a solution.

ILP. Given a system of linear inequalities, find a 0-1 solution.

$$(1-x_1) + x_2 + x_3 \ge 1$$

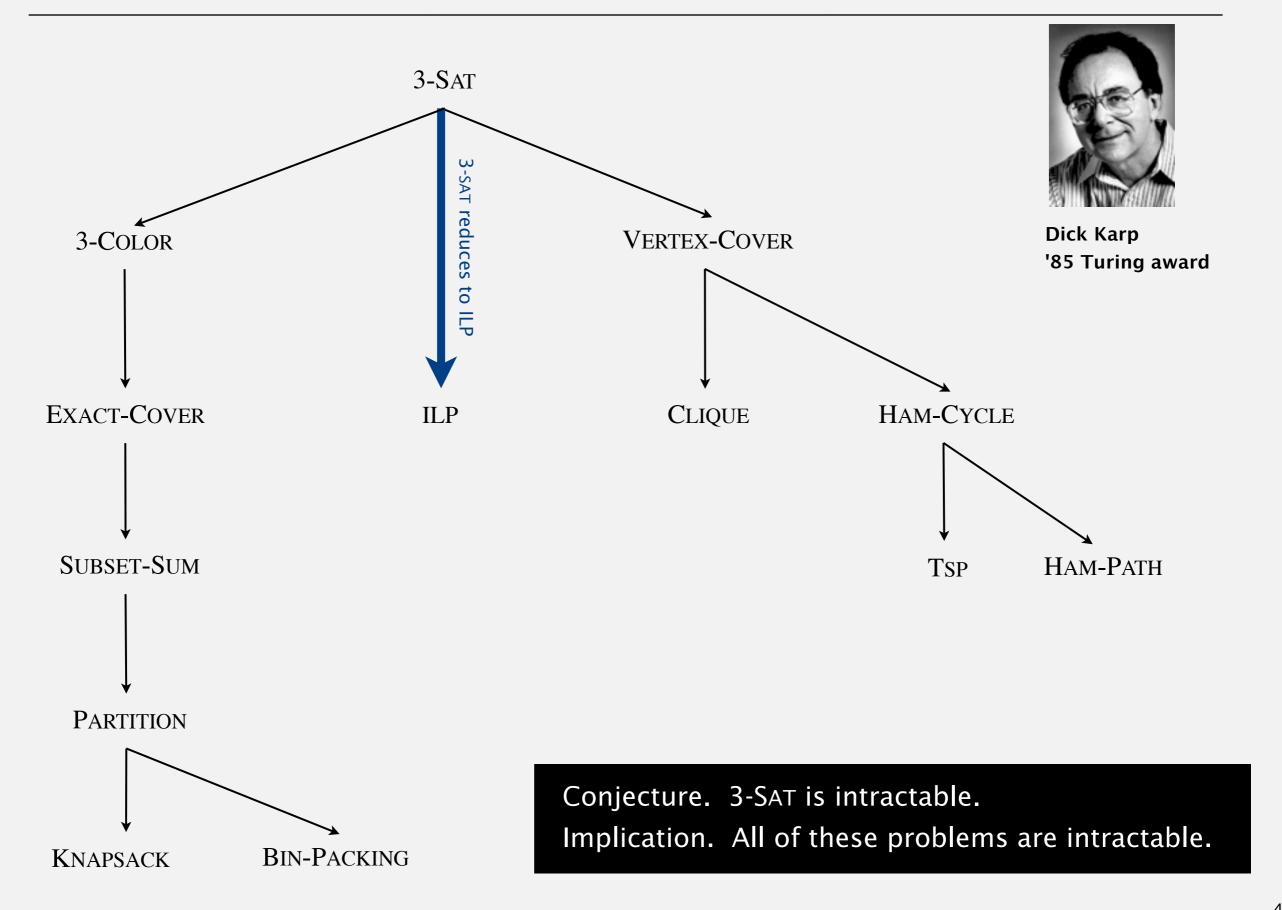
$$x_1 + (1-x_2) + x_3 \ge 1$$

$$(1-x_1) + (1-x_2) + (1-x_3) \ge 1$$

$$(1-x_1) + (1-x_2) + x_3 + x_4 \ge 1$$

$$(1-x_2) + x_3 + x_4 \ge 1$$

# More poly-time reductions from 3-satisfiability

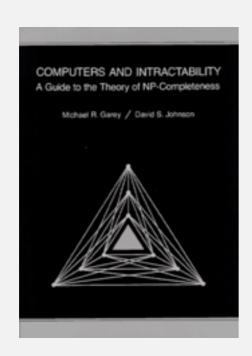


# Implications of poly-time reductions from 3-satisfiability

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

- Q. How to convince yourself that a new problem is (probably) intractable?
- A1. [hard way] Long futile search for an efficient algorithm (as for 3-SAT).
- A2. [easy way] Reduction from 3-SAT.

