

# Algorithms

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A decorative graphic in the bottom right corner consisting of a blue curved shape filled with various-sized circles in different shades of blue, creating a bubble-like or cellular pattern.

# 1. Basics of Algorithm Design and Analysis (part II)

# Contents

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- Analyzing and Designing algorithms
  - Kinds of analyses
  - Machine-independent time
  - The running time of the algorithm
  - Asymptotic performance

# Kinds of analyses

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- Worst-case (usually)
  - $T(n)$  = maximum time of algorithm on any input of size  $n$ .
  - Analysis for the worst-case input(s)
- Average-case (sometimes)
  - $T(n)$  = expected time of algorithm over all inputs of size  $n$ .  
(Analysis for all inputs)
  - Need assumption of statistical distribution of inputs
  - More difficult to analyze
- Best-case (bogus)
  - Analysis for the best-case input(s)

# Machine-independent time

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- *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.
    - Look at **growth** of  $T(n)$  as  $n$ .
- "Asymptotic Analysis"**

# Insertion sort

INSERTION-SORT ( $A, n$ )

**for**  $j \leftarrow 2$  **to**  $n$

**do**  $key \leftarrow A[j]$   $\triangleleft \triangleleft \triangleleft$   $n-1$  times

$i \leftarrow j - 1$   $\triangleleft \triangleleft \triangleleft$   $n-1$  times

**while**  $i > 0$  and  $A[i] > key$

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

$i \leftarrow i - 1$

$A[i+1] = key$

# $\Theta$ -notation

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## *Math:*

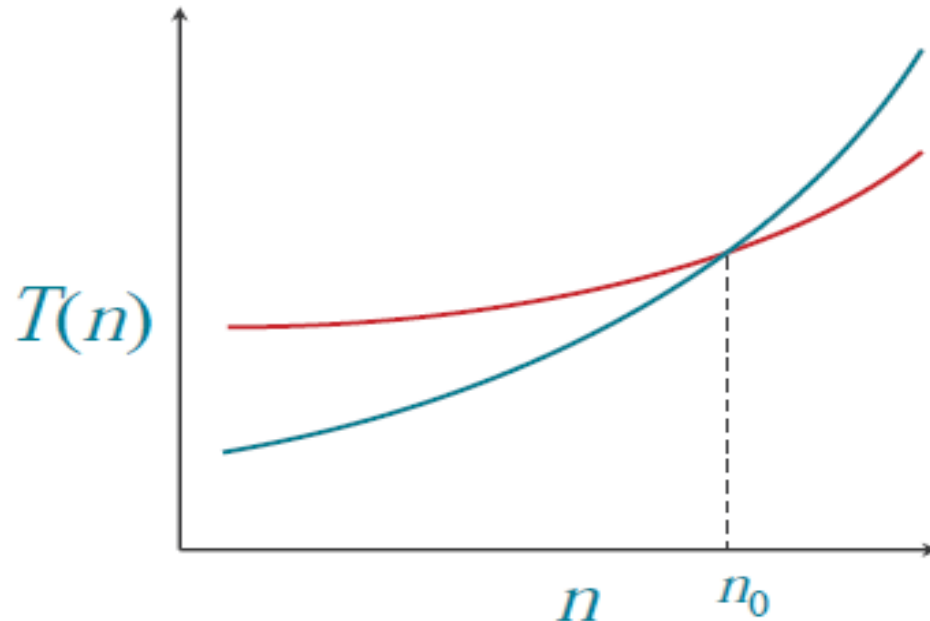
$\Theta(g(n)) = \{ f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0 \}$

## *Engineering:*

- 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.



# The running time of the algorithm

$$\sum (cost\ of\ statement) \cdot (number\ of\ times\ statement\ is\ executed).$$

INSERTION-SORT ( $A, n$ )

for  $j \leftarrow 2$  to  $n$

do  $key \leftarrow A[j]$   $\triangleleft \triangleleft \triangleleft$   $n-1$  times of Insertion sort.

$i \leftarrow j - 1$   $\triangleleft \triangleleft \triangleleft$   $n-1$  times  $- c_3(n-1) + c_4 \sum_{j=2}^n t_j + c_5 \sum_{j=2}^n (t_j - 1)$

