

8.6 Reductions

- ▶ designing algorithms
- ▶ establishing lower bounds
- ▶ intractability

Bird's-eye view

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

Desiderata'.

Suppose we could (couldn't) solve problem X efficiently.

What else could (couldn't) we solve efficiently?



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

Bird's-eye view

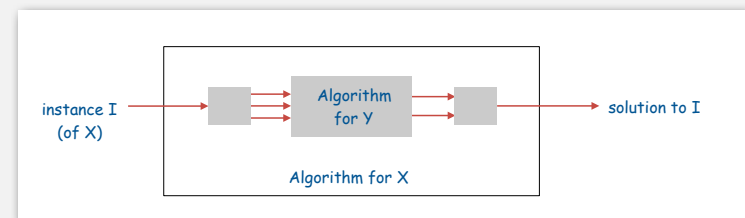
Desiderata. Classify **problems** according to computational requirements.

complexity	order of growth	examples
linear	N	min, max, median, Burrows-Wheeler transform, ...
linearithmic	$N \log N$	sorting, convex hull, closest pair, farthest pair, ...
quadratic	N^2	???
	...	
exponential	c^N	???

Frustrating news. Huge number of problems have defied classification.

Reduction

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

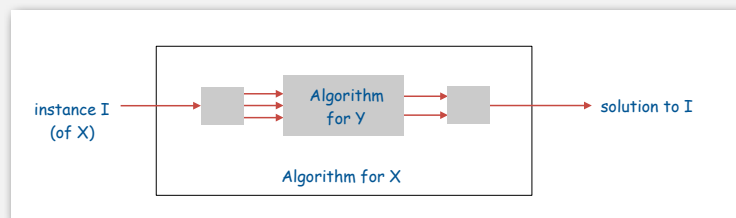


Cost of solving X = total cost of solving Y + cost of reduction.

↑ perhaps many calls to Y on problems of different sizes ↑ preprocessing and postprocessing

Reduction

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



Ex 1. [element distinctness reduces to sorting]

To solve element distinctness on N integers:

- Sort N integers.
- Check adjacent pairs for equality.

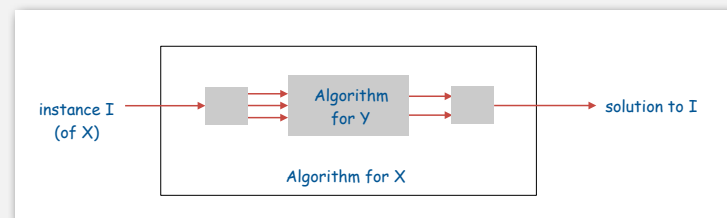
Cost of solving element distinctness. $N \log N + N$

cost of sorting (pointing to $N \log N$)
cost of reduction (pointing to $+ N$)

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Reduction

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



Ex 2. [3-collinear reduces to sorting]

To solve 3-collinear instance on N points in the plane:

- For each point, sort other points by polar angle.
 - check adjacent triples for collinearity

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

cost of sorting (pointing to $N^2 \log N$)
cost of reduction (pointing to $+ N^2$)

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- ▶ designing algorithms
- ▶ establishing lower bounds
- ▶ intractability

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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 algorithm for Y, can also solve X.

Ex.

- Element distinctness reduces to sorting.
- 3-collinear reduces to sorting.
- PERT reduces to topological sort. [see digraph lecture]
- h-v line intersection reduces to 1D range searching. [see geometry lecture]
- Burrows-Wheeler transform reduces to suffix sort. [see assignment 8]

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

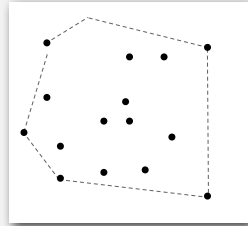
↑
programmer's version: I have code for Y. Can I use it for X?

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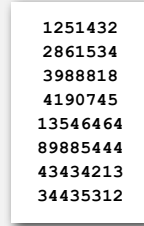
Convex hull reduces to sorting

Sorting. Given N distinct integers, rearrange them in ascending order.

Convex hull. Given N points in the plane, identify the extreme points of the convex hull (in counter-clockwise order).



convex hull



sorting

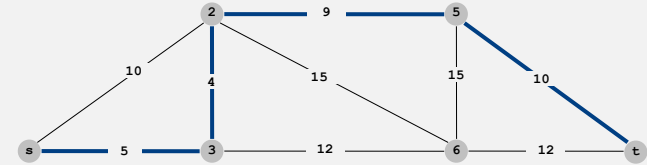
Proposition. Convex hull reduces to sorting.

Pf. Graham scan algorithm.

