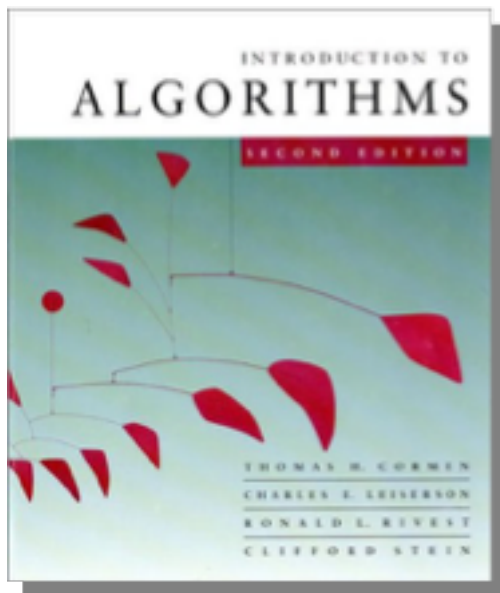


# Introduction to Algorithms



- Insertion sort
- Asymptotic analysis
- Merge sort
- Recurrences

Intuitive approach for the first lectures

# The problem of sorting

**Input:** sequence  $\langle a_1, a_2, \dots, a_n \rangle$  of numbers.

**Output:** permutation  $\langle a'_1, a'_2, \dots, a'_n \rangle$

such that  $a'_1 \leq a'_2 \leq \dots \leq a'_n$ .

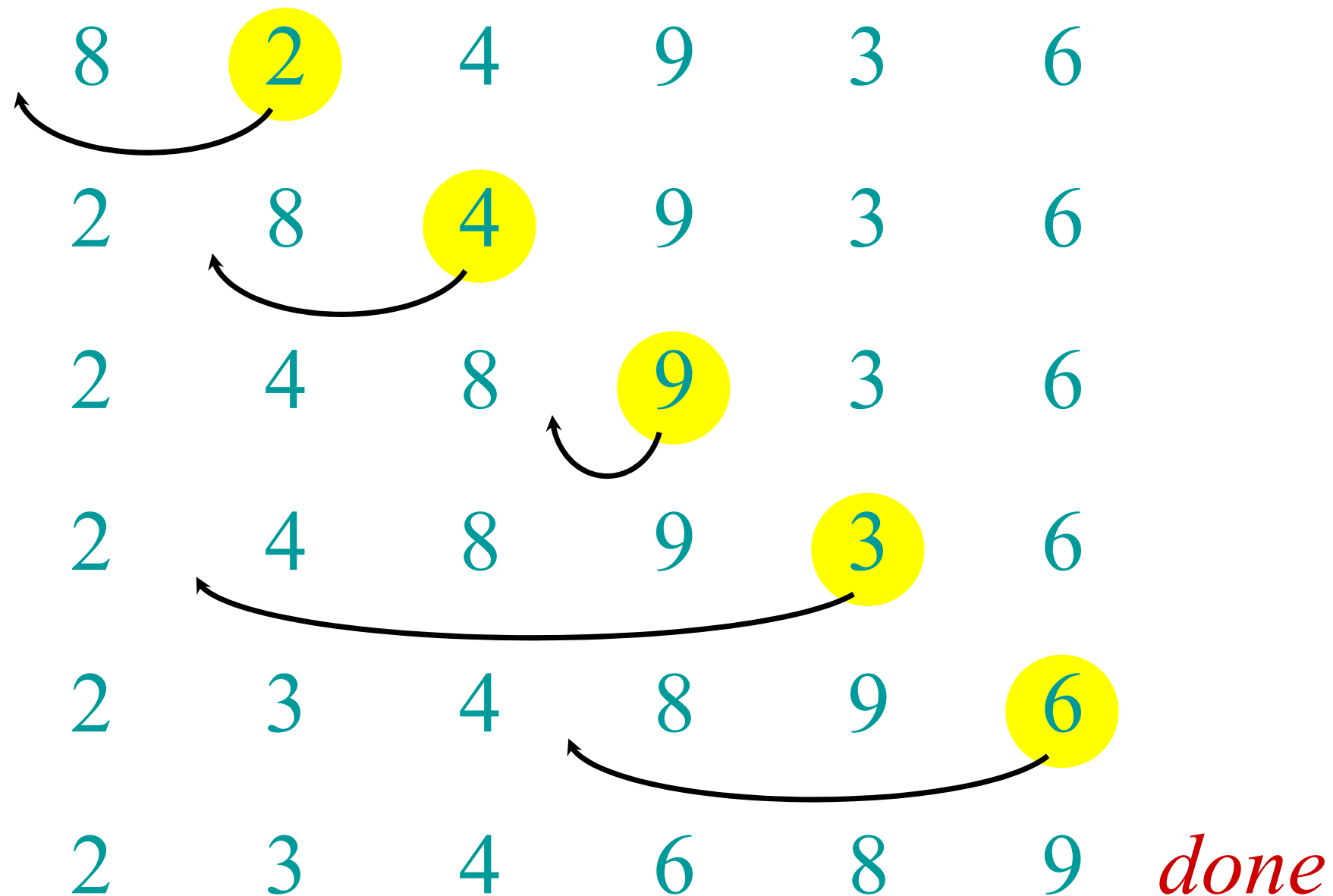
Example:

**Input:** 8 2 4 9 3 6

**Output:** 2 3 4 6 8 9

Sort: 8 2 4 9 3 6

Intuition: move smaller numbers to the left ...



notation:  $x \leftarrow y$  means assign the value  $y$  to the variable  $x$  (in most programming languages you would write  $x = y$ ).

# Insertion sort “pseudocode”

INSERTION-SORT ( $A, n$ )  $\triangleright A[1 \dots n]$

start from 2 because 1 element alone would be sorted ;-)

for  $j \leftarrow 2$  to  $n$  we will sort the first  $j$  numbers, having already sorted the first  $j-1$

do  $key \leftarrow A[j]$  read the  $j$ -th number and put it in the auxiliary variable “key”

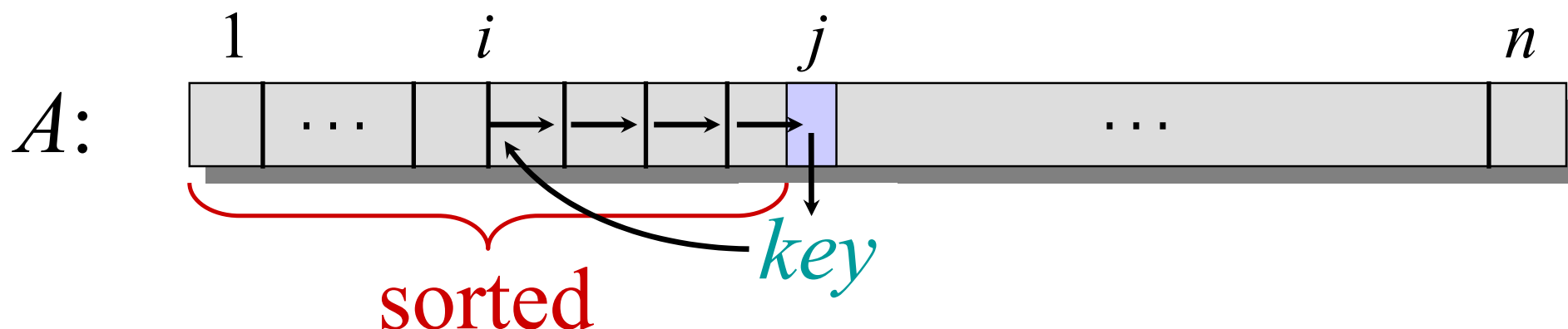
$i \leftarrow j-1$  starting position for scanning the list backwards

while  $i > 0$  and  $A[i] > key$  continue moving backwards until you find the right place for the insertion or the list ends