while  $i > 0$  and  $A[i] > key$

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

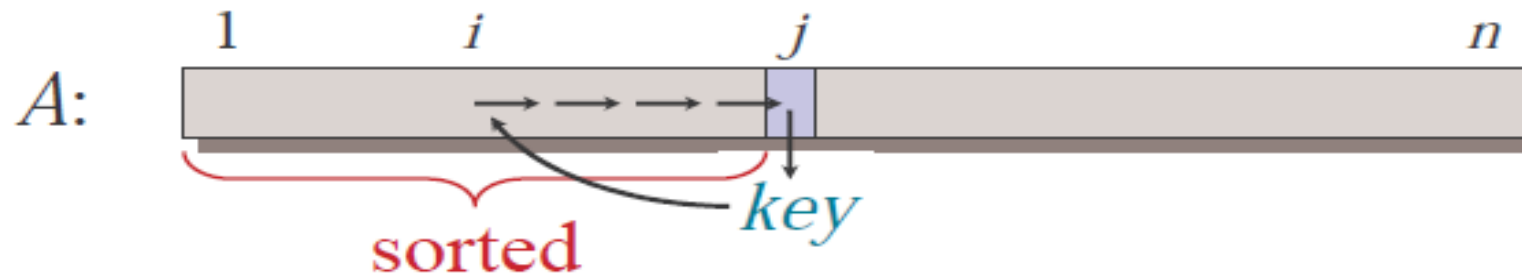
$i \leftarrow i - 1$   $+ c_7(n-1).$

$A[i+1] = key$

# Insertion sort

“pseudocode”

```
INSERTION-SORT ( $A, n$ )     $\triangleleft A[1 \dots n]$   
  for  $j \leftarrow 2$  to  $n$   
    do  $key \leftarrow A[j]$   
       $i \leftarrow j - 1$   
      while  $i > 0$  and  $A[i] > key$   
        do  $A[i+1] \leftarrow A[i]$   
           $i \leftarrow i - 1$   
       $A[i+1] = key$ 
```



# Insertion sort analysis

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***Worst case:*** Input reverse sorted.

$$T(n) = \sum_{j=2}^n \Theta(j) = \Theta(n^2) \quad [\text{arithmetic series}]$$

***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  $n$ .

# Designing algorithms

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- There are many ways to design algorithms  
For example, insertion sort is incremental:  
having sorted  $A[1 \dots j-1]$ , place  $A[j]$  correctly,  
So that  $A[1 \dots J]$  is sorted.

*Another common approach*

- Divide and Conquer
  - Divide: the problem into a number of subproblems that are smaller instances of the same problem.
  - Conquer: the subproblems by solving them recursively.
  - Combine: the subproblem solutions to give a solution to the original problem.

# Merge sort

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

***Key subroutine: MERGE***

# Merge sort

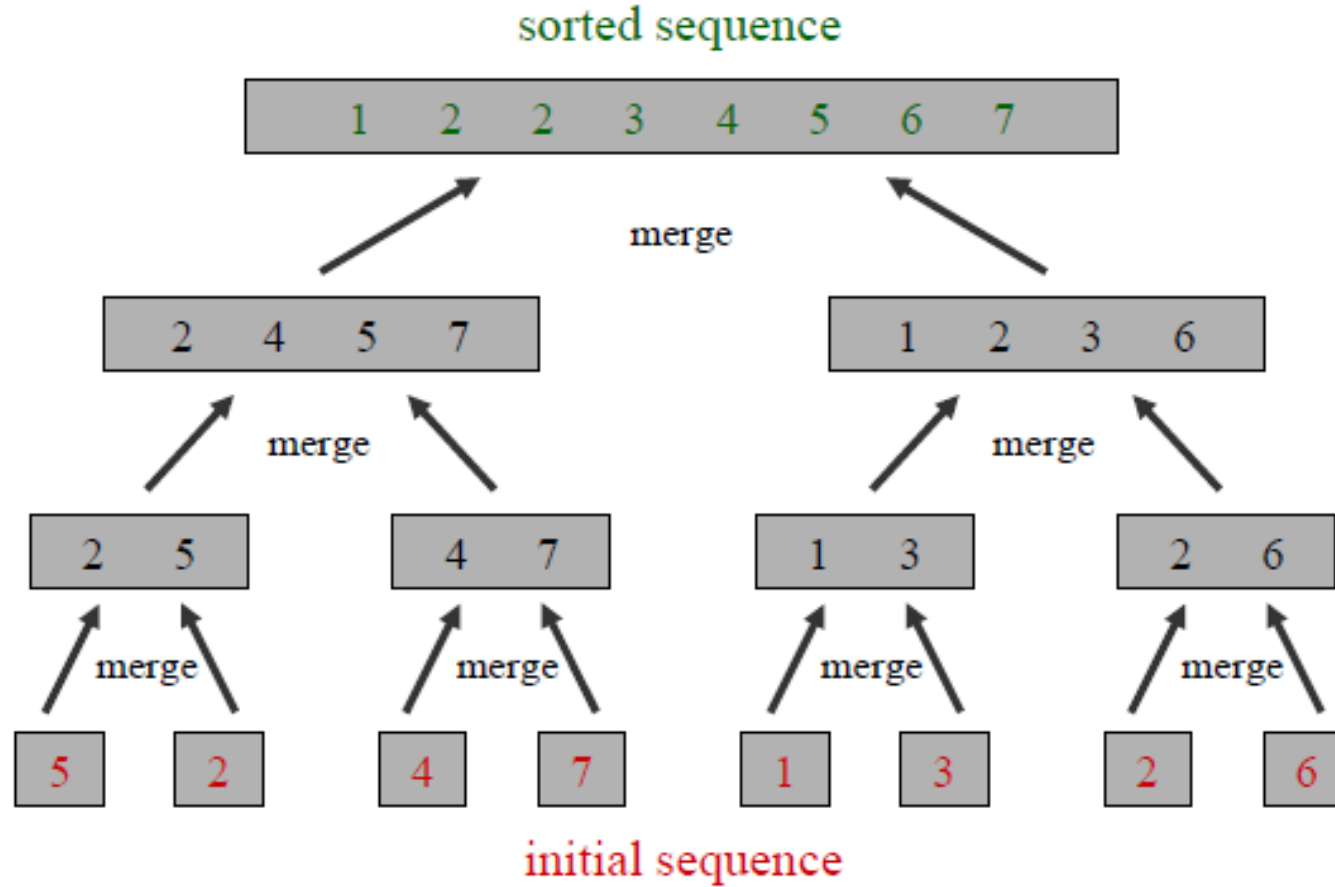
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**MERGE-SORT**  $A[1 \dots n]$