Cost of convex hull. $N \log N + N$.
cost of sorting \swarrow $N \log N$ \nwarrow cost of reduction $+ N$

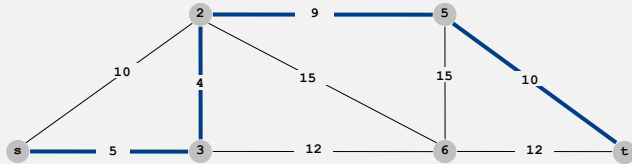
Shortest path on graphs and digraphs

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

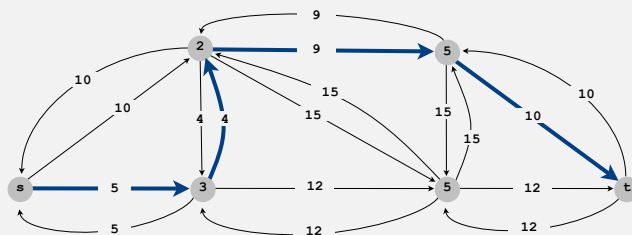


Shortest path on graphs and digraphs

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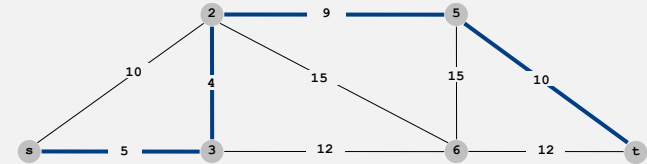


Pf. Replace each undirected edge by two directed edges.



Shortest path on graphs and digraphs

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

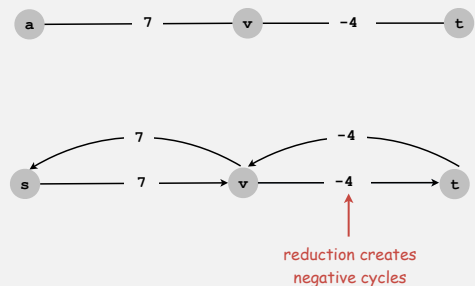


Cost of undirected shortest path. $E \log E + E$.

cost of shortest path in digraph \swarrow $E \log E$ \nwarrow cost of reduction $+ E$

Shortest path with negative weights

Caveat. Reduction is invalid in networks with negative weights (even if no negative cycles).

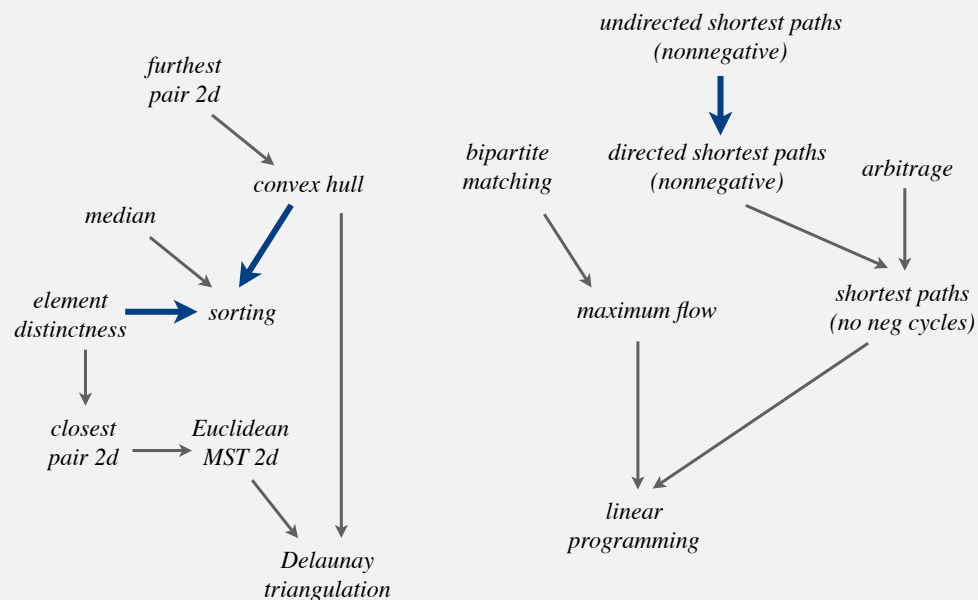


Remark. Can still solve shortest path problem in undirected graphs (if no negative cycles), but need more sophisticated techniques.

reduces to weighted non-bipartite matching (!)

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Some reductions involving familiar problems



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- ▶ designing algorithms
- ▶ **establishing lower bounds**
- ▶ intractability

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Bird's-eye view

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

Ex. $\Omega(N \log N)$ lower bound for sorting.

```

1251432
2861534
3988818
4190745
13546464
89885444
43434213
    
```

argument must apply to all conceivable algorithms

Bad news. Very difficult to establish lower bounds from scratch.

