BU CS 332 – Theory of Computation

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Lecture 20:

- P Examples
- NP

Mark Bun April 14, 2025 Reading:

Sipser Ch 7.2-7.3

Complexity class P

Definition: P is the class of languages decidable in polynomial time on a basic single-tape (deterministic) TM

$$P = \bigcup_{k=1}^{\infty} TIME(n^k)$$

- Class doesn't change if we substitute in another reasonable deterministic model (Extended Church-Turing)
- Cobham-Edmonds Thesis: Roughly captures class of problems that are feasible to solve on computers

Check your type checker: P



Consider the following computational problem: Given two numbers x, y (written in binary), output their sum x + y (in binary). Which of the following is true?

- a) This is a problem in P
- b) This problem is not in P because it cannot be solved by a Turing machine (i.e., it's undecidable)
- c) This problem is not in P because it cannot be solved in polynomial time
- d) This problem is not in P because it is not a decision problem (i.e., does not correspond to a language)

A note about encodings

We'll still use the notation () for "any reasonable" encoding of the input to a TM...but now we have to be more careful about what we mean by "reasonable"

How long is the encoding of a V-vertex, E-edge graph...

... as an adjacency matrix?

... as an adjacency list?

How long is the encoding of a natural number k

... in binary?

... in decimal?

... in unary?

Describing and analyzing polynomial-time algorithms

- Due to Extended Church-Turing Thesis, we can still use high-level descriptions on multi-tape machines
- Polynomial-time is robust under composition: poly(n) executions of poly(n)-time subroutines run on poly(n)-size inputs gives an algorithm running in poly(n) time.
 - ⇒ Can freely use algorithms we've seen before as subroutines if we've analyzed their runtime

• Need to be careful about size of inputs! (Assume inputs represented in <u>binary</u> unless otherwise stated.)

Examples of languages in P

 $PATH = \{\langle G, s, t \rangle | G \text{ is a directed graph with a directed path from } s \text{ to } t\}$

Idea: Breadth-first search

Assume G presented as adjacency matrix

"On input $\langle G, s, t \rangle$:

- 1. Mark start vertex s
- 2. For i = 1, 2, ..., |V|:
- 3. Mark all neighbors of currently marked vertices
- 4. If t is marked, accept. Else, reject."

Examples of languages in P

 $E_{\text{DFA}} = \{\langle D \rangle \mid D \text{ is a DFA that recognizes the empty language}\}$

Examples of languages in P

• $RELPRIME = \{\langle x, y \rangle \mid x \text{ and } y \text{ are relatively prime} \}$

• $PRIMES = \{\langle x \rangle \mid x \text{ is prime}\}$

2006 Gödel Prize citation







The 2006 Gödel Prize for outstanding articles in theoretical computer science is awarded to Manindra Agrawal, Neeraj Kayal, and Nitin Saxena for their paper "PRIMES is in P."

In August 2002 one of the most ancient computational problems was finally solved....

A polynomial-time algorithm for *PRIMES*?

Consider the following algorithm for PRIMES



On input $\langle x \rangle$:

For
$$b = 2, 3, 4, 5, ..., \sqrt{x}$$
:

- Try to divide x by b
- If b divides x, reject

If all b fail to divide x, accept

How many divisions does this algorithm require in terms of

$$n = |\langle x \rangle|$$
? a) $O(\sqrt{n})$ b) $O(n)$ c) $2^{O(\sqrt{n})}$

a)
$$O(\sqrt{n})$$

b)
$$O(n)$$

c)
$$2^{O(\sqrt{n})}$$

d)
$$2^{O(n)}$$

Beyond polynomial time

Definition: EXP is the class of languages decidable in exponential time on a basic single-tape (deterministic) TM

$$EXP = \bigcup_{k=1}^{\infty} TIME(2^{n^k})$$

Why study P?