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**

# Operation of merge sort



<https://www.youtube.com/watch?v=JSceec-wEyw>

# Merging two sorted arrays

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20 12

13 11

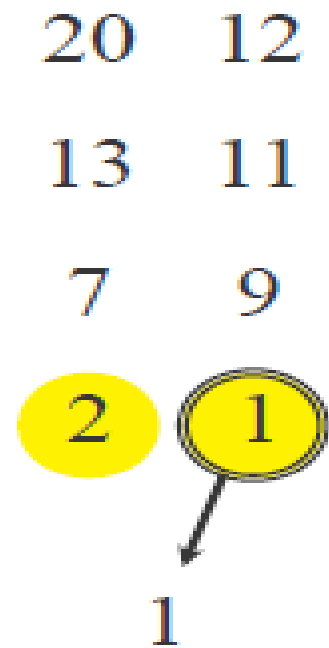
7 9

2 1



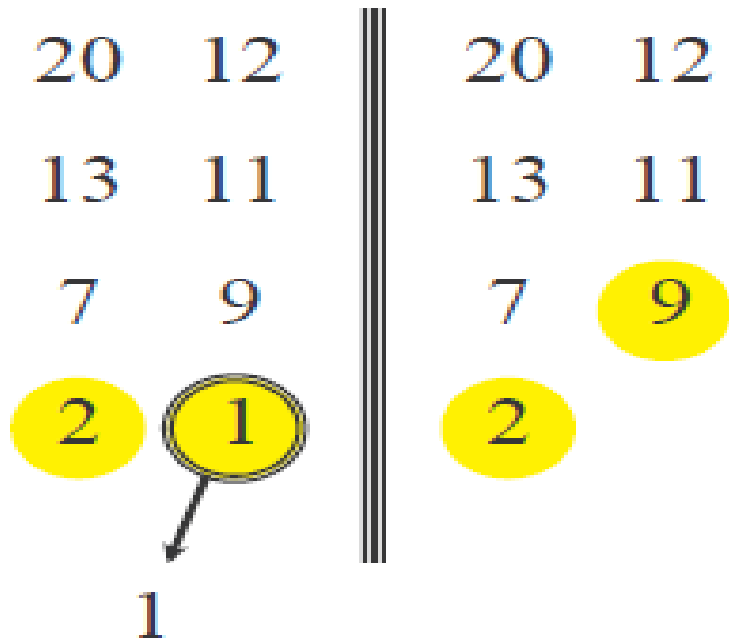
