NP and NP-completeness

CS240

Spring 2023

Rui Fan



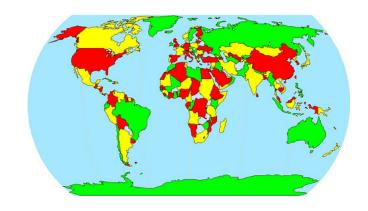
Limits of efficiency

- What is the fastest way to solve a problem?
 - □ E.g. sorting n numbers takes O(n log n) operations using mergesort.
 - □ Is there a different algorithm that sorts n numbers faster, say in O(n) time?
 - \square No. Sorting n numbers takes Ω (n log n) time if the algorithm can only compare numbers.



Limits of efficiency

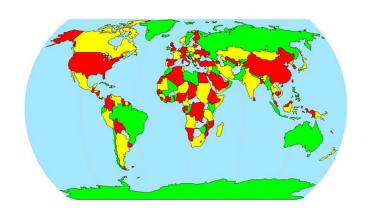
- How much time does it take to solve a more difficult problem, like coloring?
- Four Color Theorem says every map can be colored with 4 colors, s.t. adjacent regions have different colors.
- So given a map, we can always efficiently answer the question whether the map can be 4-colored. Namely, yes.
- But sometimes 3 colors are enough to color the map.
- Can we efficiently determine whether a map can be 3-colored?
- If there are n regions, we can find a 3coloring by trying all possible colorings.
 - ☐ There are 3ⁿ possible colorings.
 - □ There are 195 countries in the world, and 1.1×10^{93} possible colorings!





Limits of efficiency

- Is there a much more efficient algorithm?
- Nobody knows of one. And almost everybody thinks no such algorithm exists.
- But no one can prove it doesn't exist either.
- The theory of NP-completeness is a mathematical attempt to prove some problems have no efficient solutions.
 - So far, it's led to more questions than answers...
- We'll define P, NP, and NPcompleteness.



The class P

- A polynomial time (polytime) algorithm is one that runs in $O(n^k)$ time, for some constant k, when input has size n.
- P is the set of all problems that can be solved by a polytime algorithm.
 - □ These problems are called "efficiently computable", because a polytime algorithm is considered efficient.
 - □ In practice though, an e.g. $O(n^3)$ algorithm is quite slow, even for moderate sized n.
- If a problem takes $\omega(n^k)$ time, for any constant k, it's considered not efficiently solvable.
 - \square Ex An $\Omega(2^n)$ time or $\Omega(n!)$ time algorithm isn't efficient.
 - □ We only know how to 3-color a map in $\Omega(3^n)$ time (more or less), so 3-coloring (currently) can't be solved efficiently.
 - \square An $\Omega(3^n)$ time algorithm is much slower than an $O(n^3)$ algorithm.
 - **Ex** If n=10000, then $n^3 = 10^{12}$, but $3^n = 1.6 \times 10^{4771}$.



The class NP

- NP = Nondeterministic polynomial time.
- Def An instance of a problem consists of an input for the problem.
 - □ Ex An instance of the sorting problem is a set {3,1,2,4} that we want to sort.
 - Ex An instance of the SSSP problem is a weighted graph along with a source node.
- P is the class of problems for which all instances can be solved in polynomial time by some algorithm.
- NP is the class of problems for which the solvability of an instance can be verified in polynomial time.
 - ☐ The verification is done by a "verifier" algorithm.
 - □ The verifier needs an additional "hint" to work correctly.
 - The hint is also called a "witness" or "certificate".
 - ☐ The verifier doesn't find a solution to a problem instance, but only checks that the instance has been solved.



The class NP

- The verifier has the following properties.
 - The verifier's input is a problem instance x, and a certificate y.
 - □ The verifier's output is either "accept" or "reject".
 - □ If x has a solution, then if y is a "good" certificate, the verifier will output accept.
 - If y is not a "good" certificate, the verifier can either accept or reject.
 - □ If x has no solution, the verifier rejects no matter what y is.
 - \square Intuitively, the certificate y indicates x is solvable.
 - For example, y can be a solution to x.
 - But y can also be an indirect representation of a solution.
 - ☐ The verifier is efficient, i.e. runs in polynomial time.

