

CPS 196.2

Computational problems, algorithms, runtime,
hardness

(a ridiculously brief introduction to theoretical computer science)

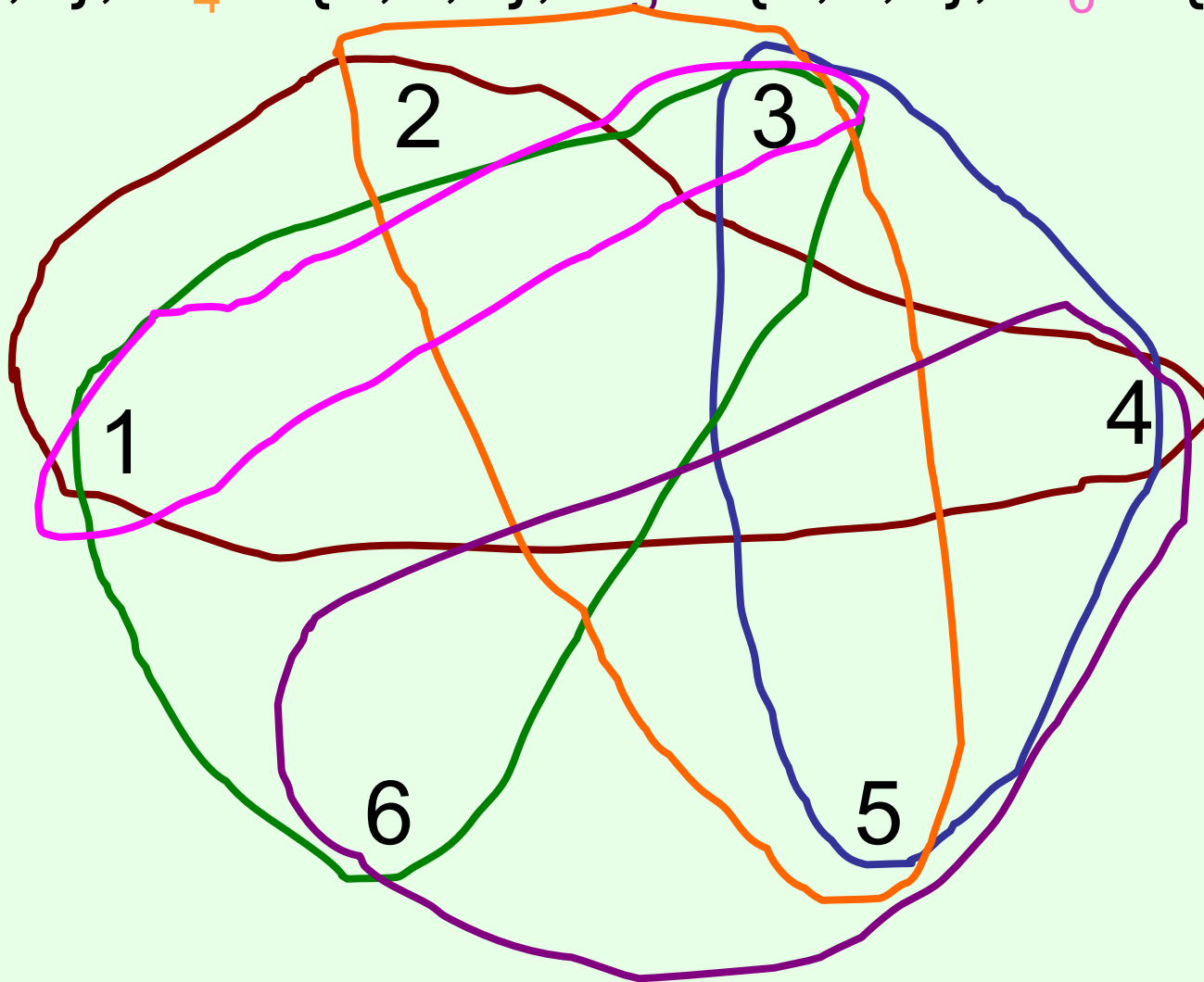
Vincent Conitzer

Set Cover (a computational problem)

