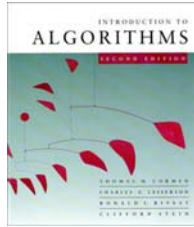




CS 5633 -- Spring 2009



Recurrences and Divide & Conquer

Carola Wenk

Slides courtesy of Charles Leiserson with small changes by Carola Wenk



Merge sort

MERGE-SORT $A[1 \dots n]$

1. If $n = 1$, done.
2. Recursively sort $A[1 \dots \lceil n/2 \rceil]$ and $A[\lceil n/2 \rceil + 1 \dots n]$.
3. “Merge” the 2 sorted lists.

Key subroutine: MERGE



Merging two sorted arrays

20 12
 13 11
 7 9
 2 1

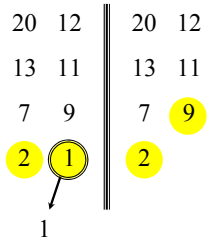


Merging two sorted arrays

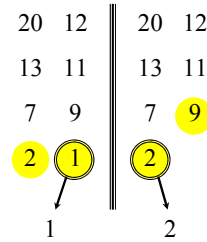
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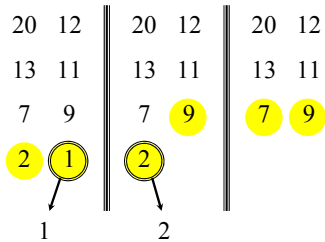
Merging two sorted arrays



