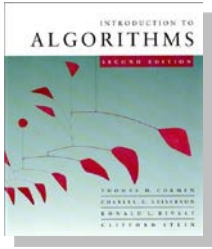


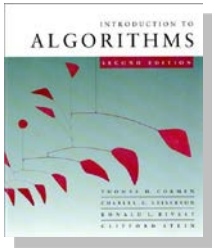
The divide-and-conquer design paradigm

1. *Divide* the problem (instance) into subproblems.
2. *Conquer* the subproblems by solving them recursively.
3. *Combine* subproblem solutions.



Merge sort

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



Merge sort

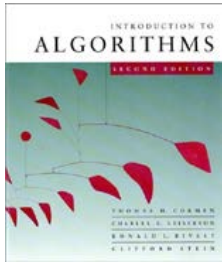
1. **Divide:** Trivial.
2. **Conquer:** Recursively sort 2 subarrays.
3. **Combine:** Linear-time merge.

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

subproblems

subproblem size

work dividing and combining

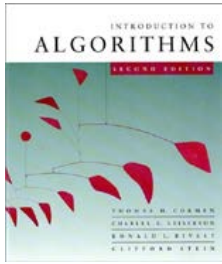


Merge sort

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

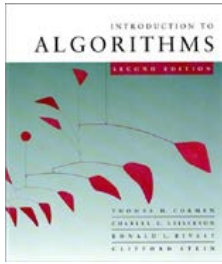
20 12

13 11

7 9

2

1

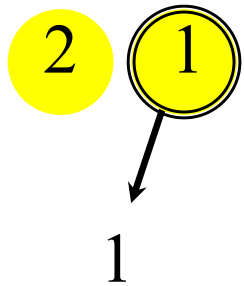


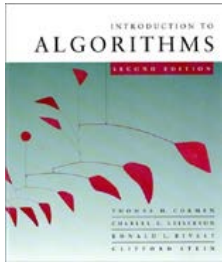
Merging two sorted arrays

20 12

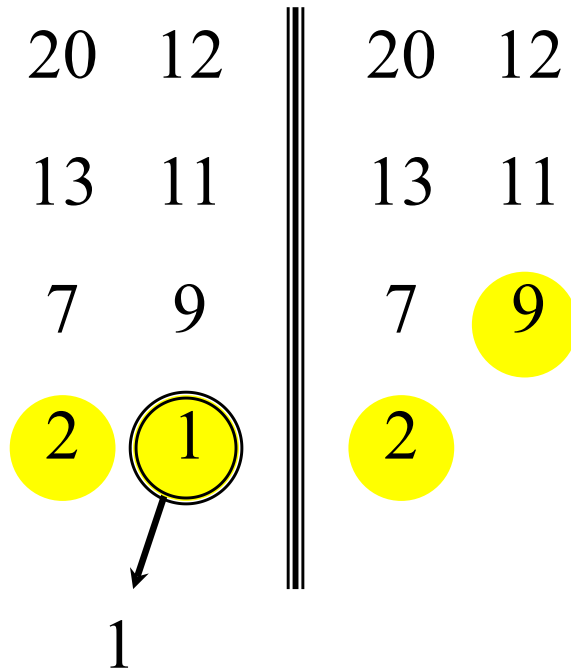
13 11

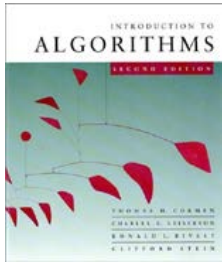
7 9



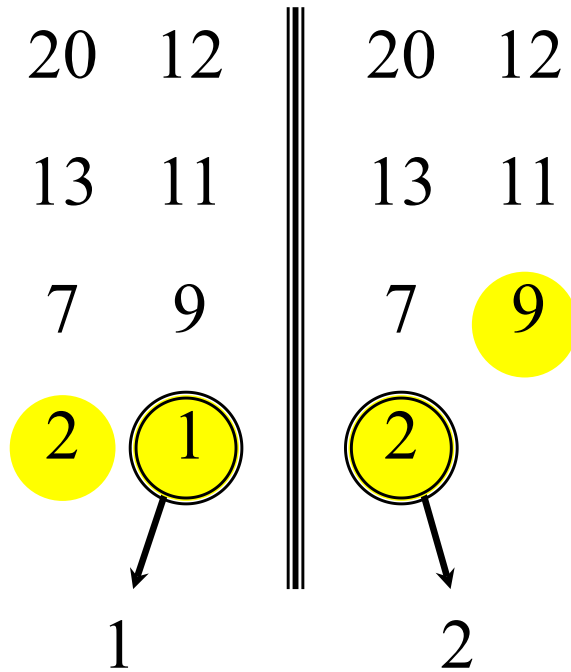


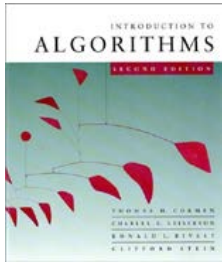
Merging two sorted arrays



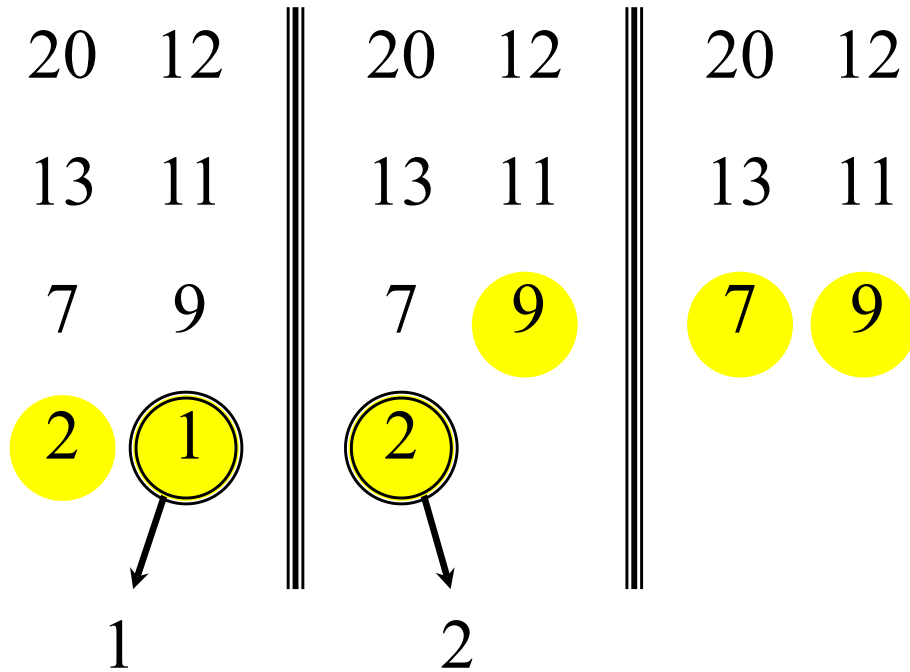


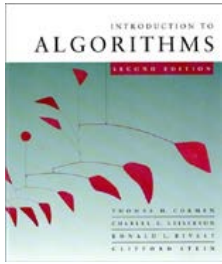
Merging two sorted arrays



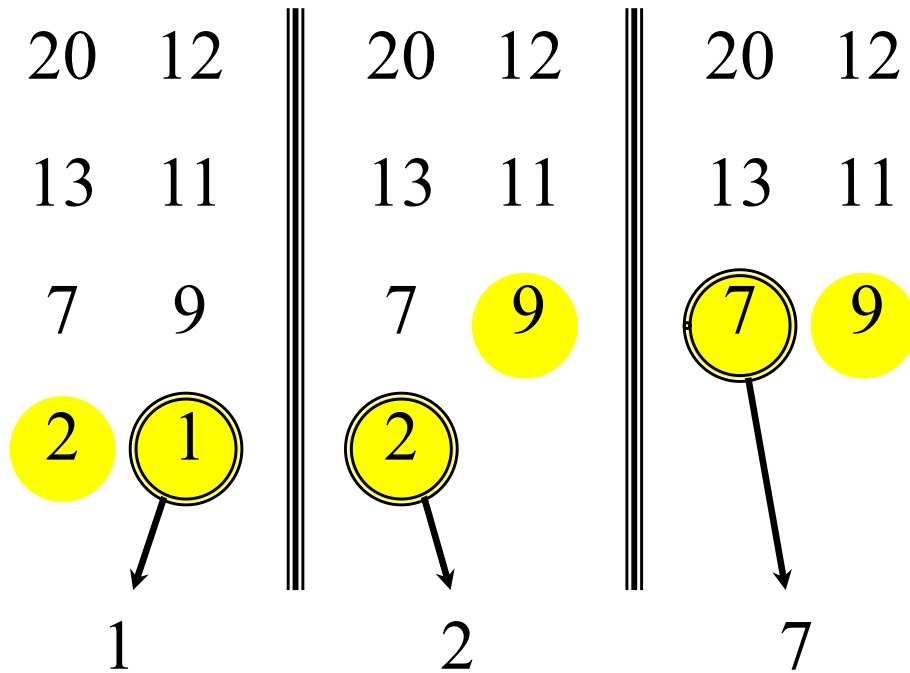


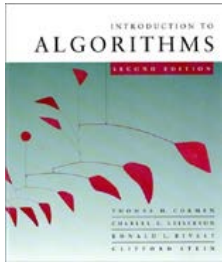