The class NP, formally

- Def A decision problem is a problem with a yes / no answer.
 - □ Ex Given a graph, is there a path from node s to t?
 - □ Ex Given a map, is there a way to 3-color it?
 - Ex Given a number, is it prime?
- Def Given a decision problem, the set of yes (resp. no) instances are the instances of the problem for which the answer is yes (resp. no).
 - □ Ex 11 is a yes instance to the prime problem, 10 is a no instance.
- Def Given a decision problem A, a polynomial time verifier V for A is an algorithm that does the following
 - □ V's input is an instance x of A, and a certificate string y.

 - □ If x is a yes instance, there exists a y for which V outputs 1, i.e. $\exists y: V(x, y) = 1$.
 - □ If x is a no instance, every y makes V output 0, i.e. $\forall y$: V(x,y) = 0.
 - □ V runs in polynomial time.
- NP is the set of all decision problems with polytime verifiers.

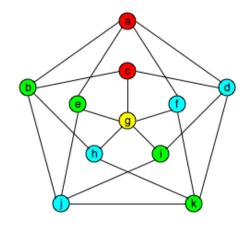


Showing a problem is in P or NP

- To show a problem is in P, give an algorithm solving the problem that runs in polynomial time.
- To show a decision problem is in NP, give a polynomial time verifier for the problem satisfying the properties on the previous slide.
 - ☐ This requires specifying what the cerficates are, and how the verifier operates, given an instance of the problem and a certificate.

4-coloring is in NP

- Given a graph, can we assign each vertex one of 4 colors, such that adjacent vertices have different colors?
- Verifier
 - \Box Certificate y is an assignment of colors to the vertices of graph x.
 - □ Check y uses at most 4 colors. If not, output no.
 - ☐ Go through all edges of x, and checks endpoints of each edge have different colors.
 - If true for all edges, output 1. Else output 0.
- If x is yes instance
 - ☐ Then x is 4-colorable.
 - So there's way to assign each vertex one of 4 colors s.t. endpoints of each edge have different colors.
 - □ Let y be this assignment, and give y to V.
 - □ Clearly V outputs 1.
- If x is no instance
 - Then x is not 4-colorable.
 - □ So no matter how we assign 4 colors to vertices of x, some edge has endpoints with the same order.
 - □ So V outputs 0, for any input y.
- V runs in polytime.
 - □ If x has n vertices, then it has $O(n^2)$ edges, so V runs in $O(n^2)$ time.





Factoring is in NP

- Given an integer x, does it have a factor $y \neq 1,x$.
- Verifier
 - Certificate y is a number.
 - \Box Check y divides x, and y \neq 1,x.
 - ☐ If so, output 1, else output 0.

999999866000004473 = 9999999929 x 999999937

- If x is yes instance
 - ☐ Then x has a nontrivial factor y.
 - ☐ Give y to V, and V outputs 1.
- If x is no instance
 - \square Then every factor of x is either 1 or x.
 - \square So for any y \neq 1,x given to V, V outputs 0.
- V runs in polytime.
 - □ Dividing x by y takes polynomial time.
- However, factoring does not seem to be in P.
 - \square Given an n digit number, there's no known way determine if it has a nontrivial factor in $O(n^k)$ time, for any constant k.

Traveling salesman is in NP

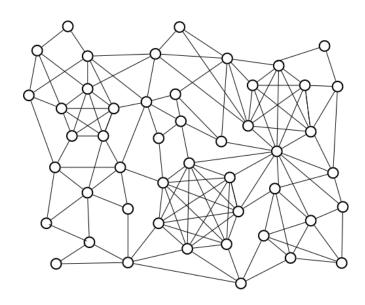
- Given a set of n cities, and distances between each pair of cities, is there a path visit each city exactly once, and has distance at most D, for a given D?
- Verifier
 - □ Certificate y is a path through the graph.
 - \square Check y goes through every vertex once, and total length of y is $\leq D$. If so, output 1, else output 0.
- If x is yes instance
 - □ Then there is a path going through each vertex once with total length $\leq D$.
 - □ Call the path y and give it to V.
 - ☐ Clearly V outputs 1.
- If x is no instance
 - Then no matter what path y you use, either y doesn't go through each city once, or y has length > D.
 - So V outputs 0, no matter what y it gets.
- V runs in polytime.
 - If the graph has n vertices, then all of V's checks can be done in O(n) time.





k-Clique is in NP

- Given a graph with n nodes and a number k, are there k nodes that form a clique, i.e. that are all connected to each other?
- Verifier
 - ☐ Certificate y is a set of k nodes in x.
 - Check each pair of the k nodes is connected by an edge. If so, output 1. Otherwise output 0.
- If x is yes instance
 - Then there are k nodes that are mutually connected.
 - ☐ Call this set y and give it to V.
 - ☐ Clearly V outputs 1.
- If x is no instance
 - Then in any set of k nodes, some 2 nodes aren't connected.
 - □ So V outputs 0, no matter what set of k nodes it gets.
- V runs in polytime.
 - Checking k nodes are mutually connected takes O(k²) time.



