

Intro to Theory of Computation

CS
332

LECTURE 21

Last time

- Recursion theorem
- Measuring complexity
- Asymptotic notation

Today

- Measuring complexity
- Relationship between models
- Class P

Sofya Raskhodnikova

In the future we will find algorithms for all computational problems, that is, problems with well-defined inputs and desired outputs.

- A.** True. I am an optimist.
- B.** It is difficult to make predictions, especially about the future. (*K.K. Steincke*)
- C.** False. Finitely many people will be able to design only finitely many algorithms.
- D.** False. There are more computational problems than algorithms.

**It is not hard to make predictions, it is hard to
make *interesting* predictions**

(of unpredictable events you don't control).

- It will be dark tonight at 11pm.
- Most people in this room will have another meal today.
- The exercise from the previous slide will appear on the final.

Running time analysis

If M is a TM and $f: \mathbb{N} \rightarrow \mathbb{N}$ then

“ M runs in time $f(n)$ ” means

for **every** input $w \in \Sigma^*$ of length n ,

M on w halts within $f(n)$ steps

- Focus on worst case:
 - upper bound on running time for all inputs of given length
- Exact time depends on computer
 - instead measure **asymptotic growth**

Time complexity classes

$\text{TIME}(f(n))$ is a class of languages.

$A \in \text{TIME}(f(n))$ means that

some 1-tape TM M

that runs in time $O(f(n))$ decides A .

How much time/memory needed to decide a language?

Example: Consider $A = \{0^m 1^m \mid m \geq 0\}$.

- $M_1 =$ “
 1. Scan input and reject if it is not of the form $0^* 1^*$.
 2. Repeat while both 0s and 1s remain on the tape:
 3. Cross off one 0 and one 1
 4. **Accept** if no 0s and no 1s left; otherwise **reject**.”
- M_1 runs in time $O(n^2)$.
- $A \in TIME(n^2)$.
- Is there a faster algorithm?

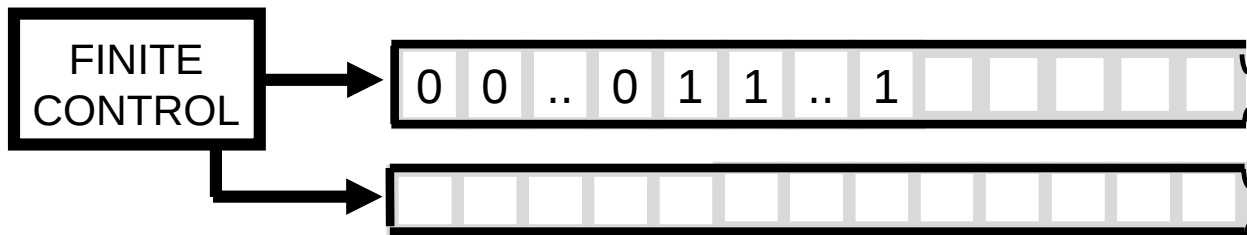
How much time/memory needed to decide a language?

Example: Consider $A = \{0^m 1^m \mid m \geq 0\}$.

- $M_2 =$ “
 1. Scan input and reject if it is not of the form $0^* 1^*$.
 2. Repeat while both 0s and 1s remain on the tape:
 3. **Reject** if total number of 0s and 1s remaining is odd.
 4. Cross off every other 0 starting from the first 0 and every other 1 starting from the first 1
 5. **Accept** if no 0s and no 1s left; otherwise **reject**.”
- M_2 runs in time $O(n \log n)$, so $A \in TIME(n \log n)$.
- **Sipser, Problem 7.49:** If language L can be decided in $o(n \log n)$ time on a 1-tape TM then L is regular.
- 1-tape TM need $\Omega(n \log n)$ time to decide A .

Two-tape TM can do it faster

Example: Consider $A = \{0^m 1^m \mid m \geq 0\}$.



- $M_3 =$ “
 1. Scan input and reject if it is not of the form 0^*1^* .
 2. Copy 0s on tape 2.
 3. Scan tape 1. For each 1 read, cross off a 0 on tape 2.
 4. **Accept** if no 0s remain on tape 2; otherwise **reject**.”
- A is decided in $O(n)$ time (linear time) on a 2-tape TM.

**Unlike decidability,
the complexity of the language depends on the model.**

Complexity relationships between models: number of tapes

Theorem. Let $t(n)$ be a function, where $t(n) \geq n$.