Good news. Can spread $\Omega(N \log N)$ lower bound to Y by reducing sorting to Y .

assuming cost of reduction is not too high

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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.

Ex. Almost all of the reductions we've seen so far. [Which one wasn't?]

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.

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Lower bound for convex hull

Proposition. In quadratic decision tree model, any algorithm for sorting N integers requires $\Omega(N \log N)$ steps.

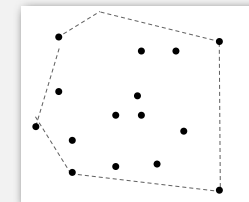
allows quadratic tests of the form:
 $x_i < x_j$ or $(x_j - x_i)(x_k - x_i) - (x_j)(x_j - x_i) < 0$

Proposition. Sorting linear-time reduces to convex hull.

Pf. [see next slide]

1251432
2861534
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43434213

sorting



convex hull

a quadratic test

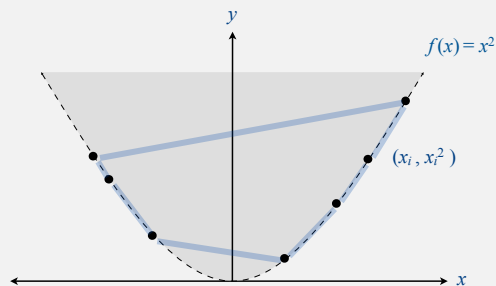
Implication. Any ccw-based convex hull algorithm requires $\Omega(N \log N)$ ccw's.

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Sorting linear-time reduces to convex hull

Proposition. Sorting linear-time reduces to convex hull.

- **Sorting instance:** x_1, x_2, \dots, x_N .
- **Convex hull instance:** $(x_1, x_1^2), (x_2, x_2^2), \dots, (x_N, x_N^2)$.



Pf.

- Region $\{x : x^2 \geq x\}$ is convex \Rightarrow all points are on hull.
- Starting at point with most negative x, counter-clockwise order of hull points yields integers in ascending order.

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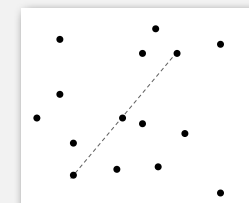
Lower bound for 3-COLLINEAR

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 that all lie on the same line? ← recall Assignment 3

1251432
-2861534
3988818
-4190745
13546464
89885444
-43434213

3-sum



3-collinear

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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 that all lie on the same line?

Proposition. 3-SUM linear-time reduces to 3-COLLINEAR.

Pf. [see next 2 slide]

Conjecture. Any algorithm for 3-SUM requires $\Omega(N^2)$ steps.

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

your $N^2 \log N$ algorithm was pretty good

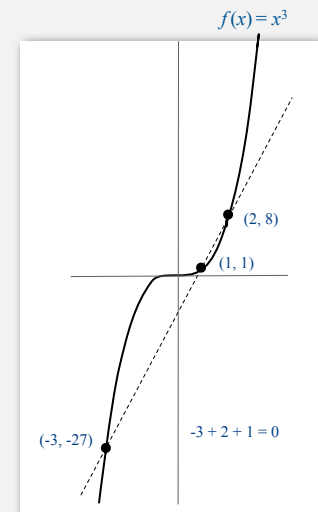
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3-SUM linear-time reduces to 3-COLLINEAR

Proposition. 3-SUM linear-time reduces to 3-COLLINEAR.

- 3-SUM instance: x_1, x_2, \dots, x_N .
- 3-COLLINEAR instance: $(x_1, x_1^3), (x_2, x_2^3), \dots, (x_N, x_N^3)$.

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.



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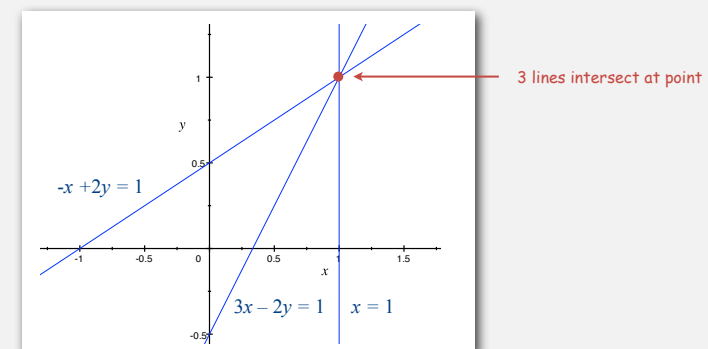
Pf. Three distinct points $(a, a^3), (b, b^3),$ and (c, c^3) are collinear iff:

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

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Concurrent lines

3-CONCURRENT. Given N distinct lines, are there 3 that intersect at a point?



