# Lecture 23: Computational Complexity

## Lecture Overview

- P, EXP, R
- Most problems are uncomputable
- NP
- Hardness & completeness
- Reductions

# **Definitions:**

- $\underline{\mathbf{P}} = \{ \text{problems solvable in polynomial } (n^c) \text{ time} \}$ (what this class is all about)
- <u>EXP</u> = {problems solvable in exponential  $(2^{n^c})$  time}
- $\underline{\mathbf{R}} = \{ \text{problems solvable in finite time} \}$  "recursive" [Turing 1936; Church 1941]



### Examples

- negative-weight cycle detection  $\in \mathbf{P}$
- $n \times n$  Chess  $\in$  EXP but  $\notin$  P Who wins from given board configuration?
- Tetris ∈ EXP but don't know whether ∈ P Survive given pieces from given board.

#### Halting Problem:

Given a computer program, does it ever halt (stop)?

- uncomputable ( $\notin \mathbb{R}$ ): no algorithm solves it (correctly in finite time on all inputs)
- decision problem: answer is YES or NO

### Most Decision Problems are Uncomputable

- program  $\approx$  binary string  $\approx$  nonneg. integer  $\in N$
- decision problem = a function from <u>binary strings</u> (≈ nonneg. integers) to {YES (1), NO (0)}
- ≈ infinite sequence of bits ≈ real number ∈ ℝ
  |ℕ| ≪ |ℝ|: no assignment of unique nonneg. integers to real numbers (ℝ uncountable)
- $\implies$  not nearly enough programs for all problems
- each program solves only one problem
- $\implies$  almost all problems cannot be solved

### $\mathbf{NP}$

 $NP = \{Decision \text{ problems solvable in polynomial time via a "lucky" algorithm}\}$ . The "lucky" algorithm can make lucky guesses, always "right" without trying all options.

- <u>nondeterministic model</u>: algorithm makes guesses & then says YES or NO
- guesses guaranteed to lead to YES outcome if possible (no otherwise)

In other words,  $NP = \{ \text{decision problems with solutions that can be "<u>checked</u>" in polynomial time \}. This means that when answer = YES, can "<u>prove</u>" it & polynomial-time algorithm can check proof$ 

#### Example

 $\mathrm{Tetris} \in \mathrm{NP}$ 

- nondeterministic algorithm: guess each move, did I survive?
- proof of YES: list what moves to make (<u>rules</u> of Tetris are easy)



# $\mathbf{P}\neq\mathbf{NP}$

Big conjecture (worth \$1,000,000)

- $\approx$  cannot engineer luck
- $\approx$  generating (proofs of) solutions can be harder than checking them

# Hardness and Completeness

### Claim:

If  $P \neq NP$ , then Tetris  $\in NP - P$ [Breukelaar, Demaine, Hohenberger, Hoogeboom, Kosters, Liben-Nowell 2004]

### Why:

Tetris is <u>NP-hard</u> = "as hard as" every problem  $\in$  NP. In fact <u>NP-complete</u> = NP  $\cap$  NP-hard.



#### Similarly

Chess is <u>EXP-complete</u> = EXP  $\cap$  <u>EXP-hard</u>. EXP-hard is as hard as every problem in EXP. If NP  $\neq$  EXP, then Chess  $\notin$  EXP  $\setminus$  NP. Whether NP  $\neq$  EXP is also an open problem but less famous/"important".

### Reductions

Convert your problem into a problem you already know how to solve (instead of solving from scratch)

- most common algorithm design technique
- unweighted shortest path  $\rightarrow$  weighted (set weights = 1)
- min-product path  $\rightarrow$  shortest path (take logs) [PS6-1]
- longest path  $\rightarrow$  shortest path (negate weights) [Quiz 2, P1k]
- shortest ordered tour  $\rightarrow$  shortest path (k copies of the graph) [Quiz 2, P5]
- cheapest leaky-tank path  $\rightarrow$  shortest path (graph reduction) [Quiz 2, P6]

All the above are <u>One-call reductions</u>: A problem  $\rightarrow$  B problem  $\rightarrow$  B solution  $\rightarrow$  A solution <u>Multicall reductions</u>: solve A using free calls to B — in this sense, every algorithm reduces problem  $\rightarrow$  model of computation

NP-complete problems are all interreducible using polynomial-time reductions (same difficulty). This implies that we can use reductions to prove NP-hardness — such as in 3-Partition  $\rightarrow$  Tetris

### Examples of NP-Complete Problems

- Knapsack (pseudopoly, not poly)
- 3-Partition: given *n* integers, can you divide them into triples of equal sum?
- Traveling Salesman Problem: shortest path that visits all vertices of a given graph decision version: is minimum weight  $\leq x$ ?
- longest common subsequence of k strings
- Minesweeper, Sudoku, and most puzzles
- SAT: given a Boolean formula (and, or, not), is it ever true? x and not  $x \to NO$
- shortest paths amidst obstacles in 3D

- 3-coloring a given graph
- find largest clique in a given graph

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