All problems in P are in NP

- Let A be a problem in P. I.e. there's a polytime algorithm S s.t. on every instance x of A
 - ☐ If x has a solution, S returns a solution.
 - ☐ If x has no solution, S returns fail.
- Verifier
 - □ V runs S. If S finds a solution, V outputs 1. Otherwise V outputs 0.
- If x is yes instance
 - □ S finds a solution, so V outputs 1.
- If x is no instance
 - ☐ S returns fail, so V outputs 0.
- V runs in polytime.
 - ☐ Because V just runs S, which runs in polytime.
- Notice that for problems in P, V doesn't need a certificate y.
 - ☐ For problems in P, it's easy to determine if they're solvable or not.
- But for hard problems (not in P), V isn't powerful enough to determine solvability by itself.
 - □ So it needs a hint / witness / certificate.
- Ex In factoring, a polytime verifier isn't powerful enough to find a nontrivial factor of an input.
 - But if it's given a nontrivial factor, it can check the factor works in polytime, and therefore verify the input is composite.

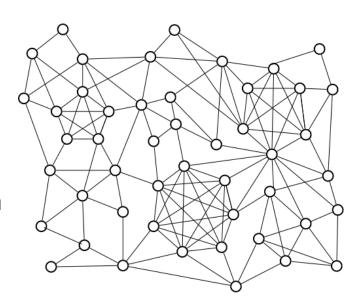
Primes is in NP

- Proving a problem is in NP isn't always so easy...
- Given a number x, is x prime?
- Verifier
 - □ What should the certificate y be?
 - \Box If y is a single number s.t. $y \nmid x$, then y doesn't certify that x is prime.
 - □ Suppose y a vector giving $\frac{x}{y}$ for every $y \le \sqrt{x}$.
 - □ V returns 1 if all these aren't integers, and 0 otherwise.
- If x is yes instance
 - ☐ I.e., x is prime. Then all the quotients are non-integer, so V returns 1.
- If x is no instance
 - □ Then x is composite, so x has a factor $y \le \sqrt{x}$, so $\frac{x}{y}$ is integer, and V outputs 0.
- V runs in polytime.
 - □ No it doesn't!
 - □ Say x has n digits. Then there are ~ $2^{n/2}$ numbers $\leq \sqrt{x}$, so y has size $O(n2^{n/2})$.
 - ☐ Since V has to check all values in y, it doesn't run in poly(n) time.
- So, this verifier is incorrect. This verifier does not show Primes is in NP.
- That doesn't mean Primes ∉ NP, it just means our verifier doesn't work.
- We can show Primes is in NP using another verifier and some number theory. This is called Pratt's Theorem, and is beyond our scope.



Incorrect verifiers

- We showed k-Clique is in NP by giving a correct verifier.
- Let's see some incorrect verifiers.
 - None of these verifiers can be used to prove k-Clique is in NP.
- Verifier 1 Always outputs 1, regardless of y.
 - □ Wrong, because when graph doesn't contain a k-clique, V is supposed to output 0.
- Verifier 2 Always output 0, regardless of y.
 - Wrong, because when the graph does contain a k-clique, V is supposed to output 1, for some y.
- Verifier 3 Check all subsets of k nodes. If any form a clique, output 1, else output 0.
 - Seems OK. When x has a k-clique, V outputs 1, and when x doesn't, it outputs 0.
 - But V is still wrong, because it doesn't run in polytime.
 - There are $O(n^k)$ subsets of k nodes, and V checks all of them.