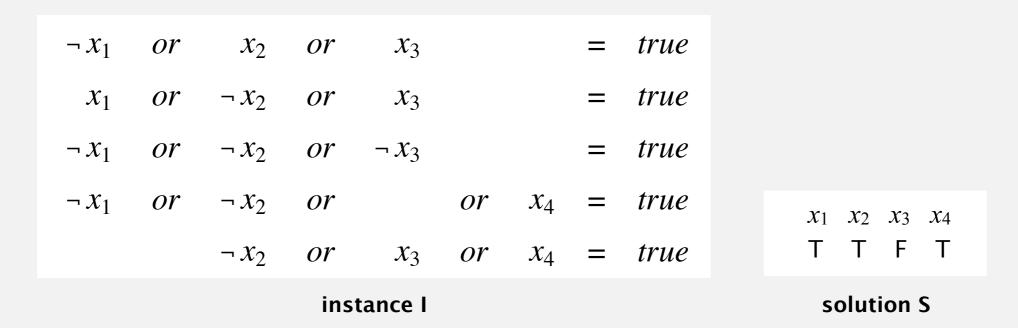
Caveat. Intricate reductions are common.



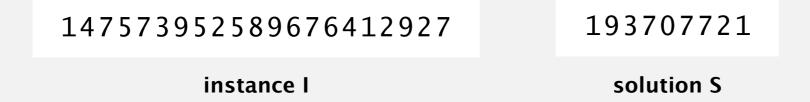
# Search problems

Search problem. Problem where you can check a solution in poly-time.

**Ex** 1. 3-SAT.



Ex 2. FACTOR. Given an N-bit integer x, find a nontrivial factor.

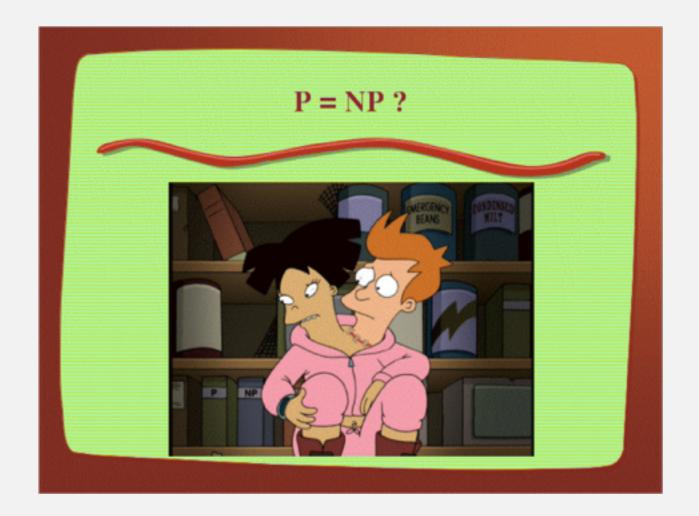


#### P vs. NP

P. Set of search problems solvable in poly-time. Importance. What scientists and engineers can compute feasibly.

NP. Set of search problems (checkable in poly-time). Importance. What scientists and engineers aspire to compute feasibly.

Fundamental question.



Consensus opinion. No.

### Cook-Levin theorem

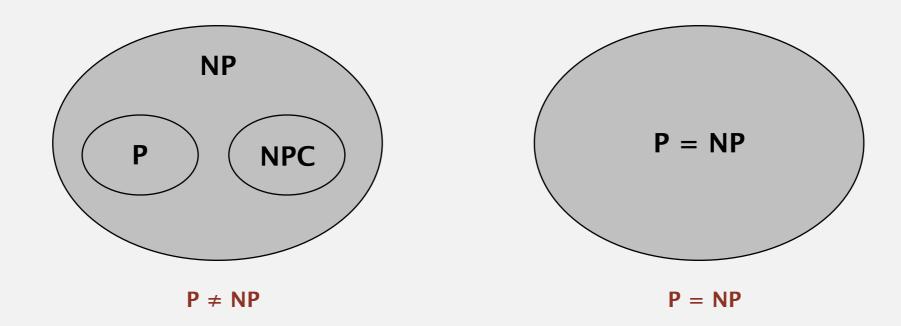
A problem is **NP-Complete** if

- It is in NP.
- All problems in NP poly-time to reduce to it.