Merging two sorted arrays



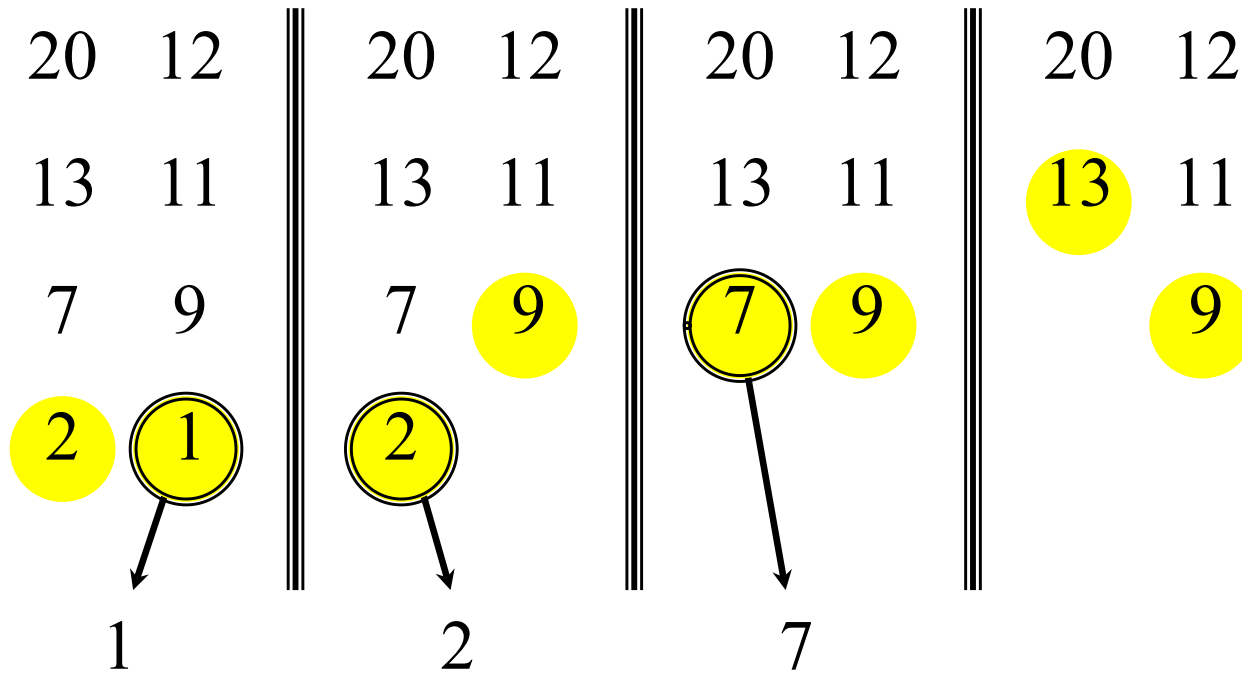


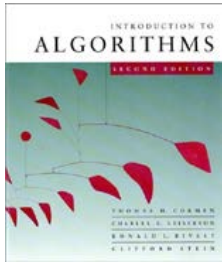
Merging two sorted arrays



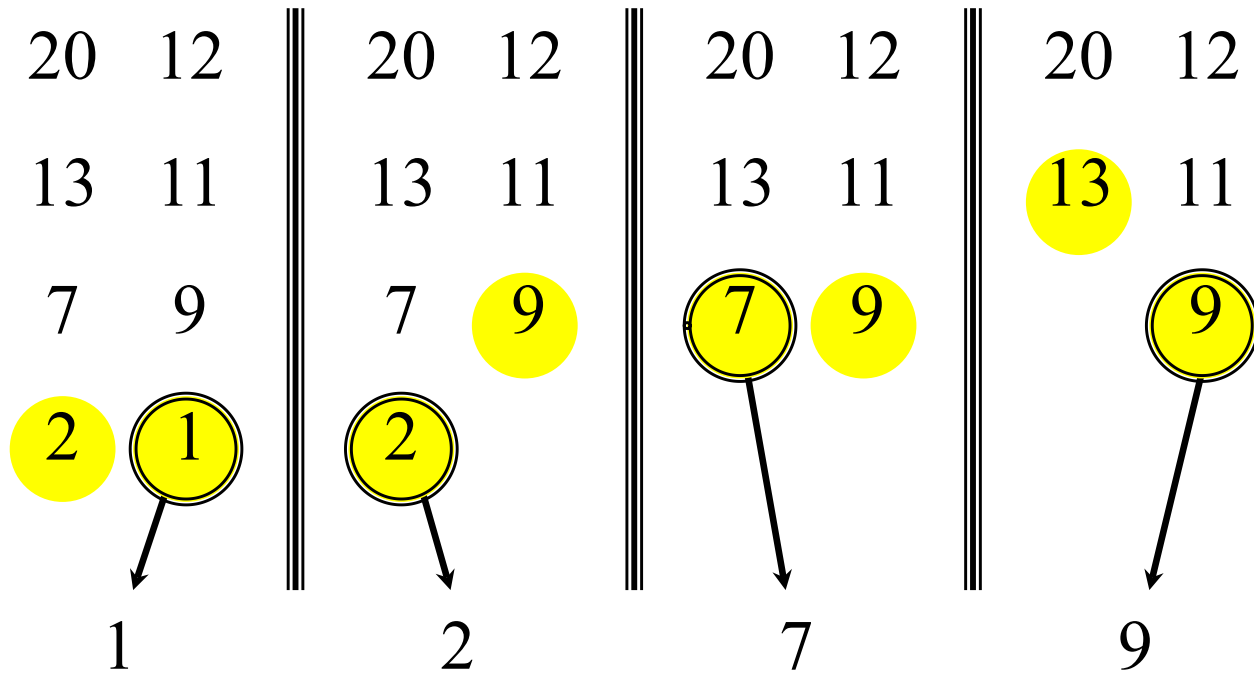


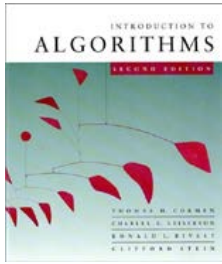
Merging two sorted arrays



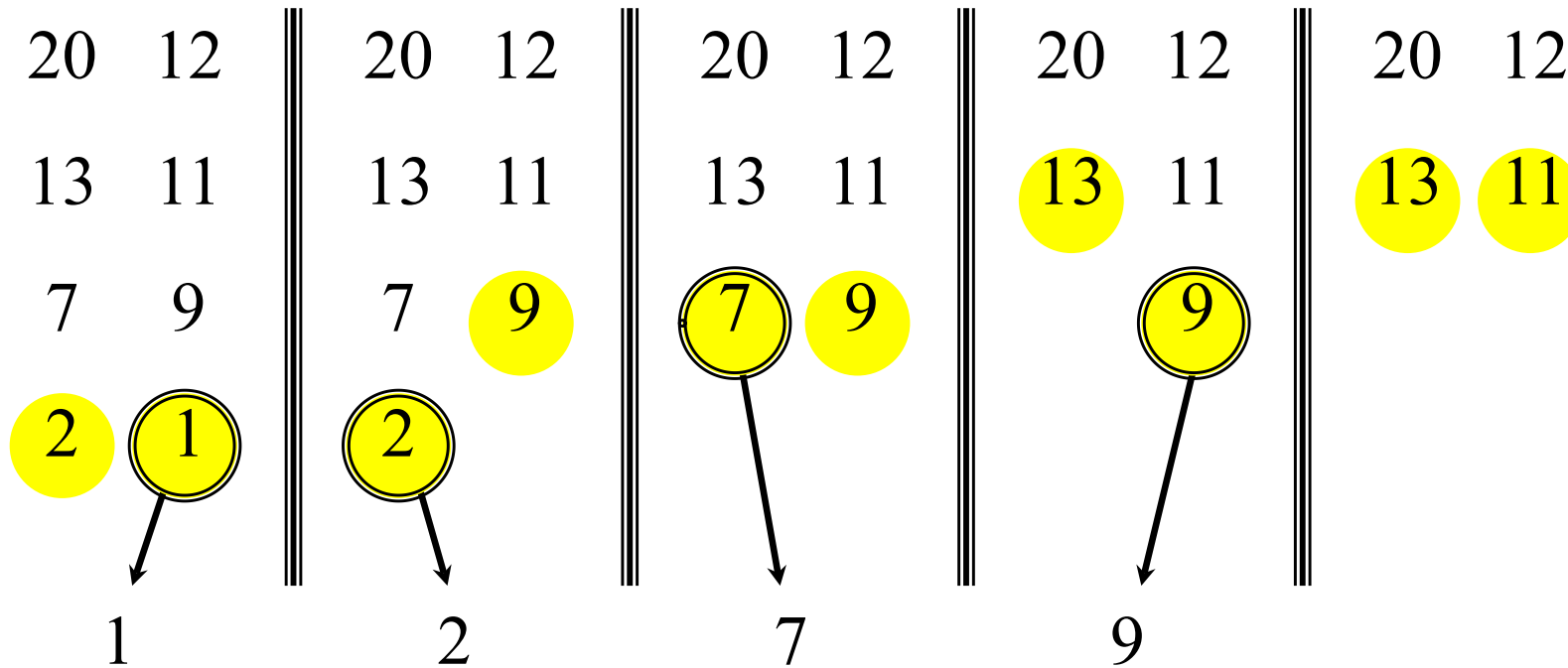


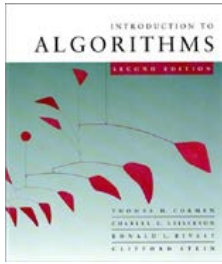
Merging two sorted arrays



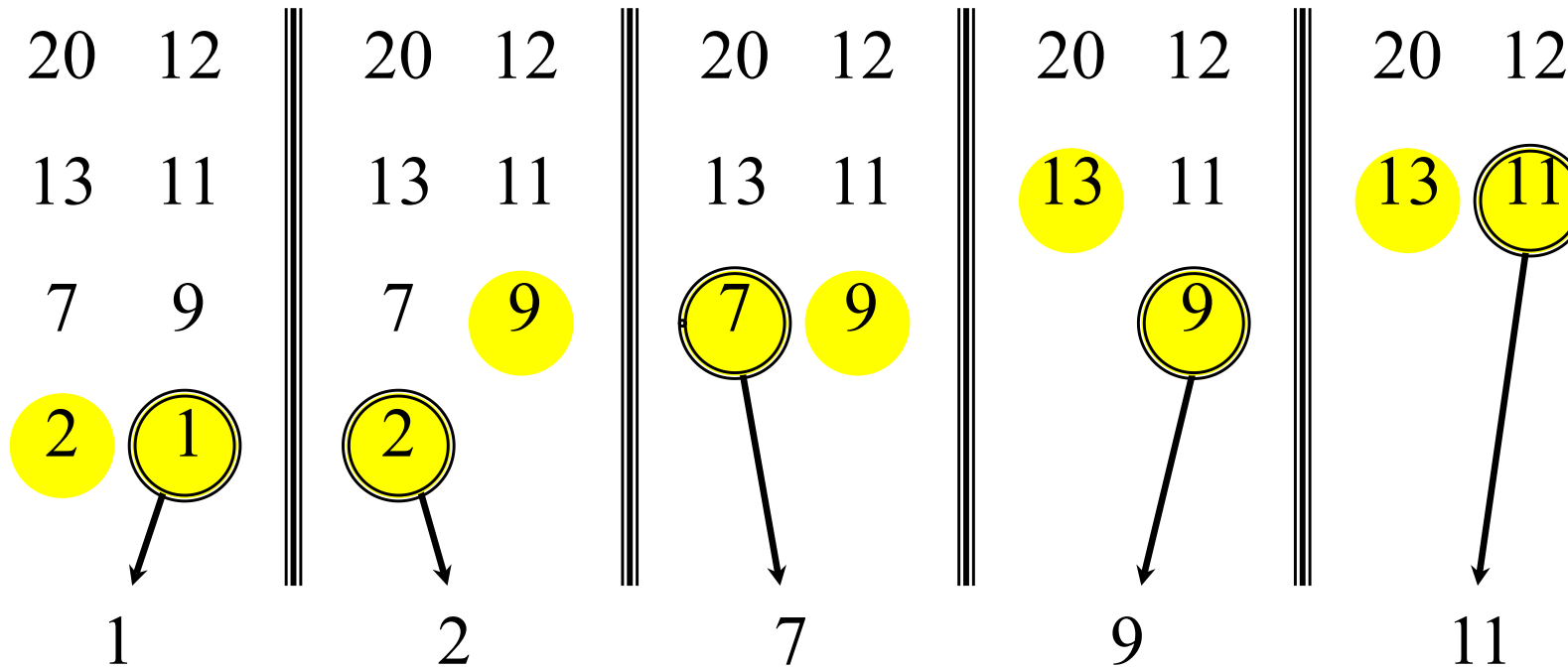


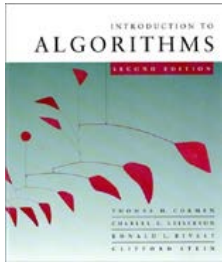
Merging two sorted arrays



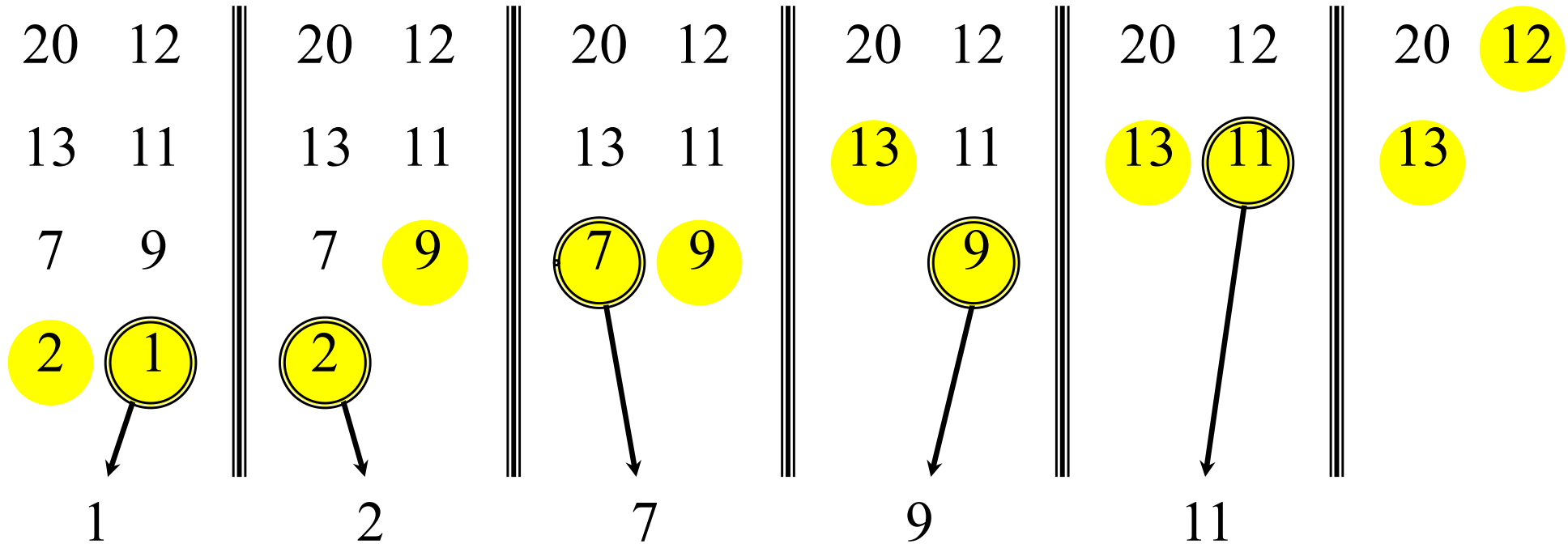


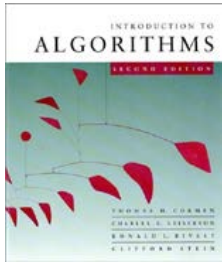
Merging two sorted arrays



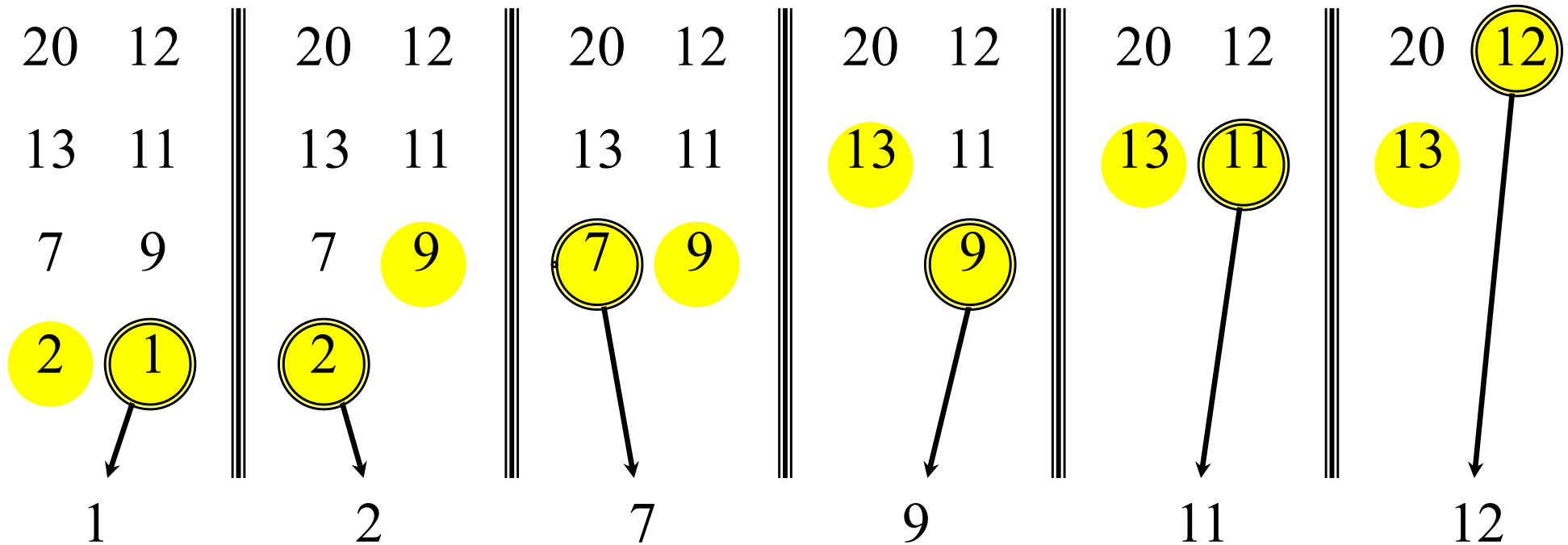


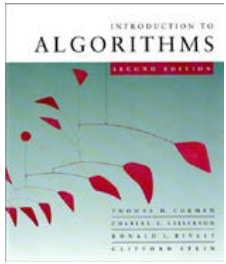
