

Lecture 21: Complexity Theory

Announcements:

- The final approaches fast!
- Three HW's: forkjoin, concurrency, complexity (all due by Wednesday afternoon!)
- Monday: no lecture, but instead time to work on your HW's/study hall.
 - I will hold off Monday instead of Friday.
- Wednesday: no lecture, still more study hall
 - I can go through P/NP section material if people are interested.
- We will try to grade HW's by Wednesday night, so you have feedback in time.
- Complexity concept check due Monday (instead of Friday)
- No section next week (day of final)
- Final Thurs 5-7 pm Sieg 134

This quarter we focused on fast algorithms and the data structures behind them.

- We could've used a linked list as our dictionary (slow)
- instead we saw how to use AVL trees or hashing to get faster runtimes!

In general, computer scientists often divide problems into two types

- ① "easy" problems: computer can solve in time $O(n^c)$ ($c = \text{constant}$)
- ② "hard" problems: computer cannot solve in time $O(n^c)$

n	$\Theta(n)$	$\Theta(n \log_2 n)$	$\Theta(n^2)$	$\Theta(n^3)$	$\Theta(1.5^n)$	$\Theta(2^n)$	$\Theta(n!)$
10	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
30	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	10^{25} years
50	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
100	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	10^{17} years	very long
1,000	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
10,000	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
100,000	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
1,000,000	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

complexity theory's goal: prove that certain problems are hard (for computers)

What is a problem? It is simply a set of inputs/outputs

ex 1 | Minimum spanning tree

input = graph output = any MST

ex 2 | "tell me if this list is sorted"

input = list $\in \mathbb{R}^n$ output = $\{0, 1\}$ ($\{ \text{yes, no} \}$)

ex 3 | "sort this array"

input = list output = list.

Def | An instance is a specific input to a problem.

Complexity Theorists focus on decision problems (output is "yes" or "no")

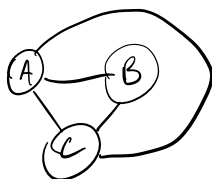
so instead of asking "find shortest path"
"is there a path of length $\leq k$?"

(this is because if you can solve the "decision version" then you can solve the "search version" :)

Examples !!!

① Eulerian Path : "Does G have a path that uses each edge exactly once?"

Theorem $\exists$ Eulerian path iff exactly 0 or 2 vertices have odd degree



intuition: Suppose every vertex is even degree

ALG: starting at u , keep following edges you haven't used yet, until you cannot anymore.

claim This process stops at u .

$\Rightarrow$ Eulerian Path is **EASY**: $\exists O(|V| + |E|)$ algo solving it (why?)

② Hamiltonian Path: "Does G have a path that touches every vertex exactly once?"

This seems like a similar problem!

ALG: for each permutation of the $|V|$ vertices, check if that path exists

runtime?

better algorithm?

... we think Hamiltonian Path is **HARD**

Def P is the class of all problems solvable in time $O(n^c)$ where c is constant

"solvable" = $\exists$ an algorithm with this runtime

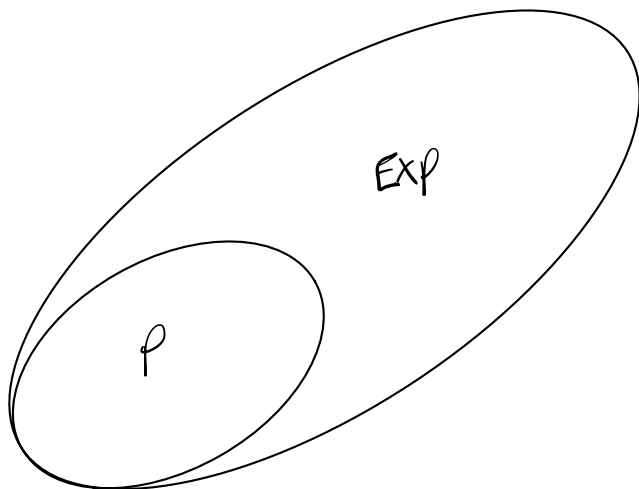
P stands for "polynomial time". It is a set of problems.