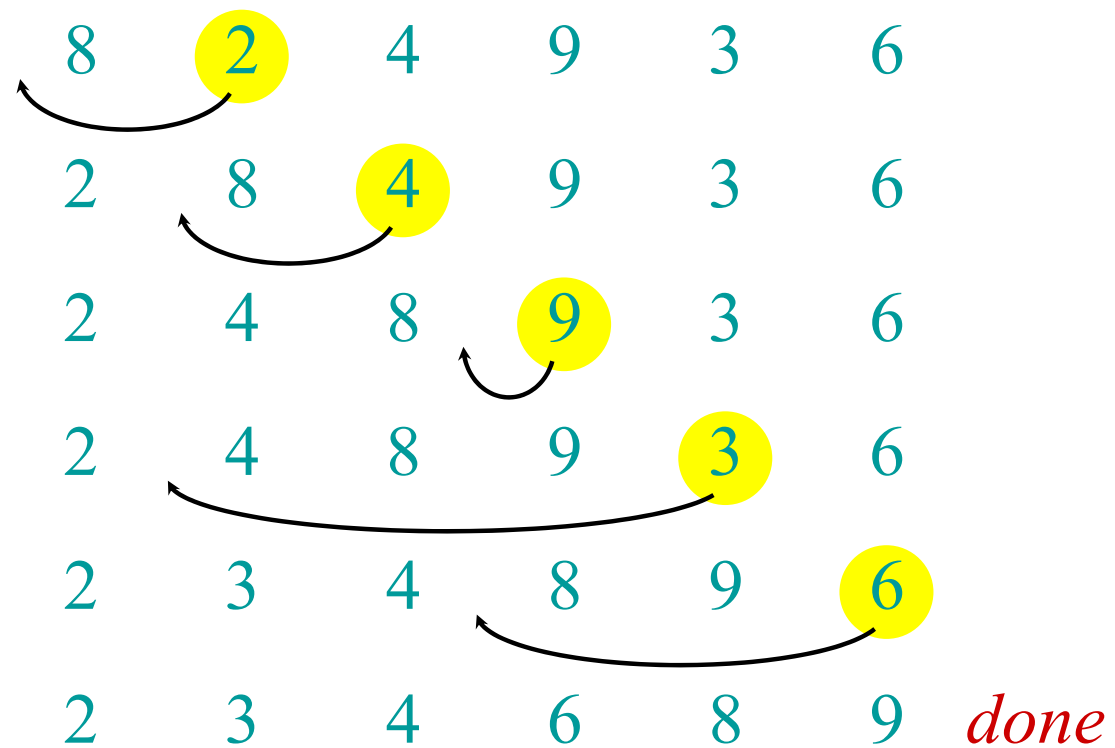
do  $A[i+1] \leftarrow A[i]$  while scanning backwards shift the elements to the right (create space the list)

$i \leftarrow i-1$  move backwards

$A[i+1] = key$  put the selected element (  $key \leftarrow A[j]$  ) in the correct position



# compare



INSERTION-SORT ( $A, n$ )  $\triangleright A[1 \dots n]$

for  $j \leftarrow 2$  to  $n$

do  $\text{key} \leftarrow A[j]$

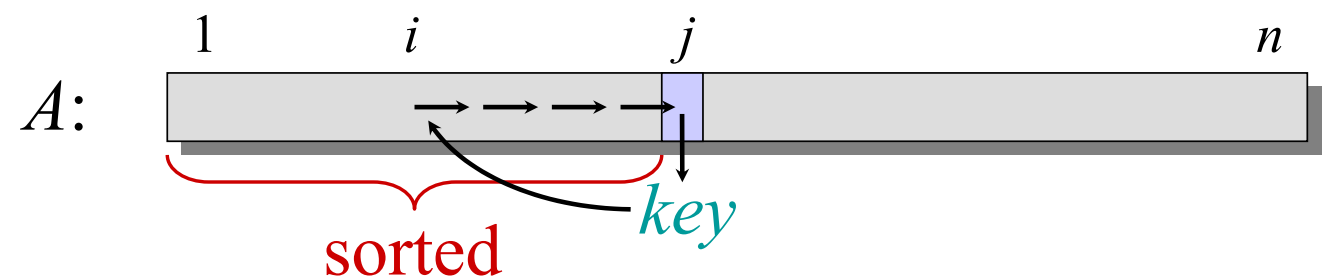
$i \leftarrow j - 1$

while  $i > 0$  and  $A[i] > \text{key}$

do  $A[i+1] \leftarrow A[i]$

$i \leftarrow i - 1$

$A[i+1] = \text{key}$



# Running time

- The running time depends on the input: an already sorted sequence is easier to sort.
- Parametrize the running time by the size of the input, since short sequences are easier to sort than long ones.
- Generally, we seek upper bounds on the running time, because everybody likes a guarantee.

# Kinds of analysis

Worst-case: (usually)

- $T(n)$  = maximum time of algorithm on any input of size  $n$ .

Average-case: (sometimes)

- $T(n)$  = expected time of algorithm over all inputs of size  $n$ .
- Needs assumption of statistical distribution of inputs.

Best-case: (bogus)

- Cheat with a slow algorithm that works fast on some input.



# Machine-independent time

What is insertion sort's worst-case time?

- It depends on the speed of our computer:
  - relative speed (on the same machine),
  - absolute speed (on different machines).

## BIG IDEA:

- Ignore machine-dependent constants (consider a conventional unit cost for any single elementary operation)
- Look at **growth** of  $T(n)$  as  $n \rightarrow \infty$ .

“Asymptotic Analysis”

# $\Theta$ -notation

Math:

$f(n) = \Theta(g(n))$  : there exist positive constants  $c_1$ ,  $c_2$ , and  $n_0$  such that

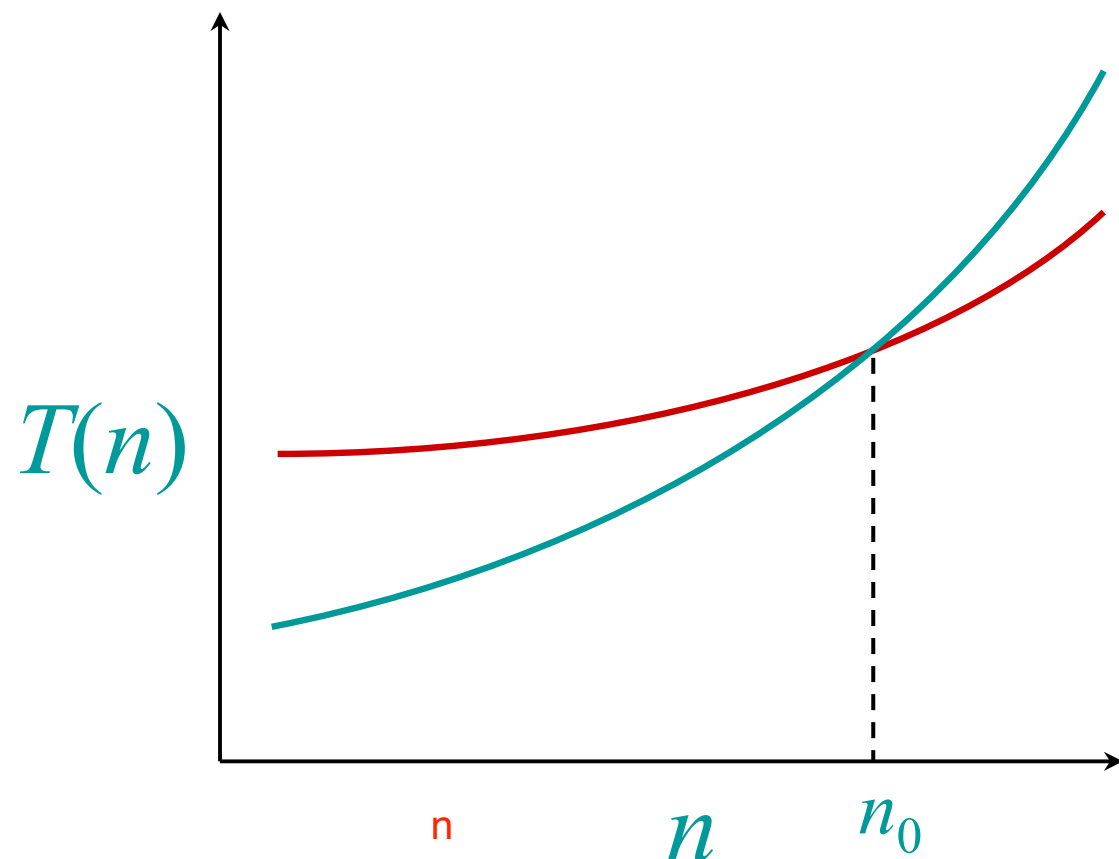
$$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \quad \text{for all } n \geq n_0$$