Merging two sorted arrays



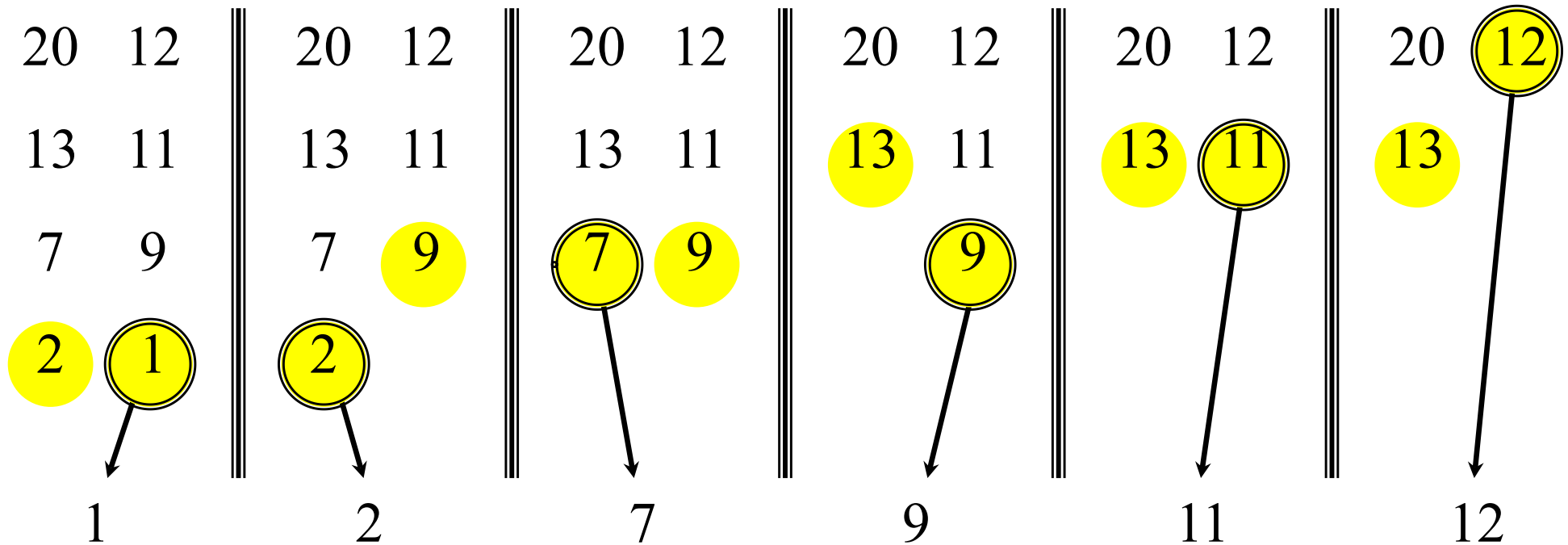


Merging two sorted arrays

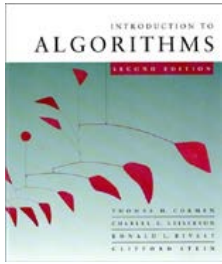




Merging two sorted arrays



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



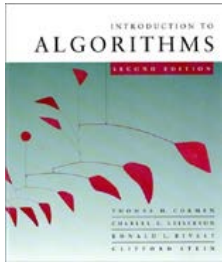
Analyzing merge sort

$T(n)$	MERGE-SORT $A[1 \dots n]$
$\Theta(1)$	
$2T(n/2)$	
$\Theta(n)$	

Abuse

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

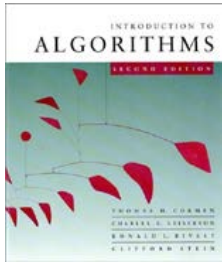
Sloppiness: Should be $T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor)$, but it turns out not to matter asymptotically.



Recurrence for merge sort

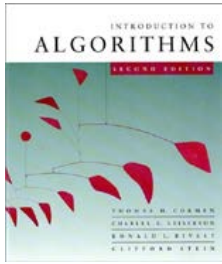