Lemma. The 3 lines $a_i x + b_i y = 1, a_j x + b_j y = 1,$ and $a_k x + b_k y = 1$ are concurrent if and only if:

$$\begin{vmatrix} a_i & b_i & 1 \\ a_j & b_j & 1 \\ a_k & b_k & 1 \end{vmatrix} = 0$$

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3-COLLINEAR linear-time reduces to 3-CONCURRENT

Proposition. 3-COLLINEAR linear-time reduces to 3-CONCURRENT.

- 3-COLLINEAR instance: $(x_1, y_1), \dots, (x_N, y_N)$.
- 3-CONCURRENT instance: $a_1x + b_1y = 1, \dots, a_Nx + b_Ny = 1$,
where $a_i = x_i$ and $b_i = y_i$.

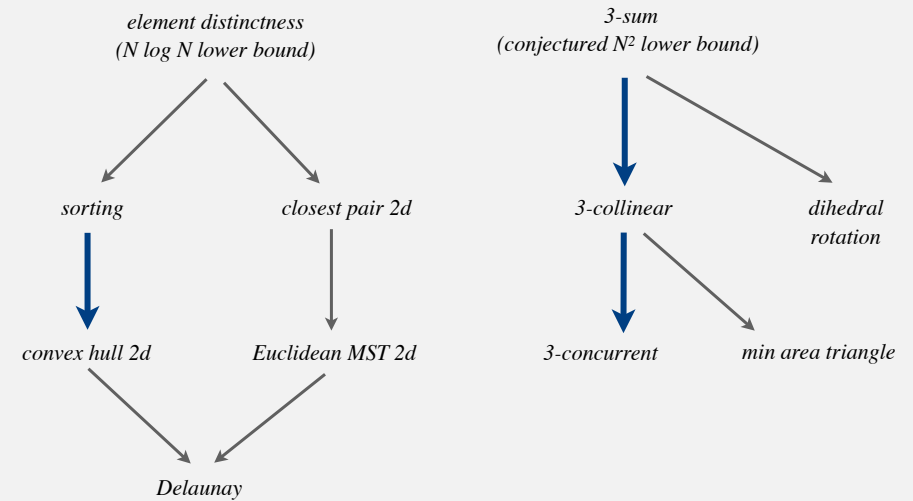
Lemma. The 3 points (x_i, y_i) , (x_j, y_j) , and (x_k, y_k) are collinear if and only if the 3 lines $a_ix + b_iy = 1$, $a_jx + b_jy = 1$, and $a_kx + b_ky = 1$ are concurrent.

Pf. [duality between points and lines]

$$\begin{vmatrix} x_i & y_i & 1 \\ x_j & y_j & 1 \\ x_k & y_k & 1 \end{vmatrix} = 0 \qquad \begin{vmatrix} a_i & b_i & 1 \\ a_j & b_j & 1 \\ a_k & b_k & 1 \end{vmatrix} = 0$$

3 points are collinear 3 lines are concurrent

More linear-time reductions and lower bounds



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 convex hull algorithm exists?

- A1.** [hard way] Long futile search for a linear-time algorithm.
- A2.** [easy way] Linear-time reduction from sorting.

Q. How to convince yourself no sub-quadratic 3-COLLINEAR algorithm exists.

- A1.** [hard way] Long futile search for a sub-quadratic algorithm.
- A2.** [easy way] Linear-time reduction from 3-SUM.

- ▶ designing algorithms
- ▶ establishing lower bounds
- ▶ **intractability**

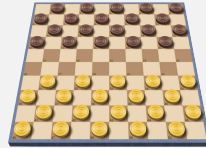
Bird's-eye view

Def. A problem is **intractable** if it can't be solved in polynomial time.

Desiderata. Prove that a problem is intractable.

Two problems that require exponential time.

- 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?



input size = $c + \lg K$

using forced capture rule

Frustrating news. Few successes.

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3-satisfiability

Literal. A boolean variable or its negation.

$$x_i \text{ OR } \neg x_i$$

Clause. An *or* of 3 distinct literals.

$$C_1 = (\neg x_1 \vee x_2 \vee x_3)$$

Conjunctive normal form. An *and* of clauses.

$$\Phi = (C_1 \wedge C_2 \wedge C_3 \wedge C_4 \wedge C_5)$$

3-SAT. Given a CNF formula Φ consisting of k clauses over n literals, does it have a satisfying truth assignment?

$$\Phi = (\neg x_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_4) \wedge (\neg x_2 \vee x_3 \vee x_4)$$

yes instance

$$\begin{array}{cccc} x_1 & x_2 & x_3 & x_4 \\ T & T & F & T \end{array}$$

$$(\neg T \vee T \vee F) \wedge (T \vee \neg T \vee F) \wedge (\neg T \vee \neg T \vee \neg F) \wedge (\neg T \vee \neg T \vee T) \wedge (\neg T \vee F \vee T)$$

Applications. Circuit design, program correctness, ...