P vs NP

- Does P=NP?
 - □ I.e. suppose there's a problem for which we can verify solvability in polynomial time. Does that mean we can actually find a solution in polynomial time?
- This is the arguably the most important question in computer science.
 - The other would be to produce general Al.
- Many real-world problems are in NP. If P=NP, we can solve them efficiently. If P≠NP, then we can't.
- Every P problem is in NP, as we saw. So $P \subseteq NP$.
- Is every NP problem in P, i.e. $NP \subseteq P$?
- After 50 years, nobody knows.
 - Most, but not all researchers think not all NP problems are in P.
 - ☐ There are probably problems we can efficiently verify but not efficiently solve.
 - □ Ex Factoring is something we can efficiently verify, but not solve.
- If you can prove $P \neq NP$, or even better, P = NP, then
 - □ you \geq Newton \geq Einstein \geq ...
 - ☐ You also get \$1M from the Clay Math Institute.
- Answering this question has vast and profound implications for CS, AI, math, physics, etc.

NP-completeness

- Out of all the NP problems, there's a subset of NP problems called NP-complete (NPC) problems that are the "hardest" NP problems.
- To determine whether P=NP, it suffices to know whether P=NPC.
 - □ If the hardest problems can be solved in polytime, then all NP problems can be solved in polytime. I.e. P=NP.
- So the study of P vs NP focuses on NPC problems.

100

Hardness and reductions

- What does it mean to say problem B is harder than problem A?
- It means if you can solve B, you can also solve A.
 - □ Ex Algebra is harder than arithmetic, because if you can do algebra, you can also do arithmetic.
 - □ So if I have an algorithm for solving B, I can use it to solve A.
- We say A reduces to B.
 - \square Write $A \leq_R B$.
 - □ Read this as "A is equally or less difficult than B".

Example

- FACTOR-ALL(n) finds all the factors of a number n.
- FACTOR-1(n) finds one factor.
- Of course, FACTOR-1 \leq_R FACTOR-ALL.
 - □ If we can find all the factors, we can certainly find one.
- FACTOR-ALL \leq_R FACTOR-1.
 - □ We use FACTOR-1(n) to find one factor m of n.
 - ☐ Then divide n by m, and run FACTOR-1 on the result, to find another factor of n.
 - Keep repeating the previous steps until we get all the factors.
- The hard part about factoring a number, is just to find one factor.
 - □ Since FACTOR-1 \leq_R FACTOR-ALL, FACTOR-ALL \leq_R FACTOR-1, these problems have the same hardness.

Reductions, formally

- Let A and B be two decision problems.
- Let X and Y be the set of yes instances for A and B, resp.
- Ex Say A = PRIME and B = k-CLIQUE.
 - □ X is the set of prime numbers.
 - ☐ Y is the set of graphs containing a k-clique.
- Let f be a function that maps instances of A to instances of B.
- Def A reduces to B if there exists $f: A \to B$ s.t. for all instances x of A, $x \in X \Leftrightarrow f(x) \in Y$.
 - \square We write $A \leq_R B$.
- To show $A \leq_R B$, just give the mapping f.
- If $A \leq_R B$, then we can use an algorithm for B to solve A.
 - □ To solve an instance of A, first map it to an instance of B using f.
 - ☐ Then run the B algorithm.
 - □ Return the same answer for A as the B algorithm gives.
 - \square By definition, A is true \Leftrightarrow f(A) is true.



Example

- Suppose we want to show PRIME \leq_R k-CLIQUE.
- This means there's some mapping f such that.
 - ☐ Given an instance of PRIME, i.e. a number n.
 - \Box f(n) is an instance of k-CLIQUE, i.e. f(n) is a graph G.
 - \square n is prime if and only if f(n) contains a k-clique.
- If we have an algorithm to solve k-CLIQUE, we can use it solve PRIME.
 - □ To tell if n is prime, map n to a graph G and run the k-CLIQUE algorithm on G.
 - ☐ If it returns true, n is prime. Otherwise n isn't.

Polynomial time reductions

- If the mapping function from A to B runs in polynomial time, then it's a polynomial time reduction, and we write $A \leq_P B$.
 - □ Ex If we're reducing PRIME to k-CLIQUE, then the function to generate a graph from a number must run in polytime.
- Thm 1 Let A, B and C be three problems, and suppose $A \leq_P B$ and $B \leq_P C$. Then $A \leq_P C$.
- Proof Since $A \leq_P B$, there's a polytime mapping f from instances of A to instances of B.
 - □ Since $B \leq_P C$, there's a polytime mapping g from instances of B to instances of C.
 - \square Given an instance X of A, let Y = f(X), and Z = g(Y) = g(f(X)).
 - □ Then X is a yes instance of A \Leftrightarrow Y is a yes instance of B \Leftrightarrow Z is a yes instance of C.
 - \square So $g \circ f$ is a valid mapping of A to C.
 - \square Since f and g are both polytime, $g \circ f$ is also polytime.