# Merging two sorted arrays

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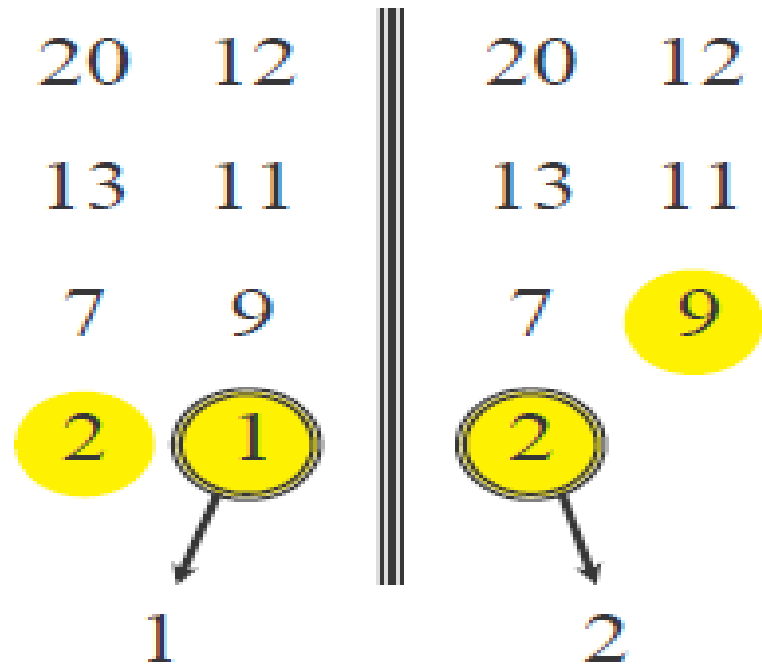
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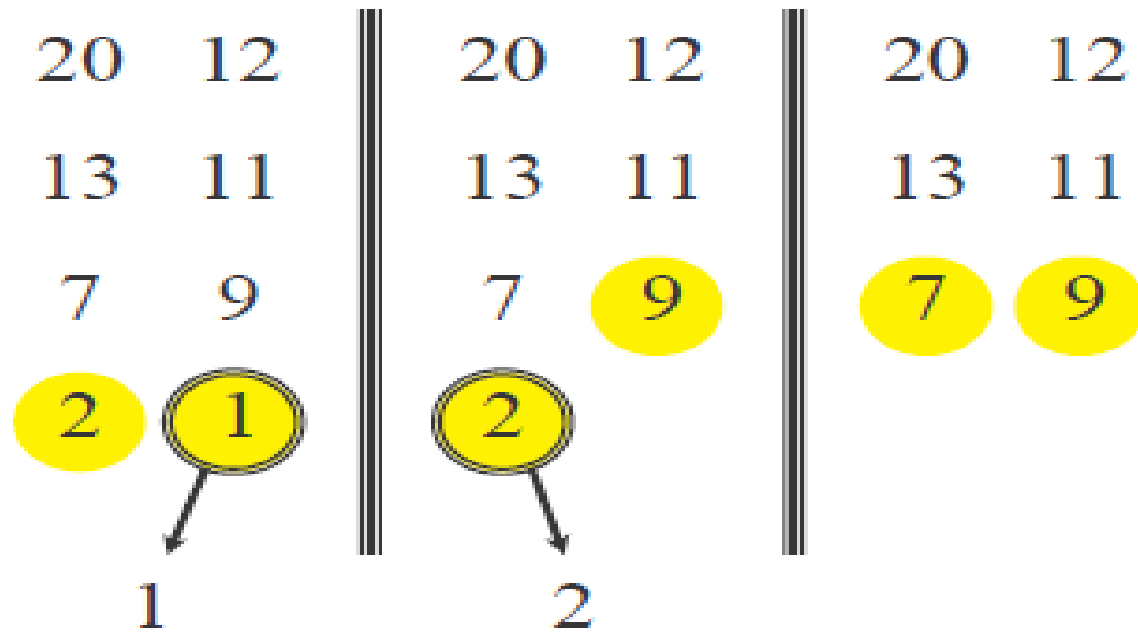


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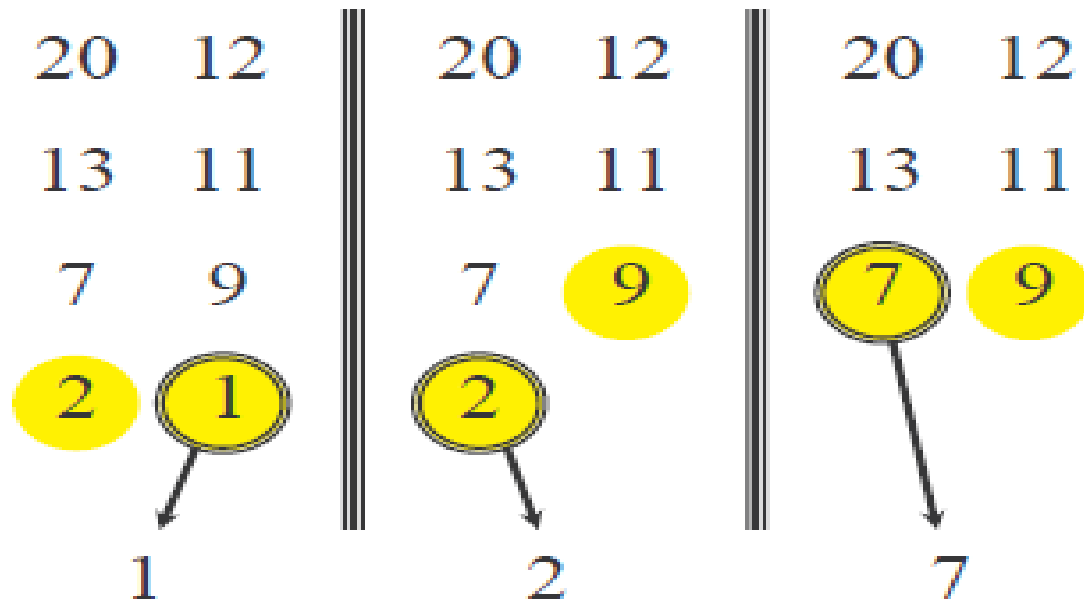