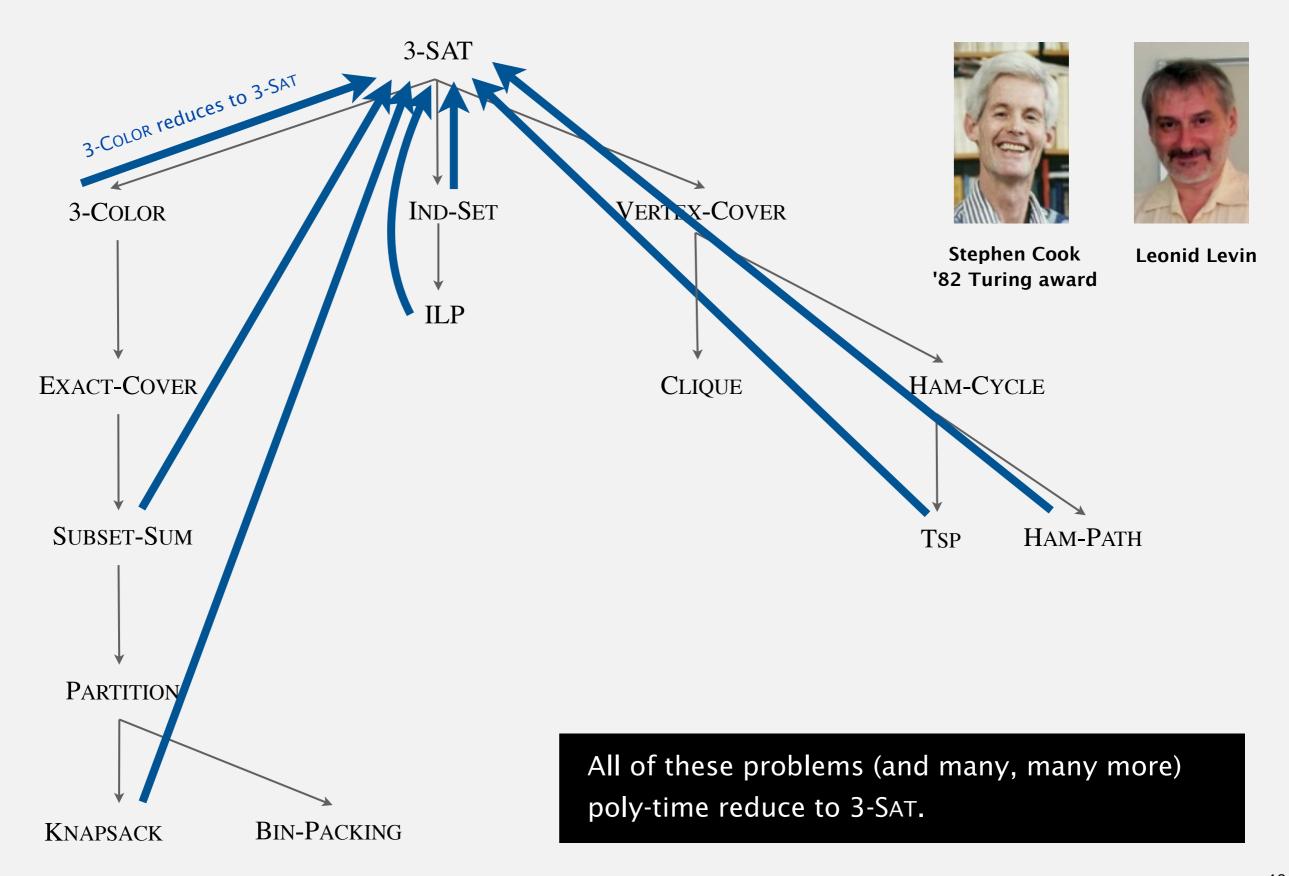
Cook-Levin theorem. 3-SAT is NP-COMPLETE.

Corollary. 3-SAT is tractable if and only if P = NP.

