

Lecture 22: Complexity 2

Task: factor 65941

ANNOUNCEMENTS

- 1) +2 points on final for filling out official course evals
• due Sunday Aug 23

Task 2: verify $65941 = 23 \cdot 47 \cdot 61$

factoring $\in NP$!

If $P = NP$:

- All emails immediately hacked
encryption based on assumption that factoring is hard!
- You get \$1,000,000 from Clay Math Institute.
- put mathematicians out of work.

Is $P \neq NP$:

- Says something fundamental about our universe!
- For many questions, there is no clever alg/solution

Reminders:

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, $c = \text{constant}$.

Def NP is the class of ^{decision} 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: $L \in NP$ iff $\exists$ algorithm $V(x, y)$ and polynomials p and q , s.t.

① $\forall x, y$, $V(x, y)$ runs in time $p(|x|)$

$V =$ "verifier"

② $\forall x \in L(1) \exists y, |y| \leq q(|x|)$, s.t. $V(x, y) = 1$.

$y =$ "proof/certificate"

③ $\forall x \in L^c(1)$ and $\forall y$ of length $\leq q(|x|)$, $V(x, y) = 0$.

I kept saying: "If Hamiltonian path / 3-COLOR / ... has a polynomial time algorithm, then $P = NP$ "

Why is this true? Why doesn't that simply mean that specific problem is in P , and nothing more?

To understand why this is, need to understand reductions

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!

another example | On your homework, you showed a reduction

"shortest path w/ ^(small) integer weights" $\leq_p$ "BFS"

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 $\forall B \in NP, B \leq_p A$.

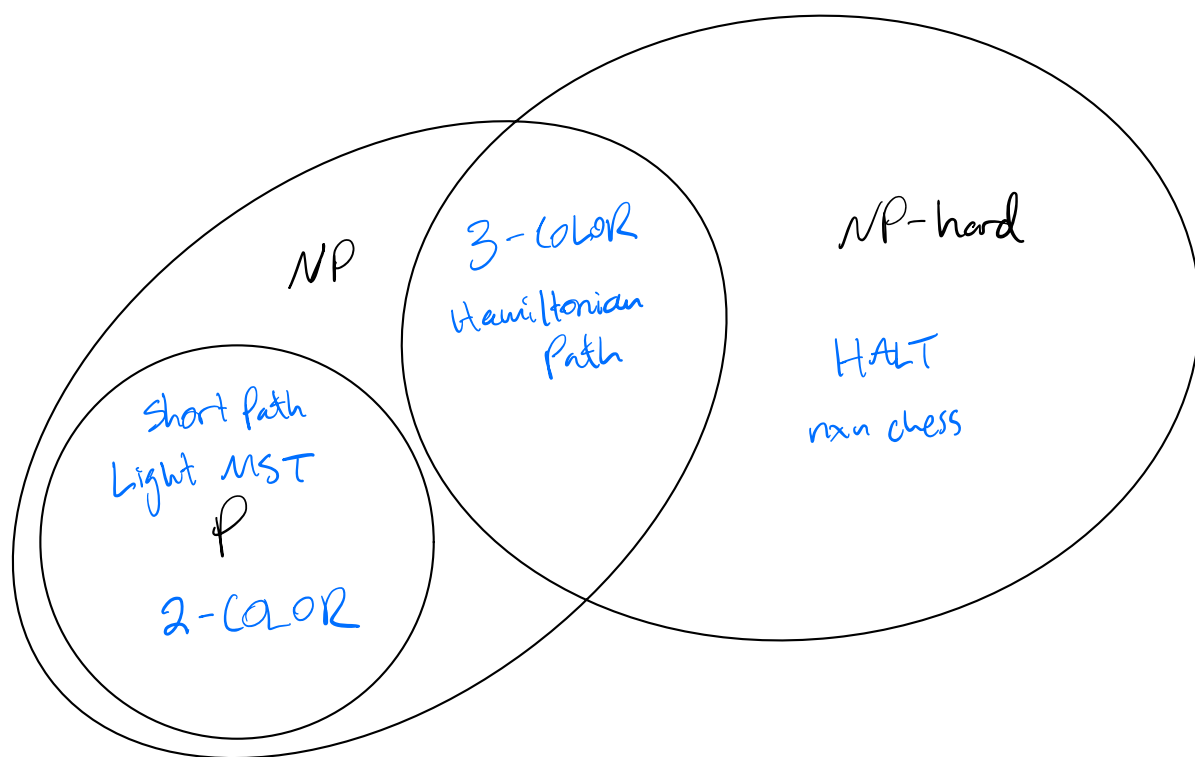
Remark: many problems are NP hard, but not in NP

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 .