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3-satisfiability is believed 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?



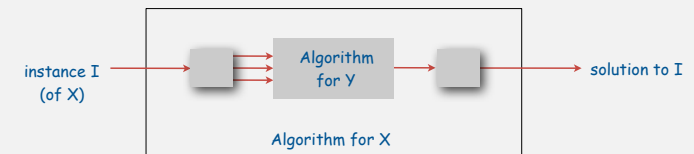
Conjecture ($P \neq NP$). 3-SAT is intractable (no poly-time algorithm).

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Polynomial-time reductions

Def. 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.

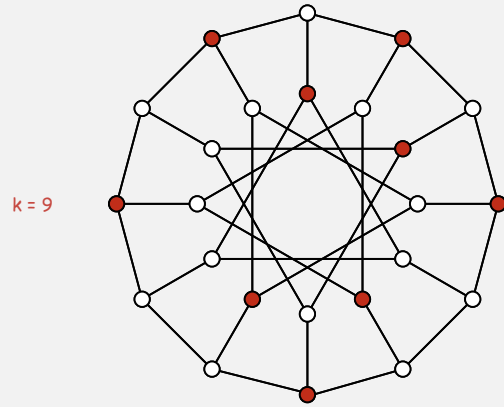
- If I could solve Y in poly-time, then I could also solve 3-SAT in poly-time.
- 3-SAT is believed to be intractable.
- Therefore, so is Y.

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Independent set

Def. An **independent set** is a set of vertices, no two of which are adjacent.

IND-SET. Given a graph G and an integer k , find an independent set of size k .



Applications. Scheduling, computer vision, clustering, ...

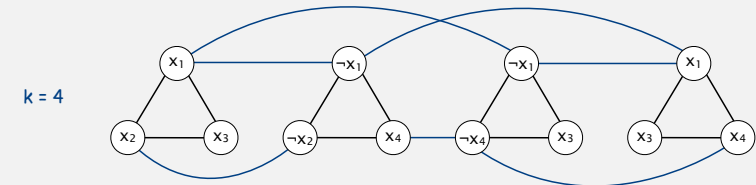
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3-satisfiability reduces to independent set

Proposition. 3-SAT poly-time reduces to IND-SET.

Pf. Given an instance Φ of 3-SAT, create an instance G of IND-SET:

- For each clause in Φ , create 3 vertices in a triangle.
- Add an edge between each literal and its negation.



$$\Phi = (x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_4) \wedge (\neg x_1 \vee x_3 \vee \neg x_4) \wedge (x_1 \vee x_3 \vee x_4)$$

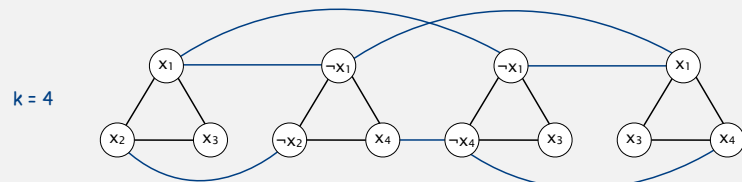
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- G has independent set of size $k \Rightarrow \Phi$ satisfiable.

↑
set literals corresponding to vertices in independent to true;
set remaining literals in consistent manner

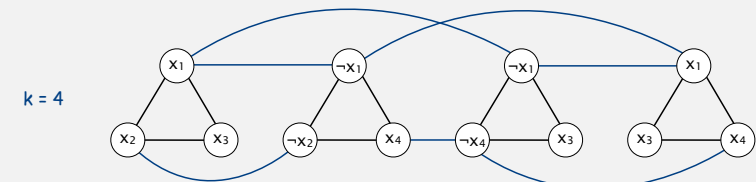
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- G has independent set of size $k \Rightarrow \Phi$ satisfiable.
- Φ satisfiable $\Rightarrow G$ has independent set of size k .