Def EXP is the class of problems that can be solved in $O(2^{n^c})$ time, $c = \text{constant}$.

questions: 1) is Eulerian Path $\in P$?

2) is Hamiltonian Path $\in P$?

3) If $f \in P$ then $f \in EXP$?



Def NP is the class of problems for which YES instances have a "proof" or "certificate" that can be verified in polynomial time $O(n^c)$, $c = \text{const.}$

more formally: $f \in NP$ iff $\exists$ algorithm $A(x, y)$ and polynomials p and q , s.t.

① $\forall x \in f^{-1}(1)$, $A(x, y)$ runs in time $p(|x|)$

- ② $\forall x \in f^{-1}(1) \exists y, |y| \leq q(|x|), \text{ s.t. } A(x,y) = 1.$
 ③ $\forall x \in f^{-1}(0) \text{ and } \forall y \text{ of length } \leq q(|x|), A(x,y) = 0.$

examples: **Hamiltonian path** $\in NP$ because if G has a Hamiltonian path, then the "proof" can be the path itself
 the "verifier" checks that the path its given

- ① is a valid path
- ② touches each vertex once

3-COLOR: "can I color vertices of G with ≤ 3 colors such that all neighbors are colored differently?"

3-COLOR $\in NP$: proof = assignment of colors to vertices

verifier = checks if ① only 3 colors used

② each edge has endpoints colored differently.

2-COLOR: "can I color vertices of G with ≤ 2 colors such that all neighbors are colored differently?"

2-COLOR $\in P$

ALG: run BFS. If layer of vertex is even, color 0

If layer is odd, color 1

(if you must color vertex with a different color than it already has, output NO)

Theorem **$P \subseteq NP$**

proof | let $f \in P$. So there exists an algorithm that runs in polynomial time deciding f .

if $f(k) = 1$, then the "proof" is empty, and the verifier

Just runs the algorithm to see whether $f(x) = 1$.

ex | "Is there an MST of weight $\leq k$ " $\in NP$

① we could have the "proof" be a MST of weight $\leq k$
and the "verifier" check that the proof a) is a tree, and
b) has cost $\leq k$.

② we could also have the "proof" be nothing at all,
and let the verifier run Prim's / Kruskal's algo.

Theorem | $NP \subseteq EXP$

proof Let $f \in NP$. given x , consider the following algorithm:

ALG: ① write down all possible "proofs" of length $\leq n^c$ that $f(x) = 1$

② run verifier on all these proofs

if verifier succeeds on any, output 1. Otherwise, output 0.

runtime of ALG?