Yet another example ... Boolean Satisfiability (SAT)

$$\Phi = (\bar{x}_1 \vee x_3 \vee x_7) \wedge (x_2 \vee \bar{x}_3 \vee x_1) \wedge (x_{12} \vee \bar{x}_4 \vee x_1) \\ \wedge \dots \wedge (x_{77} \vee x_{13} \vee \bar{x}_5)$$

SAT: "does Φ have a satisfying assignment"

3-SAT: CNF w/ ≤ 3
variables per clause

Theorem | (Cook-Levin) SAT is NP-Complete

proof idea | 1) $SAT \in NP$

* actually, even 3SAT is too!

2) goal: $\forall A \in NP, A \leq_p SAT$

suppose $A \in NP$. fix instance x s.t. $A(x) = 1$.

we know: $\exists$ "proof" y and verifier V s.t. $V(x, y) = 1$

rough idea: using x, V , create a SAT instance Φ s.t.

Φ satisfiable $\iff \exists y, V(x, y) = 1$

inputs of Φ will be y !

Corollary 1 | if $SAT \in P$, then $P = NP$.

why?

Corollary 2 | If $SAT \leq_p B$ and $B \in NP$, then
 B is NP -Complete.

proof | $A_1 \leq_p A_2 \leq_p A_3 \Rightarrow A_1 \leq_p A_3$

why?

- Main Theorem.** All the problems on the following list are complete.
- SATISFIABILITY**
COMMENT: By duality, this problem is equivalent to determining whether a disjunctive normal form expression is a tautology.
 - 0-1 INTEGER PROGRAMMING**
INPUT: integer matrix C and integer vector d
PROPERTY: There exists a 0-1 vector x such that $Cx = d$.
 - CLIQUE**
INPUT: graph G , positive integer k
PROPERTY: G has a set of k mutually adjacent nodes.
 - SET PACKING**
INPUT: Family of sets $\{S_i\}$, positive integer k
PROPERTY: $\{S_i\}$ contains k mutually disjoint sets.
 - NODE COVER**
INPUT: graph G' , positive integer k
PROPERTY: There is a set $R \subseteq V$ such that $|R| \leq k$ and every arc is incident with some node in R .
 - SET COVERING**
INPUT: finite family of finite sets $\{S_i\}$, positive integer k
PROPERTY: There is a subfamily $\{T_i\} \subseteq \{S_i\}$ containing $\leq k$ sets such that $\bigcup T_i = \bigcup S_i$.
 - FEEDBACK NODE SET**
INPUT: digraph H , positive integer k
PROPERTY: There is a set $R \subseteq V$ such that every (directed) cycle of H contains a node in R .
 - FEEDBACK ARC SET**
INPUT: digraph H , positive integer k
PROPERTY: There is a set $S \subseteq E$ such that every (directed) cycle of H contains an arc in S .
 - DIRECTED HAMILTON CIRCUIT**
INPUT: digraph H
PROPERTY: H has a directed cycle which includes each node exactly once.
 - UNDIRECTED HAMILTON CIRCUIT**
INPUT: graph G
PROPERTY: G has a cycle which includes each node exactly once.