- We are given:
 - A finite set $S = \{1, \dots, n\}$
 - A collection of subsets of S : S_1, S_2, \dots, S_m
- We are asked:
 - Find a subset T of $\{1, \dots, m\}$ such that $\bigcup_{j \in T} S_j = S$
 - Minimize $|T|$
- **Decision variant** of the problem:
 - we are additionally given a target size k , and
 - asked whether a T of size at most k will suffice
- One **instance** of the set cover problem:
 $S = \{1, \dots, 6\}$, $S_1 = \{1, 2, 4\}$, $S_2 = \{3, 4, 5\}$, $S_3 = \{1, 3, 6\}$, $S_4 = \{2, 3, 5\}$, $S_5 = \{4, 5, 6\}$, $S_6 = \{1, 3\}$

Visualizing Set Cover

- $S = \{1, \dots, 6\}$, $S_1 = \{1, 2, 4\}$, $S_2 = \{3, 4, 5\}$, $S_3 = \{1, 3, 6\}$, $S_4 = \{2, 3, 5\}$, $S_5 = \{4, 5, 6\}$, $S_6 = \{1, 3\}$



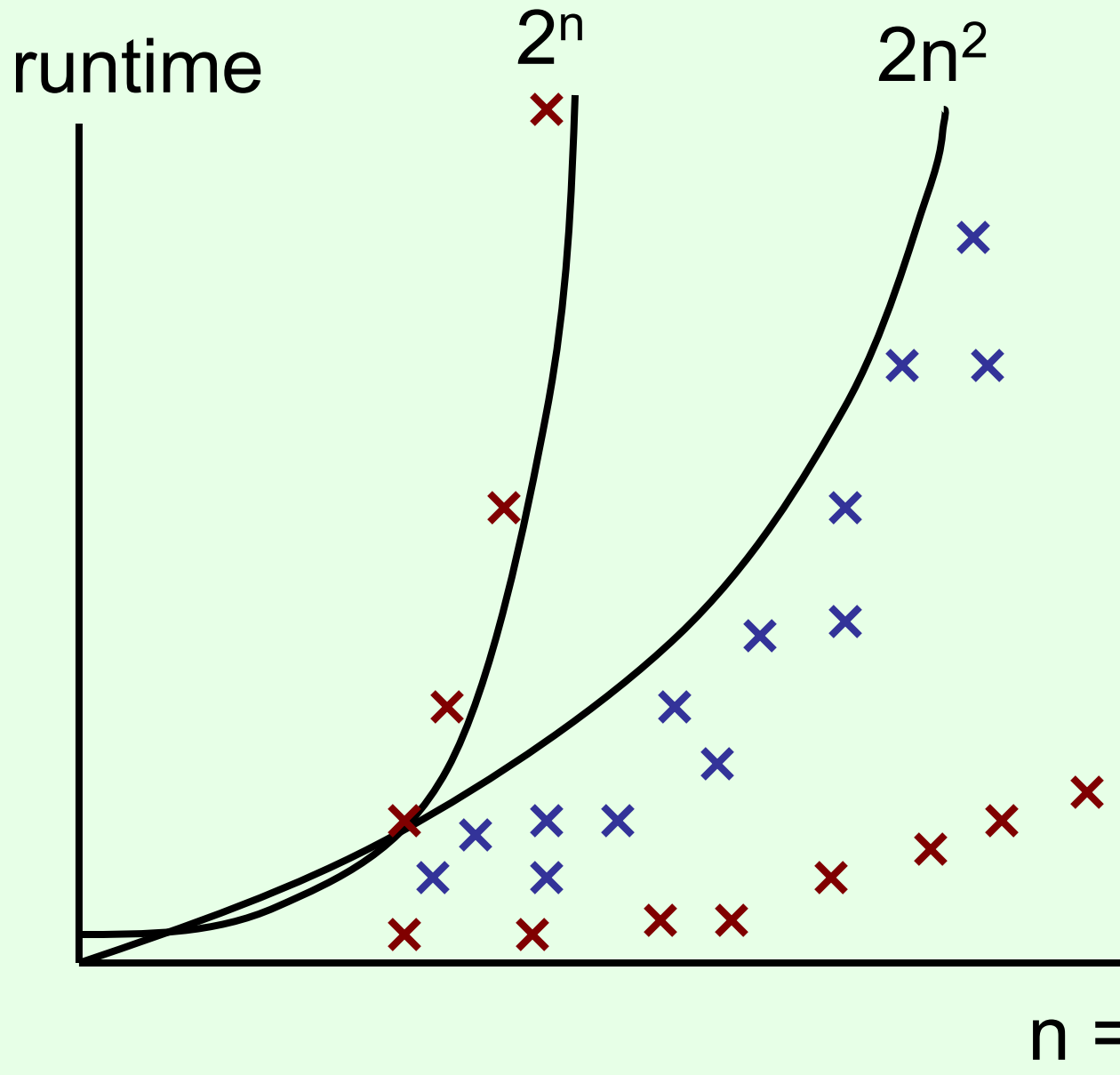
Algorithms and runtime

- We saw:
 - the **runtime** of glpsol on set cover instances increases rapidly as the instances' sizes increase
 - if we drop the integrality constraint, can scale to larger instances
- Questions:
 - Using glpsol on our integer program **formulation** is but one **algorithm** – maybe other algorithms are faster?
 - different formulation; different optimization package (e.g. CPLEX); simply going through all the combinations one by one; ...
 - What is “fast enough”?
 - Do (mixed) integer programs always take more time to solve than linear programs?
 - Do set cover instances **fundamentally** take a long time to solve?

Polynomial time

- Let $|x|$ be the **size** of problem instance x (e.g. the size of the file in the .lp language)
- Let a be an algorithm for the problem
- Suppose that for **any** x , $\text{runtime}(a,x) < cf(|x|)$ for some constant c and function f