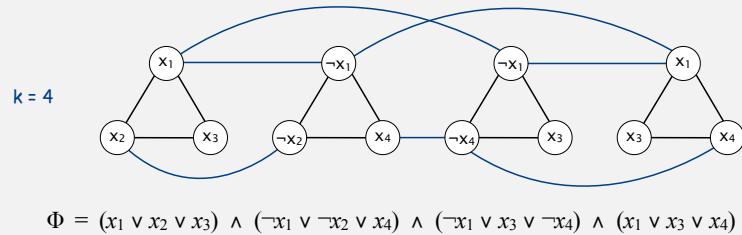
↑
for each clause, take vertex corresponding to one true literal

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3-satisfiability reduces to independent set

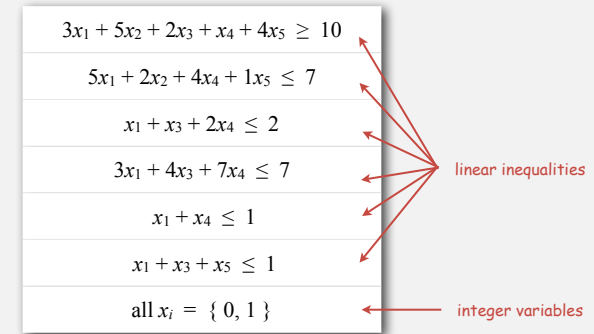
Proposition. 3-SAT poly-time reduces to IND-SET.

Implication. Assuming 3-SAT is intractable, so is IND-SET.



Integer linear programming

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



Context. Cornerstone problem in operations research.

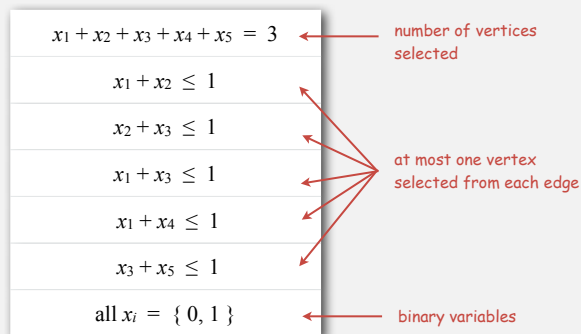
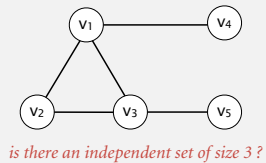
Remark. Finding a real-valued solution is tractable (linear programming).

Independent set reduces to integer linear programming

Proposition. IND-SET poly-time reduces to ILP.

Pf. Given an instance G, k of IND-SET, create an instance of ILP as follows:

Intuition. $x_i = 1$ if and only if vertex v_i is in independent set.



is there a feasible solution?

3-satisfiability reduces to integer linear programming

Proposition. 3-SAT poly-time reduces to IND-SET.

Proposition. IND-SET poly-time reduces to ILP.

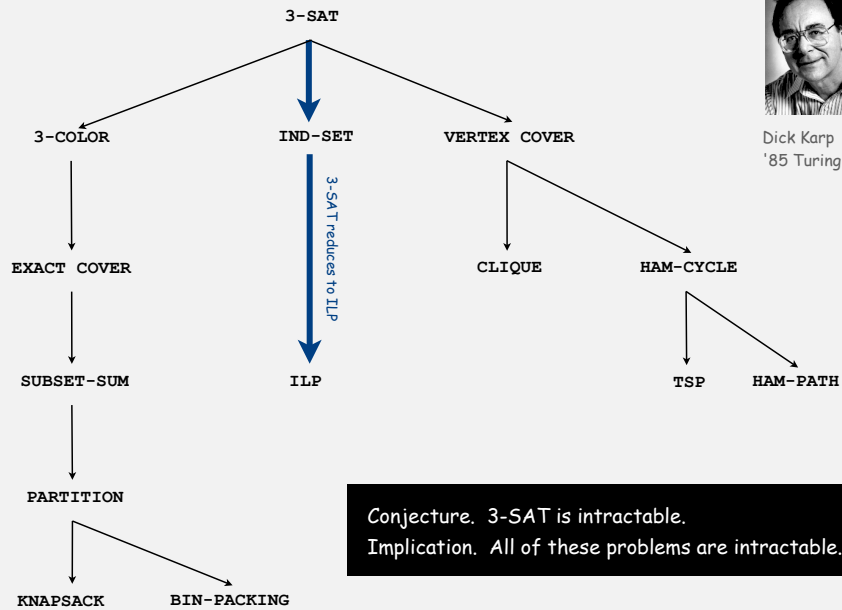
Transitivity. If X poly-time reduces to Y and Y poly-time reduces to Z, then X-poly-time reduces to Z.

Implication. Assuming 3-SAT is intractable, so is ILP.

More poly-time reductions from 3-satisfiability



Dick Karp
'85 Turing award



Conjecture. 3-SAT is intractable.
Implication. All of these problems are intractable.