Practitioners:

- Drop low-order terms; ignore leading constants.
- Example:  $3n^3 + 90n^2 - 5n + 6046 = \Theta(n^3)$

# Asymptotic performance

When  $n$  gets large enough, a  $\Theta(n^2)$  algorithm **always** beats a  $\Theta(n^3)$  algorithm.



- We shouldn't ignore asymptotically slower algorithms, however.
- Real-world design situations often call for a careful balancing of engineering objectives.
- Asymptotic analysis is a useful tool to help to structure our thinking.

# Insertion sort analysis

**Worst case:** Input reverse sorted.

$$T(n) = \sum_{j=2}^n \Theta(j) = \Theta(n^2)$$

[arithmetic series (Gauss 6 years old)]

**Average case:** All permutations equally likely.

$$T(n) = \sum_{j=2}^n \Theta(j/2) = \Theta(n^2)$$

Is insertion sort a fast sorting algorithm?

- Moderately so, for small  $n$ .
- Not at all, for large  $n$ .

# Merge sort (recursive)

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” MERGE-SORT the 2 sorted lists.

Key subroutine: MERGE-SORT

this is an example of the approach

# Divide-and-Conquer

Mergesort:

1. **Divide:**  $n=1$ , divide the given  $n$ -element array  $A$  into 2 subarrays of  $n/2$  elements each
2. **Conquer:** recursively sort the two subarrays
3. **Combine:** merge 2 sorted subarrays into 1 sorted array.

# Beautiful “ultrashort” recursive pseudocode

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

{ if  $n > 1$

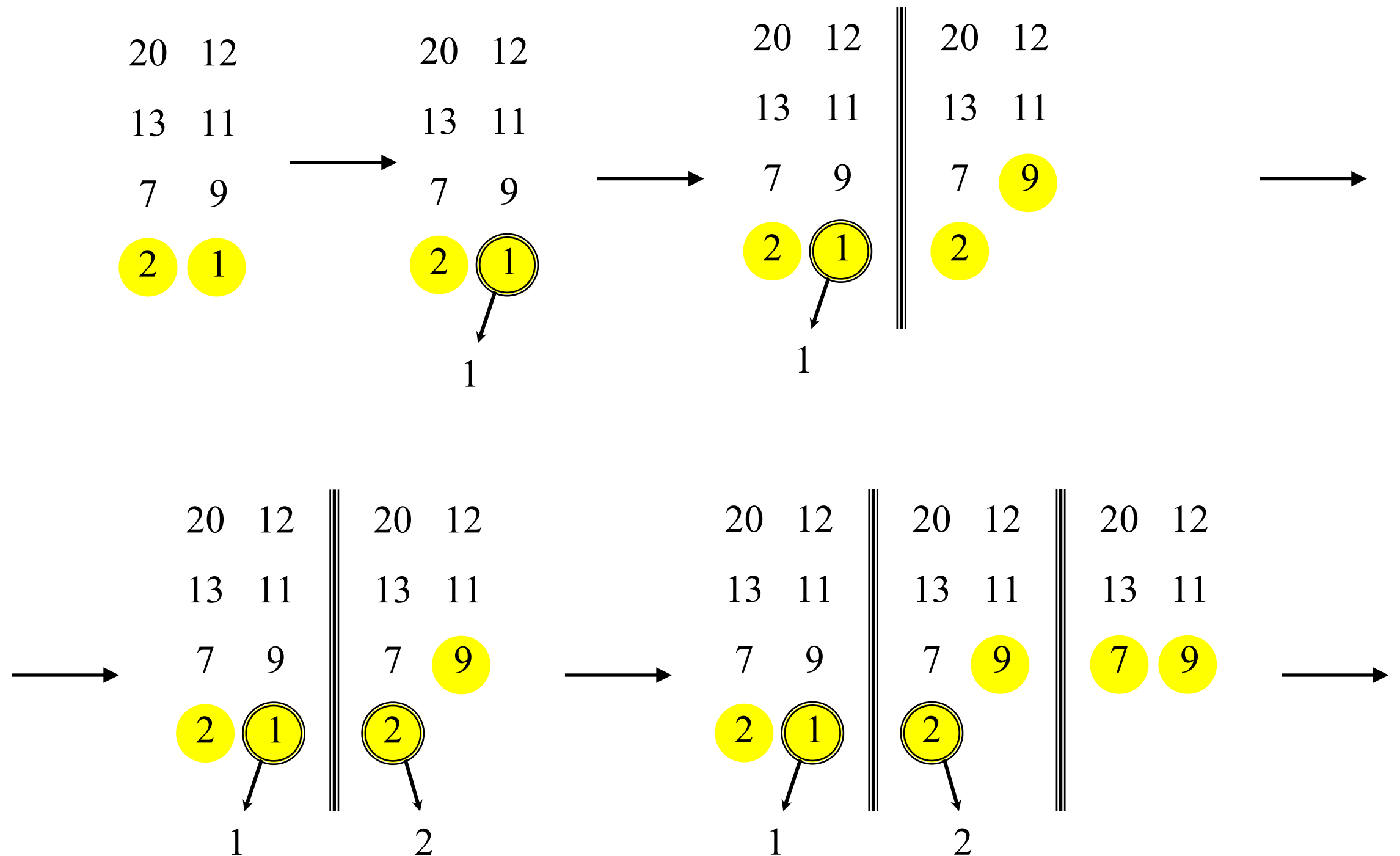
    MERGE-SORT ( $A[1 \dots \lceil n/2 \rceil]$ )

    MERGE-SORT ( $A[\lceil n/2 \rceil + 1 \dots n]$ )