Then we say algorithm a 's runtime is $O(f|x|)$
- a is a **polynomial-time algorithm** if it is $O(f(|x|))$ for some **polynomial** function f
- **P** is the class of all problems that have at least one polynomial-time algorithm
- Many people consider an algorithm **efficient** if and only if it is polynomial-time

Two algorithms for a problem



- × run of algorithm 1
- × run of algorithm 2

Algorithm 1 is $O(n^2)$
(a polynomial-time algorithm)

Algorithm 2 is not $O(n^k)$
for any constant k
(not a polynomial-time algorithm)

The problem is in P

Linear programming and (mixed) integer programming

- LP and (M)IP are also computational problems
- LP is in P
 - Ironically, the most commonly used LP algorithms are not polynomial-time (but “usually” polynomial time)
- (M)IP is not known to be in P
 - Most people consider this unlikely

Reductions

- Sometimes you can reformulate problem A in terms of problem B (i.e. **reduce** A to B)
 - E.g. we have seen how to formulate several problems as linear programs
- In this case problem A is **at most** as hard as problem B
 - Since LP is in P, all problems that we can formulate using LP are in P
 - Caveat: only true if the linear program itself can be created in polynomial time!

NP

- Recall: **decision problems** require a yes or no answer
- **NP**: the class of all decision problems such that if the answer is yes, there is a simple proof of that
- E.g. “does there exist a set cover of size k ?”
- If yes, then just show which subsets to choose!
- Technically:
 - The proof must have polynomial length
 - The correctness of the proof must be verifiable in polynomial time

