

# Harvard CS 121 and CSCI E-207

## Lecture 22: The P vs. NP Question and NP-completeness

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- Reading: Sipser §7.4, §7.5.
- For “culture”: *Computers and Intractability: A Guide to the Theory of NP-completeness*, by Garey & Johnson.

## P vs. NP

- We would like to solve problems in NP efficiently.
- We know  $P \subseteq NP$ .
- Problems in P can be solved “fairly” quickly.
- What is the relationship between P and NP?

## NP and Exponential Time

Claim:  $NP \subseteq \bigcup_k \text{TIME}(2^{n^k})$

Proof:

Of course, this gets us nowhere near P.

Is  $P = NP$ ?

i.e., do all the NP problems have polynomial time algorithms?

It doesn't "feel" that way but as of today there is no NP problem that has been proven to require exponential time!

## The Strange, Strange World if $P = NP$

Thousands of important languages can be decided in polynomial time, e.g.

- SATISFIABILITY
- TRAVELLING SALESMAN
- HAMILTONIAN CIRCUIT
- MAP COLORING
- ⋮

## If $P = NP$ , then Searching becomes easy

Every “reasonable” search problem could be solved in polynomial time.

- “reasonable”  $\equiv$  solutions can be recognized in polynomial time (and are of polynomial length)
- SAT SEARCH: Given a satisfiable boolean formula, find a satisfying assignment.
- FACTORING: Given a natural number (in binary), find its prime factorization.
- NASH EQUILIBRIUM: Given a two-player “game”, find a Nash equilibrium.
- $\vdots$

## If $P = NP$ , Optimization becomes easy

Every “reasonable” optimization problem can be solved in polynomial time.

- Optimization problem  $\equiv$  “maximize (or minimize)  $f(x)$  subject to certain constraints on  $x$ ” (AM 121)
- “Reasonable”  $\equiv$  “ $f$  and constraints are poly-time”
- MIN-TSP: Given a TSP instance, find the shortest tour.
- SCHEDULING: Given a list of assembly-line tasks and dependencies, find the maximum-throughput scheduling.
- PROTEIN FOLDING: Given a protein, find the minimum-energy folding.
- CIRCUIT MINIMIZATION: Given a digital circuit, find the smallest equivalent circuit.

## If $P = NP$ , Secure Cryptography becomes impossible

Every polynomial-time encryption algorithm can be “broken” in polynomial time.

- “Given an encryption  $z$ , find the corresponding decryption key  $K$  and message  $m$ ” is an NP search problem.
- Thus modern cryptography seeks to design encryption algorithms that cannot be broken under the *assumption* that certain NP problems are hard to solve (e.g. FACTORING).
- Take CS 220r.

## If $P = NP$ , Artificial Intelligence becomes easy

Machine learning is an NP search problem

- Given many examples of some concept (e.g. pairs (image1, “dog”), (image2, “person”), ...), classify new examples correctly.
- Turns out to be equivalent to finding a short “classification rule” consistent with examples.
- Take CS228.



## If $P = NP$ , Even Mathematics Becomes Easy!

Mathematical proofs can always be found in polynomial time (in their length).

- **SHORT PROOF:** Given a mathematical statement  $S$  and a number  $n$  (in unary), decide if  $S$  has a proof of length at most  $n$  (and, if so, find one).
- An NP problem!
- cf. letter from Gödel to von Neumann, 1956.



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## Gödel's Letter to Von Neumann, 55 years ago

[ $\phi(n)$  = time required for a TM to determine whether a mathematical statement has a proof of length  $n$ ]

...

If there really were a machine with  $\phi(n) \sim k \cdot n$  (or even  $\sim k \cdot n^2$ ) this would have consequences of the greatest importance. Namely, it would obviously mean that in spite of the undecidability of the Entscheidungsproblem, the mental work of a mathematician concerning Yes-or-No questions could be completely replaced by a machine. . . .

It would be interesting to know, for instance, the situation concerning the determination of primality of a number and how strongly in general the number of steps in finite combinatorial problems can be reduced with respect to simple exhaustive search. . . .