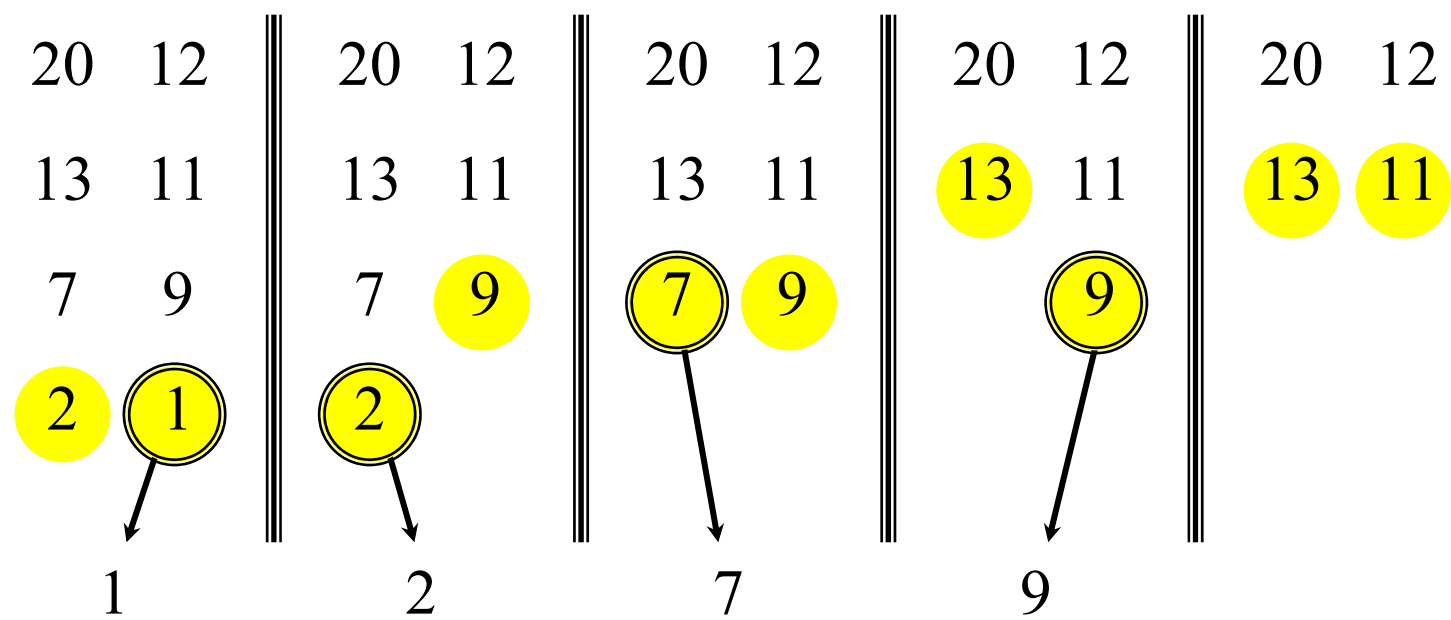
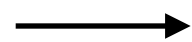
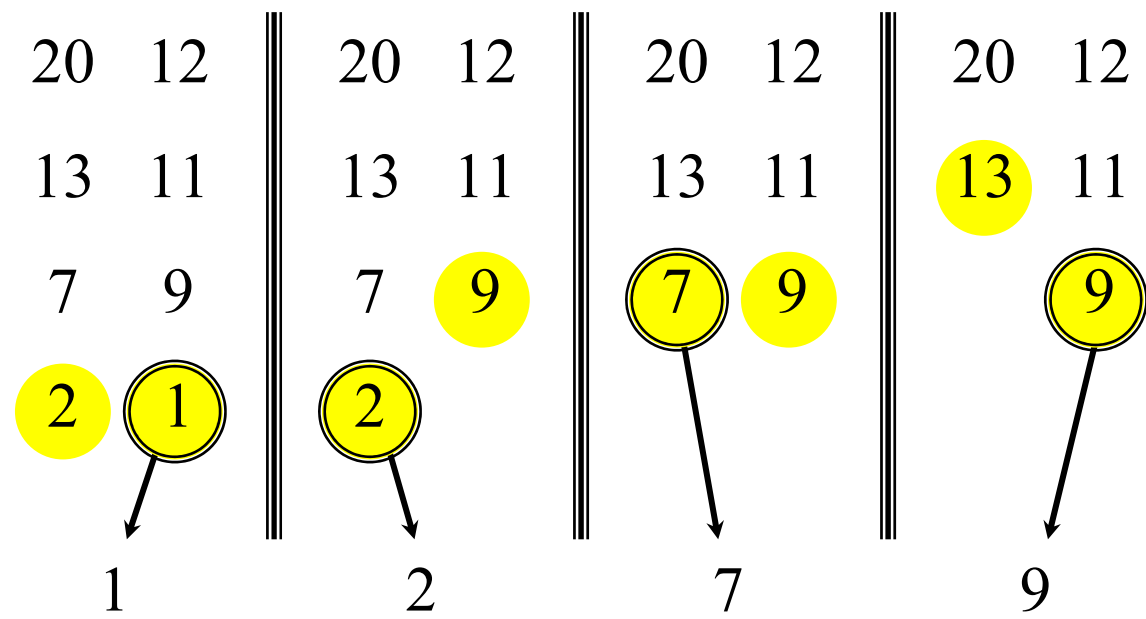
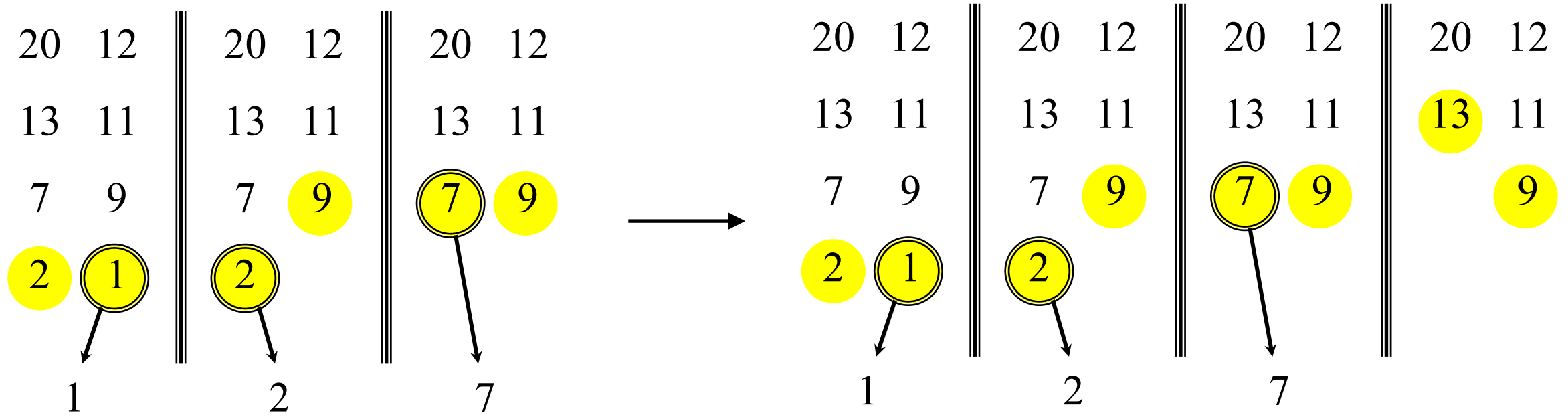
    MERGE( $A[1 \dots n]$ )

