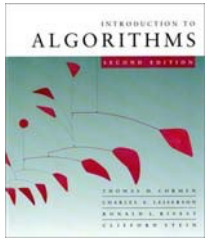


CS60020: Foundations of Algorithm Design and Machine Learning

Sourangshu Bhattacharya

DIVIDE AND CONQUER

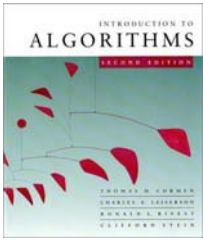


Merge sort

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

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

Key subroutine: **MERGE**



Merging two sorted arrays

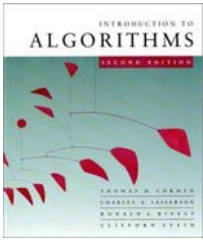
20 12

13 11

7 9

2

1

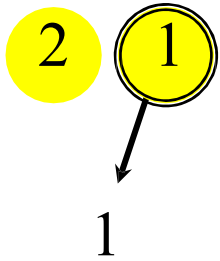


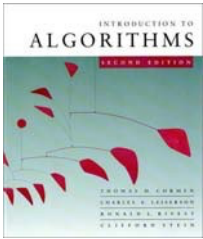
Merging two sorted arrays

20 12

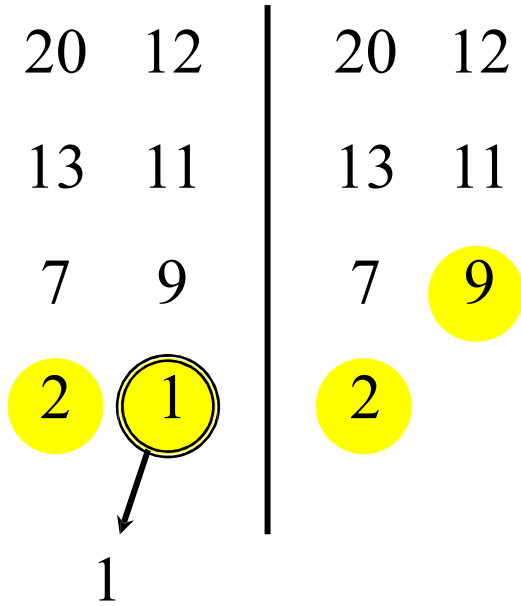
13 11

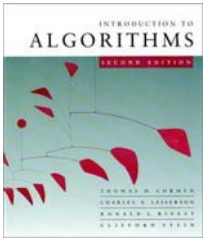
7 9



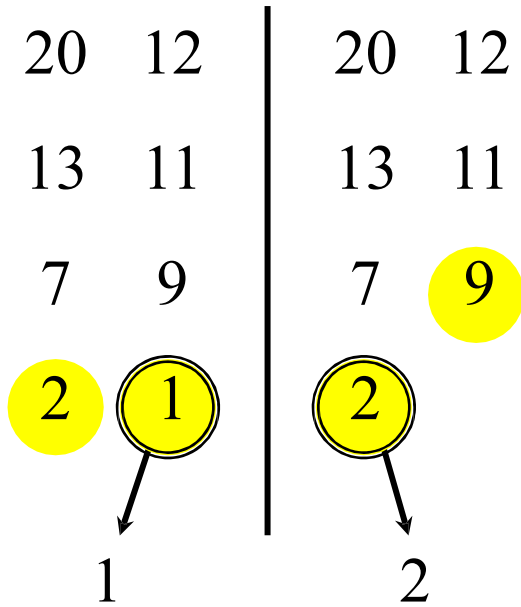


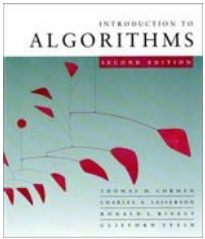