$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

- 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

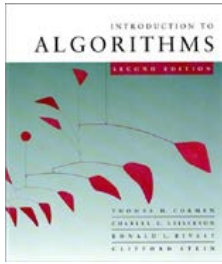
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.



Recursion tree

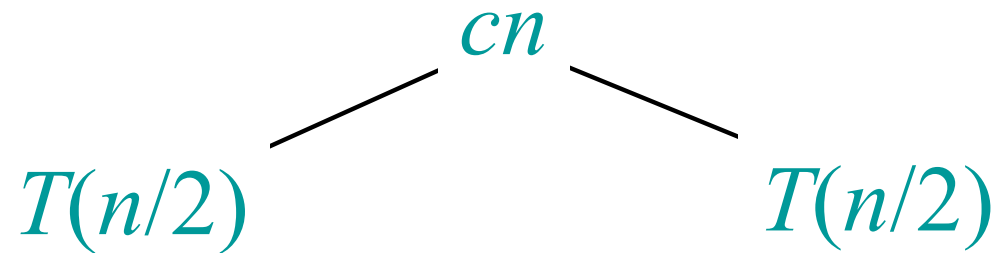
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

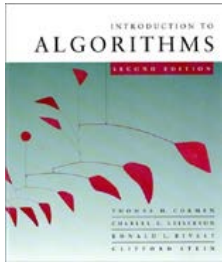
$$T(n)$$



Recursion tree

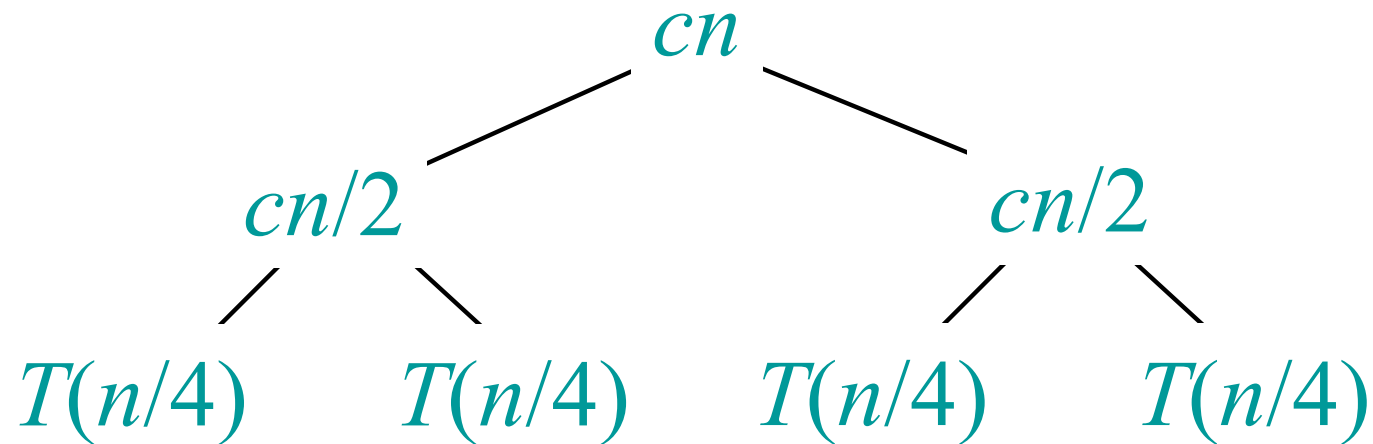
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

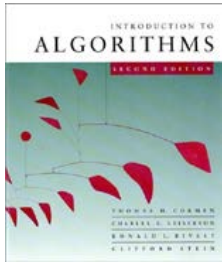




Recursion tree

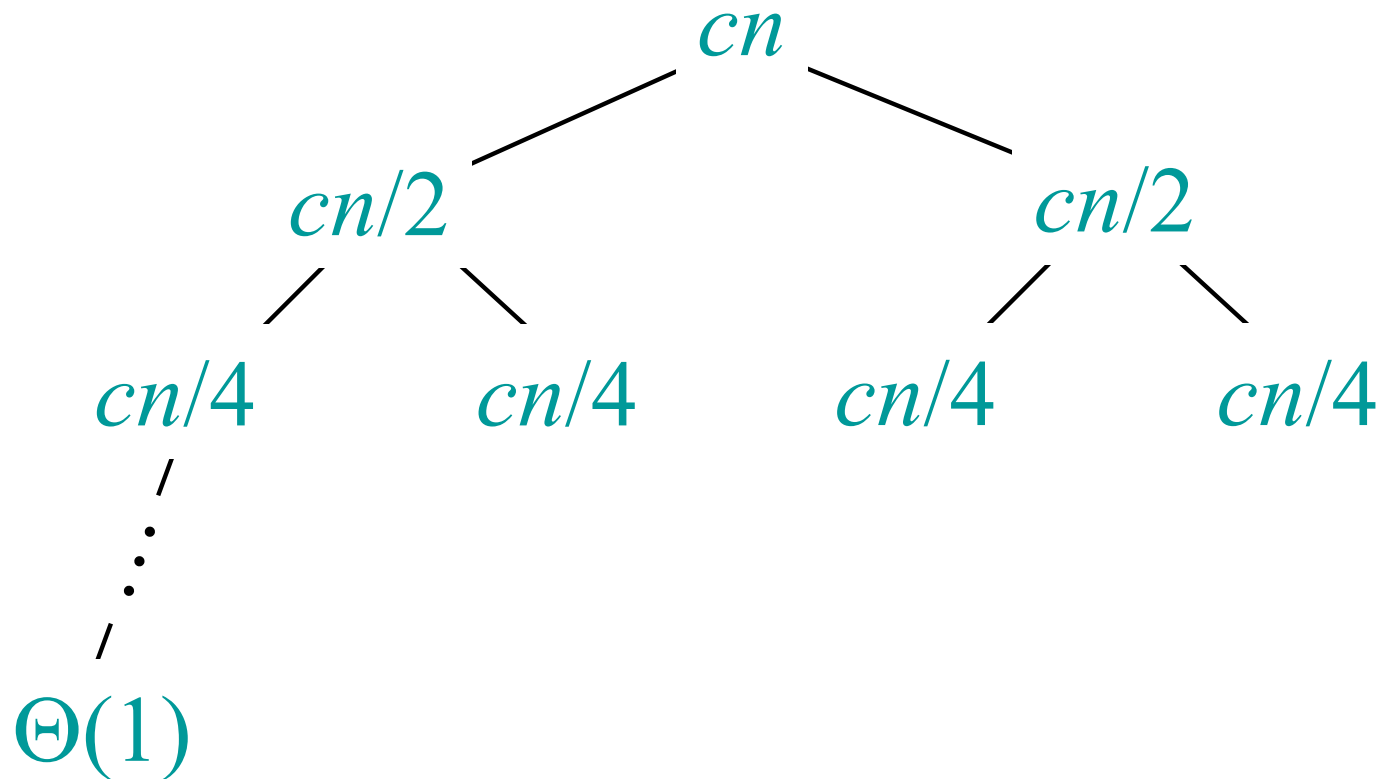
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