NP-completeness

- Def A problem A is NP-complete (NPC) if the following are true.
 - $\square A \in NP$.
 - \square Given any other problem $B \in NP$, $B \leq_P A$.
- Thus, a NP-complete problem is an NP problem that can be used to solve any other NP problem.
 - □ It's a "hardest" NP problem.

NP-completeness and SAT

- Do NP-complete problems really exist?
 - Can we really find an NP problem that can be used to solve every other NP problem?
 - □ One problem to rule them all?
- Yes! Steve Cook and Leonid Levin proved around 1970 that SAT is NP-complete.
- SAT = satisfiable Boolean formulas.
 - ☐ Given a Boolean formula, is there any setting for the variables which makes the formula true?
 - $\square \, \, \mathsf{Ex} \, (A \vee B \vee \neg C) \wedge (A \vee \neg B \vee C) \wedge (\neg A \vee \neg B \vee \neg D) \in SAT.$
 - Setting A=B=C=true, D=false makes the formula true.
 - $\square \to A \land \neg A \notin SAT$.
 - The formula's false for all settings of A.

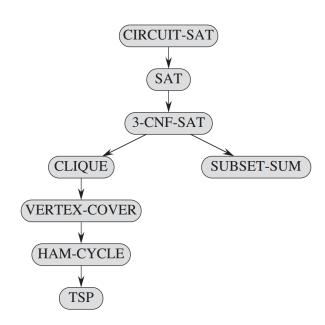
NP-completeness and SAT

- Cook-Levin theorem says 2 things.
 - □ SAT∈NP.
 - Prove this yourself.
 - \square Every NP problem reduces to SAT. I.e. every problem A in NP can be mapped to an SAT formula ϕ in polytime, such that
 - If A is true, then ϕ is satisfiable.
 - If A is false, then ϕ is not satisfiable.
- Basic idea of the theorem is to use the logical operations in a SAT formula to emulate the logical operations in any algorithm.
 - \square Any NP problem X has a polytime verifier V. The Cook-Levin theorem uses a SAT formula ϕ to emulate the verifier's operations.
 - □ For a yes instance of X, there's some certificate making V return 1.
 - The certificate can be transformed to a satisfying truth setting for ϕ .
 - \square Any certificate making V return 0 corresponds to a non-satisfying truth setting for ϕ .
 - \square So $\phi \in SAT$ if and only if X is a yes instance, and $X \leq_P SAT$.



The web of NP-completeness

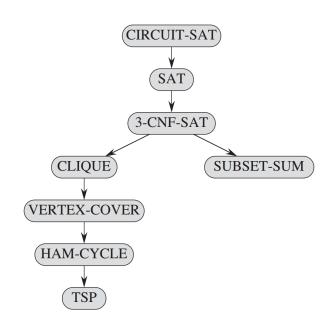
- For every problem in the picture, if A points to B, it means $A \leq_P B$.
 - □ So A can be solved using B.
- CIRCUIT-SAT was the original problem that Cook-Levin proved was NP-complete.
- So every problem in NP can be solved using CIRCUIT-SAT.
- But CIRCUIT-SAT can be solved using SAT, because CIRCUIT-SAT ≤_PSAT.
 - So every problem in NP can be solved using SAT.
 - ☐ So SAT is also NP-complete!
- SAT can be solved using 3-CNF-SAT.
 - So every NP problem can be solved using 3-CNF-SAT.
 - So 3-CNF-SAT is also NP-complete.
- All problems in the diagram are NP-complete.
- Of course, each of the reductions requires a proof, which is sometimes tricky.
 - □ We'll see some reduction proofs next lecture.
- There are thousands of other NPC problems.





