

CS5371

Theory of Computation

Lecture 17: Complexity II
(Relationship among models)

Objectives

- Complexity relationship among models
 - Single-Tape versus Multi-Tape
 - NTM versus DTM

Single-Tape versus Multi-Tape

Theorem: Let $t(n)$ be a function, where $t(n) \geq n$. Then every $t(n)$ time k -tape TM has an equivalent $O(k^2 t(n)^2)$ time single-tape TM.

Proof: Let M be a k -tape TM that runs in $t(n)$ time. We construct a single-tape TM S that runs in $O(k^2 t(n)^2)$ time.

Single-Tape vs Multi-Tape (2)

Recall that we learnt one way of how S can simulate M (in Lecture 11):

- S uses its single tape to represent the contents of all k tapes in M
- The k tapes are stored consecutively, separated by $\#$
- Positions of tape heads are represented by "marked" symbols

Here, S uses the same way to simulate M

Single-Tape vs Multi-Tape (3)

Recall that to perform a step in M , S will do:

- Scan the tape to collect the characters under each of the tape heads in M
- Scan the tape again, update the symbol under the tape heads of M , and update the positions of the tape heads
- Special case: when a tape head of M moves rightward onto an unread portion, we add a space in the corresponding place in S 's tape (by shifting)

Single-Tape vs Multi-Tape (4)

Since M runs in $t(n)$ time, each of its tape head can access only the first $t(n)$ cells. Thus, S will use (and access) only the first $k \times t(n) + k + 1 = O(k t(n))$ cells.

We call these $O(k t(n))$ cells the **active portion** of S 's tape

Single-Tape vs Multi-Tape (5)

S simulates M for $O(t(n))$ steps.

- Each step in the worst case needs to (1) scan the tape first, (2) add a space to each of the k tapes, and (3) update the tape contents and tape heads.
- (1) and (3) accesses only the active portion of S 's tape $\rightarrow O(k t(n))$ time
- (2) scans the active portion for at most k times, which is $O(k^2 t(n))$ time

In total, it takes $O(k^2 t(n)^2)$

Polynomial Time Bounds

If the running time $t(n)$ of a machine M is $O(n^c)$ for some fixed constant $c > 0$, the running time is called **polynomial bounded**, or we say M **runs in polynomial time**. This gives the following corollary.