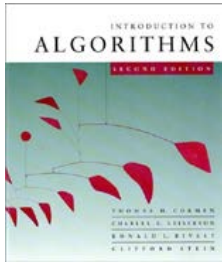




Recursion tree

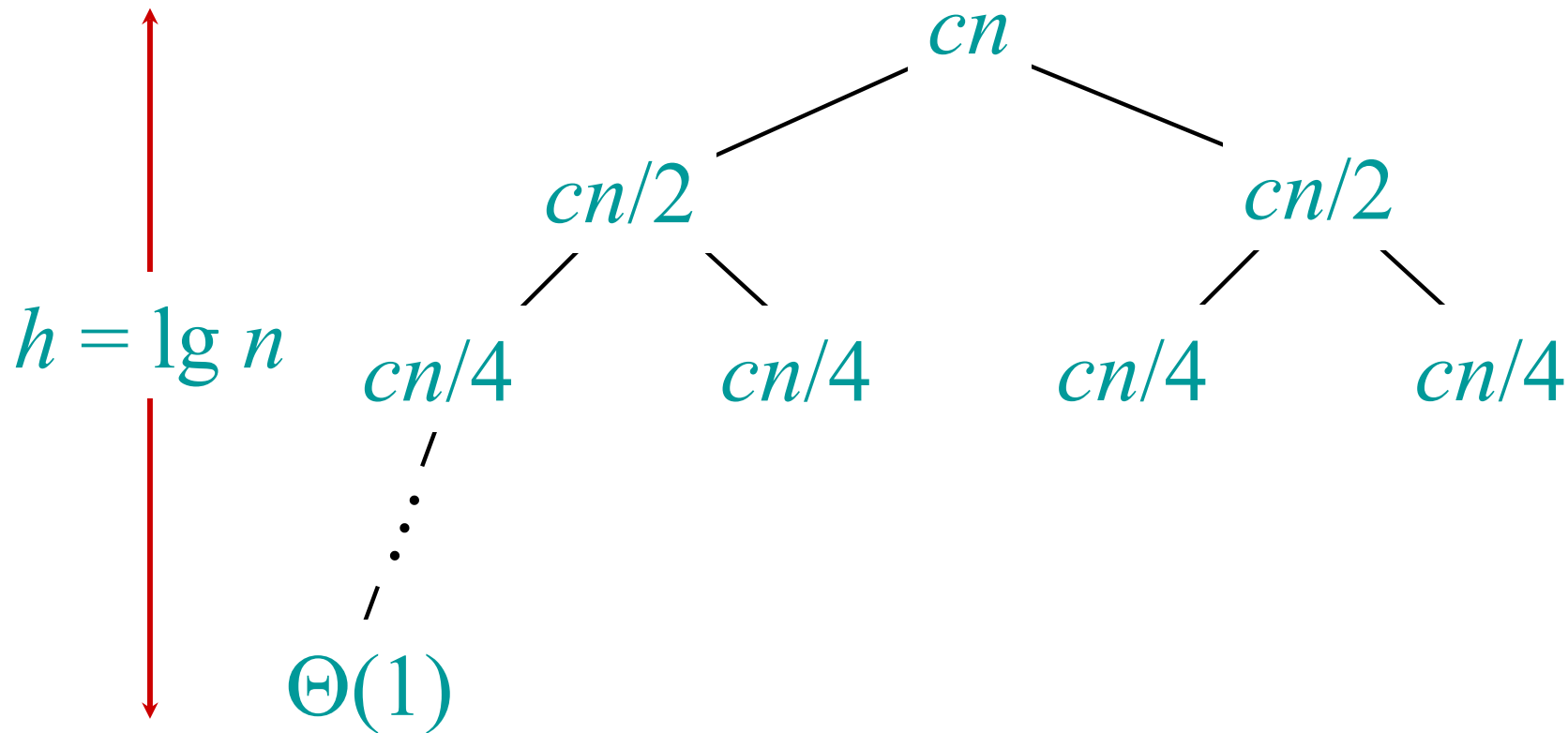
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

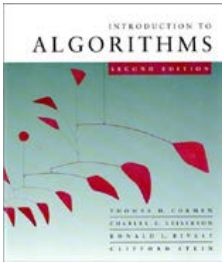




Recursion tree

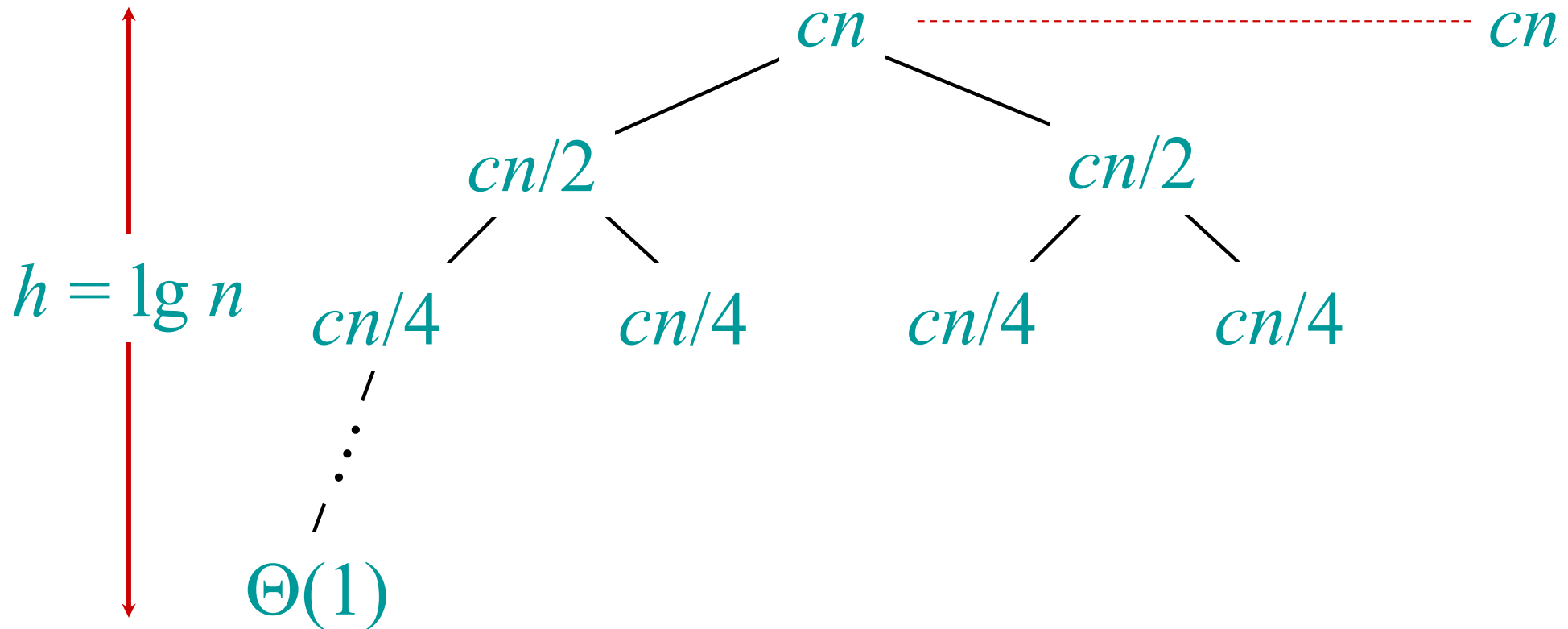
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