The web of NP-completeness

- Thm Given two NP problems A and B, suppose A is NP-complete, and $A \leq_P B$. Then B is also NP-complete.
- Proof Let C be any NP problem. Then $C \leq_P A$, since A is NP-complete.
 - □ Since $A \leq_P B$, then by Theorem 1, we have $C \leq_P A \leq_P B$.
 - □ Since also $B \in NP$, then B is NPC.
- To prove a problem B is NP-complete
 - □ Take a problem A you know is NPC, and prove $A \leq_P B$.
 - □ E.g., A can be any problem in the previous diagram.
 - □ To prove $A \leq_P B$, you need to give a polytime reduction from A to B.
 - This can sometimes be quite challenging.
 - □ You also have to prove $B \in NP$, but that's usually not hard.

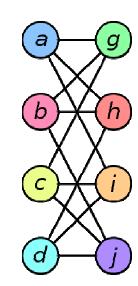


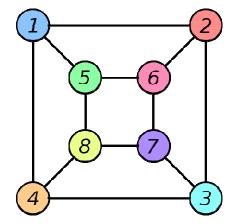
NP-completeness and P vs NP

- Thm 2 Suppose a problem A is NP-complete, and $A \in P$. Then P=NP.
- Proof Consider any other NP problem B. We'll show $B \in P$.
 - ☐ Since *A* is NPC, there's a polytime mapping f from B to A.
 - ☐ Given an instance X of B, run f on X to get an instance Y of A.
 - \square Since $A \in P$, there's a polytime algorithm g to solve A.
 - \square Run g(Y), and return the same answer for X.
 - \square By the definition of \leq_P , g(Y) is true \Leftrightarrow X is true.
 - ☐ Running f and g both take polytime. So we can solve B in polytime.
- Cor Suppose a problem A is NP-complete, and $A \notin P$. Then for any NP-complete problem B, $B \notin P$.
 - □ If $B \in P$, then since B is NPC, we have P = NP by Theorem 2. So since $A \in NP$, we have $A \in P$, a contradiction.
- To prove $P \neq NP$ (which is what most people think), it's enough to show one NPC problem is not solvable in polytime, by the corollary.
 - □ But after 50 years, no one has any such proof.
 - Nor has anyone shown a polytime algorithm for any NPC problem.

Beyond NP

- NP includes many important and practical problems.
- But not all problems are in NP.
- In the graph isomorphism (GRAPH-ISO) problem, we ask whether two graphs simply relabelings of each other.
 - I.e. Given two graphs G = (V, E) and G' = (V, E'), is there a permutation $f: V \to V$ s.t. $(u, v) \in E \Leftrightarrow (f(u), f(v)) \in E'$.
 - This is in NP, because on "yes" instances, we just give the relabeling to the verifier.
- The graph non-isomorphism (GRAPH-NONISO) problem is the opposite: are two graphs really different, and not relabelings of each other.
 - □ We don't know if GRAPH-NONISO \in NP.
 - If two graphs are really different, how do we produce a certificate to prove this to a polytime verifier?
 - □ We could give the verifier a list of all possible relabelings, and show the graphs are different under each.
 - But this isn't polytime because there are n! relabelings.
 - \square We also don't know if GRAPH-NONISO \in P.





More complexity classes

- co-NP All problems whose "complement" is in NP.
 - □ E.g. GRAPH-ISO∈ NP, so GRAPH-NONISO∈ co-NP.
- PSPACE All problems whose computation takes polynomial amount of space.
 - □ Includes all problems in P.
- EXPTIME All problems whose computation takes at most exponential amount of time.
 - □ Includes all problems in P and NP.
- NEXPTIME All problems whose correct answer can be verified in at most exponential amount of time.
 - □ NEXPTIME is to EXPTIME what NP is to P.
- Each of these is called a complexity class.
 - □ Many, many other complexity classes, some very obscure.
 - □ "Complexity zoo": https://complexityzoo.uwaterloo.ca/Complexity Zoo



Complexity theory

- A central goal of complexity theory and theoretical computer science is to study the relationship between complexity classes.
- We know some trivial things, like P⊆ NP, or NP⊆EXPTIME.
- We know a few nontrivial things, like PSPACE=NPSPACE=IP, and NL=coNL.
- Beyond this, we really know hardly anything!
- P?=NP, P ?=co-NP, NP?=co-NP, NP?=PSPACE, NP?=EXPTIME.
- In the last 50 years, we haven't gotten much closer.
- Maybe our techniques are wrong?
- Understanding any of the relationships would have many profound implications.