Criticism of the Cobham-Edmonds Thesis:

- Algorithms running in time n^{100} aren't really efficient Response: Runtimes improve with more research
- Does not capture some physically realizable models using randomness, quantum mechanics

Response: Randomness may not change P, useful principles



TIME(n) vs. $TIME(n^2)$



P vs. EXP



decidable vs. undecidable

Nondeterministic Time and NP

Extended Church-Turing Thesis

Every "reasonable" (physically realizable) model of computation can be simulated by a basic, single-tape TM with only a **polynomial** slowdown.

E.g., doubly infinite TMs, multi-tape TMs, RAM TMs

Does not include nondeterministic TMs (not reasonable)

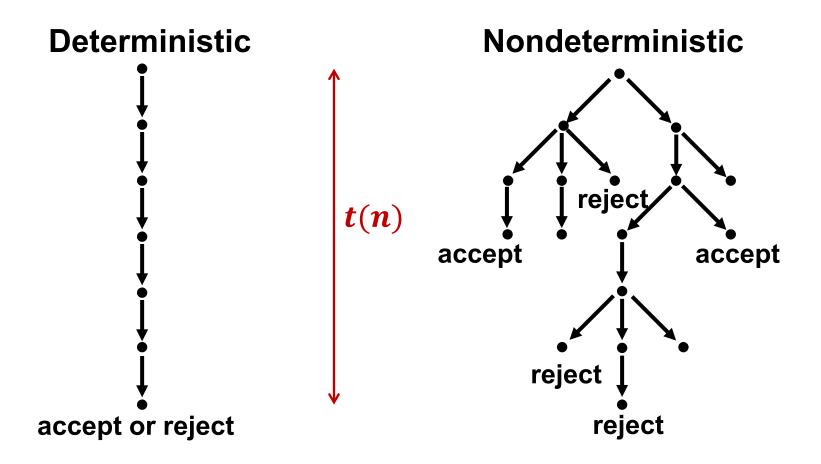
Nondeterministic time

Let $t: \mathbb{N} \to \mathbb{N}$

NTM M runs in time t(n) if:

For every n and every input $w \in \Sigma^n$, M halts on w within at most t(n) steps on every computational branch

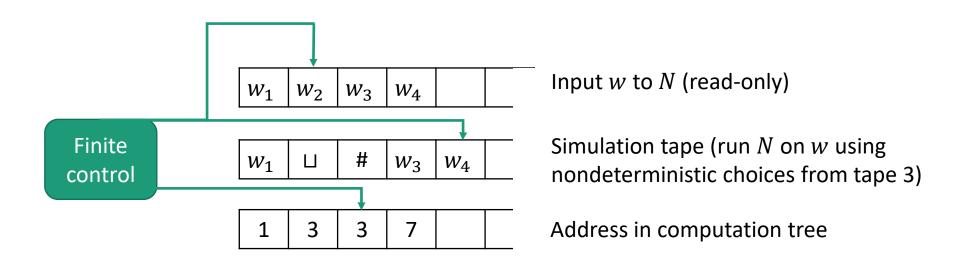
Deterministic vs. nondeterministic time



Deterministic vs. nondeterministic time

Theorem: Let $t(n) \ge n$ be a function. Every NTM running in time t(n) has an equivalent deterministic single-tape TM running in time $2^{O(t(n))}$

Proof: Simulate NTM by 3-tape TM

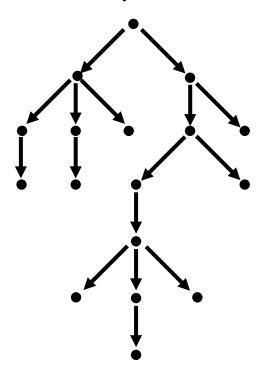


Counting leaves



What is an upper bound on the maximum number of nodes in a tree with branching factor b and depth t?