## The World if $P \neq NP$ ?

**Q:** If  $P \neq NP$ , can we conclude anything about any specific problems?

**Idea:** Try to find a “hardest” NP language.

- Just like  $A_{TM}$  was the “hardest” Turing-recognizable language.
- Want  $L \in NP$  such that  $L \in P$  iff every NP language is in  $P$ .

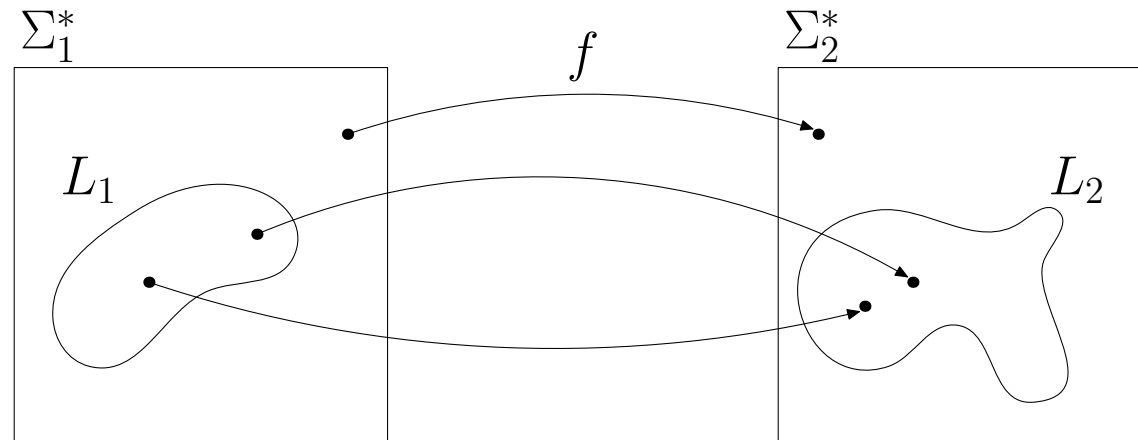
## Polynomial-time Reducibility

**Def:**  $L_1 \leq_P L_2$  iff there is a polynomial-time computable function  $f : \Sigma_1^* \rightarrow \Sigma_2^*$  s.t. for every  $x \in \Sigma_1^*$ ,  $x \in L$  iff  $f(x) \in L_2$ .

**Proposition:** If  $L_1 \leq_P L_2$  and  $L_2 \in P$ , then  $L_1 \in P$ .

**Proof:**

$$L_1 \leq_{\mathbf{P}} L_2$$



$$x \in L_1 \Rightarrow f(x) \in L_2$$

$$x \notin L_1 \Rightarrow f(x) \notin L_2$$

$f$  computable in polynomial time

$$L_2 \in \mathbf{P} \Rightarrow L_1 \in \mathbf{P}.$$

## NP-Completeness

**Def:**  $L$  is NP-complete iff

1.  $L \in \text{NP}$  and
2. Every language in NP is reducible to  $L$  in polynomial time.  
(“ $L$  is NP-hard”)

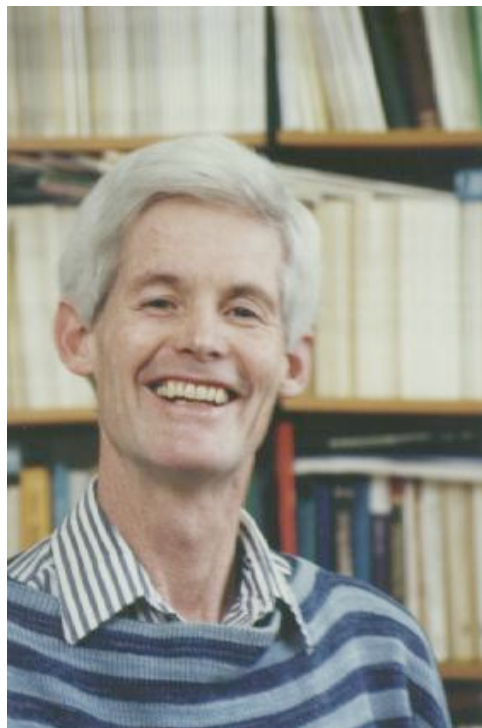
**Prop:** Let  $L$  be any NP-complete language.  
Then  $P = \text{NP}$  *if and only if*  $L \in P$ .