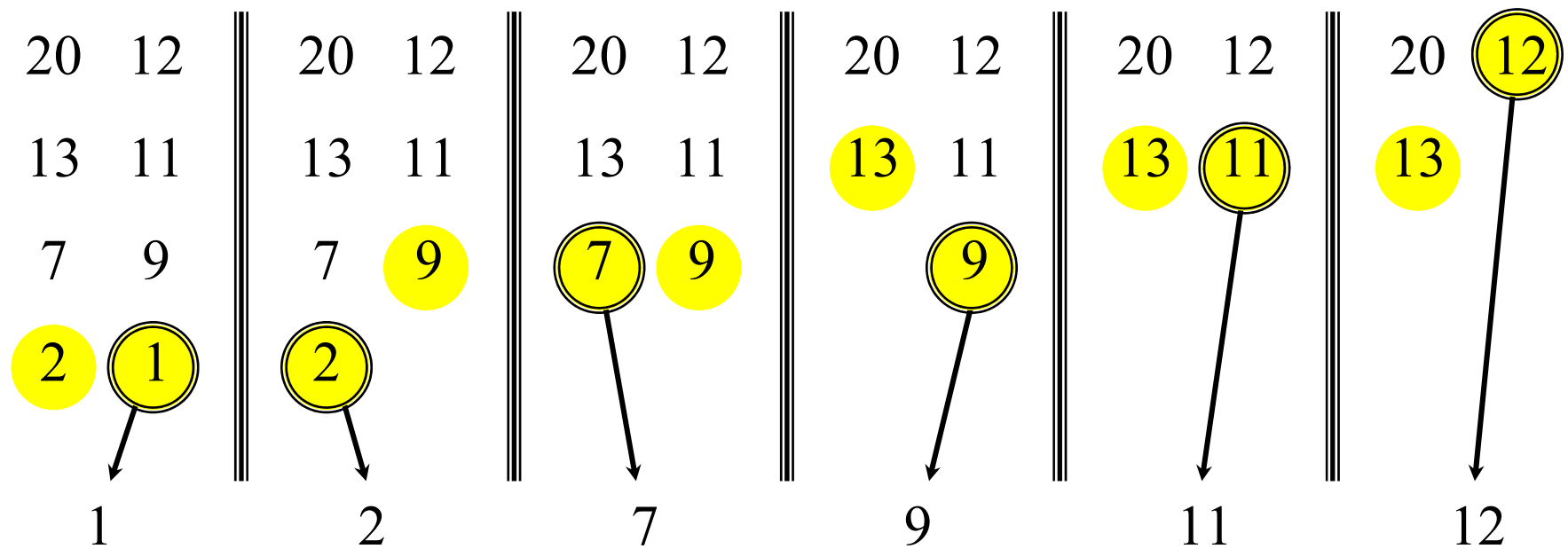
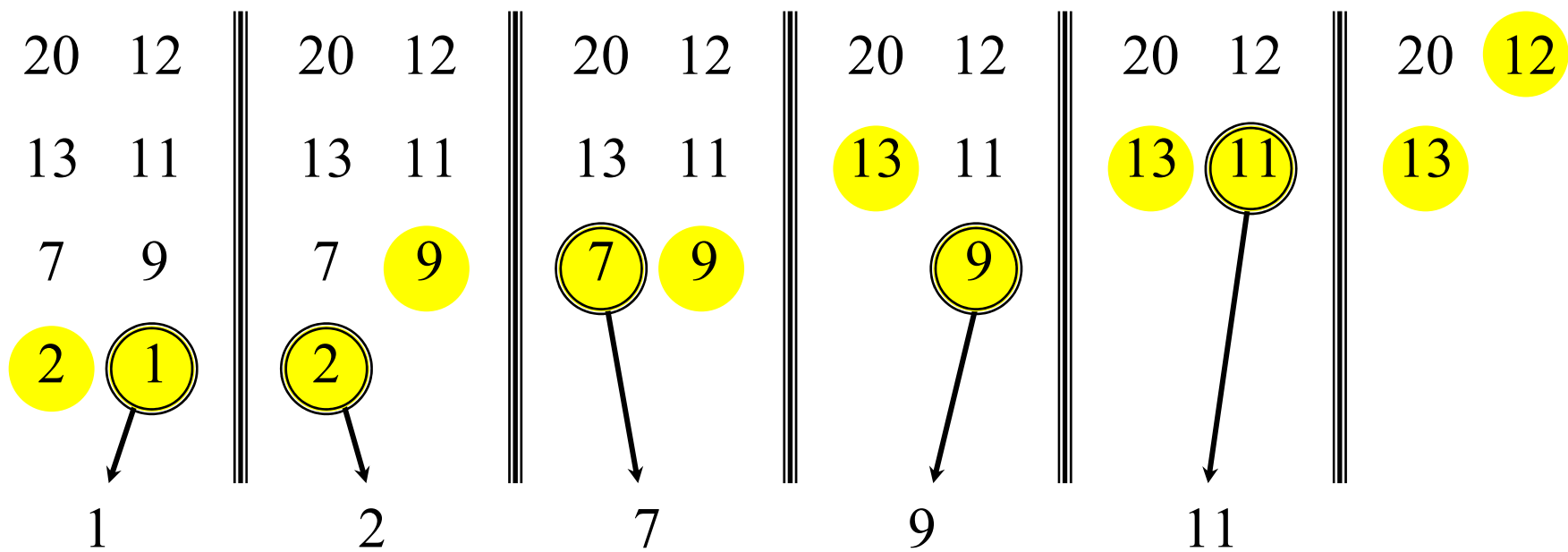
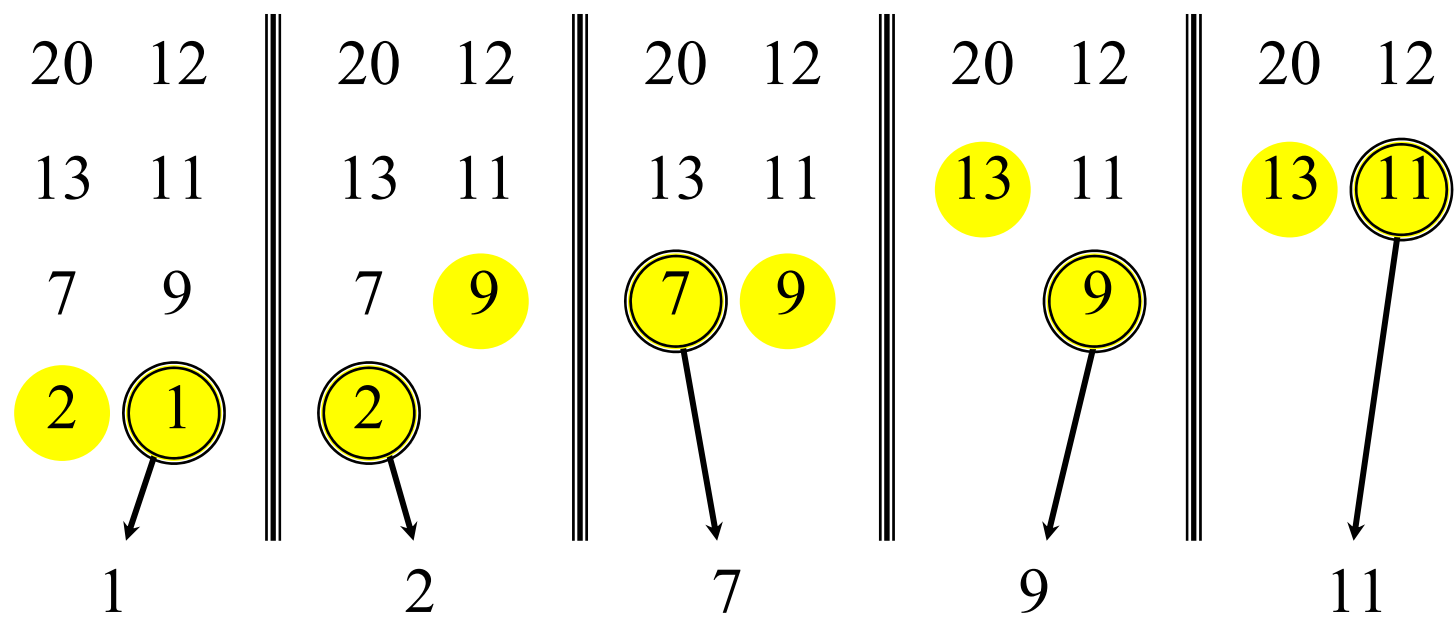
}