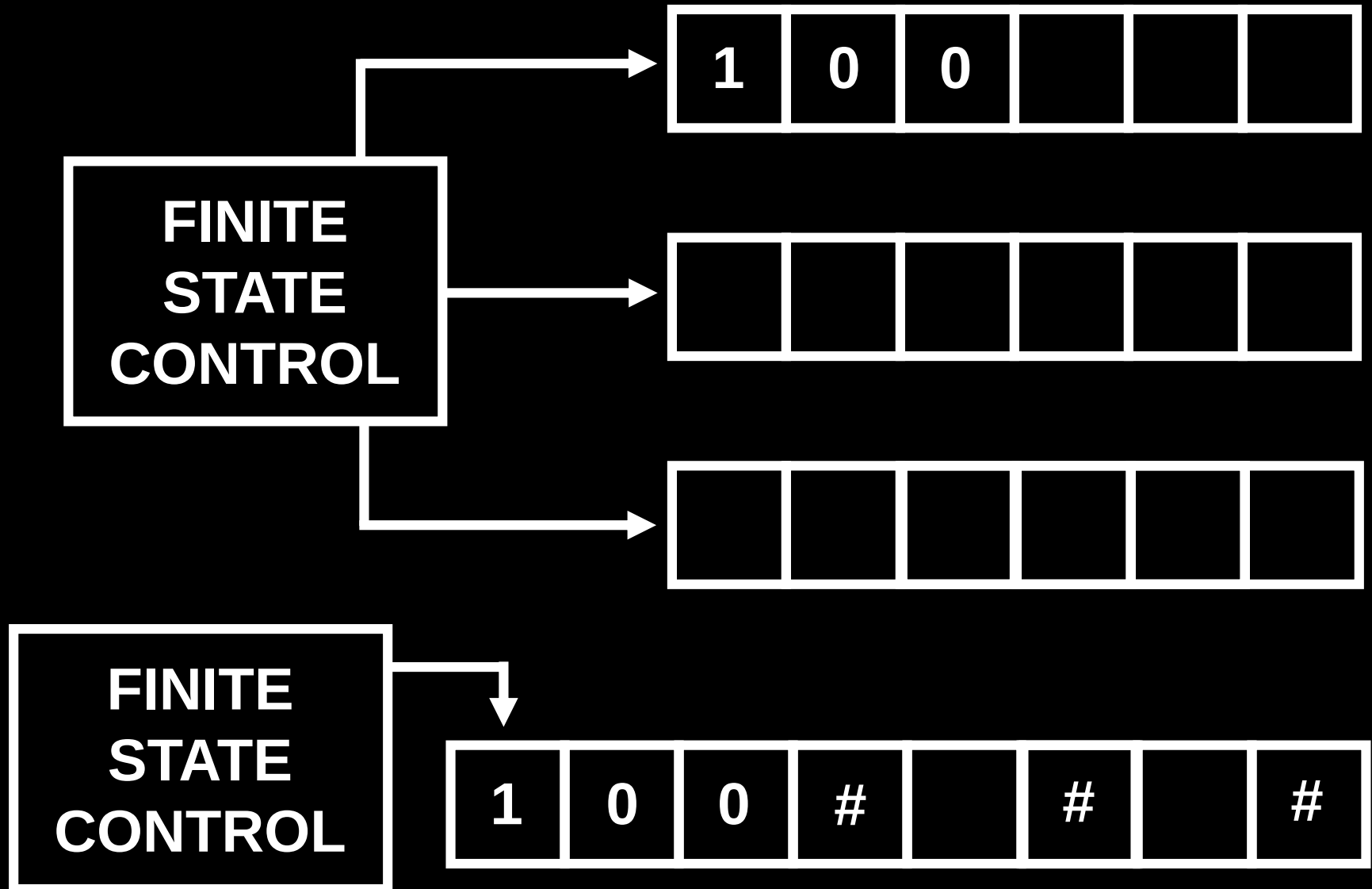
Every $t(n)$ time multitape TM has

an equivalent $O\left((t(n))^2\right)$ time 1-tape TM.

Proof:

- Recall: we already showed how to simulate multitape TMs by 1-tape TMs.
- Need time analysis of the simulation.

Theorem: Every Multitape Turing Machine can be transformed into a single-tape Turing Machine



SIMULATING MULTIPLE TAPES

L#100# 0 # 1 #R

 q_{jRS} q_{i1} q_{i1} $q_{i101RSS}$

1. “Format” tape.
2. For each move of the k-tape TM:
 - Scan left-to-right, finding current symbols
 - Scan left-to-right, writing new symbols
 - Scan left-to-right, moving each tape head.
3. If a tape head goes off right end, insert blank.
 - If tape head goes off left end, move back right.

Complexity relationships between models: number of tapes

Theorem. Let $t(n)$ be a function, where $t(n) \geq n$.