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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.

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Search problems

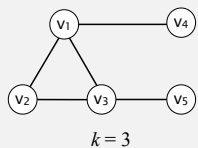
Search problem. Problem where you can check a solution in poly-time.

Ex 1. 3-SAT.

$$\Phi = (x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_4) \wedge (\neg x_1 \vee x_3 \vee \neg x_4) \wedge (x_1 \vee x_3 \vee x_4)$$

$x_1 = \text{true}, x_2 = \text{true}, x_3 = \text{true}, x_4 = \text{true}$

Ex 2. IND-SET.



$\{v_2, x_4, v_5\}$

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P vs. NP

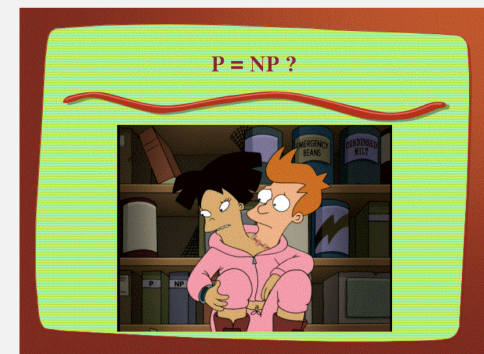
P. Set of search problems solvable in poly-time.

Importance. What scientists and engineers can compute feasibly.

NP. Set of search problems.

Importance. What scientists and engineers aspire to compute feasibly.

Fundamental question.



Consensus opinion. No.

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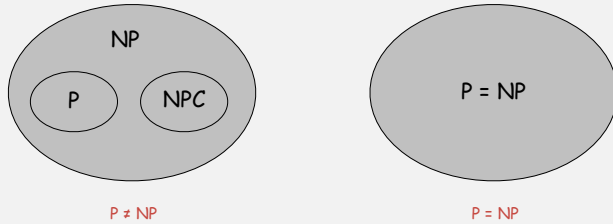
Cook's theorem

Def. An NP is **NP-complete** if all problems in NP poly-time to reduce to it.

Cook's theorem. 3-SAT is NP-complete.

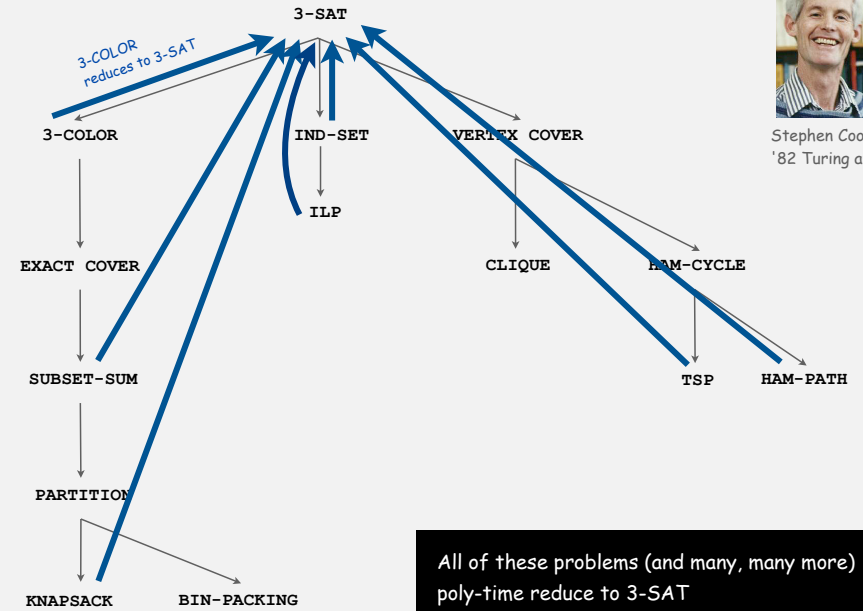
Corollary. 3-SAT is tractable if and only if $P = NP$.

Two worlds.



