CMPS 2200 – Fall 2014

Divide-and-Conquer Carola Wenk

Slides courtesy of Charles Leiserson with changes and additions by Carola Wenk

The divide-and-conquer design paradigm

- 1. Divide the problem (instance) into subproblems of sizes that are fractions of the original problem size.
- 2. Conquer the subproblems by solving them recursively.
- 3. Combine subproblem solutions.

Find an element in a sorted array:

- 1. Divide: Check middle element.
- 2. Conquer: Recursively search 1 subarray.
- 3. Combine: Trivial.

Example: Find 9

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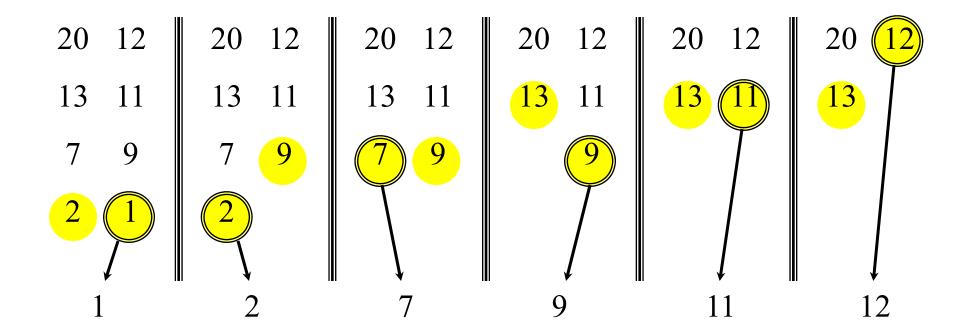
Merge sort

- 1. Divide: Trivial.
- 2. Conquer: Recursively sort 2 subarrays of size n/2
- 3. Combine: Linear-time key subroutine MERGE

MERGE-SORT (A[0 ... n-1])

- 1. If n = 1, done.
- 2. Merge-Sort $(A[0...\lceil n/2\rceil -1])$
- 3. Merge-Sort $(A \lceil \lceil n/2 \rceil ... n-1 \rceil)$
- 4. "Merge" the 2 sorted lists.

Merging two sorted arrays



Time $dn \in \Theta(n)$ to merge a total of n elements (linear time).

Analyzing merge sort

```
T(n)MERGE-SORT (A[0 ... n-1])d_01. If n = 1, done.T(n/2)2. MERGE-SORT (A[0 ... \lceil n/2 \rceil + 1])T(n/2)3. MERGE-SORT (A[\lceil n/2 \rceil ... n-1])dn4. "Merge" the 2 sorted lists.
```

Sloppiness: Should be $T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor)$, but it turns out not to matter asymptotically.

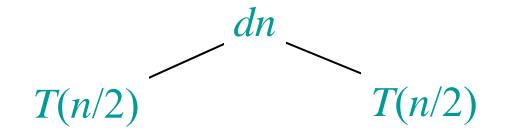
Recurrence for merge sort

$$T(n) = \begin{cases} d_0 \text{ if } n = 1; \\ 2T(n/2) + dn \text{ if } n > 1. \end{cases}$$

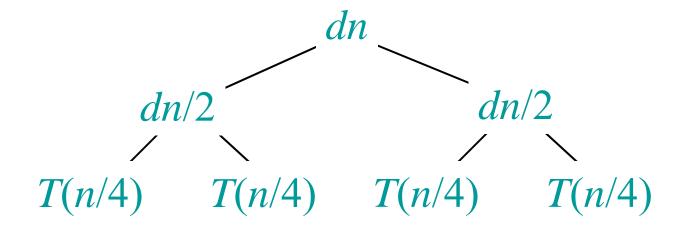
• But what does T(n) solve to? I.e., is it O(n) or $O(n^2)$ or $O(n^3)$ or ...?

Solve
$$T(n) = 2T(n/2) + dn$$
, where $d > 0$ is constant.
$$T(n)$$

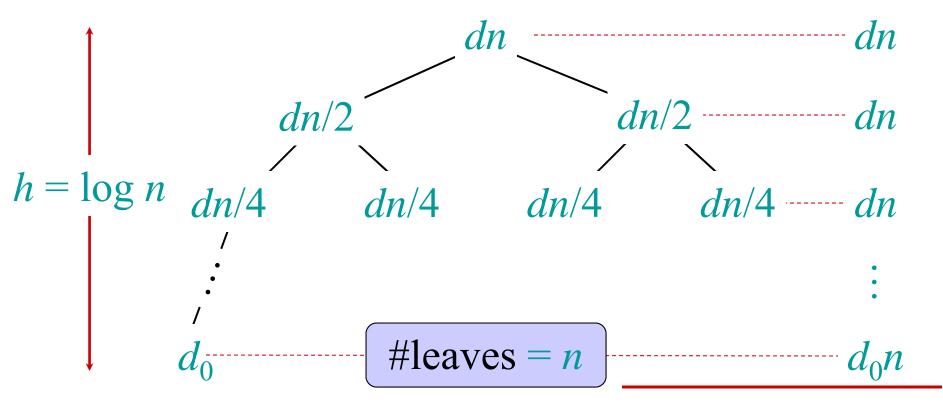
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Total $dn \log n + d_0 n$

Mergesort Conclusions

- Merge sort runs in $\Theta(n \log n)$ time.
- $\Theta(n \log n)$ grows more slowly than $\Theta(n^2)$.
- Therefore, merge sort asymptotically beats insertion sort in the worst case.
- In practice, merge sort beats insertion sort for n > 30 or so. (Why not earlier?)

Recursion-tree method

- A recursion tree models the costs (time) of a recursive execution of an algorithm.
- The recursion-tree method can be unreliable, just like any method that uses ellipses (...).
- It is good for generating **guesses** of what the runtime could be.

But: Need to verify that the guess is correct.

→ Induction (substitution method)

Substitution method

The most general method to solve a recurrence (prove O and Ω separately):

- 1. Guess the form of the solution:(e.g. using recursion trees, or expansion)
- 2. Verify by induction (inductive step).
- 3. Solve for O-constants n_0 and c (base case of induction)