# Merging two sorted arrays



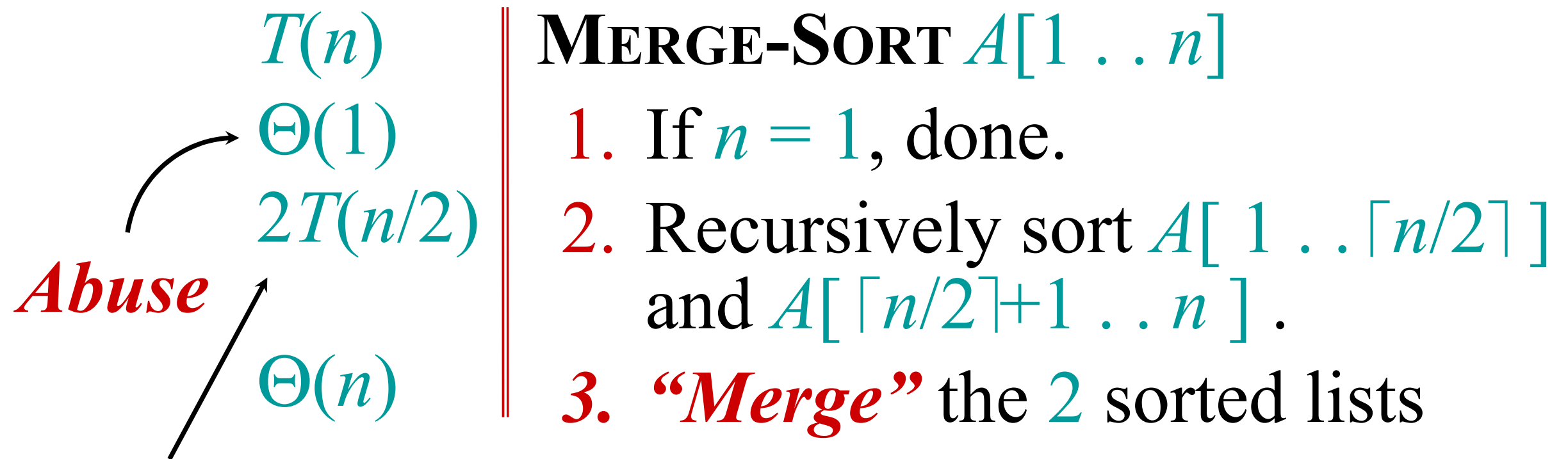






Time =  $\Theta(n)$  to **merge** a total of  $n$  elements (linear time).

# Analyzing merge sort



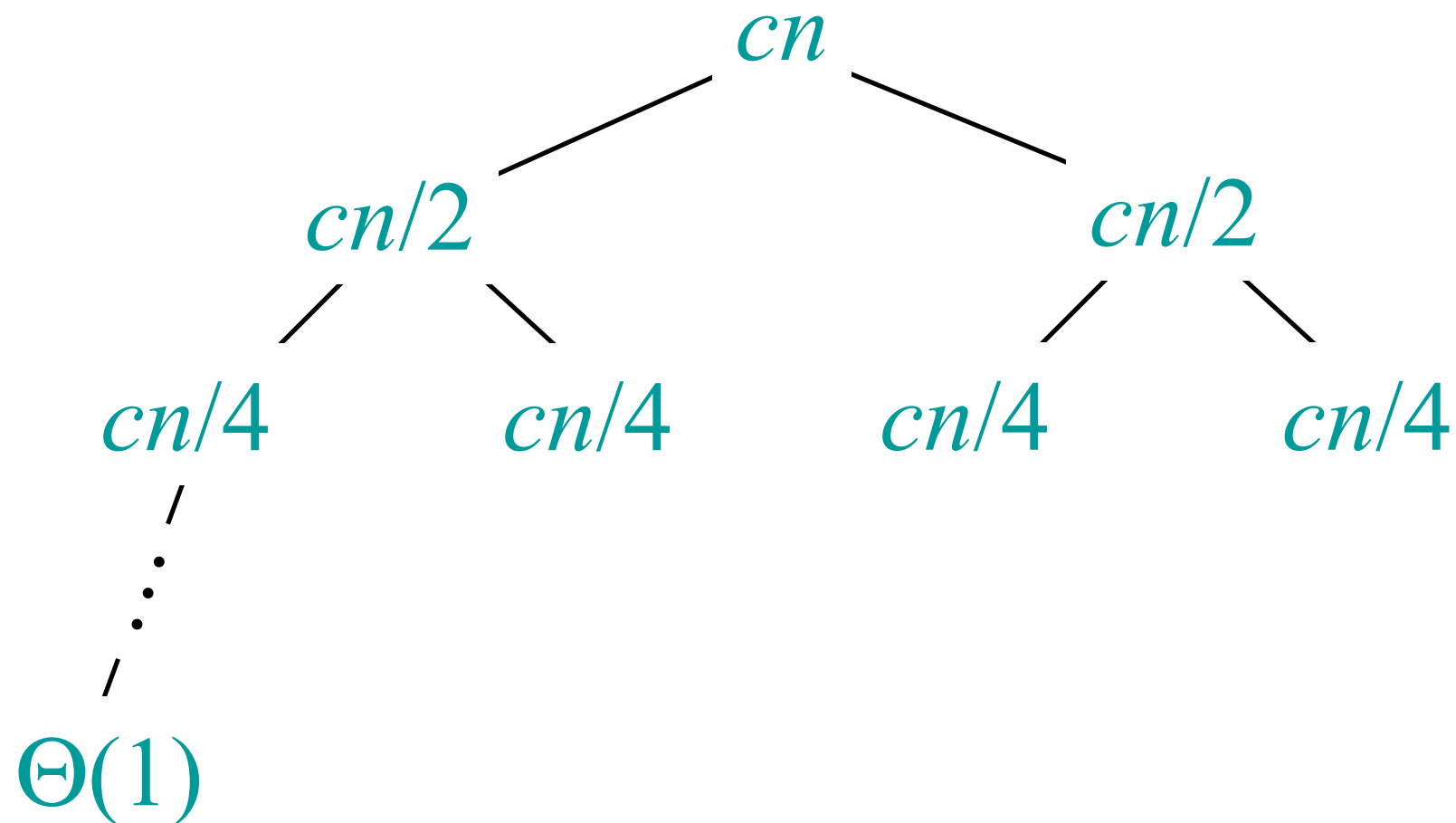
*Sloppiness:* Should be  $T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor)$ , but it turns out not to matter asymptotically.