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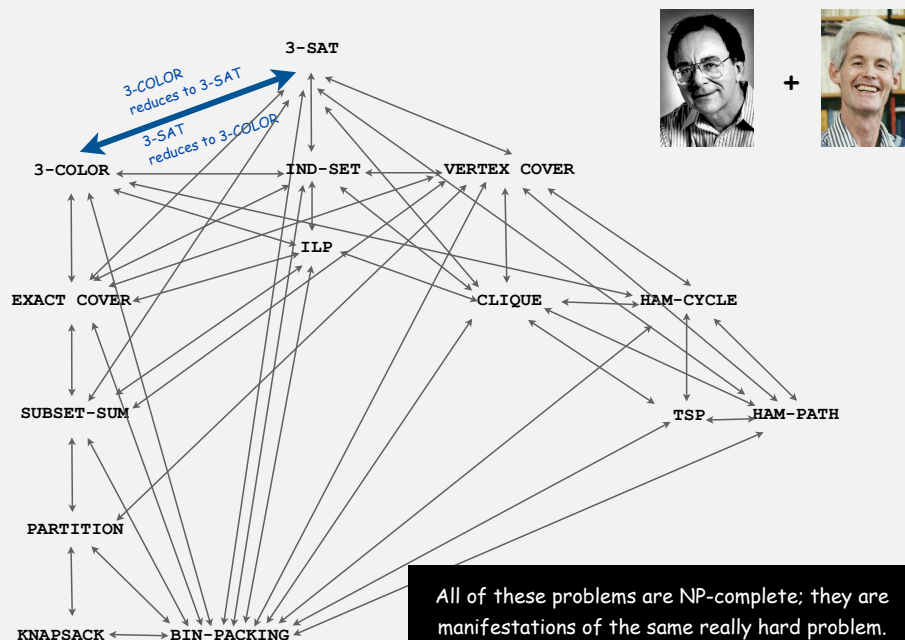
Implications of Cook's theorem



Stephen Cook
'82 Turing award

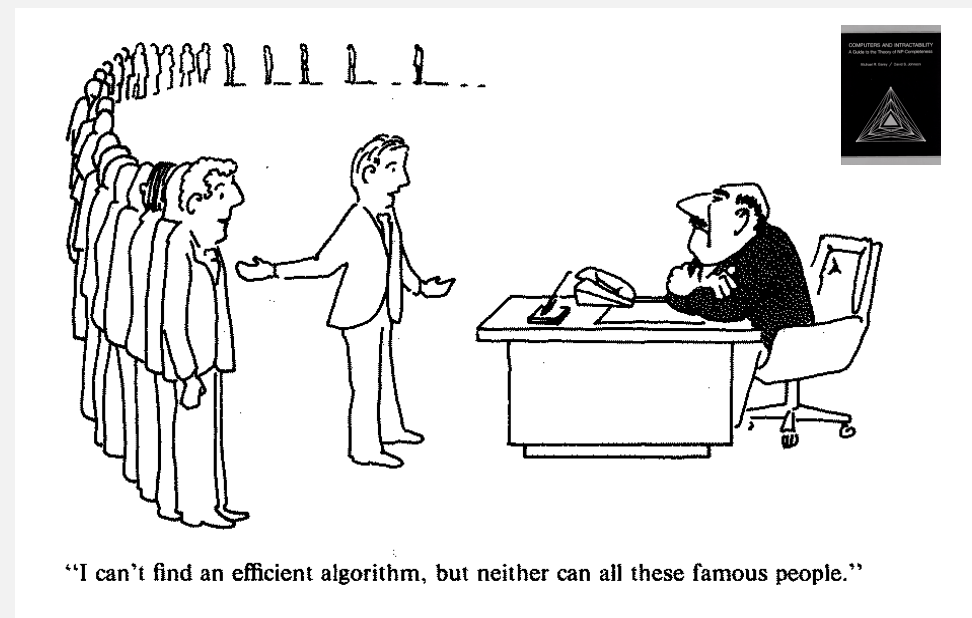
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Implications of Karp + Cook



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Implications of NP-completeness



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

complexity	order of growth	examples
linear	N	min, max, median, Burrows-Wheeler transform, ...
linearithmic	$N \log N$	sorting, convex hull, closest pair, farthest pair, ...
quadratic	N^2	???
	...	
exponential	c^N	???

Frustrating news. Huge number of problems have defied classification.

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

complexity	order of growth	examples
linear	N	min, max, median, Burrows-Wheeler transform, ...
linearithmic	$N \log N$	sorting, convex hull, closest pair, farthest pair, ...
3-SUM complete	probably N^2	3-SUM, 3-COLLINEAR, 3-CONCURRENT, ...
	...	
NP-complete	probably c^N	3-SAT, IND-SET, ILP, ...

Good news. Can put problems in equivalence classes.

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Summary

Reductions are important in theory to:

- Establish tractability.
- Establish intractability.
- Classify problems according to their computational requirements.

Reductions are important in practice to:

- Design algorithms.
- Design reusable software modules.
 - stack, queue, priority queue, symbol table, set, graph
 - sorting, regular expression, Delaunay triangulation
 - minimum spanning tree, shortest path, maximum flow, linear programming
- Determine difficulty of your problem and choose the right tool.
 - use exact algorithm for tractable problems
 - use heuristics for intractable problems

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