# Cook–Levin Theorem

## (Stephen Cook 1971, Leonid Levin 1973)

**Theorem:** SAT (Boolean satisfiability) is NP-complete.

**Proof:** Need to show that every language in NP reduces to SAT (!) Proof later.



## More NP-complete problems

From now on we prove NP-completeness using:

**Lemma:** If we have the following

- $L$  is in NP
- $L_0 \leq_P L$  for some NP-complete  $L_0$

Then  $L$  is NP-complete.

**Proof:**



## 3-SAT

**Def:** A Boolean formula is in 3-CNF if it is of the form:

$$C_1 \wedge C_2 \wedge \dots \wedge C_n$$

where each clause  $C_i$  is a disjunction (“or”) of 3 literals:

$$C_i = (C_{i1} \vee C_{i2} \vee C_{i3})$$

where each literal  $C_{ij}$  is either

- a variable  $x$ , or
- the negation of a variable,  $\neg x$ .

e.g.  $(x \vee y \vee z) \wedge (\neg x \vee \neg u \vee w) \wedge (u \vee u \vee u)$

3-SAT is the set of satisfiable 3-CNF formulas.

## 3-SAT is NP-complete

**Proof:** Show that  $\text{SAT} \leq_P \text{3-SAT}$ .

1. Given an arbitrary Boolean formula, e.g.

$$F = (\underbrace{\neg}_{1}(\underbrace{(x \vee \neg y)}_{2\ 3}) \wedge \underbrace{(z \vee w)}_{4\ 5})) \vee \underbrace{\neg x}_{6\ 7}.$$

2. Number the operators.
3. Select a new variable  $a_i$  for each operator.  
The variable  $a_i$  is supposed to mean “the subformula rooted at operator  $i$  is true.”
4. Write a formula stating the relation between each subformula and its children subformulas.

## Reduction of SAT to 3-SAT, continued

For example, where

$$F = (\underbrace{\neg}_{1}(\underbrace{(x \vee \neg y)}_{2\ 3}) \wedge \underbrace{(z \vee w)}_{4\ 5}) \vee \underbrace{\neg x}_{6\ 7},$$

$$F_1 = \left( \begin{array}{l} (a_3 \equiv \neg y) \quad \wedge \quad (a_7 \equiv \neg x) \\ \wedge \quad (a_2 \equiv x \vee a_3) \quad \wedge \quad (a_1 \equiv \neg a_4) \\ \wedge \quad (a_5 \equiv z \vee w) \quad \wedge \quad (a_6 \equiv a_1 \vee a_7) \\ \wedge \quad (a_4 \equiv a_2 \wedge a_5) \end{array} \right)$$

- Let  $k$  be the number of the main operator/subformula of  $F$ .  
(Note:  $k = 6$  in the example)

## Write $F_1$ in 3-CNF to obtain $F_2$

- **Fact:** Every function  $f : \{0, 1\}^k \rightarrow \{0, 1\}$  can be written as a  $k$ -CNF and as a  $k$ -DNF (OR of ANDs).  
[albeit with possibly  $2^k$  clauses]
- **Proof:**

Output of the reduction:  $a_k \wedge F_2$ .

**Q:** Does this prove that every Boolean formula can be converted to 3-CNF?

## In contrast, 2-SAT $\in$ P

Method (resolution):

1. If  $x$  and  $\neg x$  are both clauses, then not satisfiable

e.g.  $(x) \wedge (z \vee y) \wedge (\neg x)$

2. If  $(x \vee y) \wedge (\neg y \vee z)$  are both clauses, add clause  $(x \vee z)$  (which is implied).

3. Repeat. If no contradiction emerges  $\Rightarrow$  satisfiable.

$O(n^2)$  repetitions of step 2 since only 2 literals/clause.

Proof of correctness: omitted