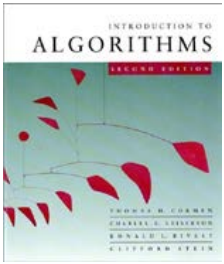




Recursion tree

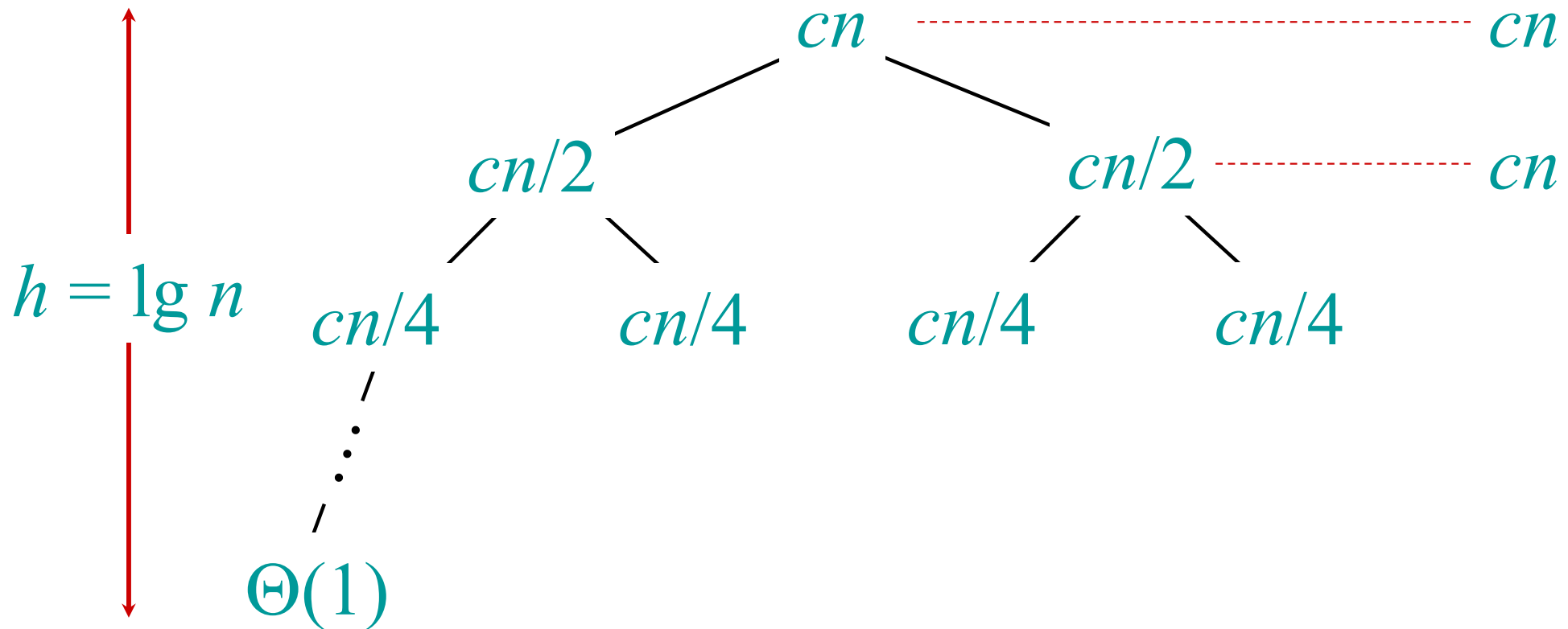
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

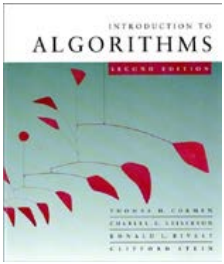




Recursion tree

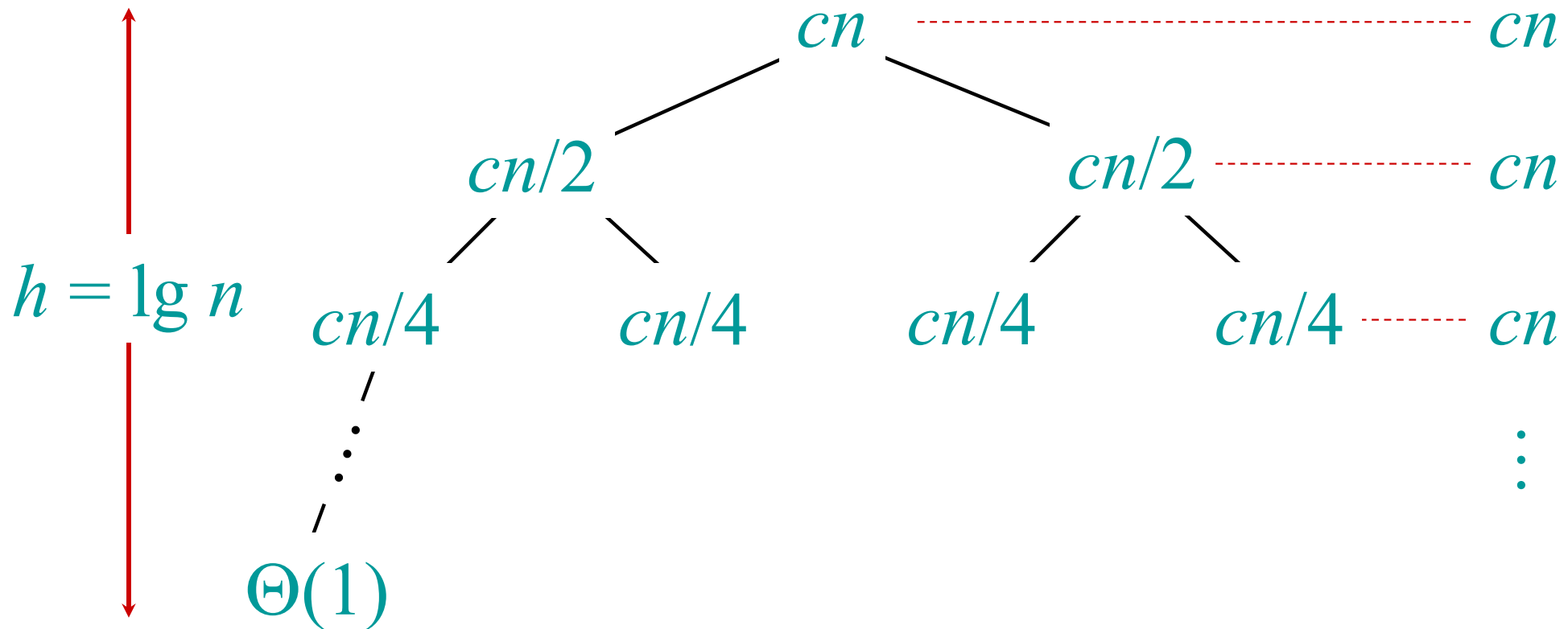
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