Merging two sorted arrays



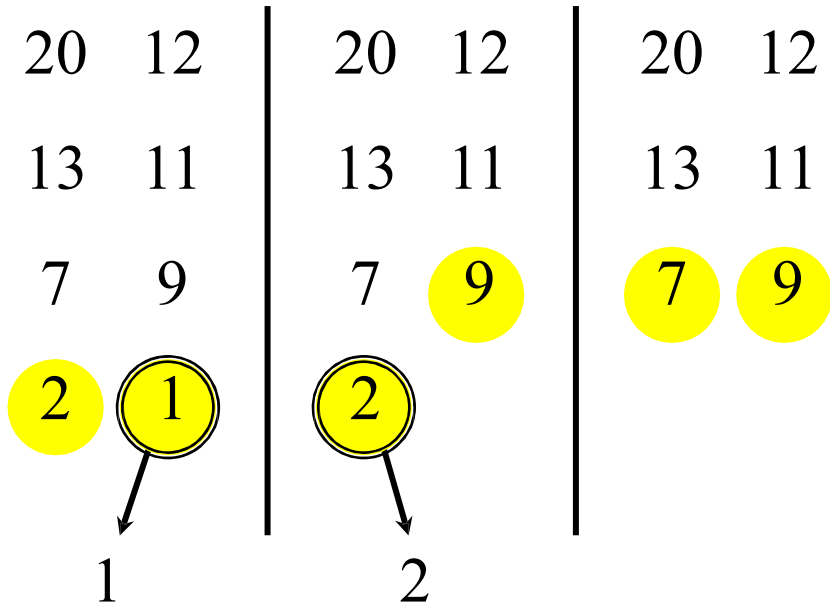


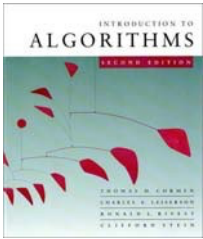
Merging two sorted arrays



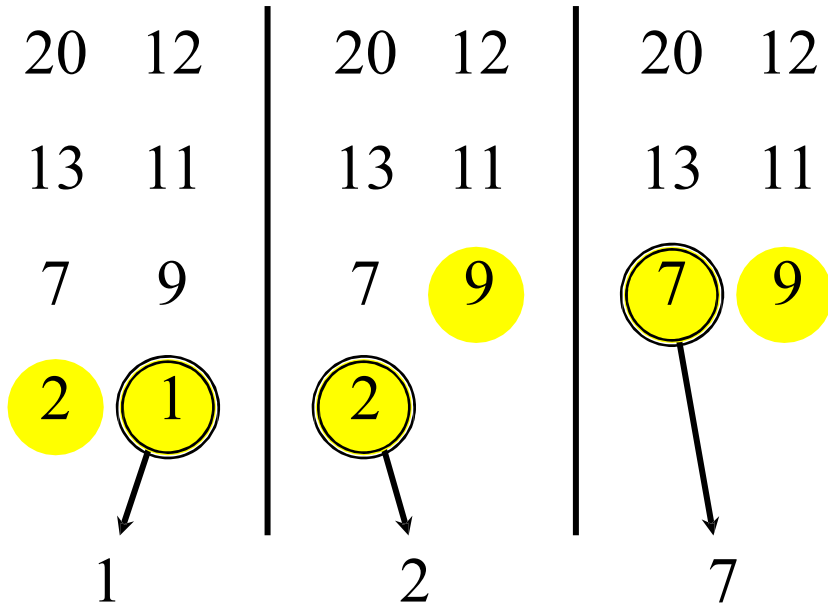


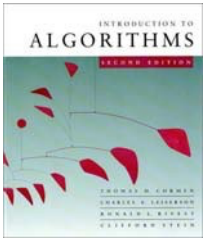
Merging two sorted arrays



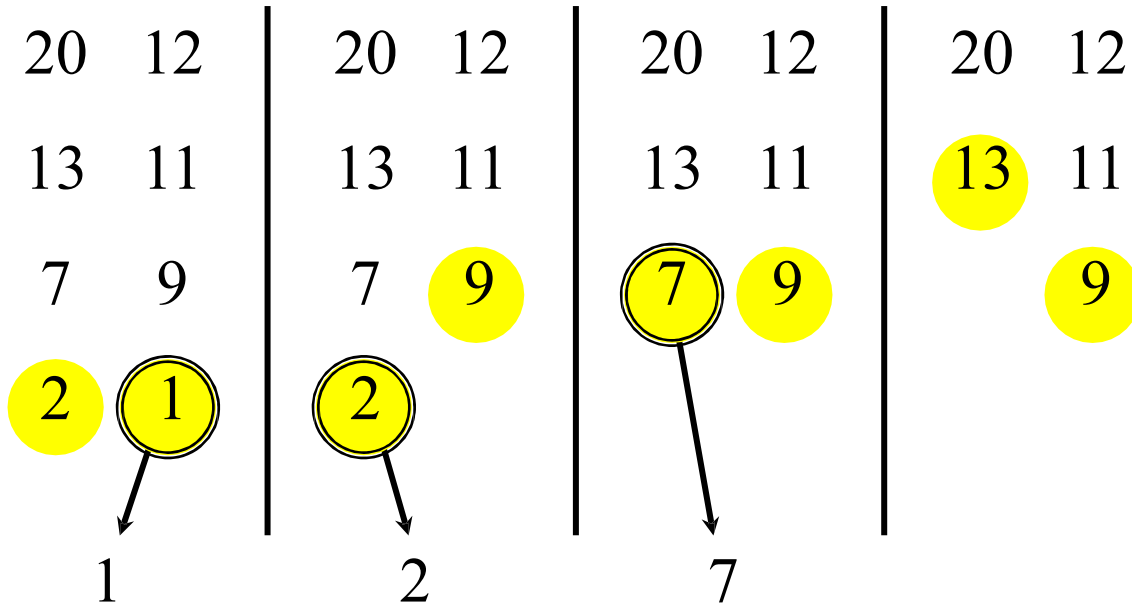


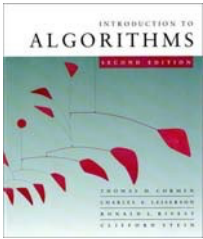
Merging two sorted arrays



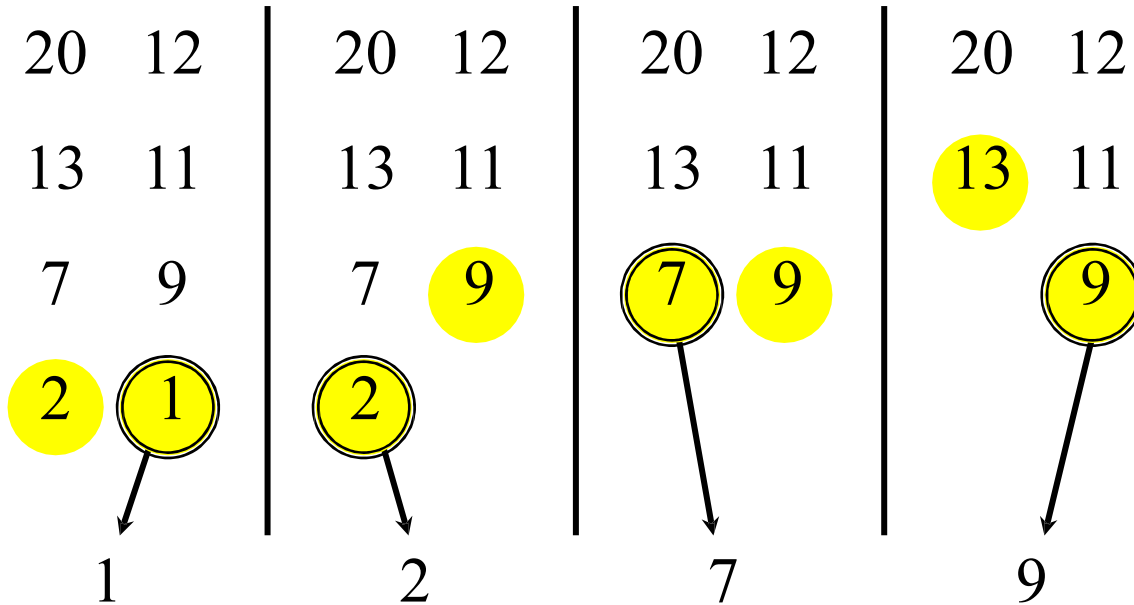


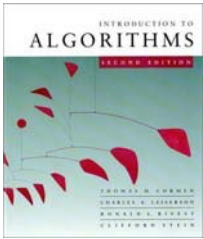
Merging two sorted arrays



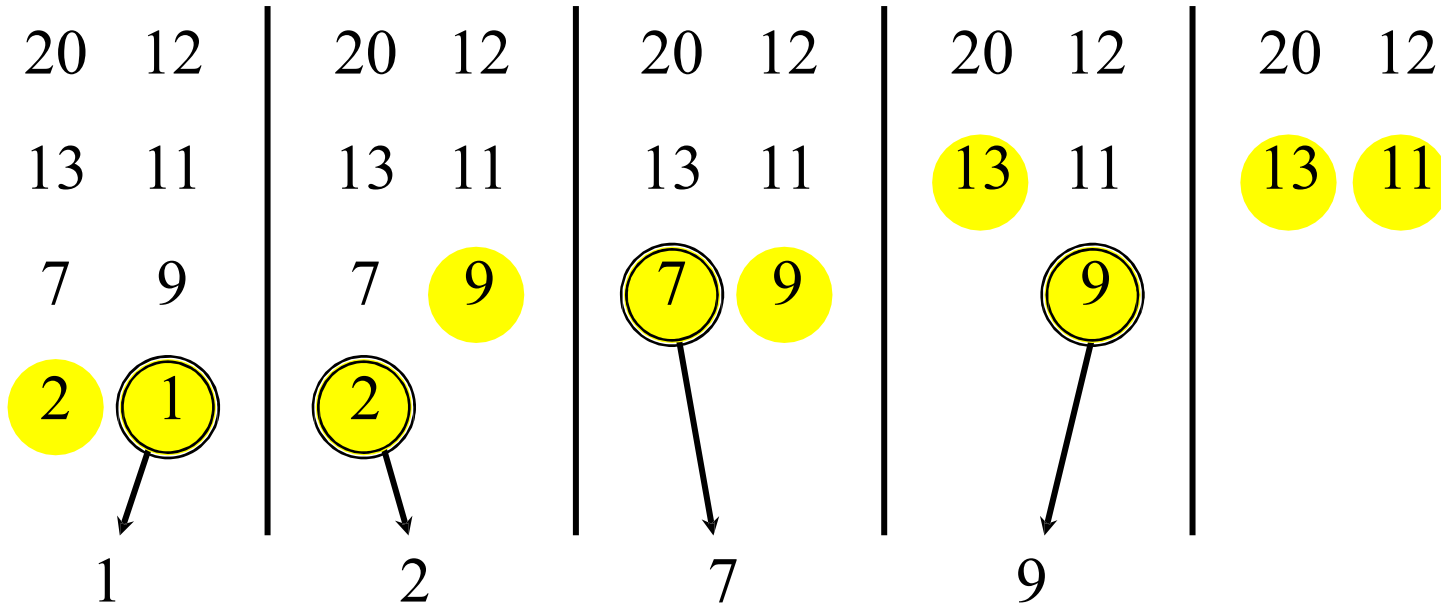


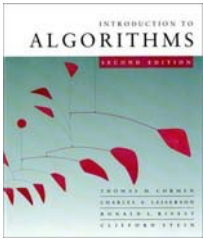
Merging two sorted arrays



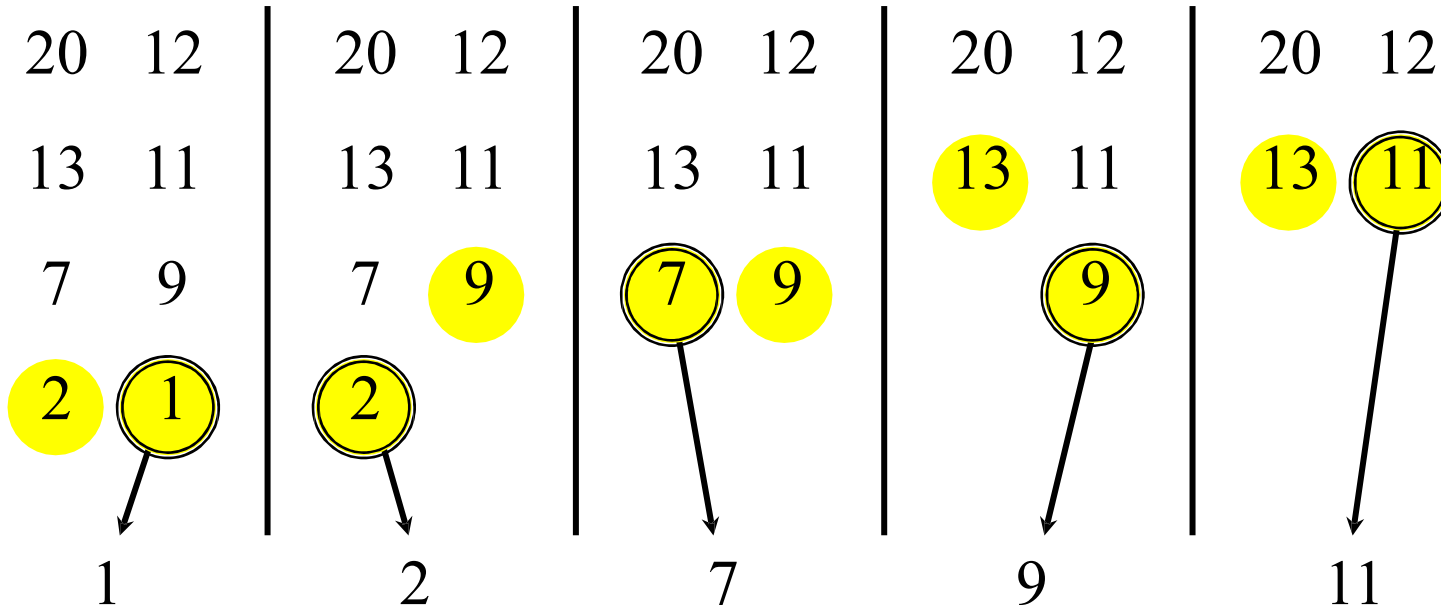


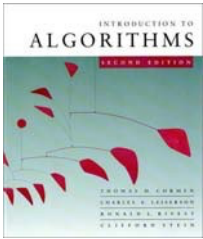
Merging two sorted arrays



