

COMS W3261 - Lecture 11, Part 2: Time Complexity.

Recall: Asymptotic Analysis: "roughly comparing functions."
time it takes to run on certain input

Def. Let f, g be some functions $f, g: \mathbb{N} \rightarrow \mathbb{R}^+$.

- $f(n) = O(g(n))$ if there exist positive integers n_0 and c such that $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

$f(n) = O(g(n)) \approx$ "In the long run, f is at most some constant times g ."

"In the long run, f is ~smaller~ than g ."

- $f(n) = \Omega(g(n))$ if there exist positive integers n_0 and c s.t. $f(n) \geq c \cdot g(n)$ for all $n \geq n_0$.

$(f(n) = \Omega(g(n)) \approx$ "In the long run, f is no ~smaller~ than g ")

(Leipser 7.1)

Examples:

$$10 = O(\log_2(n)).$$

$$\log_2(n) = O(n).$$

$$n = \Omega(\log_2(n)).$$

$$5n = O(n) = \Omega(n).$$

$$16n^2 + n + 4 = O(n^2).$$

$$2^n = O(3^n).$$

$$\log_2(3^n) = n \cdot \underline{\log_2(3)} = O(n).$$

Def. Let M be a deterministic single-tape TM that halts on all inputs. The running time or time complexity of M is a function $f: \mathbb{N} \rightarrow \mathbb{N}$ that indicates the maximum number of steps that

M takes to halt on any input of length n .
 set of languages.

Def. We define the complexity class $\text{TIME}(t(n))$ to be the set of all languages decided by some TM that runs in time $O(t(n))$.

Example. TMs that decide $A = \{0^k 1^k \mid k \geq 0\}$.

One approach:

$M_1 = \text{"On input } w \text{ of length } n:$

- 1. Scan and reject if any 0 occurs to the right of a 1.
- 2. Shuttle back and forth crossing off 0's and 1's, accept iff there are the same number of 0's and 1's.

this takes
 $\approx n$ steps.

shuttling:
 scan the tape once for every input symbol.

$\approx n$ to scan

$\approx n$ input symbols

$\overline{\approx n^2}$ time to shuttle.

Total time complexity: $a \cdot n (\text{step 1}) + b \cdot n^2 (\text{step 2}) + c$

Second approach (2-tape TM): $= O(n^2)$.

$M_2 = \text{"On input } w:$

- $\approx n$ 1. Scan for formatting (same as before.)
- $O(n)$ steps 2. Scan left to right until we see a 1, copying all 0's to the second tape.
- $O(n)$ steps 3. Scan the 1's, crossing off a 0 on the second tape for each 1. Accept/reject depending on whether the number of 0's equals the number of 1's.

Total runtime: $O(n)$ total.

Multiple tapes don't increase the set of languages we can decide, but they may increase how fast we can do it!

P: Polynomial Time.

Def. P is the class of languages that some deterministic TM decides in polynomial time. In math,

$$P = \bigcup_k \text{TIME}(n^k).$$

Why this class?

Idea: polynomial functions << exponential functions.

At $n=1000$, $n^3 = 1$ billion

$2^n >$ # of atoms in the universe.

Idea: polynomial \cdot polynomial = polynomial.

(can use polynomial functions as subroutines and stay in P.)

Problem A decidable in time $n^c \in P$.

Problem B can be decided by deciding A n^d times.

(Runtime for B: now $O(n^d \cdot n^c) = O(n^{c+d})$.)

Idea: "In practice," problems in P tend to have algorithms that run in 'small polynomials'!

Sipser: "All reasonable deterministic computational models are polynomially equivalent."

For example: computers (and TMs) can simulate each other with at most polynomial-time slowdown.

Takeaway: If a problem is poly-time solvable on a TM, it's poly-time

solvable on all computers. 'Model independence.'

Takeaway: "P is the class of efficiently decidable languages."

Problems in P: All Context-free languages.

PATH - does the input of a directed graph, $S, t,$ contain a path between s and $t?$

Relatively prime - $\langle x, y \rangle$, are x, y relatively prime?

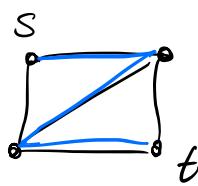
Many, many, many others.



NP: Nondeterministic Polynomial Time.

Idea: A problem is verifiable if you can show me some certificate (evidence) that proves a given string is in the language.

Example. HAMPATH = $\{ \langle G, s, t \rangle \mid G \text{ is a graph with a Hamiltonian path - a path that passes through every vertex exactly once - from } s \text{ to } t. \}$



- Finding a path is hard.
- Showing a path is easy, if we can find it.

Def. A verifier for a language A is an algorithm V , where

$$A = \{ w \mid V \text{ accepts } \langle w, c \rangle \text{ for some certificate } c \}$$

V is a polynomial-time verifier if it runs in time polynomial in the length of w .

Def. NP is the class of efficiently verifiable languages - all languages

that have polynomial time verifiers.

Examples of languages in NP:

Sudoku = $\{\langle P \rangle \mid P \text{ is a sudoku puzzle and } P \text{ is solvable}^2\}$
(certificate for any puzzle - valid solution.)

k-clique = $\{\langle G \rangle \mid G \text{ has a complete subgraph of size } k\}$
(certificate: subgraph)

Subset sum = $\{\langle S, t \rangle \mid S \text{ is a set of numbers with a subset that adds up to } t\}$.
(certificate: subset.)

(By the way: $NP = \text{all languages decidable by a nondeterministic TM in polynomial time.}$)

P and NP.

efficiently decidable efficiently verifiable

$P \subseteq NP$. (Certificate for any string $w \in L$, where $L \in P$?
Just show off the accepting computation of a decider on w .)

$NP \subseteq P$? Seems unlikely. Does being able to efficiently certify a solution imply that you can efficiently find one?

Conjectured that $NP \not\subseteq P$.

We don't know if P=NP.

CSOR 4231. - Algorithms.

COMS 4236 - Complexity. NP, P, and beyond:

COMS 4252 - Computational Learning Theory.



Reminders: HW #6 due Monday. Final ^(8/4) 8/10, 8/11

Readings: 5.1 (Undecidability, reductions)

7.1 - 7.3 (Time complexity.)