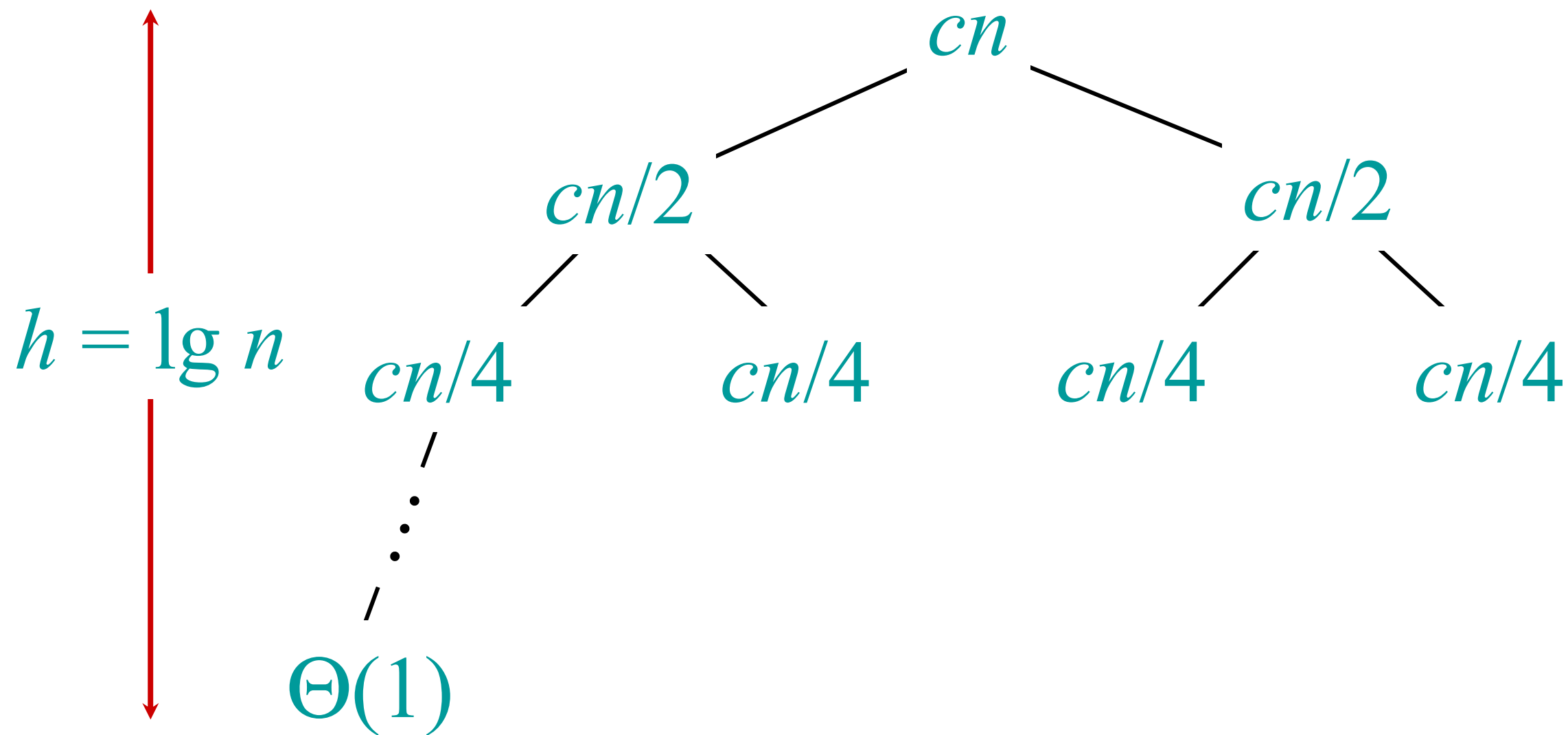
# Recurrence for merge sort

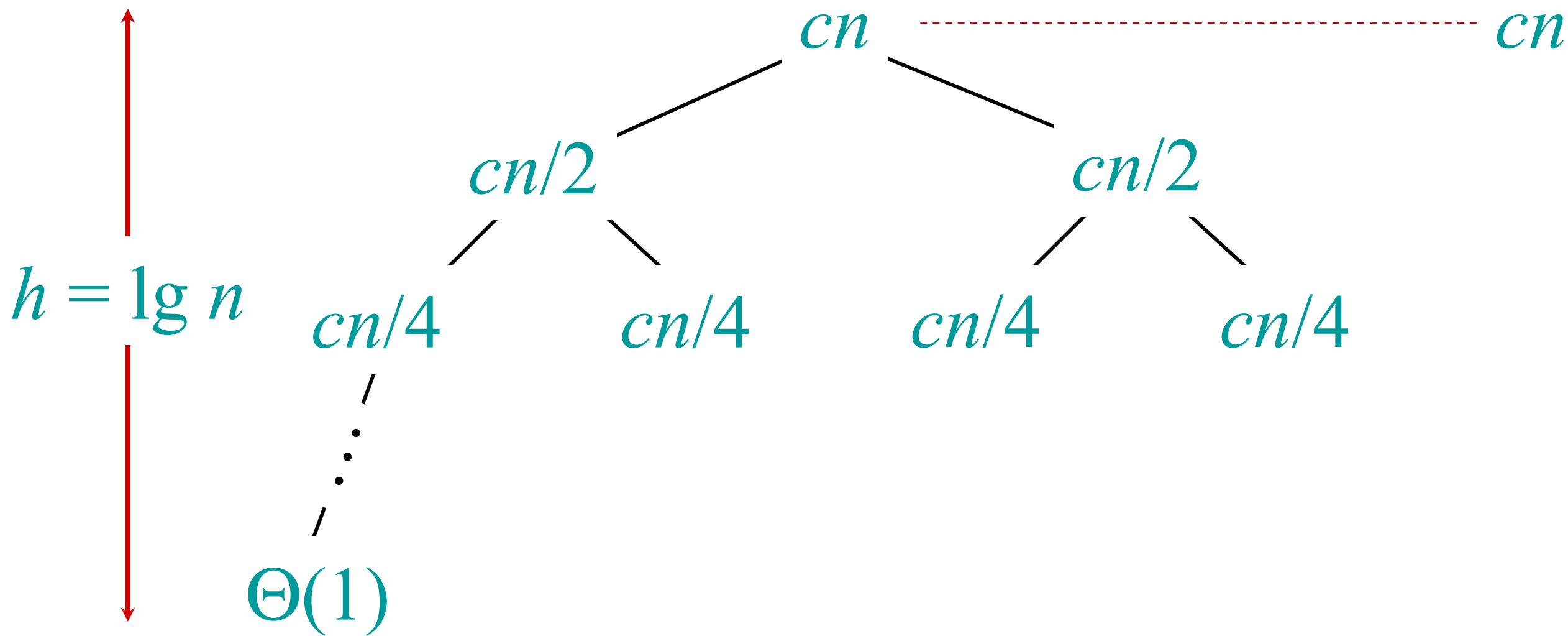
$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

- 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 (Textbook) provide several ways to find a good upper bound on  $T(n)$ .

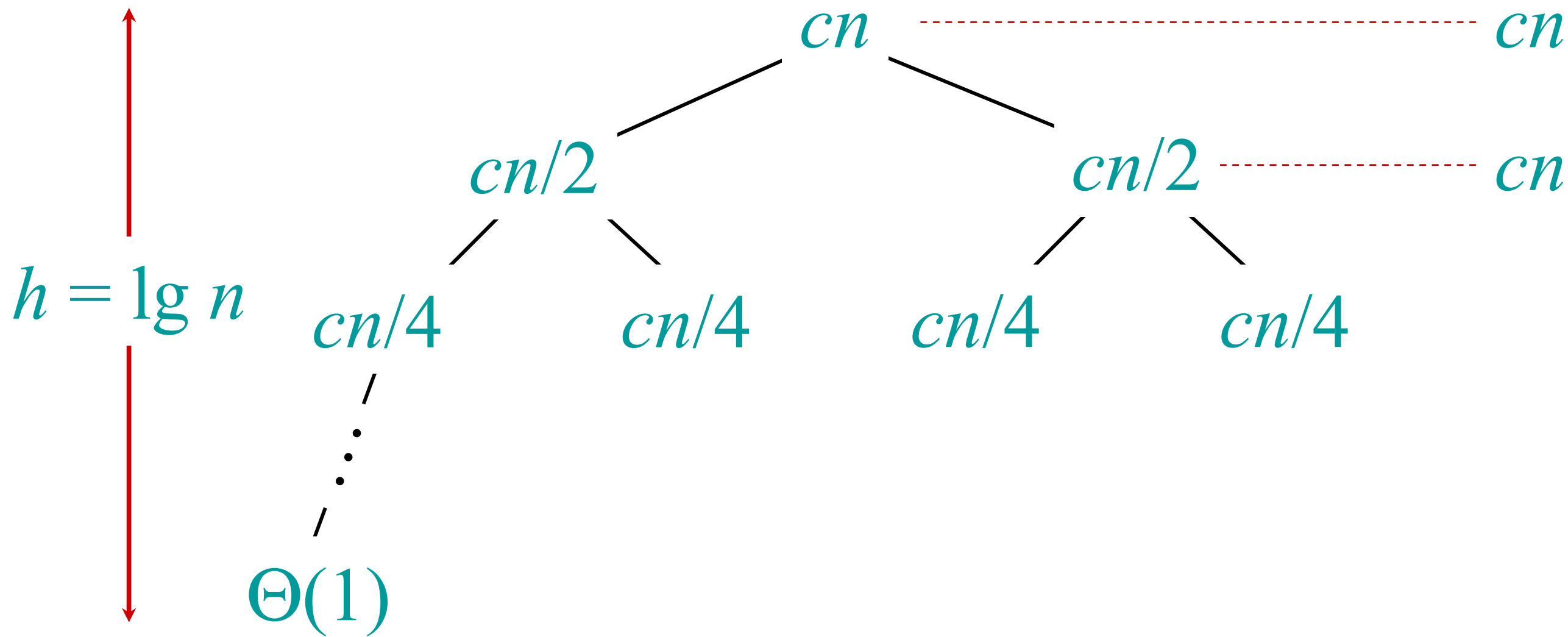
Solve  $T(n) = 2T(n/2) + cn$ , where  $c > 0$  is constant.