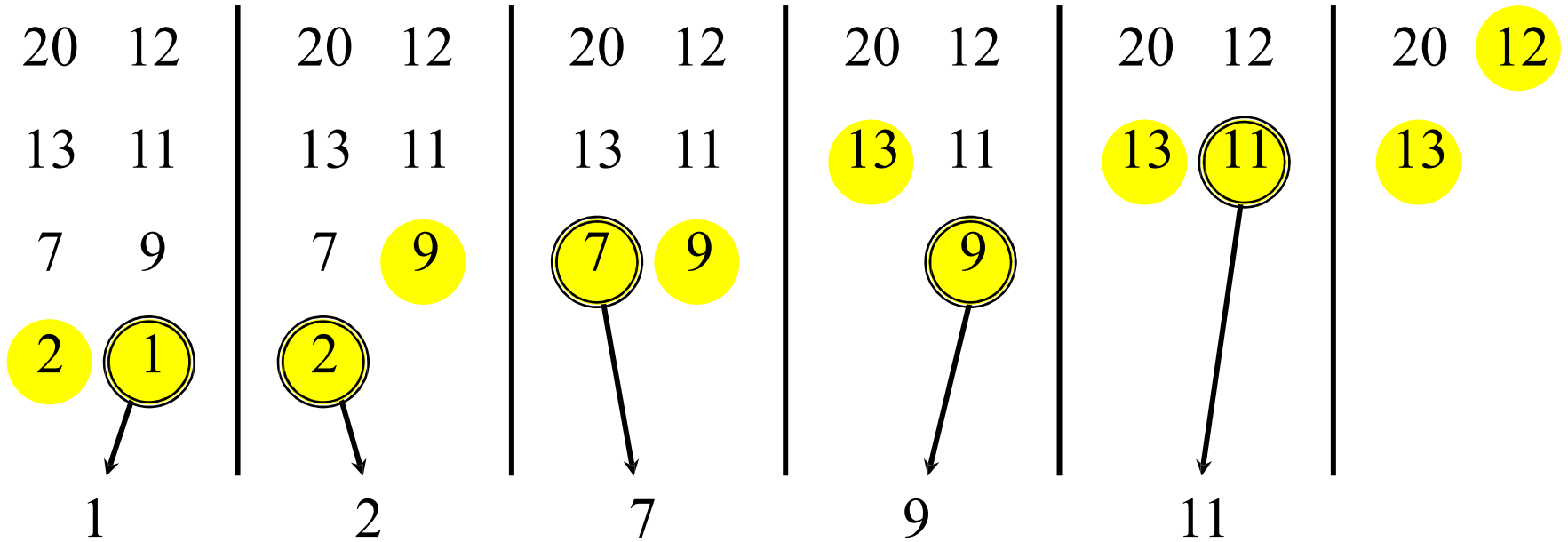


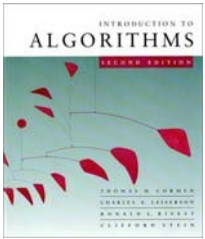
Merging two sorted arrays



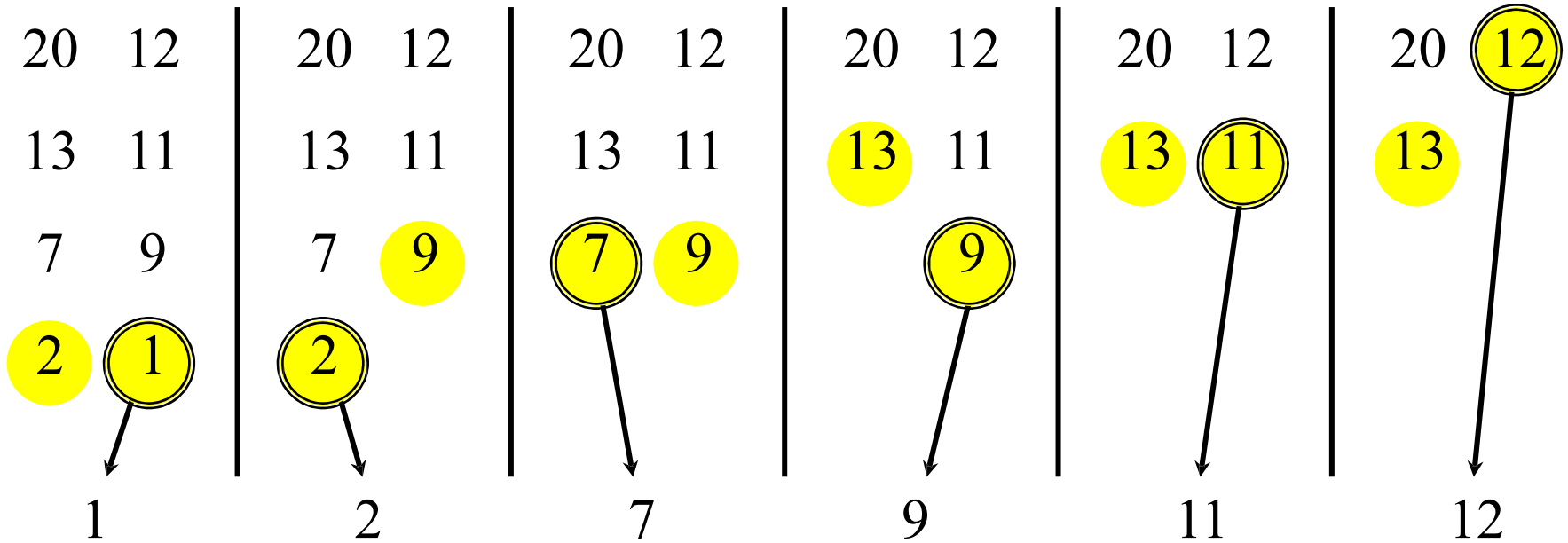


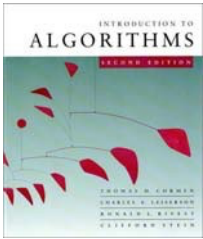
Merging two sorted arrays



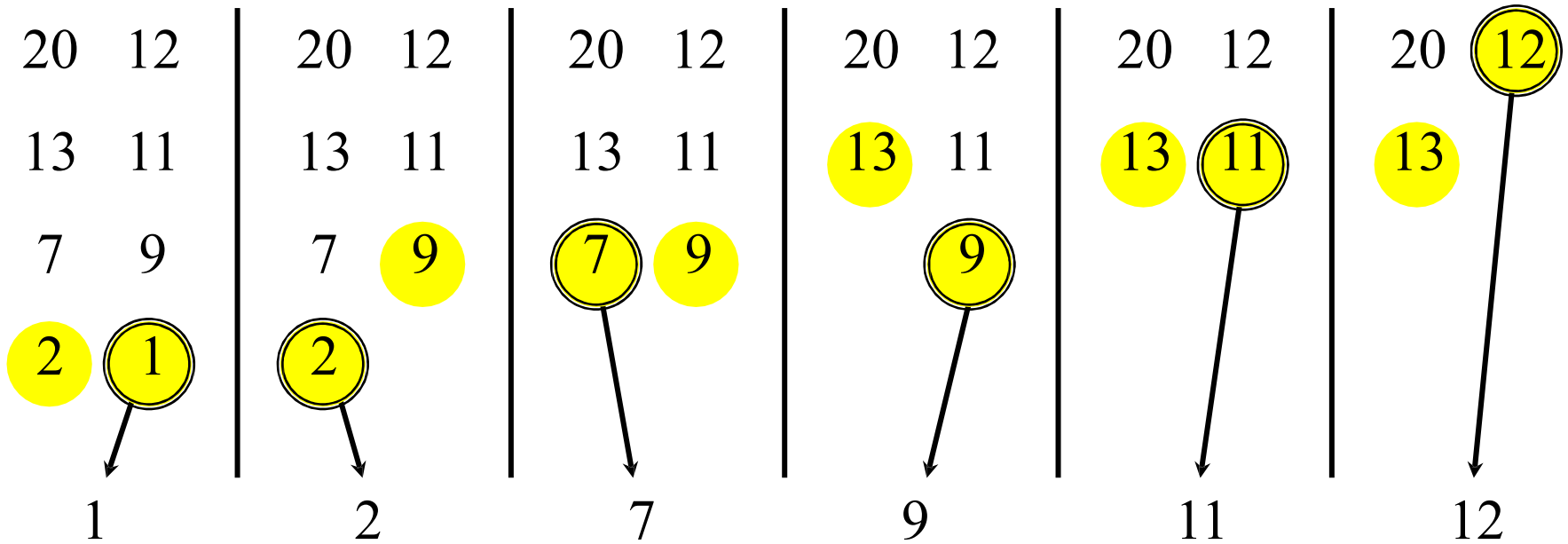


Merging two sorted arrays





Merging two sorted arrays

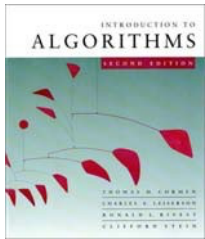


Time = $\Theta(n)$ to merge a total of n elements (linear time).

Merge

MERGE(A, p, q, r)

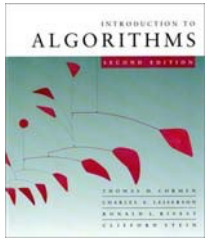
```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else  $A[k] = R[j]$ 
17          $j = j + 1$ 
```



Analyzing merge sort

	$T(n)$		MERGE-SORT $A[1 \dots n]$
	$\Theta(1)$		1. If $n = 1$, done.
<i>Abuse</i>	$2T(n/2)$		2. Recursively sort $A[1 \dots \lfloor n/2 \rfloor]$ and $A[\lfloor n/2 \rfloor + 1 \dots n]$.
	$\Theta(n)$		3. “Merge” the 2 sorted lists

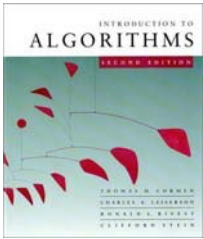
Sloppiness: Should be $T(\lfloor n/2 \rfloor) + T(\lfloor n/2 \rfloor)$, but it turns out not to matter asymptotically.



Recurrence for merge sort

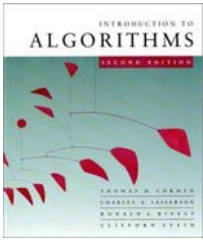
$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

- We shall usually 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.
- CLRS and Lecture 2 provide several ways to find a good upper bound on $T(n)$.



Recursion tree

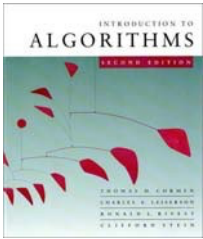
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.