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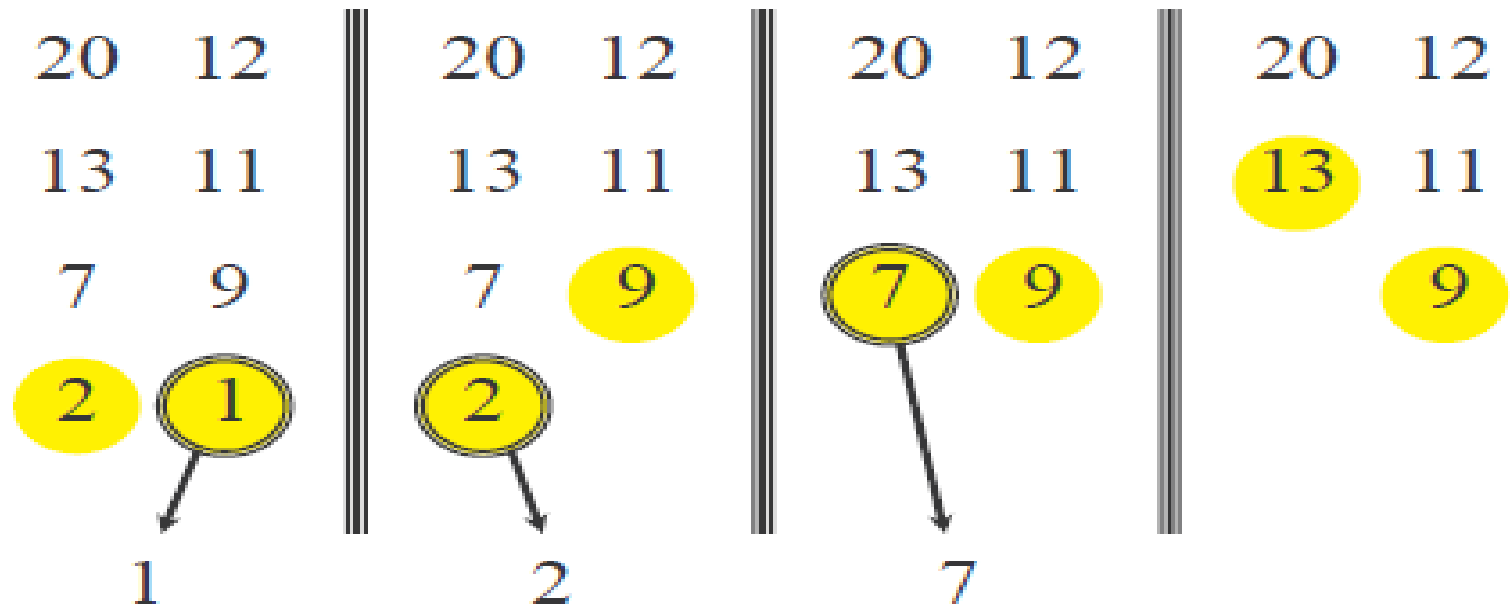