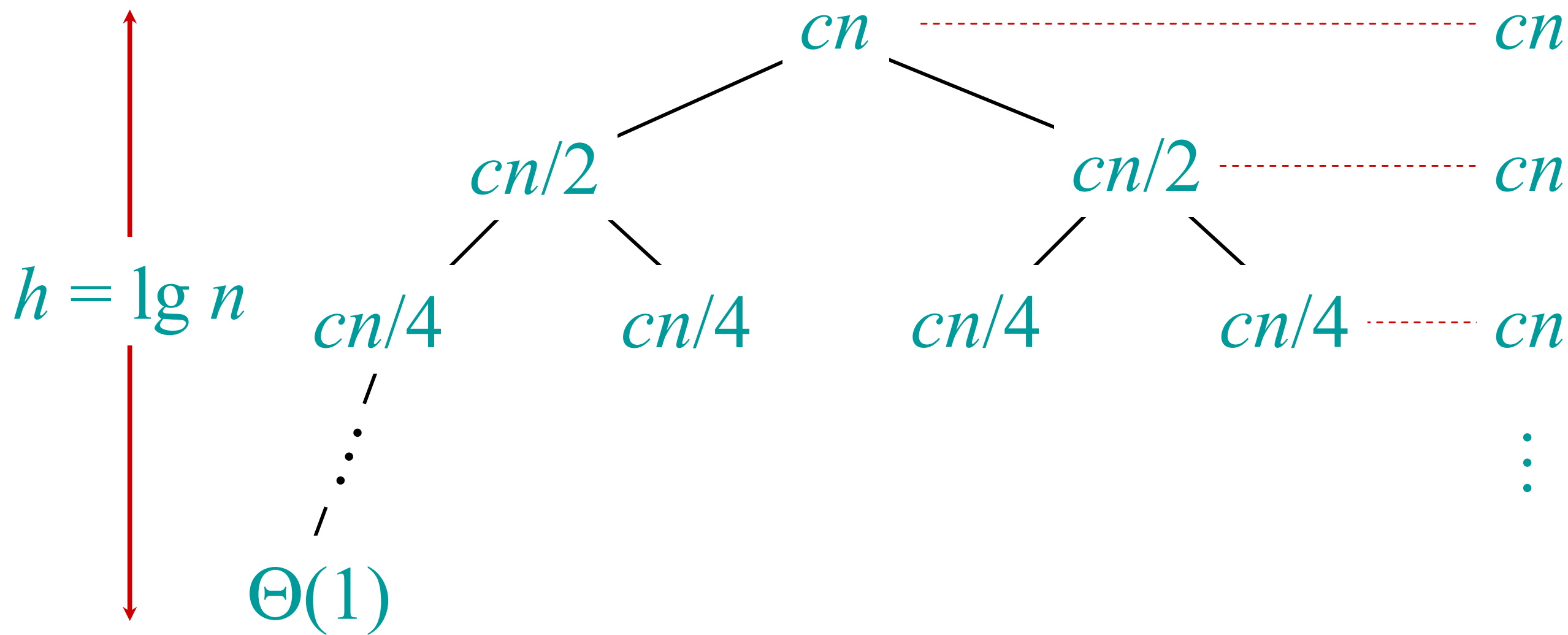


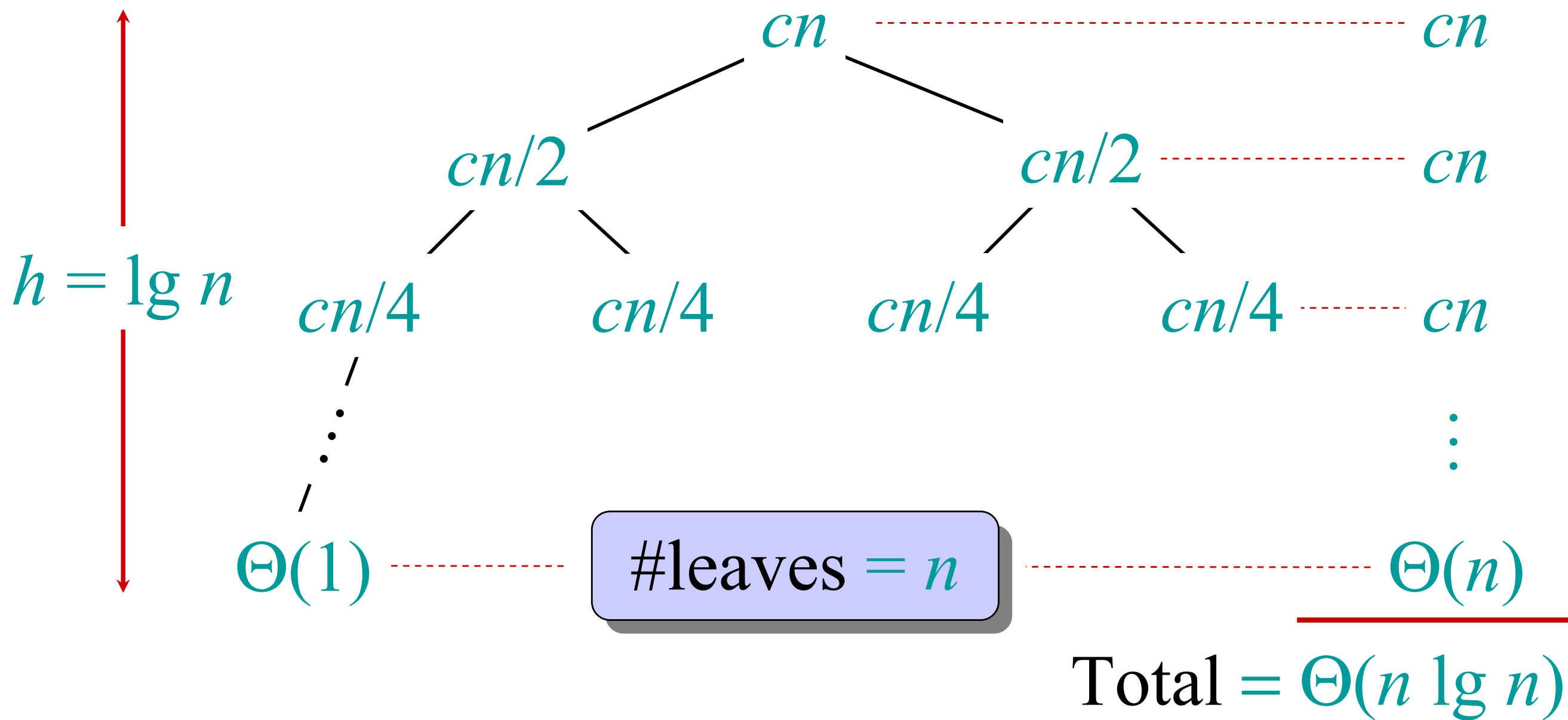












# Conclusions

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