Recursion tree

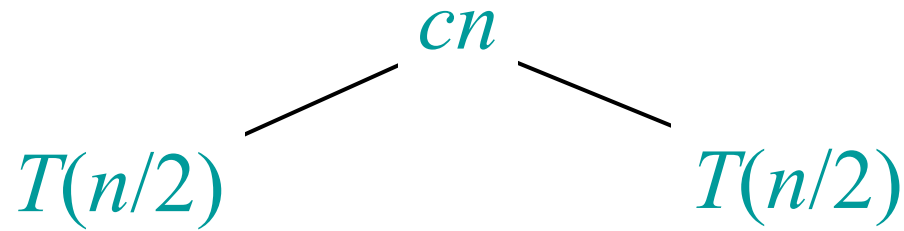
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

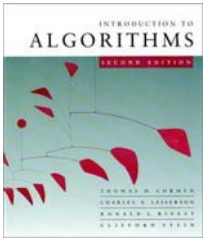
$$T(n)$$



Recursion tree

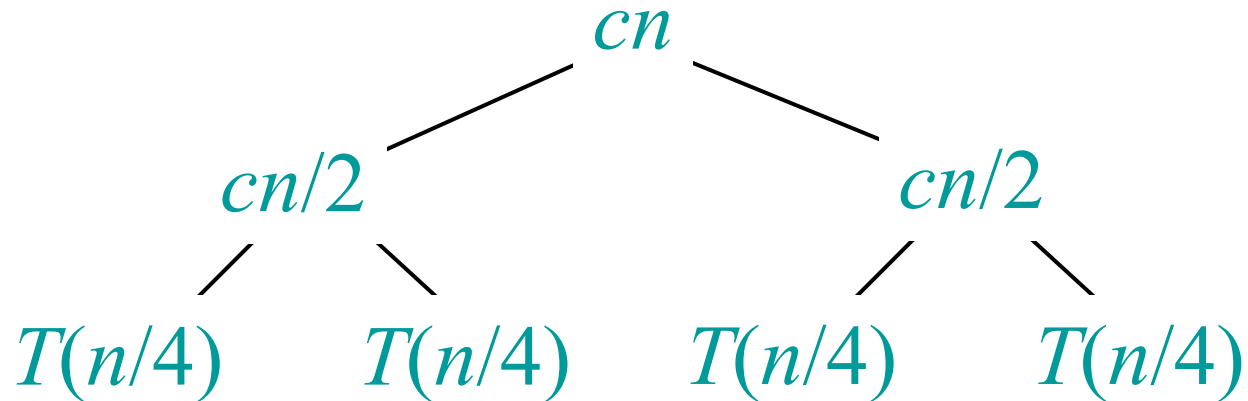
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

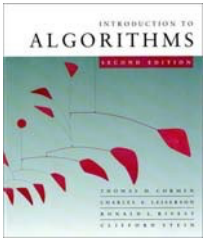




Recursion tree

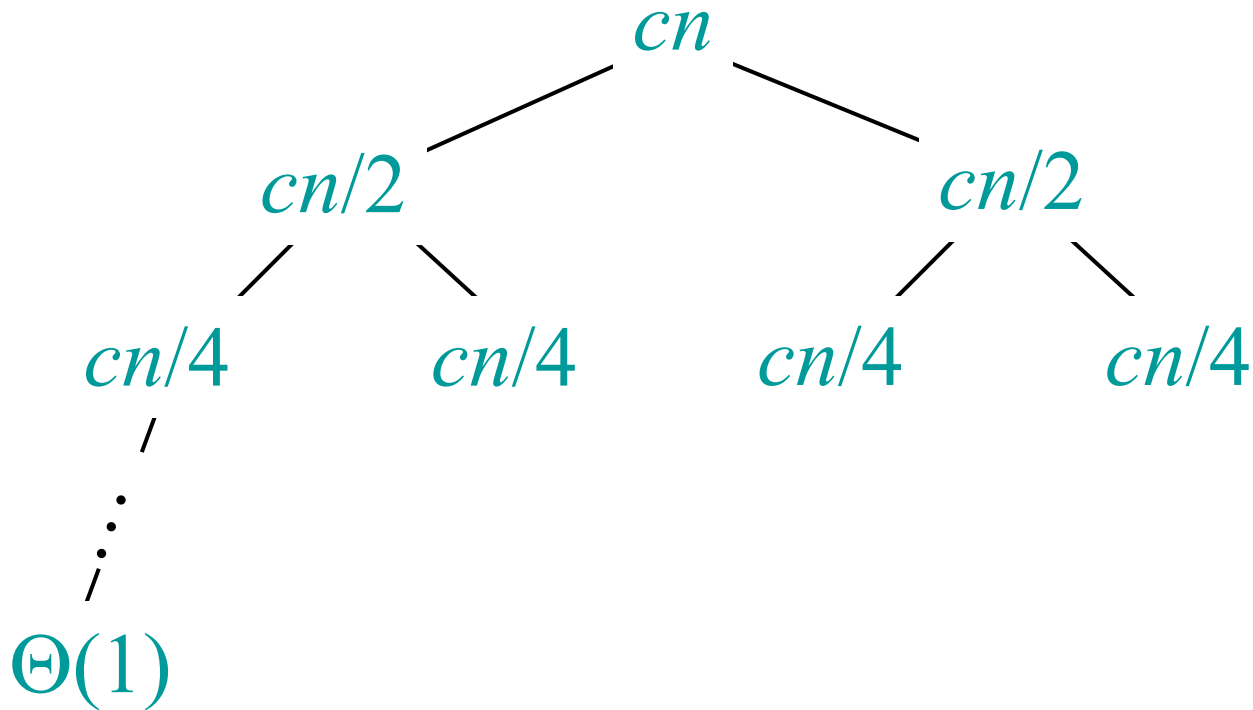
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