Every $t(n)$ time multitape TM has

an equivalent $O\left((t(n))^2\right)$ time 1-tape TM.

Proof: Time analysis of the simulation.

- Time initialize tape: $O(n + k) = O(n)$
- Time to simulate one step of the multitape TM: $O(t(n))$
(at any point $\leq t(n)$ nonblank squares on each tape)
- Number of steps to simulate: $t(n)$

Total time: $O(n) + O(t(n))t(n) = O((t(n))^2)$

Let $t(n)$ be a function, where $t(n) \geq n$.

Every 3-tape TM that runs in time $O(t(n))$ can be simulated by a 1-tape TM that runs in time

- A.** $O(t(n))$
- B.** $O(t(n^2))$
- C.** $O(t(n^3))$
- D.** $O\left((t(n))^2\right)$
- E.** Some 3-tape TMs can't be simulated by 1-tape TMs

The class P

P is the class of languages decidable in polynomial time on a *deterministic* 1-tape TM:

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

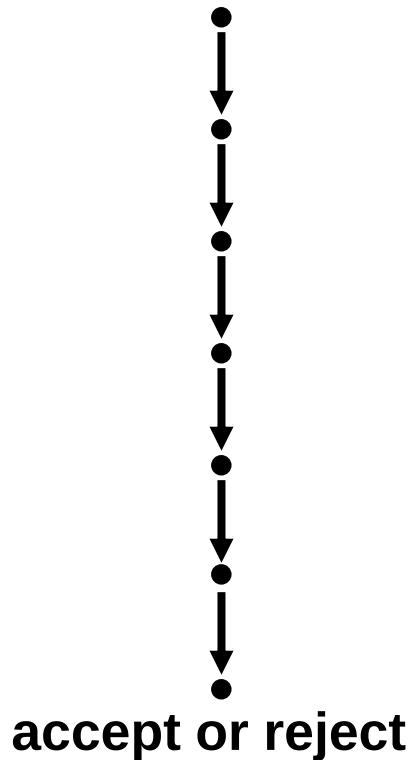
- The same class even if we substitute another reasonable deterministic model.
- Roughly the class of problems realistically solvable on a computer.

Time complexity of NTMs

The **running time** a nondeterministic **decider** N is $t(n)$ if on **all** inputs of length n , NTM N takes **at most** $t(n)$ steps on the **longest** nondeterministic branch.