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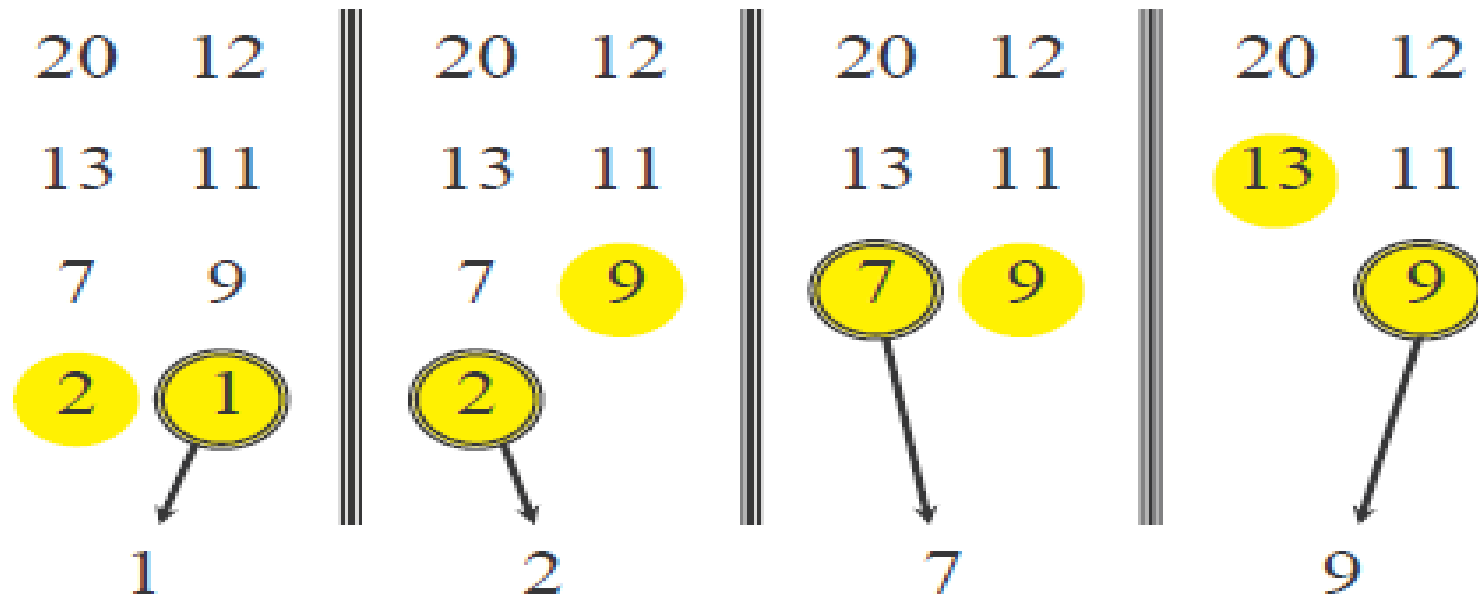
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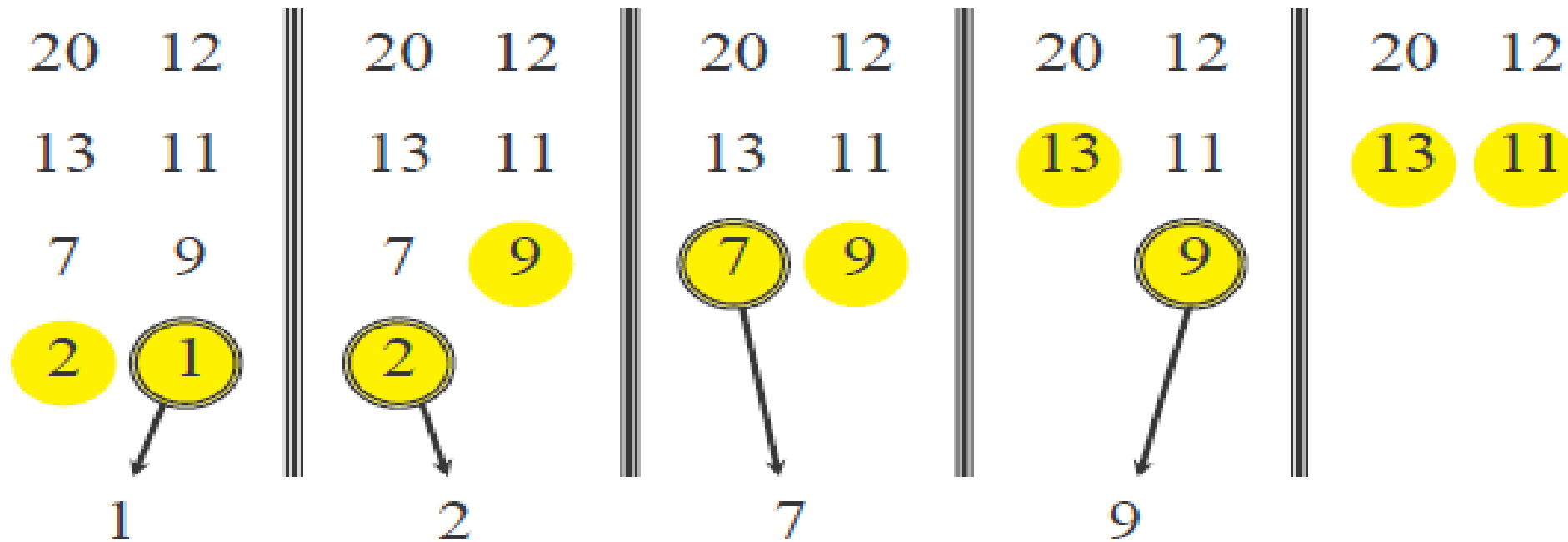
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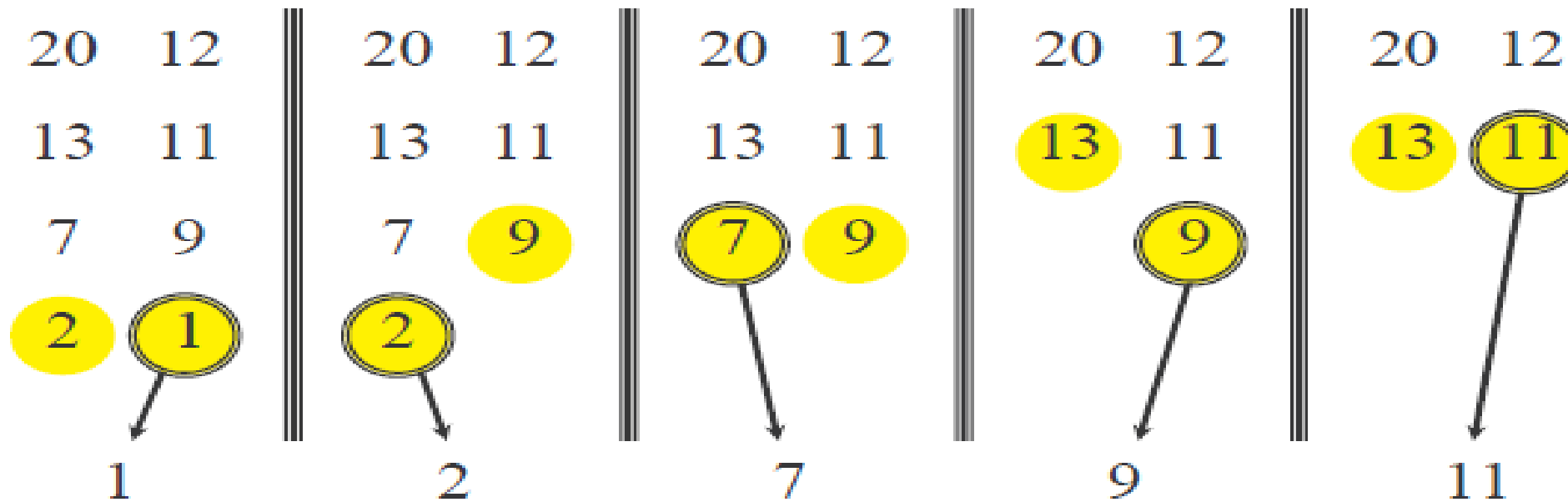