- a) bt
- b) h^t
- c) t^b
- d) 2^t



Deterministic vs. nondeterministic time

Theorem: Let $t(n) \ge n$ be a function. Every NTM running in time t(n) has an equivalent deterministic single-tape

TM running in time $2^{O(t(n))}$

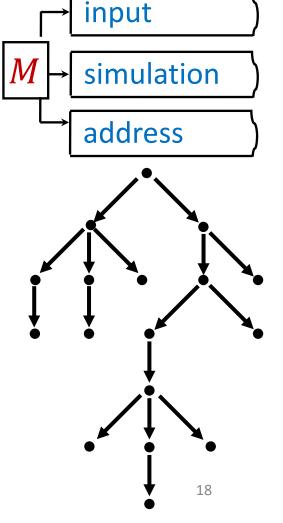
Proof: Simulate NTM by 3-tape TM

• # nodes:

Running time:

To simulate one root-to-node path:

Total time:



Deterministic vs. nondeterministic time

Theorem: Let $t(n) \ge n$ be a function. Every NTM running in time t(n) has an equivalent deterministic single-tape TM running in time $2^{O(t(n))}$

Proof: Simulate NTM by 3-tape TM in time $2^{O(t(n))}$

We know that a 3-tape TM can be simulated by a singletape TM with quadratic overhead, hence we get running time

$$(2^{O(t(n))})^2 = 2^{2 \cdot O(t(n))} = 2^{O(t(n))}$$

Difference in time complexity

Extended Church-Turing Thesis:

At most polynomial difference in running time between all (reasonable) deterministic models

At most exponential difference in running time between deterministic and nondeterministic models

Nondeterministic time

Let $f: \mathbb{N} \to \mathbb{N}$

NTM M runs in time f(n) if:

For every n and every input $w \in \Sigma^n$, M halts on w within at most f(n) steps on every computational branch

NTIME(f(n)) is a class (i.e., set) of languages:

A language $A \in NTIME(f(n))$ if there exists an NTM M that

- 1) Decides A, and
- 2) Runs in time O(f(n))

NTIME explicitly

A language $A \in \text{NTIME}(f(n))$ if there exists an NTM M such that, on every input $w \in \Sigma^*$

- 1. Every computational branch of M halts in either the accept or reject state within O(f(|w|)) steps
- 2. If $w \in A$, then there exists an accepting computational branch of M on input w

3. If $w \notin A$, then every computational branch of M rejects on input w

Complexity class NP



Definition: NP is the class of languages decidable in polynomial time on a nondeterministic TM

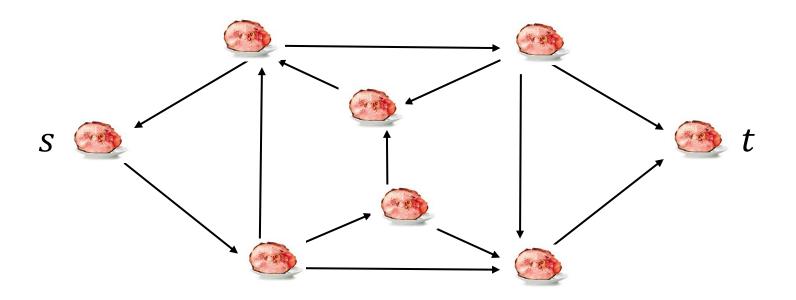
$$NP = \bigcup_{k=1}^{\infty} NTIME(n^k)$$

Which of the following are definitely true about NP?

- a) $P \subseteq NP$
- b) $NP \subseteq P$
- c) NP ⊈ P
- d) NP⊆EXP
- e) $EXP \subseteq NP$

Hamiltonian Path

 $HAMPATH = \{\langle G, s, t \rangle | G \text{ is a directed graph and there}$ is a path from s to t that passes through every vertex exactly once}



$HAMPATH \in NP$

The following algorithm decides *HAMPATH* in **nondeterministic** polynomial time:

On input $\langle G, s, t \rangle$: (Vertices of G are numbers 1, ..., k)

- 1. Nondeterministically guess a sequence $c_1, c_2, ..., c_k$ of numbers 1, ..., k
- 2. Check that $c_1, c_2, ..., c_k$ is a permutation: Every number 1, ..., k appears exactly once
- 3. Check that $c_1 = s$, $c_k = t$, and there is an edge from every c_i to c_{i+1}
- 4. Accept if all checks pass, otherwise, reject.

Analyzing the algorithm

Need to check:

1) Correctness

2) Running time