P vs. NP

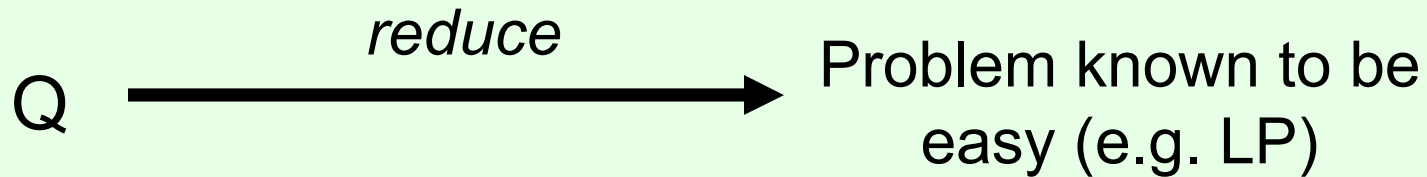
- **Open problem:** is it true that $P=NP$?
- The most important open problem in theoretical computer science (maybe in mathematics?)
- \$1,000,000 Clay Mathematics Institute Prize
- Most people believe P is not NP
- If P **were** equal to NP ...
 - Current cryptographic techniques can be broken in polynomial time
 - Computers can probably solve many difficult mathematical problems

NP-hardness

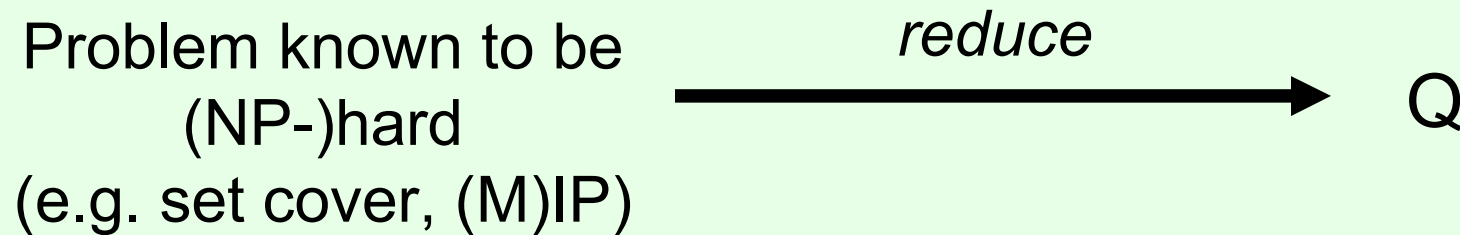
- A problem is **NP-hard** if the following is true:
 - Suppose that it is in P
 - Then $P=NP$
- So, trying to find a polynomial-time algorithm for it is like trying to prove $P=NP$
- Set cover is NP-hard
- Typical way to prove problem Q is NP-hard:
 - Take a known NP-hard problem Q'
 - Reduce it to your problem Q
 - (in polynomial time)
- E.g. (M)IP is NP-hard, because we have already reduced set cover to it
 - (M)IP is more general than set cover, so it can't be easier
- A problem is **NP-complete** if it is 1) in NP, and 2) NP-hard

Reductions:

To show problem Q is easy:



To show problem Q is (NP-)hard:

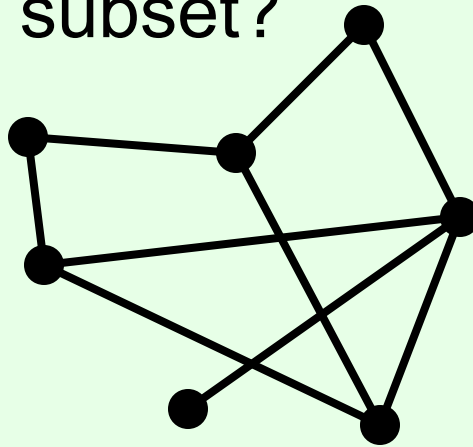


ABSOLUTELY NOT A PROOF OF NP-HARDNESS:



