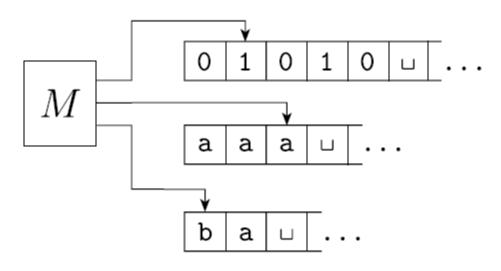
# Complexity Theory ...

Multi-tape, NTM, ...

#### MULTITAPE TURING MACHINES

A k tape Turing Machine will have k tapes (a 3 tape TM is shown in the figure)



$$\delta: Q \times \Gamma^k \longrightarrow Q \times \Gamma^k \times \{L, R\}^k$$

where k is the number of tapes. The expression

$$\delta(q_i, a_1, \dots, a_k) = (q_j, b_1, \dots, b_k, L, R, \dots, L)$$

THEOREM 3.13 ------

Every multitape Turing machine has an equivalent single-tape Turing machine.

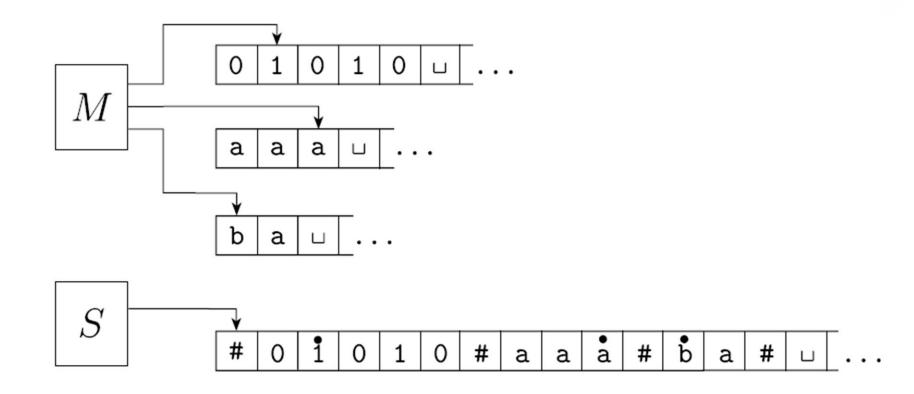


FIGURE 3.14

Representing three tapes with one

## Single-Tape versus Multi-Tape

Theorem: Let t(n) be a function, and  $t(n) \ge n$ . Then for every t(n)-time TM that works with k tapes, there is an equivalent  $O(k^2 t(n)^2)$ -time TM that works with 1 tape.

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)

How the simulation of a k tape TM (M) can be done over a single tape TM (S):

- 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 5'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, where each step (in the worst case) does the following:
  - (1) scan the tape first
  - (2) add a space to each of the k tapes,
  - (3) update tape heads and its contents
  - For (1) and (3), S accesses only the active portion of S's tape  $\rightarrow O(k t(n))$  time
  - For (2), S scans the active portion for at most k times, which is  $O(k^2 t(n))$  time
- $\rightarrow$  In total, it thus takes  $O(k^2 t(n)^2)$  time

## 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.

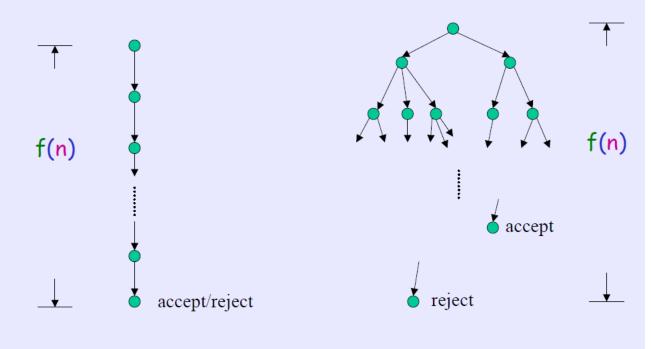
Corollary: For any k-tape TM that runs in polynomial time, it has an equivalent single-tape TM that runs in polynomial time.

#### NTM decider

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→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



Non-deterministic time

Deterministic time

#### DTM versus NTM decider

Theorem: Let t(n) be a function,  $t(n) \ge n$ . 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 → number of leaves is at most O(b<sup>t(n)</sup>)
- 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 always travel starting from the root
  - $\rightarrow$  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