- SATISFIABILITY WITH AT MOST 3 LITERALS PER CLAUSE**
INPUT: Clauses $D_1, D_2, \dots, D_r$, each consisting of at most 3 literals from the set $\{u_1, u_2, \dots, u_n\} \cup \{\bar{u}_1, \bar{u}_2, \dots, \bar{u}_n\}$
PROPERTY: The set $\{D_1, D_2, \dots, D_r\}$ is satisfiable.
- CHROMATIC NUMBER**
INPUT: graph G , positive integer k
PROPERTY: There is a function $\phi: V \rightarrow Z_k$ such that, if u and v are adjacent, then $\phi(u) \neq \phi(v)$.
- CLIQUE COVER**
INPUT: graph G' , positive integer k
PROPERTY: G' is the union of k or fewer cliques.
- EXACT COVER**
INPUT: family $\{S_i\}$ of subsets of a set $\{u_i, i = 1, 2, \dots, t\}$
PROPERTY: There is a subfamily $\{T_i\} \subseteq \{S_i\}$ such that the sets T_i are disjoint and $\bigcup T_i = \bigcup S_i = \{u_i, i = 1, 2, \dots, t\}$.
- HITTING SET**
INPUT: family $\{U_i\}$ of subsets of $\{a_j, j = 1, 2, \dots, r\}$
PROPERTY: There is a set W such that, for each i , $|W \cap U_i| = 1$.
- STEINER TREE**
INPUT: graph G , $R \subseteq V$, weighting function $w: A \rightarrow Z$, positive integer k
PROPERTY: G has a subtree of weight $\leq k$ containing the set of nodes in R .
- 3-DIMENSIONAL MATCHING**
INPUT: set $U \subseteq T \times T \times T$, where T is a finite set
PROPERTY: There is a set $W \subseteq U$ such that $|W| = |T|$ and no two elements of W agree in any coordinate.
- KNAPSACK**
INPUT: $(a_1, a_2, \dots, a_n, b) \in Z^{n+1}$
PROPERTY: $\sum a_i x_i = b$ has a 0-1 solution.
- JOB SEQUENCING**
INPUT: "execution time vector" $(T_1, \dots, T_p) \in Z^p$, "deadline vector" $(D_1, \dots, D_p) \in Z^p$, "penalty vector" $(P_1, \dots, P_p) \in Z^p$, positive integer k
PROPERTY: There is a permutation π of $\{1, 2, \dots, p\}$ such that $\sum_{j=1}^p \{ \text{if } T_{\pi(j)} + \dots + T_{\pi(j)} > D_{\pi(j)} \text{ then } P_{\pi(j)} \text{ else } 0 \} \leq k$.

- PARTITION**
INPUT: $(c_1, c_2, \dots, c_p) \in Z^p$
PROPERTY: There is a set $I \subseteq \{1, 2, \dots, p\}$ such that $\sum_{h \in I} c_h = \sum_{h \notin I} c_h$.
- MAX CUT**
INPUT: graph G , weighting function $w: A \rightarrow Z$, positive integer W
PROPERTY: There is a set $S \subseteq V$ such that $\sum_{\substack{(u,v) \in A \\ u \in S \\ v \notin S}} w((u,v)) \geq W$.

Karp's Theorem (1972)
A lot of problems people care about are NP-complete

3SAT
Hamiltonian Path
3-COLOR
K-CLIQUE
Knapsack
:
:
:

All NP-complete!

Scenario: You have a problem in NP, but need to solve it. What to do??

options:

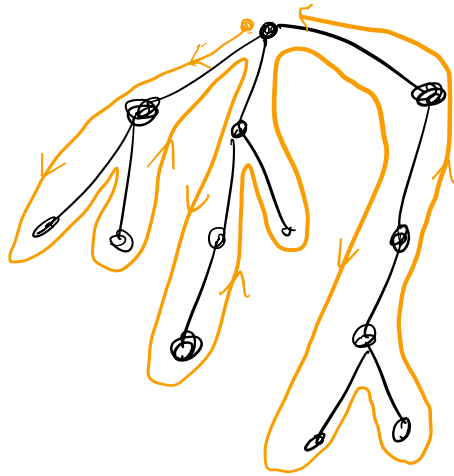
- ① give up.
- ② make sure it's not actually in P
- ③ Convert it to a SAT problem, use a SAT-solver
- ④ Check and see if you are dealing with a special instance
(some problems have fast algorithms for "nice" instances!)
- ⑤ Use an approximation algorithm

Approximation Alg Example

TSP: given weighted graph G , find a tour
(walk that visits every vertex and returns to start)
of minimum total weight.

approximation algorithm:

① Compute a MST



② let your tour be the orange walk

claim this produces a tour of weight at most
twice as much as the best possible tour.

$\Rightarrow$ I can get a pretty good "approximation" of the answer!

why?

Aside: Why is it "nondeterministic" polynomial time?

a "nondeterministic" program can have multiple
possible next states

(generalization of an NFA from 311!)

a nondeterministic (turning machine) program "computes"

ℓ if $\forall x \in \ell^{-1}(1)$, $\exists$ a valid sequence of states
ending in "accept"

and $\forall x \in \ell^{-1}(0)$, all valid sequences of states
end in "reject".

takeaway: this ^{leads to} an alternate (less intuitive) definition of NP,
equivalent to our "verifier" one.