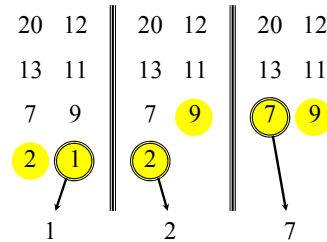
Merging two sorted arrays



Merging two sorted arrays

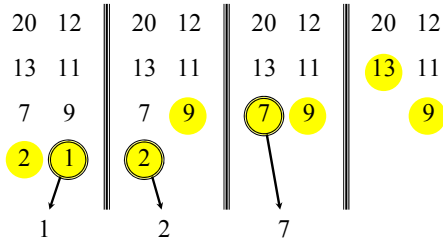


Merging two sorted arrays

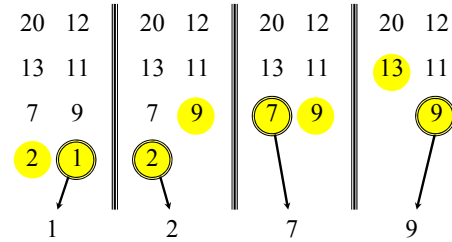




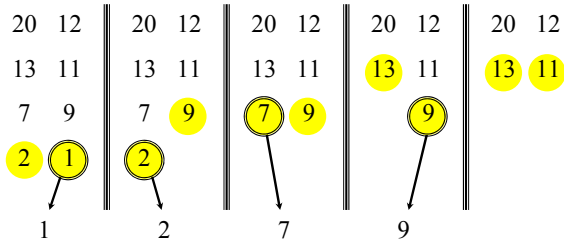
Merging two sorted arrays



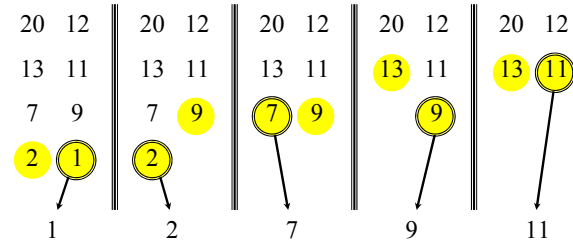
Merging two sorted arrays



Merging two sorted arrays

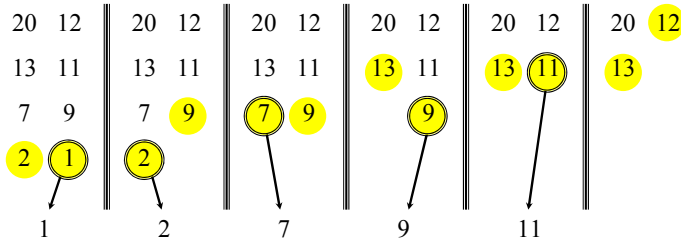


Merging two sorted arrays

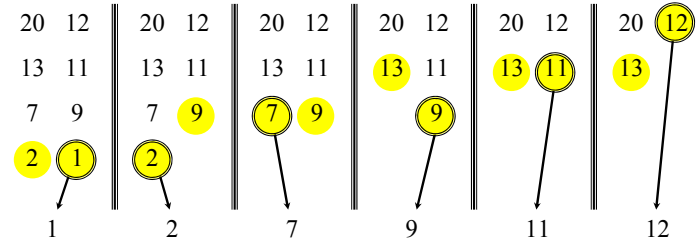




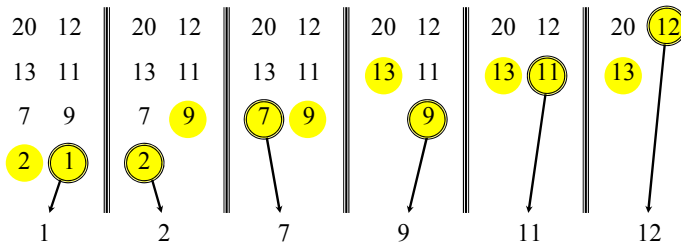
Merging two sorted arrays



Merging two sorted arrays



Merging two sorted arrays



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



Analyzing merge sort

- $T(n)$ | **MERGE-SORT** $A[1 \dots n]$
- d_0 | 1. If $n = 1$, done.
- $2T(n/2)$ | 2. Recursively sort $A[1 \dots \lceil n/2 \rceil]$ and $A[\lceil n/2 \rceil + 1 \dots n]$.
- dn | 3. **“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

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

- Later we shall often omit stating the base case when $T(n) = \Theta(1)$ for sufficiently small n , but only when it has no effect on the asymptotic solution to the recurrence.
- But what does $T(n)$ solve to? I.e., is it $O(n)$ or $O(n^2)$ or $O(n^3)$ or ...?



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.



Example: merge sort

1. **Divide**: Trivial.
2. **Conquer**: Recursively sort 2 subarrays.
3. **Combine**: Linear-time merge.

$$T(n) = 2T(n/2) + \Theta(n)$$

subproblems subproblem size work dividing and combining



Binary search

Find an element in a sorted array:

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

Example: Find 9

3 5 7 8 9 12 15



Binary search

Find an element in a sorted array:

- 1. Divide:** Check middle element.
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Example: Find 9

3 5 7 8 **9** 12 15



Recurrence for binary search

$$T(n) = 1T(n/2) + \Theta(1)$$

subproblems subproblem size work dividing and combining



Recurrence for merge sort

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

- How do we solve $T(n)$? I.e., how do we find out if it is $O(n)$ or $O(n^2)$ or ...?



Recursion tree

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



Recursion tree

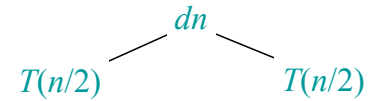
Solve $T(n) = 2T(n/2) + dn$, where $d > 0$ is constant.

$$T(n)$$



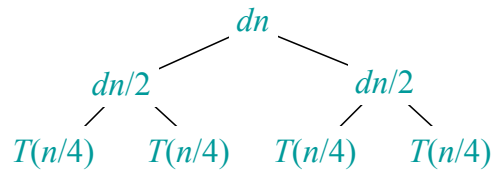
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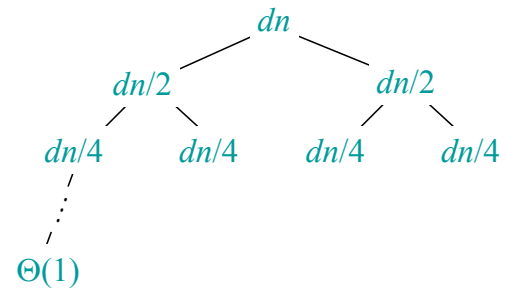
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Recursion tree

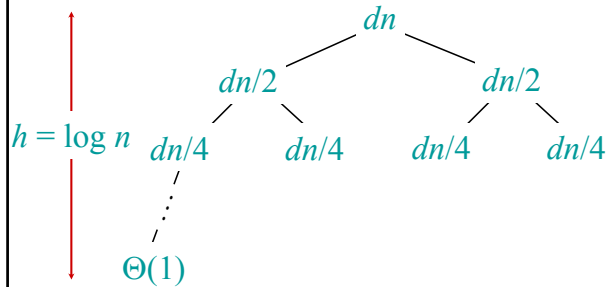
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Recursion tree