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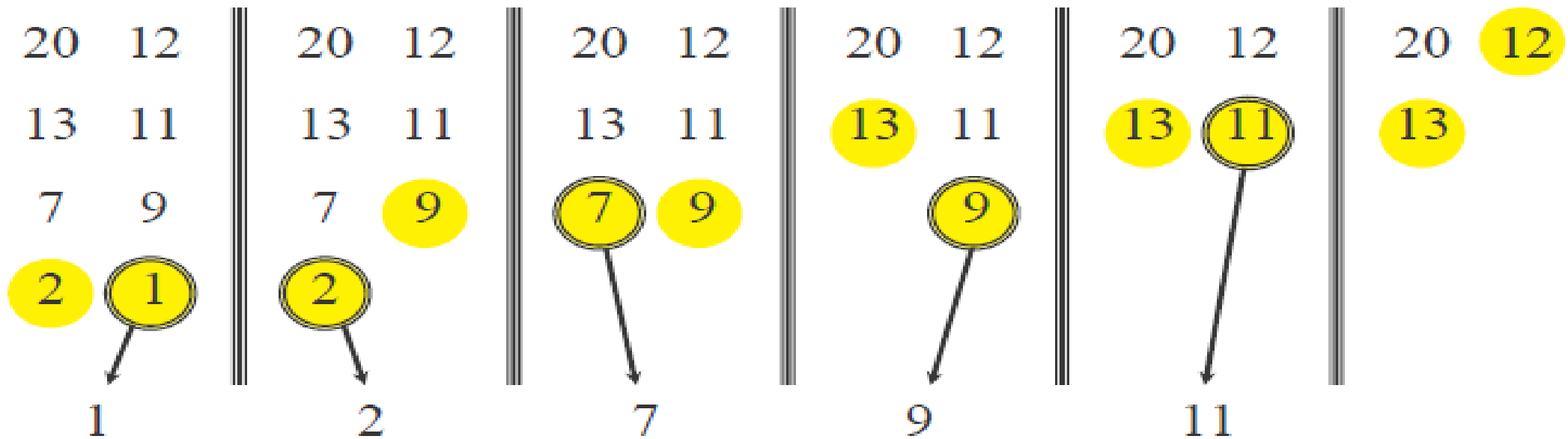