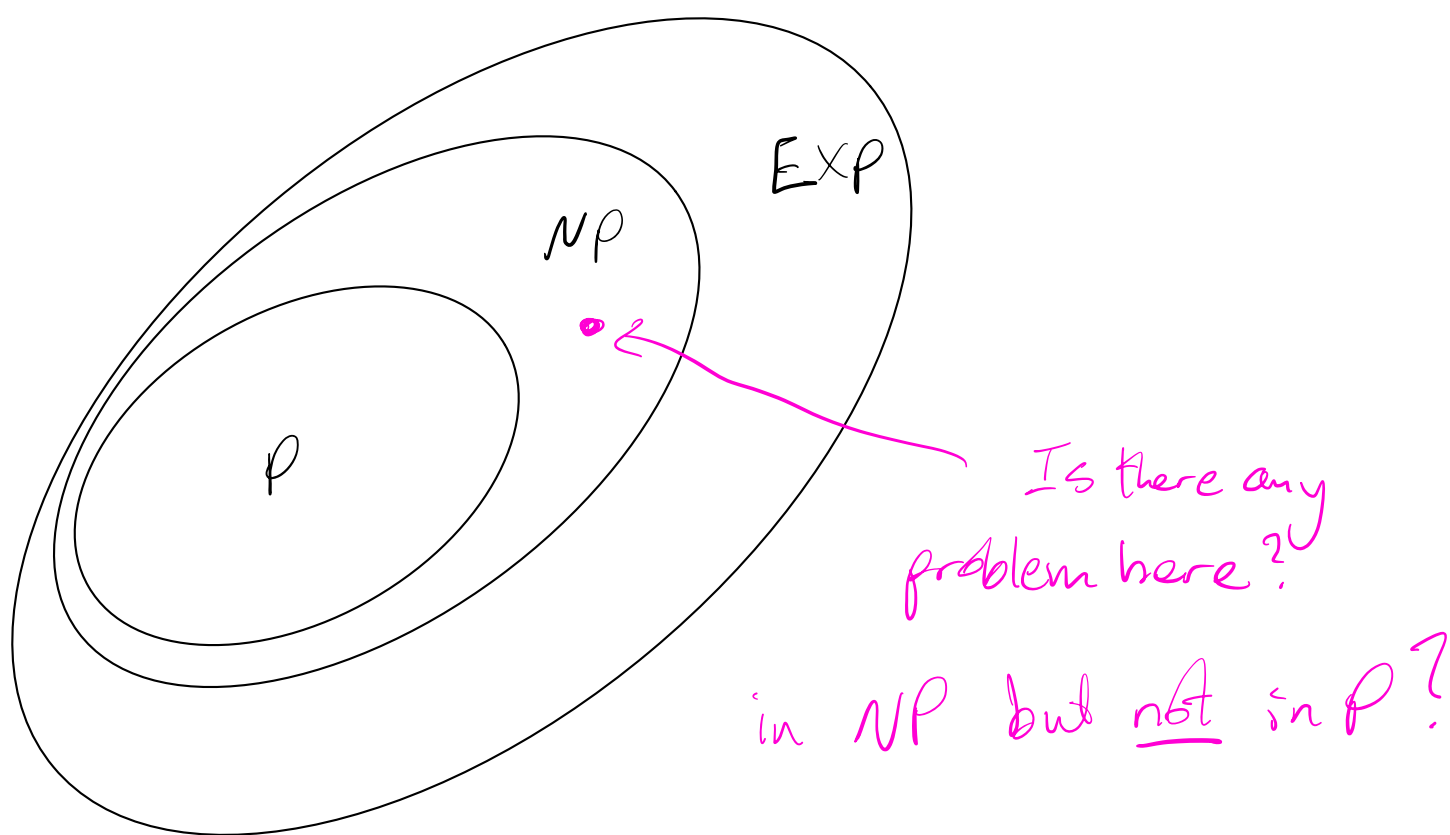
$\Rightarrow f \in EXP$

QED

BIG COMPLEXITY THEORY QUESTION: $P \stackrel{?}{=} NP$

intuitively: If I can "verify" a good solution quickly, can I use this to come up with a fast algorithm myself?

... most people think the answer is no.



One more fun example

"is there MST of weight $\leq k$ " $\in P$

"is there a walk that visits each vertex exactly once, and returns to start, and has weight $\leq k$ " $\in NP$

electric company can optimize its grid, but Amazon doesn't know

how to optimally route it's drivers !! (without solving P vs. NP, that is)

REDUCTIONS

Def We say that problem A is polynomial time reducible to B if there is an algorithm which, using a (hypothetical) polynomial time algorithm for B , solves A in polynomial time.

notation: $A \leq_p B$

intuition: A is no harder to solve than B , because if we can solve B , then the reduction tells us we can also solve A !

Warning: Don't get it backwards!

Silly example: Suppose $A =$ I have ants in my kitchen, I want them gone.
 $B =$ I blow up my house, and rebuild it.

Then $A \leq_p B$, because I can run my algo for B , and it gets rid of the ants.