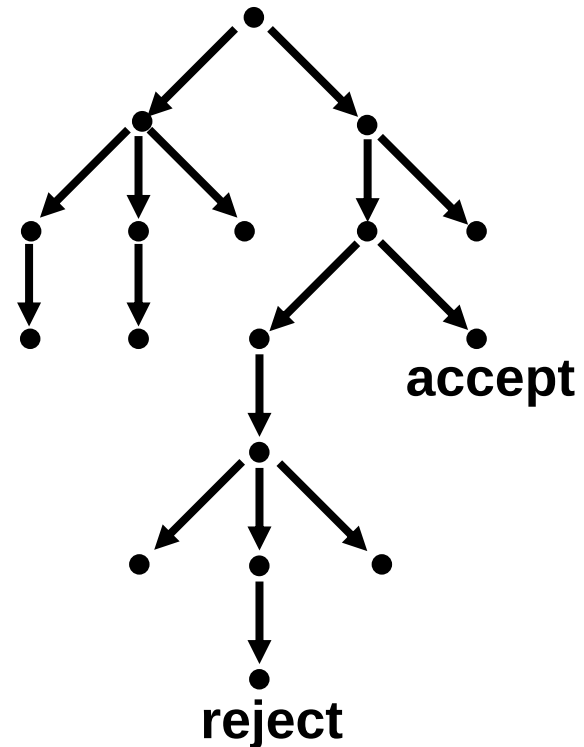
Time complexity of NTMs

Deterministic



$t(n)$

Nondeterministic



- Length of the longest computational branch, even if accepts before

Complexity relationships between models: nondeterminism

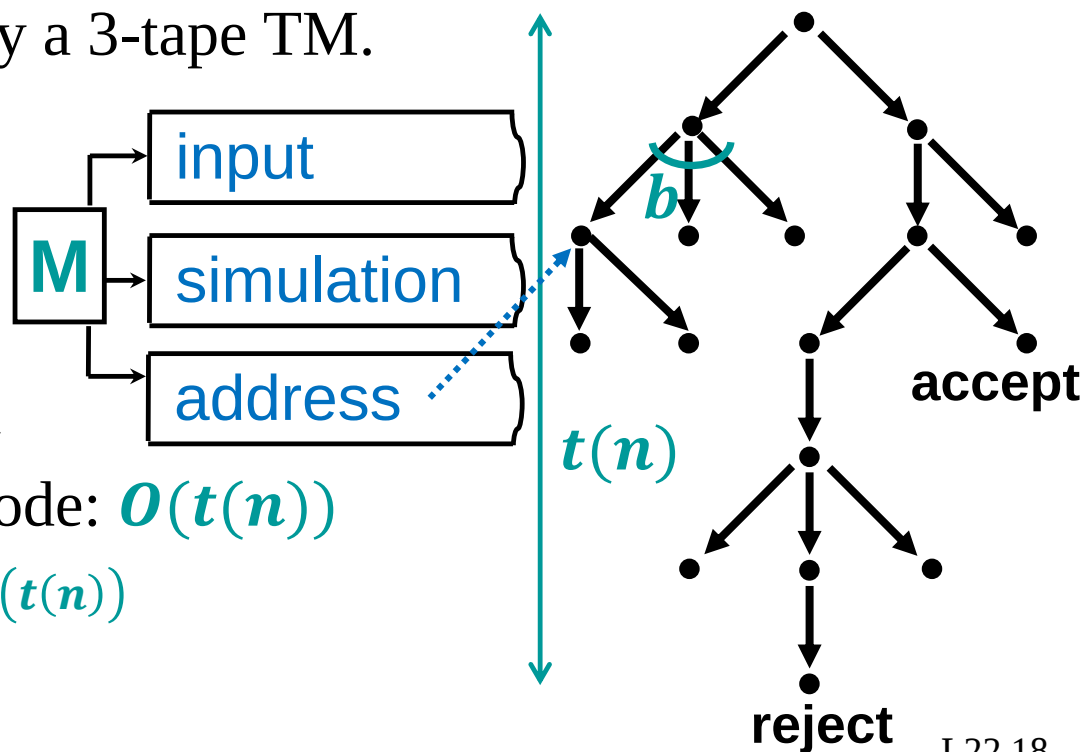
Theorem. Let $t(n)$ be a function, where $t(n) \geq n$. Every $t(n)$ time nondeterministic TM has an equivalent $2^{O(t(n))}$ time 1-tape deterministic TM.

Proof: Simulate an NTM by a 3-tape TM.

- # of leaves $\leq b^{t(n)}$
- # of nodes $\leq 2b^{t(n)}$

Time

- increment the address and simulate from the root to a node: $O(t(n))$
- Total: $O(t(n)b^{t(n)}) = 2^{O(t(n))}$



Complexity relationships between models: nondeterminism

Theorem. Let $t(n)$ be a function, where $t(n) \geq n$.
Every $t(n)$ time nondeterministic TM has
an equivalent $2^{O(t(n))}$ time 1-tape deterministic TM.

Proof: So, a 3-tape TM can simulate an NTM in $2^{O(t(n))}$ time.

Converting to a 1-tape TM at most squares the running time:

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

Difference in time complexity

At most *polynomial* difference between *all reasonable* deterministic models.

At most *exponential* difference between deterministic and nondeterministic models.

The class P

P is the class of languages decidable in polynomial time on a *deterministic* 1-tape TM:

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

- The same class even if we substitute another reasonable deterministic model.
- Roughly the class of problems realistically solvable on a computer.

Examples of languages in P

- $\text{PATH} = \{\langle G, s, t \rangle \mid G \text{ is a directed graph that has a directed path from } s \text{ to } t\}$
- $\text{RELPRIME} = \{\langle x, y \rangle \mid x \text{ and } y \text{ are relatively prime}\}$
- $\text{PRIMES} = \{x \mid x \text{ is a prime number}\}$ [2002]
- Every context-free language (On the board)

Recall: Chomsky Normal Form for CFGs

- Can have a rule $S \rightarrow \varepsilon$.
- All remaining rules are of the form
$$A \rightarrow BC \qquad A, B, C \in V$$
$$A \rightarrow a \qquad a \in \Sigma$$
- Cannot have S on the RHS of any rule.

Lemma. Any CFG can be converted into an equivalent CFG in Chomsky normal form.

Lemma. If G is in Chomsky normal form, any derivation of string w of length n in G has $2n - 1$ steps.

A decider for a CFL

- Let L be a CFL generated by a CFG G in CNF

M = ``On input $\langle w \rangle$, where w is a string:

1. Let $n = |w|$.
2. Test all derivations with $2n - 1$ steps.
3. **Accept** if any derived w . O.w. **reject**."

- How long does it take? **(exponential time)**
- Idea: use dynamic programming **(on the board)**
 - Solve smaller subproblems
 - Record results in a table
 - Construct solution for each subproblem from smaller solved instances