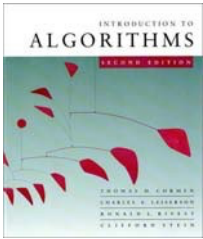




Recursion tree

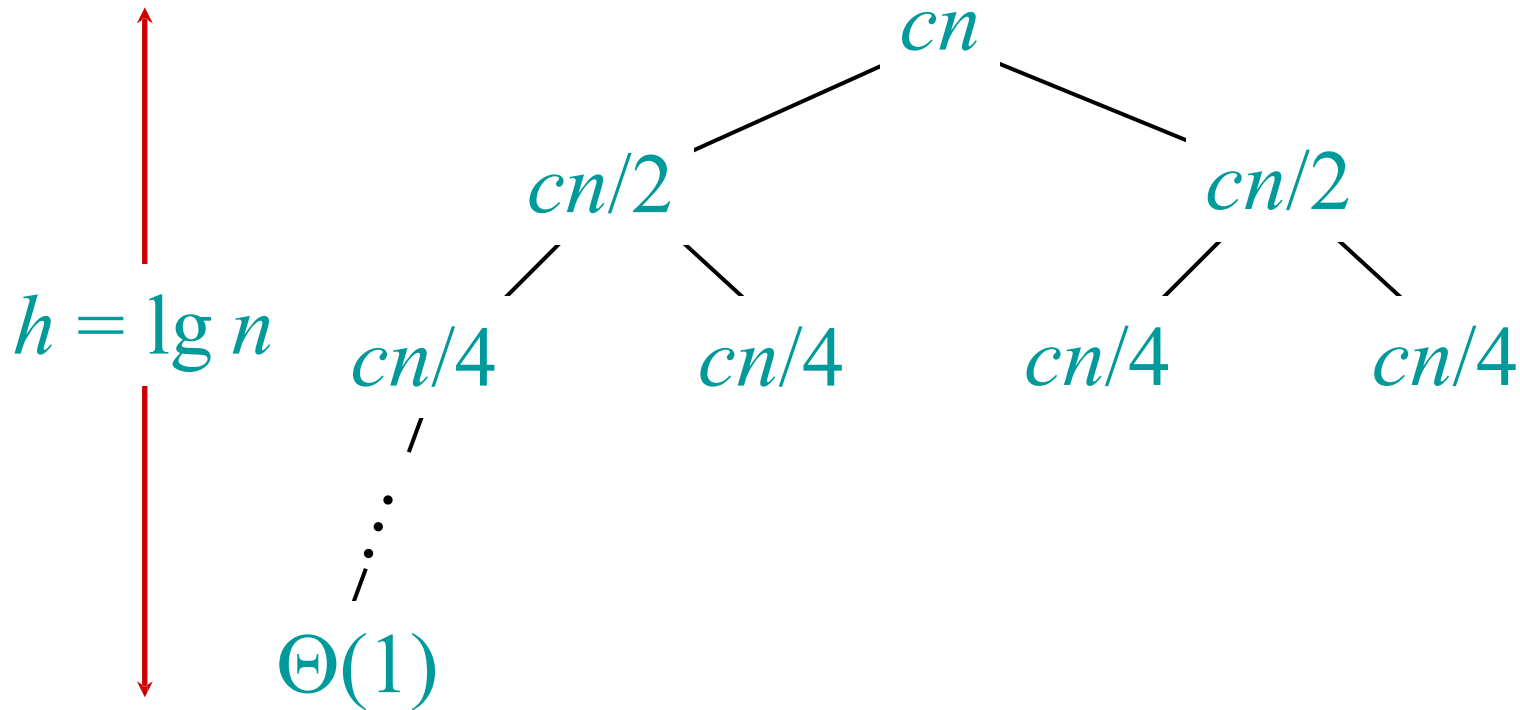
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

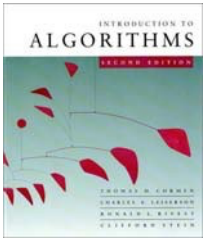




Recursion tree

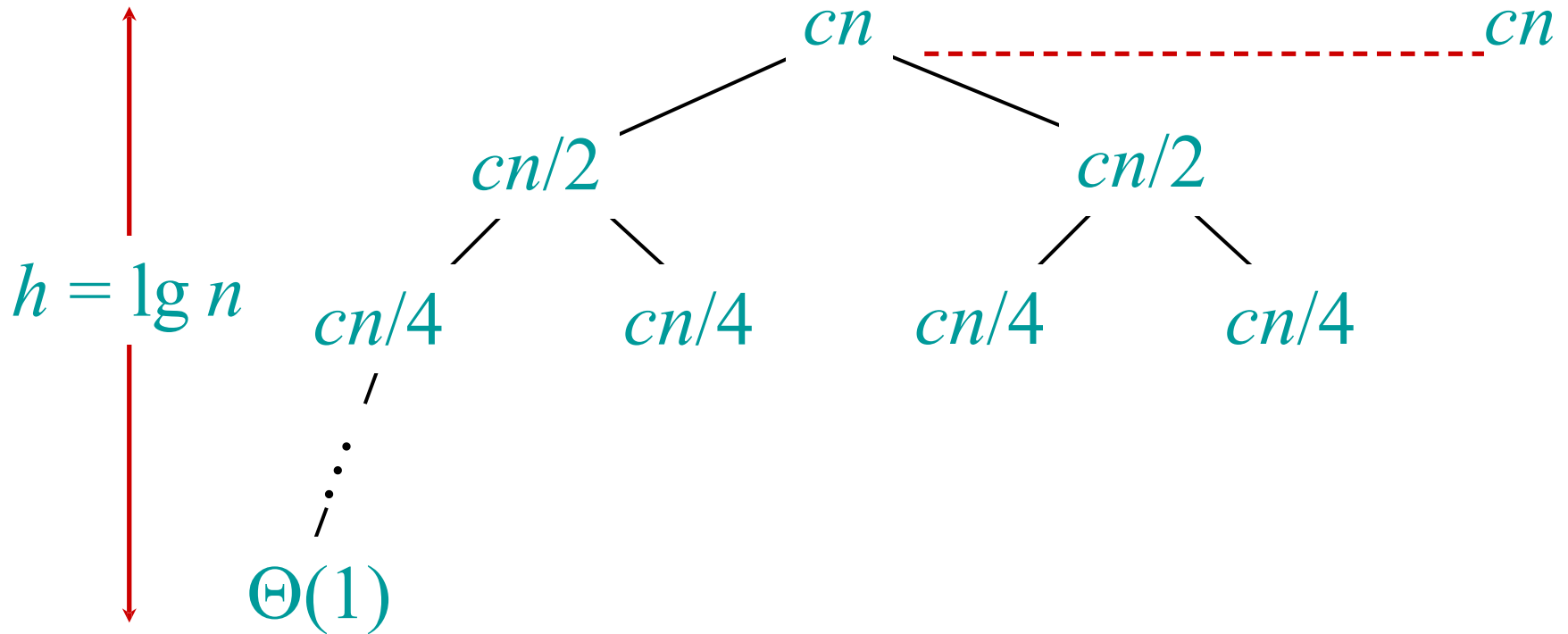
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