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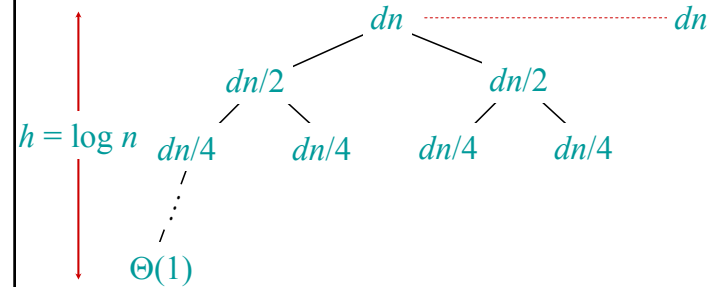
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Recursion tree

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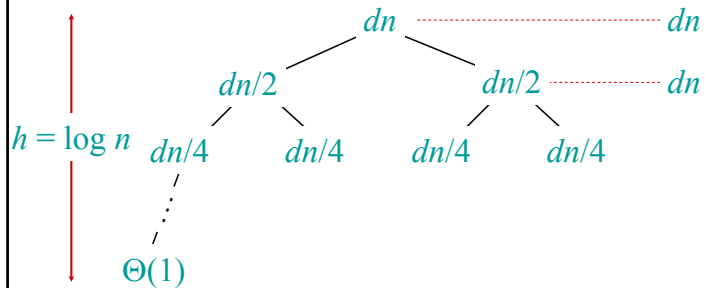
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Recursion tree

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



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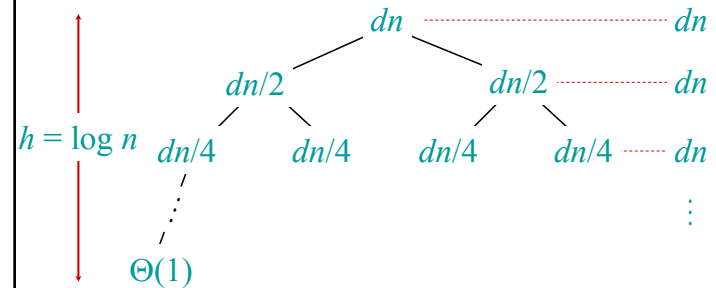
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Recursion tree

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



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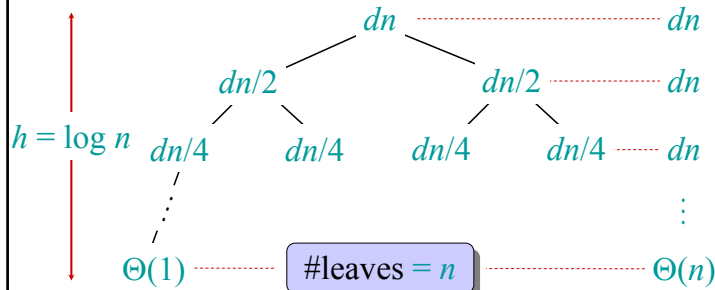
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Recursion tree

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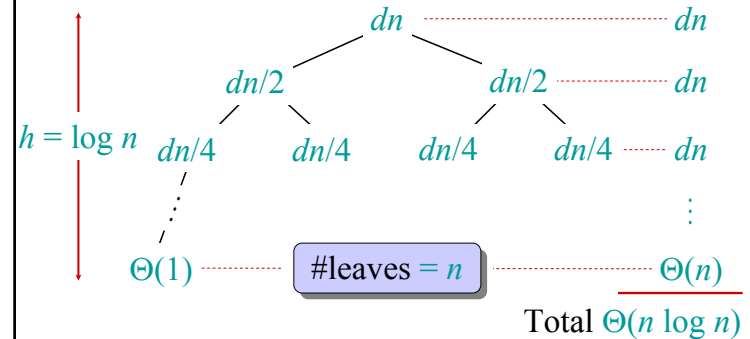
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Recursion tree

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



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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?)

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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 right.
 → Induction (substitution method)

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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)