A is no harder than B .

another example | $3\text{-COLOR} \leq_p 4\text{-COLOR}$

ALG: on input G , add a new vertex, and connect it to all other vertices
call this graph G'
return $4\text{-COLOR}(G')$

need to check: 1) if G 3-colorable, then G' 4-colorable
2) if G not 3-colorable, then G' not 4-colorable
3) ALG runs in polynomial time, assuming 4-COLOR does

} correctness
] poly-time

Def | A is **NP-complete** iff ① $\forall B \in NP, B \leq_p A$
② and $A \in NP$.

motivation: these are the "hardest" problems in NP

so if we want to show $P \neq NP$, these would be the best candidates!
(most likely not to have a poly time algo)

Def | A is **NP-hard** iff $\forall B \in NP, B \leq_p A$.

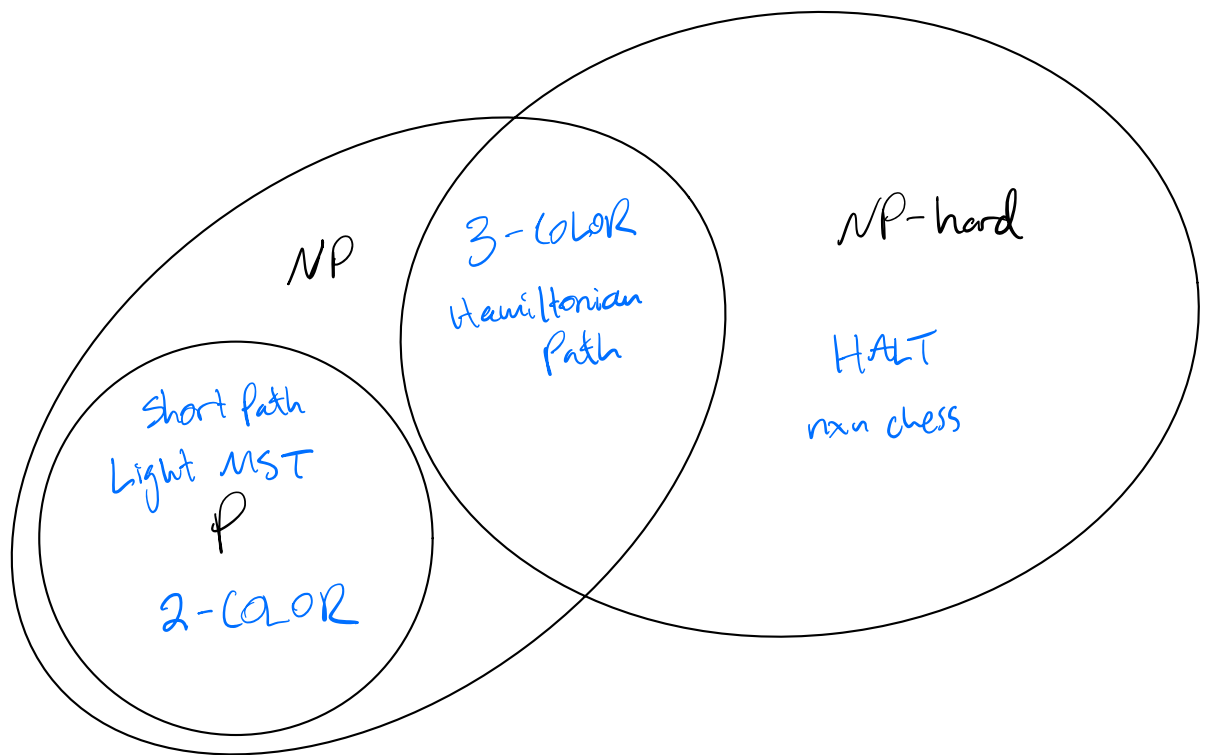
Remark: many problems are NP hard, but not in NP

ex] halting problem is NP-hard (but certainly not NP-complete!)

$n \times n$ chess: is there an optimal strategy for white?

not easy to verify!

What we think the world looks like



if $P=NP$, then P "grows" and "consumes" all of NP.

History / Why You Should Care