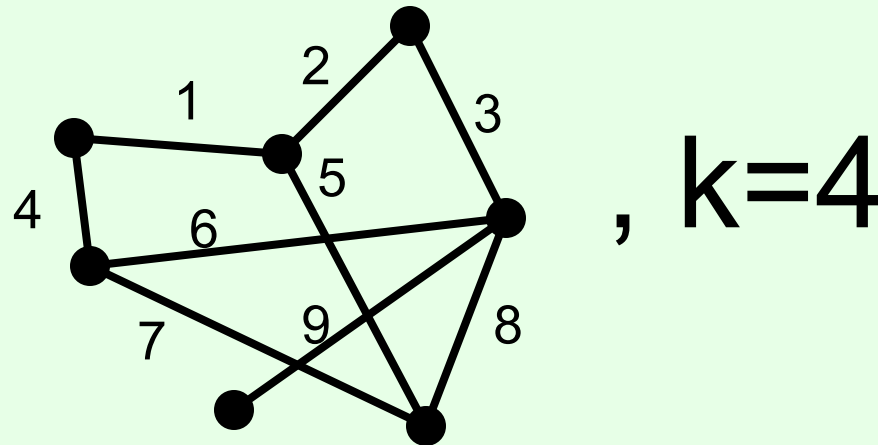
Independent Set

- In the below graph, does there exist a subset of **vertices**, of size 4, such that there is no **edge** between members of the subset?



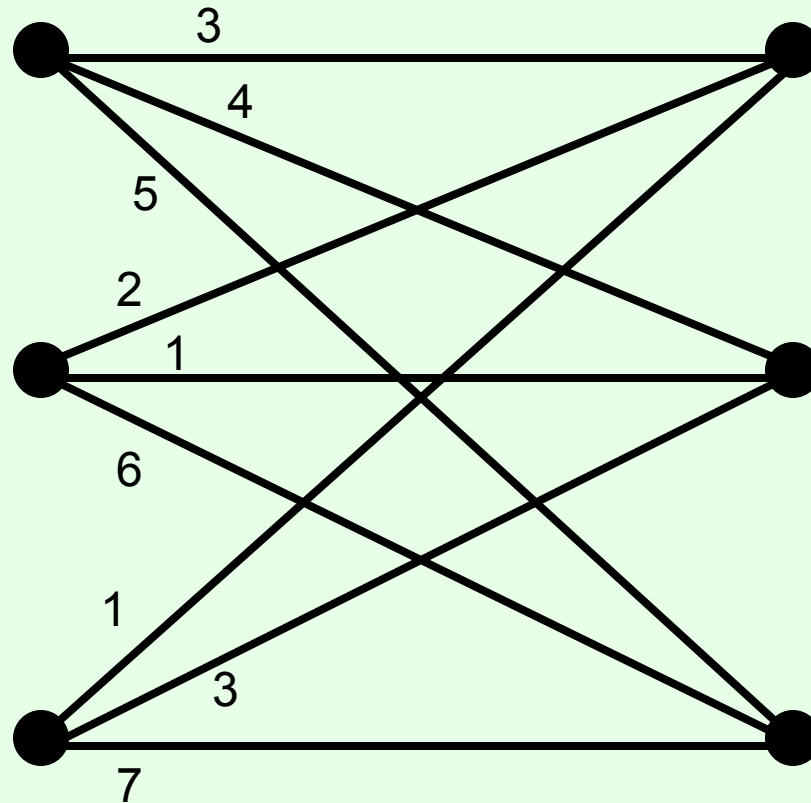
- General problem (decision variant): given a graph and a number k , are there k vertices with no edges between them?
- NP-complete

Reducing independent set to set cover



- In set cover instance (decision variant),
 - let $S = \{1,2,3,4,5,6,7,8,9\}$ (set of edges),
 - for each vertex let there be a subset with the vertex's adjacent edges: $\{1,4\}$, $\{1,2,5\}$, $\{2,3\}$, $\{4,6,7\}$, $\{3,6,8,9\}$, $\{9\}$, $\{5,7,8\}$
 - target size = #vertices - $k = 7 - 4 = 3$
- Claim: answer to both instances is the same (why??)
- So which of the two problems is harder?

Weighted bipartite matching



- Match each node on the left with one node on the right (can only use each node once)
- Minimize total cost (weights on the chosen edges)

Weighted bipartite matching...

- minimize $\sum_{ij} c_{ij} x_{ij}$
- subject to
- for every i , $\sum_j x_{ij} = 1$
- for every j , $\sum_i x_{ij} = 1$
- for every i, j , $x_{ij} \geq 0$

- Theorem [Birkhoff-von Neumann]: this linear program always has an optimal solution consisting of just integers
 - and typical LP solving algorithms will return such a solution

- So weighted bipartite matching is in P