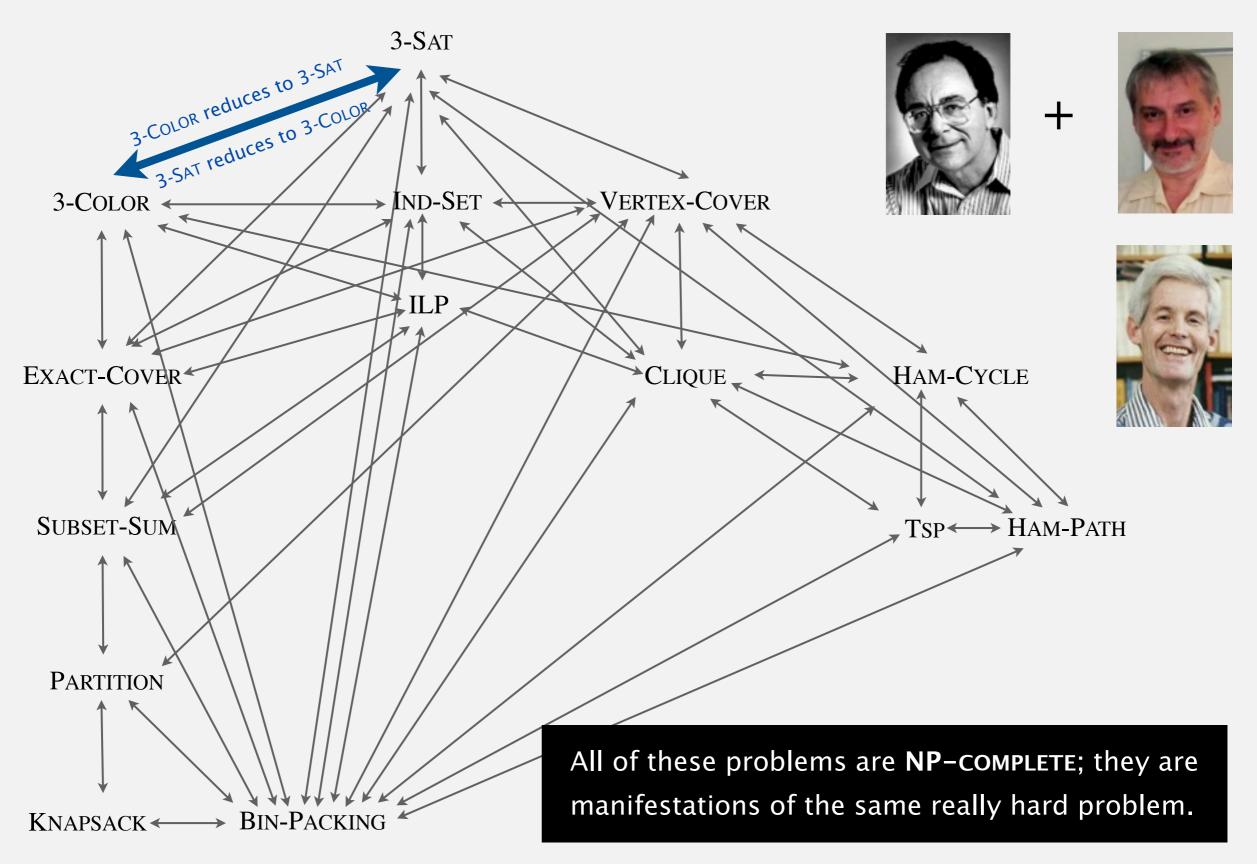
#### Two worlds.



# Implications of Cook-Levin theorem



# Implications of Karp + Cook-Levin



### Birds-eye view: review

Goal. Classify problems according to computational requirements.

| complexity   | order of growth | examples                                     |
|--------------|-----------------|--|
| linear       | N               | min, max, median, Burrows-Wheeler transform, |
| linearithmic | $N \log N$      | sorting, element distinctness,               |
| quadratic    | N <sup>2</sup>  | ?  |
| <b>:</b>     | ÷               | <b>:</b>                                     |
| exponential  | $c^N$           | ?  |

Frustrating news. Huge number of problems have defied classification.

# Birds-eye view: revised

Goal. Classify problems according to computational requirements.

| complexity   | order of growth             | examples   |
|--------------|-----------------------------|--|
| linear       | N                           | min, max, median, Burrows-Wheeler transform,                 |
| linearithmic | $N \log N$                  | sorting, element distinctness,                               |
| M(N)         | ?                           | integer multiplication,<br>division, square root,            |
| MM(N)        | ?                           | matrix multiplication, $Ax = b$ , least square, determinant, |
| <b>:</b>     | <b>:</b>                    | :  |
| NP-complete  | probably not N <sup>b</sup> | 3-SAT, IND-SET, ILP,   |

Good news. Can put many problems into equivalence classes.

### Complexity zoo

Complexity class. Set of problems sharing some computational property.



https://complexityzoo.uwaterloo.ca

Bad news. Lots of complexity classes (496 animals in zoo).

### Summary

#### Reductions are important in theory to:

- Design algorithms.
- Establish lower bounds.
- Classify problems according to their computational requirements.

#### Reductions are important in practice to:

- Design algorithms.
- Design reusable software modules.
  - stacks, queues, priority queues, symbol tables, sets, graphs
  - sorting, regular expressions, suffix arrays
  - MST, shortest paths, maxflow, linear programming
- Determine difficulty of your problem and choose the right tool.