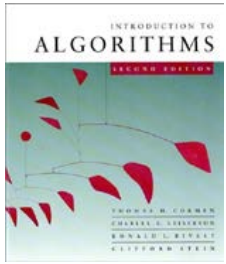




Recursion tree

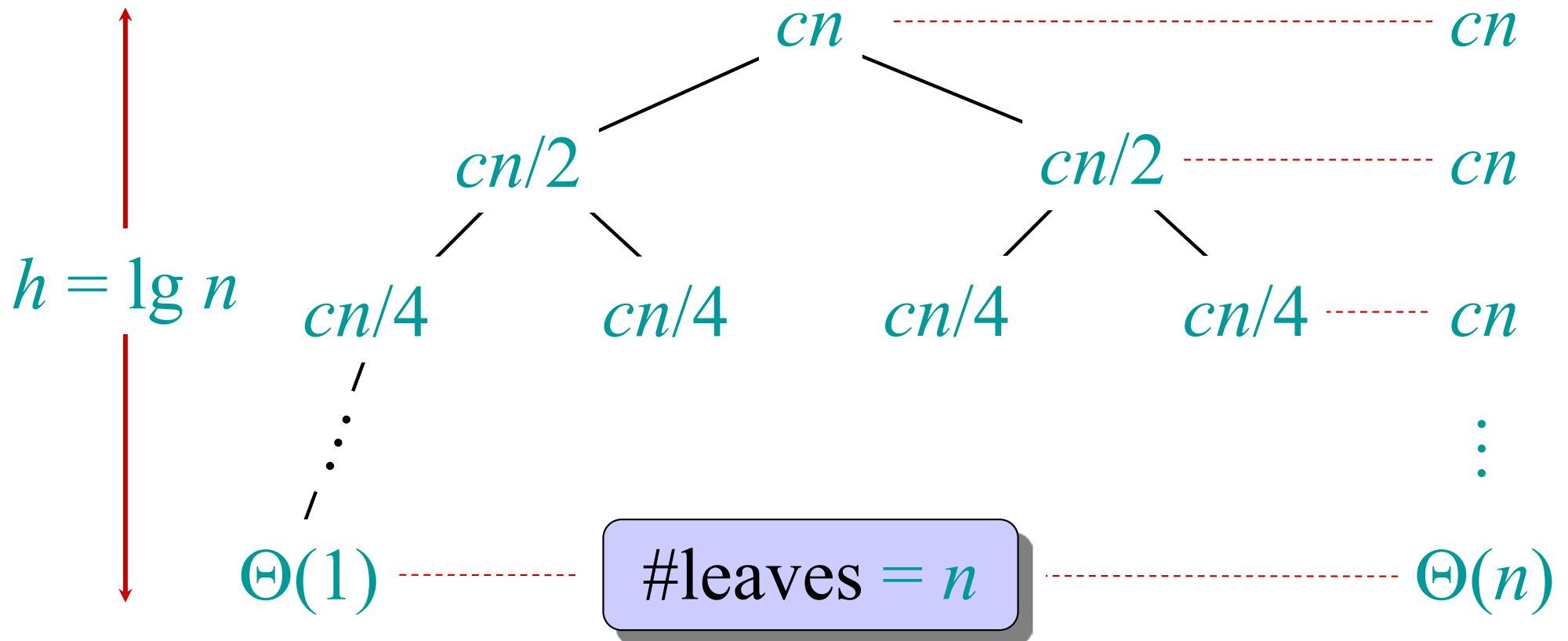
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

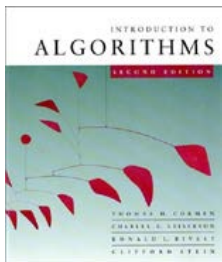




Recursion tree

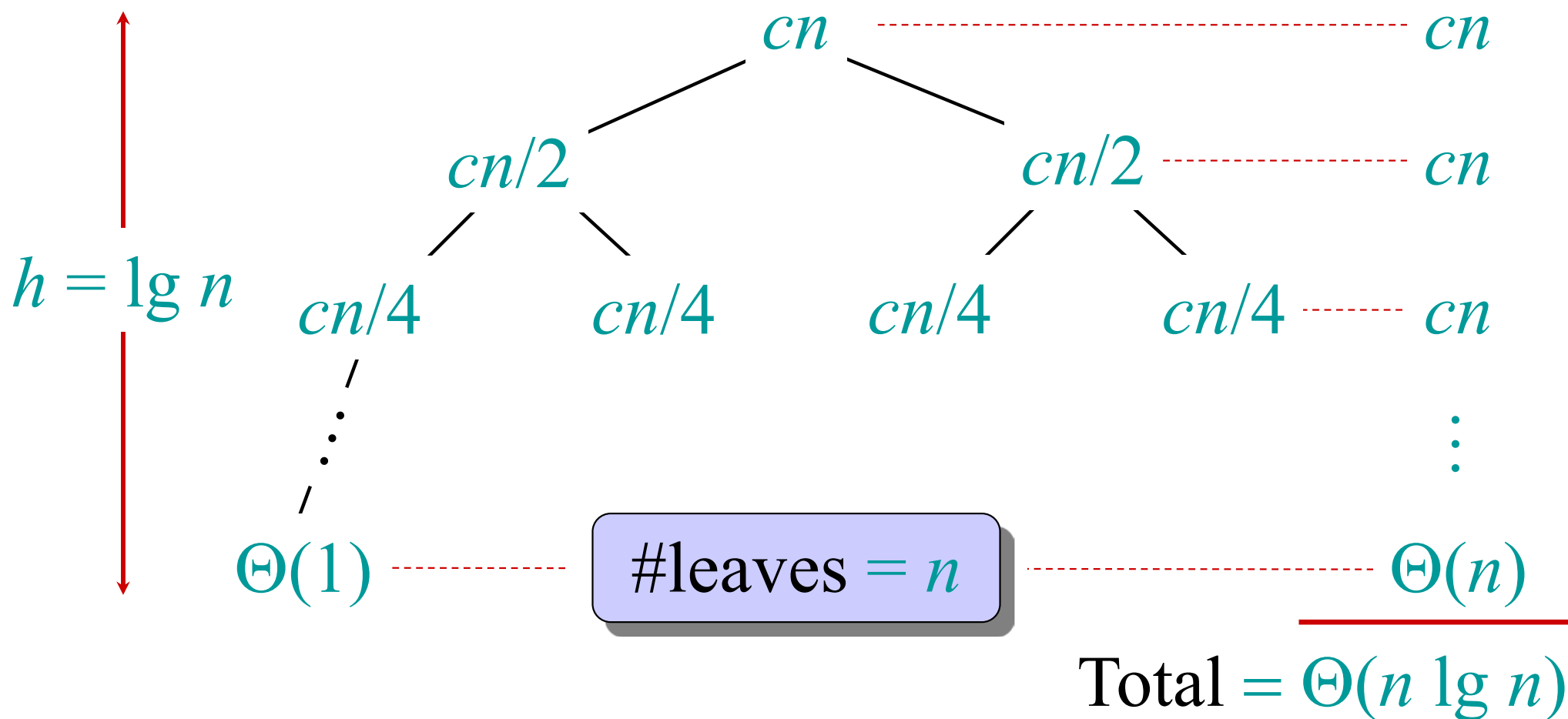
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

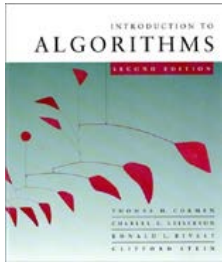




Recursion tree

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





Conclusions

- $\Theta(n \lg n)$ 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.
- Go test it out for yourself!