# NP and NP-completeness

CS240

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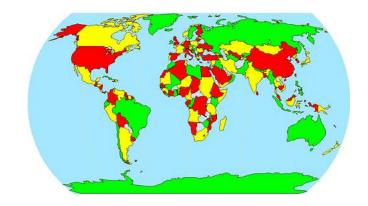


### Limits of efficiency

- 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?
  - □ No. Sorting n numbers takes  $\Omega$ (n log n) time if the algorithm can only compare numbers.

# Limits of efficiency

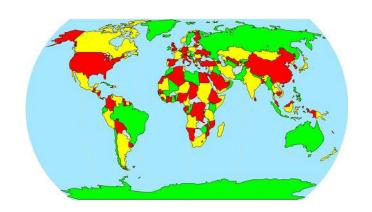
- How much time does it take to solve a more difficult problem, like coloring?
- Four Color Theorem says every map can be colored with 4 colors, s.t. adjacent regions have different colors.
- So given a map, we can always efficiently answer the question whether the map can be 4-colored. Namely, yes.
- 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 3<sup>n</sup> possible colorings.
  - □ There are 195 countries in the world, and  $1.1 \times 10^{93}$  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.
  - □ 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.

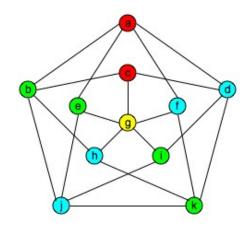


#### Showing a problem is in P or NP

- 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
  - 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 ≠ 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

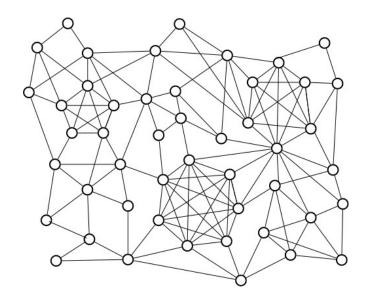
- Given a set of n cities, and distances between each pair of cities, is there a path visit each city exactly once, and has distance at most D, for a given D?
- Verifier
  - □ Certificate y is a path through the graph.
  - $\square$  Check y goes through every vertex once, and total length of y is  $\leq D$ . 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 ever	Ъ
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.

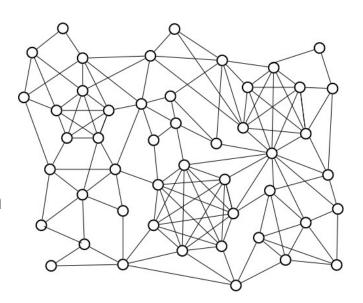
#### Primes is in NP

- 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!
  - $\square$  Say x has n digits. Then there are  $\sim 2^{n/2}$  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 ≥ Newton ≥ Einstein ≥ ...
  - □ 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.

## b/A

#### 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".

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#### 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.

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#### 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.
  - □ 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.

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## 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.
  - □ 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.
  - □ 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.

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#### NP-completeness

- Def A problem A is NP-complete (NPC) if the following are true.
  - $\square A \in NP$ .
  - □ 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.

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## 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.

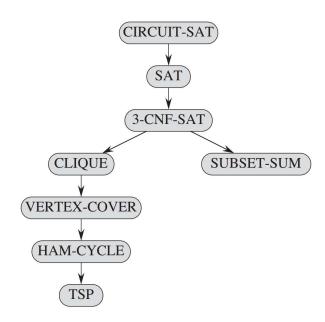
# NP-completeness and SAT

- 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  $\phi$  in polytime, such that
    - If A is true, then  $\phi$  is satisfiable.
    - If A is false, then  $\phi$  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  $\phi$  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  $\phi$ .
  - $\square$  Any certificate making V return 0 corresponds to a non-satisfying truth setting for  $\phi$ .
  - □ So  $\phi \in SAT$  if and only if X is a yes instance, and  $X \leq_P SAT$ .



## The web of NP-completeness

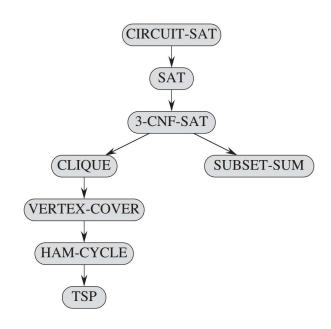
- For every problem in the picture, if A points to B, it means  $A \leq_P B$ .
  - □ So A can be solved using B.
- CIRCUIT-SAT was the original problem that Cook-Levin proved was NP-complete.
- So every problem in NP can be solved using CIRCUIT-SAT.
- But CIRCUIT-SAT can be solved using SAT, because CIRCUIT-SAT ≤<sub>P</sub>SAT.
  - 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  $A \le_P B$ . 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.

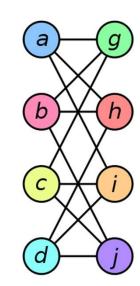


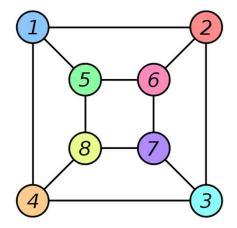
# NP-completeness and P vs NP

- 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.
  - □ Since  $A \in P$ , there's a polytime algorithm g to solve A.
  - $\square$  Run g(Y), and return the same answer for X.
  - □ 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.
  - □ We also don't know if GRAPH-NONISO  $\in$  P.





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### 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.net/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.