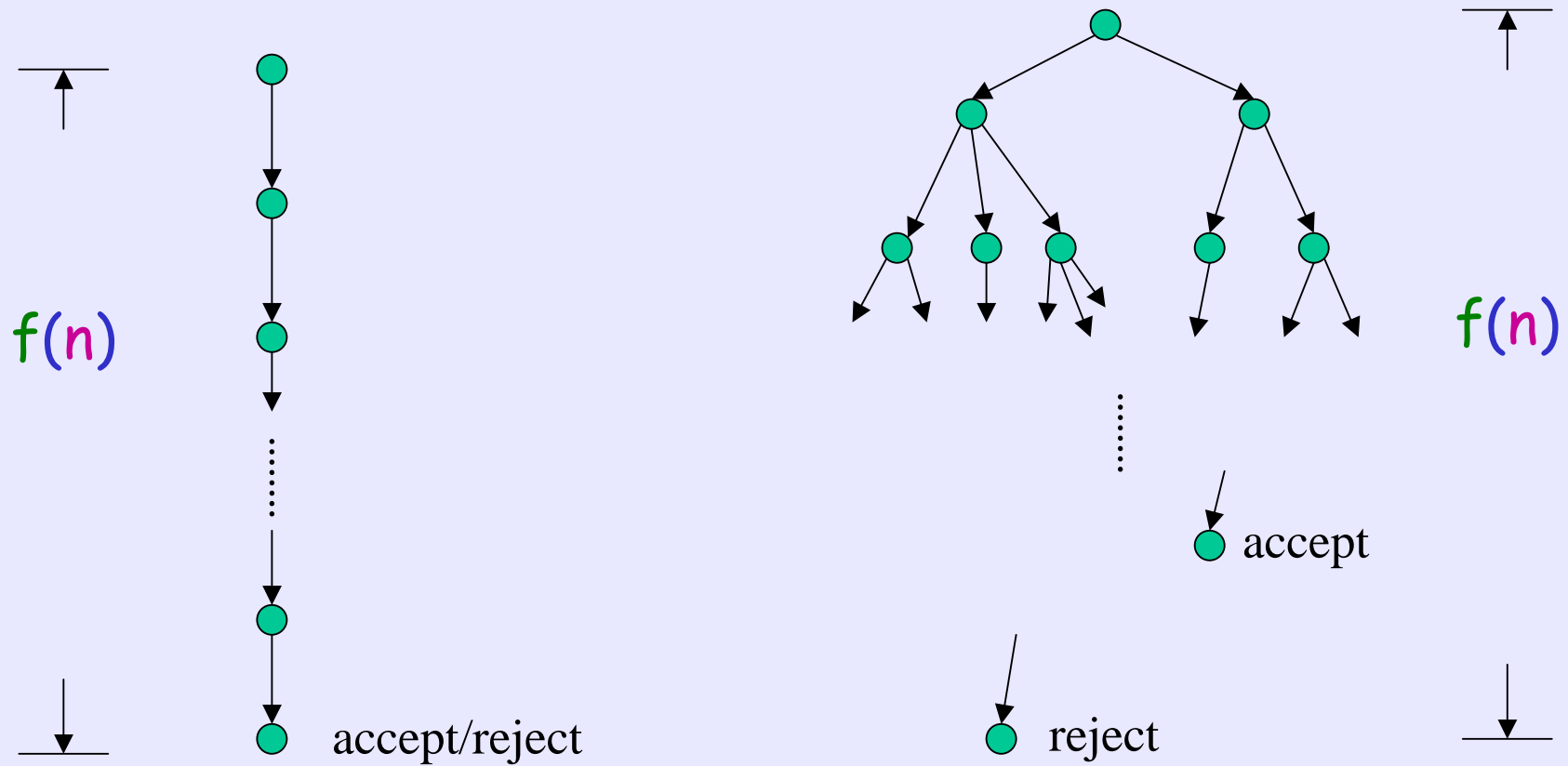
Corollary: For any **k**-tape TM that runs in polynomial time, it has an equivalent single-tape TM that runs in polynomial time.

NTM decider

Recall that an NTM is a **decider** if all its computation branches halt on all inputs.

Definition: Let M be an NTM decider. The **running time** of M is the function $f:N \rightarrow N$, where $f(n)$ is the maximum number of steps that M uses on **any branch of its computation** on any input of length n

Comparison of Running Times



Deterministic time

Non-deterministic time

DTM versus NTM decider

Theorem: Let $t(n)$ be a function, where $t(n) \geq n$. Then every $t(n)$ time single-tape NTM decider has an equivalent $2^{O(t(n))}$ time single-tape DTM.

Proof: Let M be a NTM that runs in $t(n)$ time. We construct a DTM D that simulates M by searching M 's computation tree, as described in Lecture 11. We now analyze D 's simulation.

DTM versus NTM decider (2)

- On an input of length n , every branch of computation of M has at most $t(n)$ steps
- Every node in the computation tree has at most b children, where b is the maximum number of choices in M 's transition \rightarrow number of leaves is at most $O(b^{t(n)})$
- Also, total number of nodes + leaves is at most $O(t(n) b^{t(n)})$ (why??)

DTM versus NTM decider (3)

- The simulation proceeds by visiting the nodes (including leaves) in BFS order. Here, when we visit a node v , we travel from the root to $v \rightarrow$ time to visit v is $O(t(n))$

Thus, the total time for D to simulate M is $O(t(n)^2 b^{t(n)}) = 2^{O(t(n))}$ (why??)

Next Time

- P and NP
 - Two important classes of problems in time complexity theory