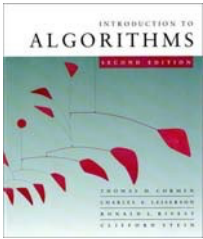




Recursion tree

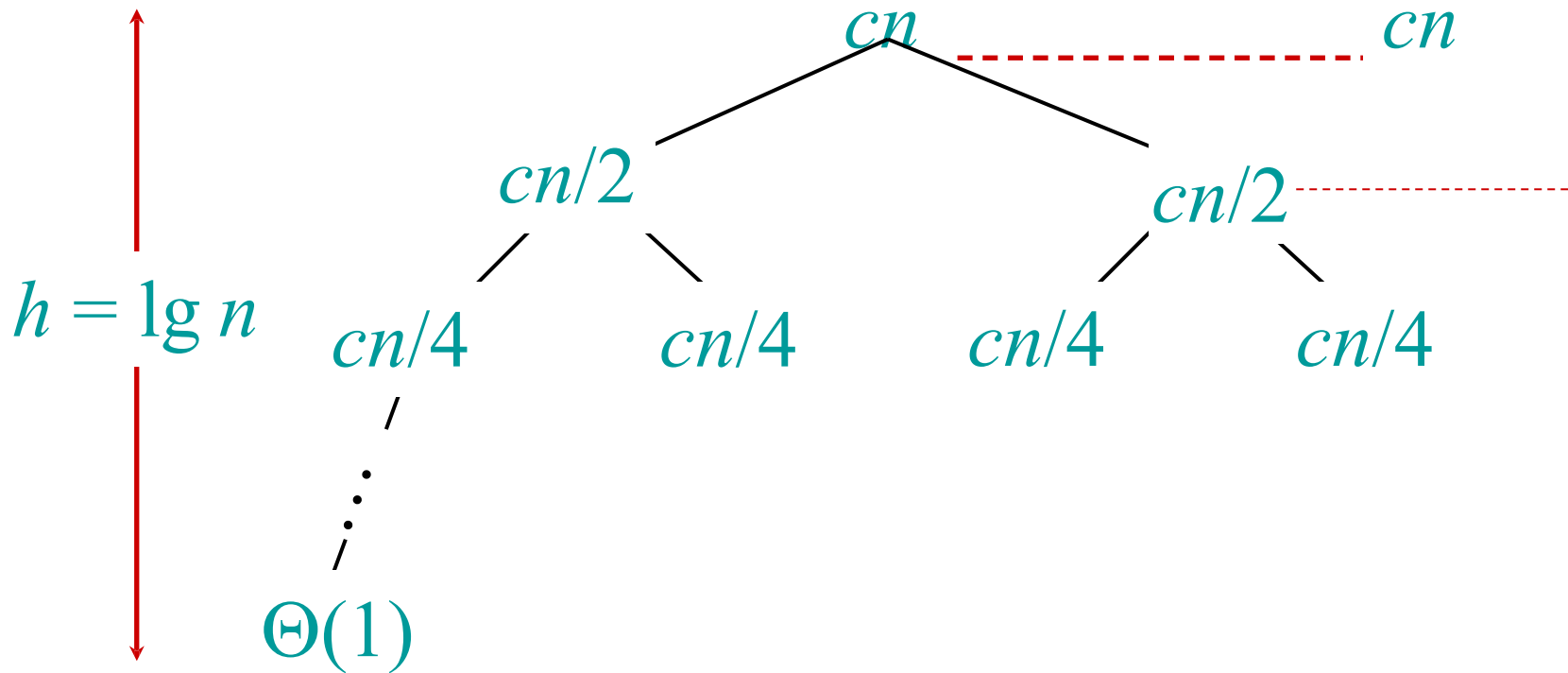
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

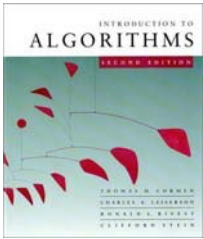




Recursion tree

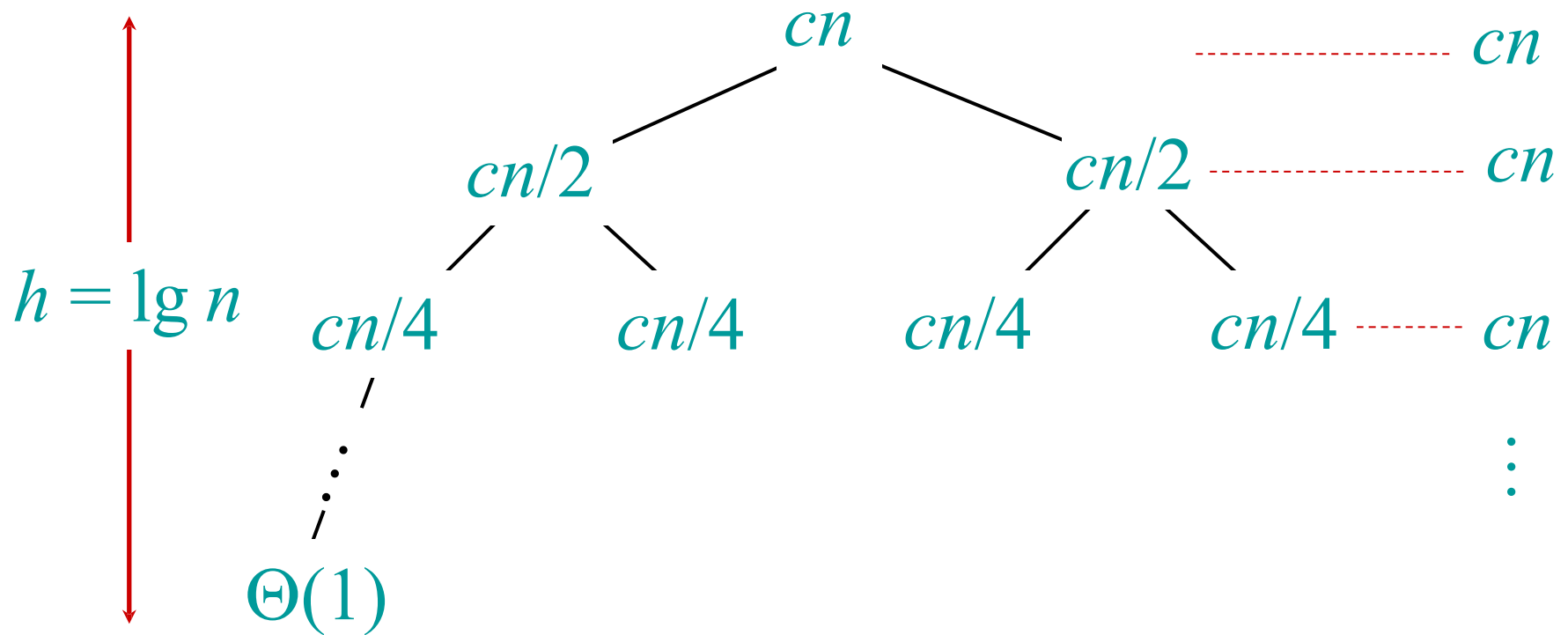
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

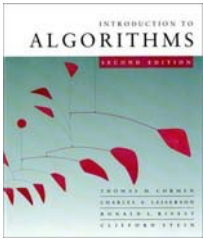




Recursion tree

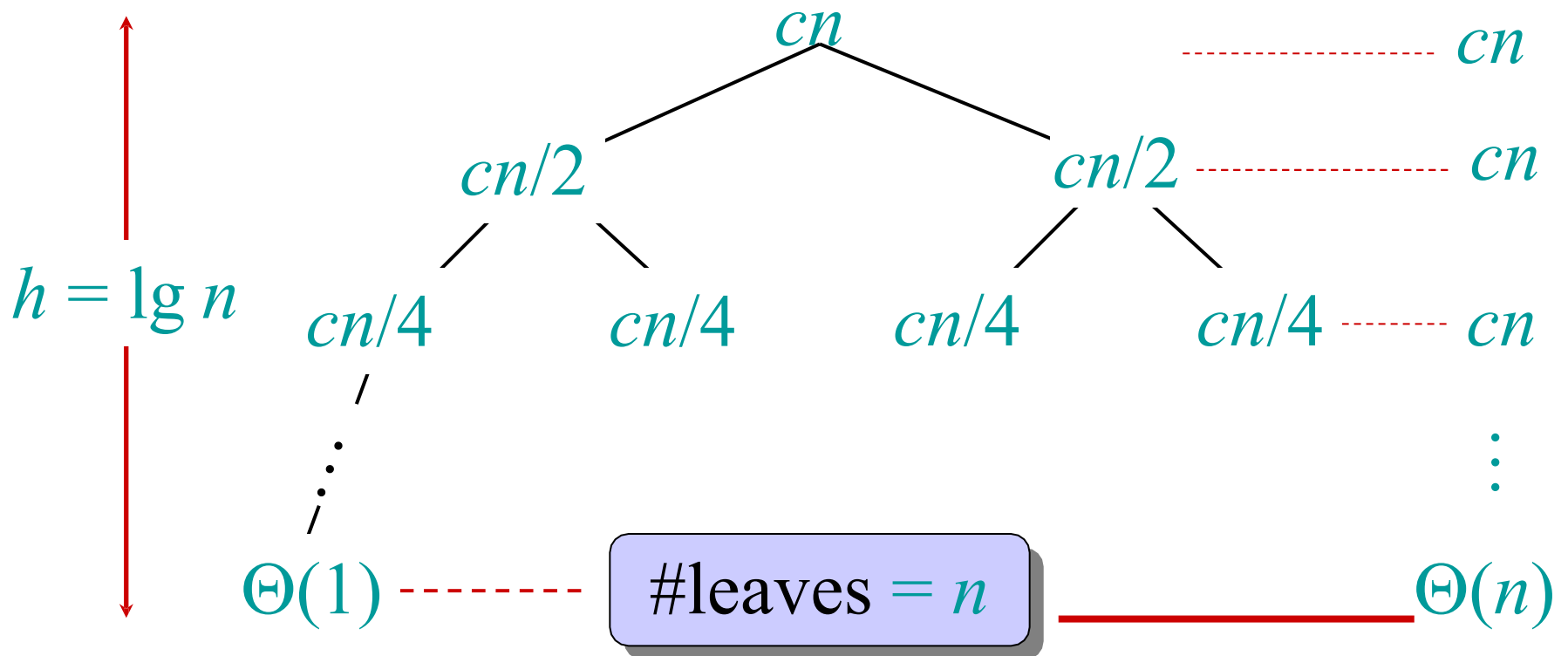
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

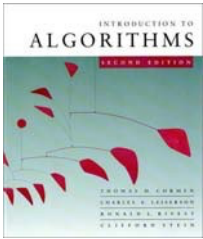




Recursion tree

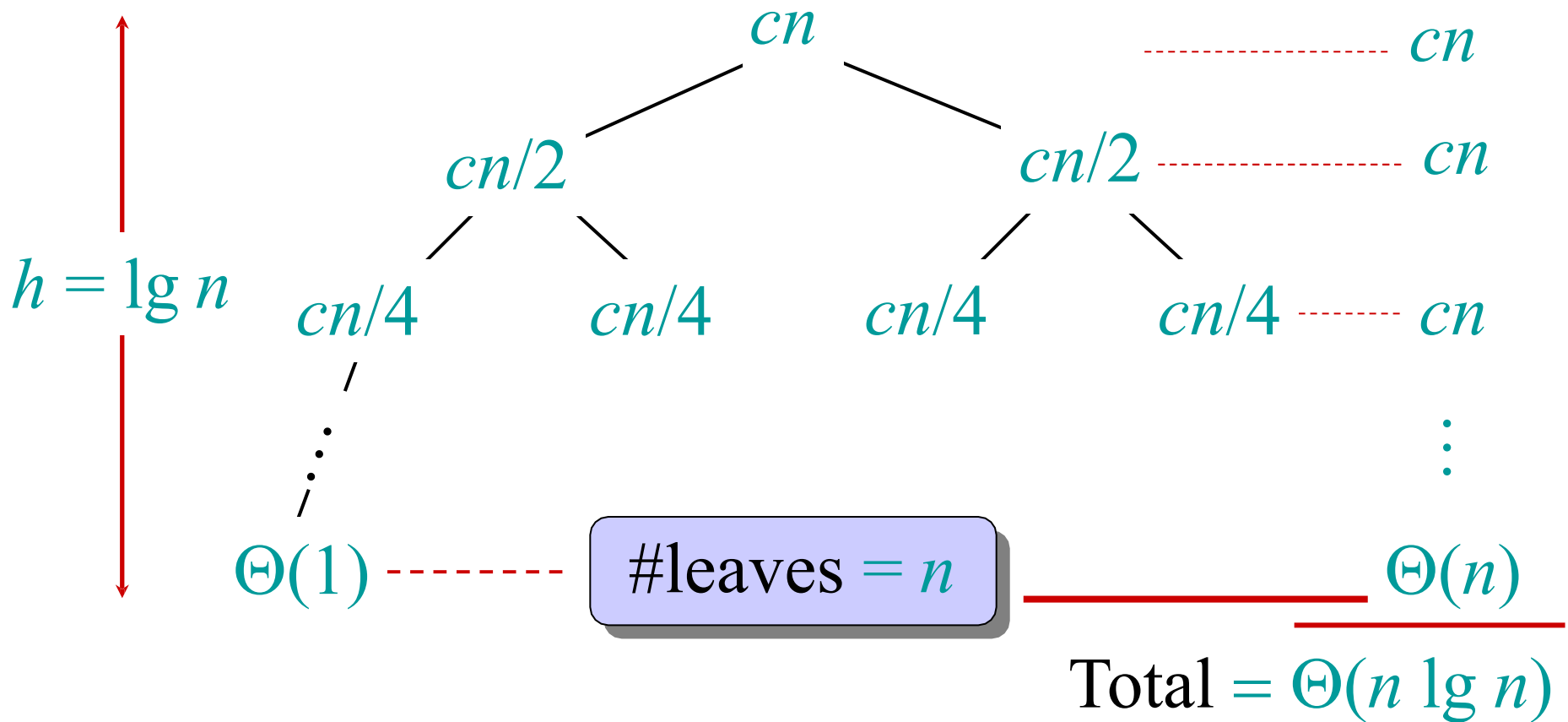
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

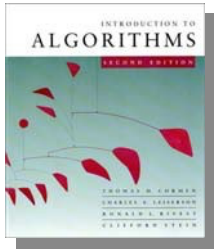




Recursion tree

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





Merge sort

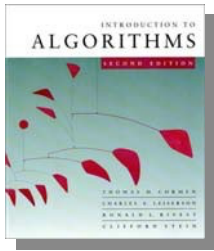
- 1. *Divide*:** Trivial.
- 2. *Conquer*:** Recursively sort 2 subarrays.
- 3. *Combine*:** Linear-time merge.

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

subproblems

subproblem size

work dividing and combining



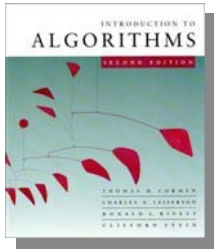
Master theorem

$$T(n) = a T(n/b) + f(n)$$

CASE 1: $f(n) = O(n^{\log_b a - \varepsilon})$, constant $\varepsilon > 0$
 $\Rightarrow T(n) = \Theta(n^{\log_b a})$.

CASE 2: $f(n) = \Theta(n^{\log_b a})$
 $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$.

CASE 3: $f(n) = \Omega(n^{\log_b a + \varepsilon})$, constant $\varepsilon > 0$,
and regularity condition
 $\Rightarrow T(n) = \Theta(f(n))$.



Master theorem

$$T(n) = a T(n/b) + f(n)$$

CASE 1: $f(n) = O(n^{\log_b a - \varepsilon})$, constant $\varepsilon > 0$
 $\Rightarrow T(n) = \Theta(n^{\log_b a})$.

CASE 2: $f(n) = \Theta(n^{\log_b a})$
 $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$.

CASE 3: $f(n) = \Omega(n^{\log_b a + \varepsilon})$, constant $\varepsilon > 0$,
and regularity condition
 $\Rightarrow T(n) = \Theta(f(n))$.

Merge sort: $a = 2, b = 2 \Rightarrow n^{\log_b a} = n^{\log_2 2} = n$
 \Rightarrow **CASE 2** $\Rightarrow T(n) = \Theta(n \lg n)$.