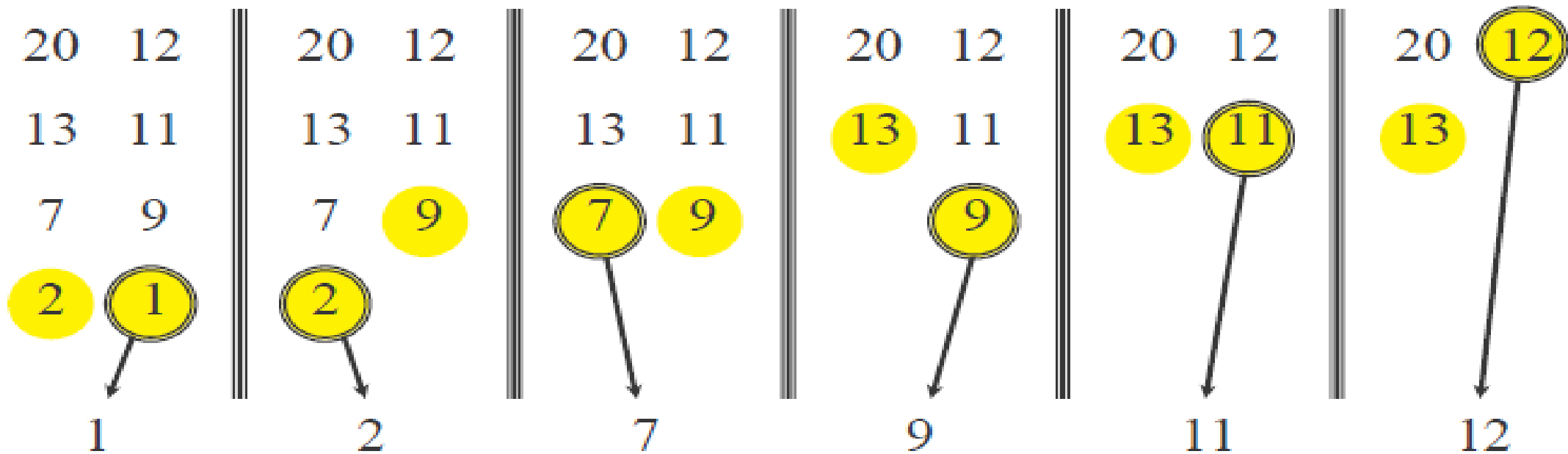
# Merging two sorted arrays



# Merging two sorted arrays

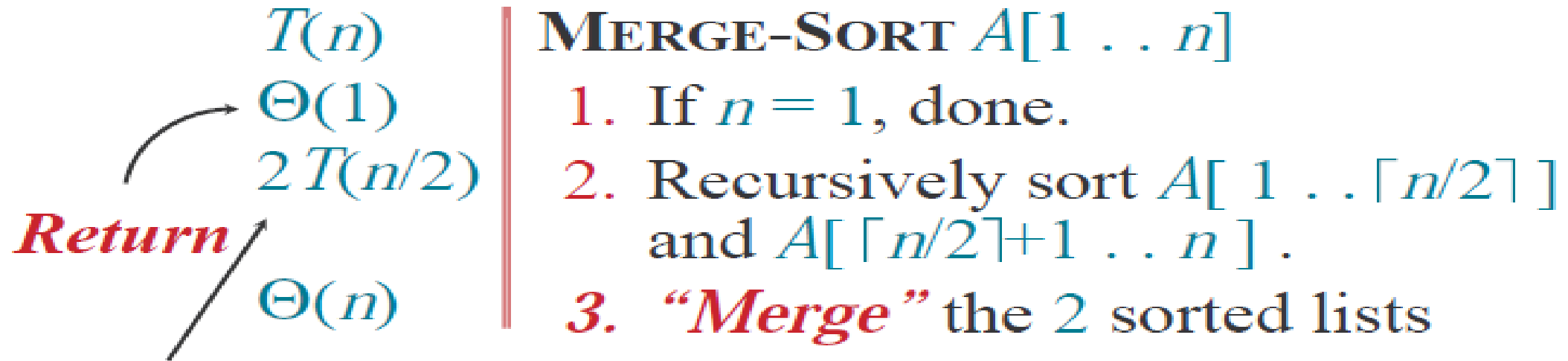


# Merging two sorted arrays



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

# Analyzing merge sort



**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

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

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.

# Recursion tree

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

# Recursion tree

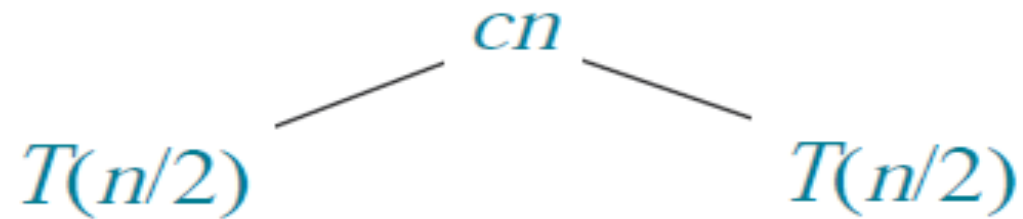
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 $T(n)$

# Recursion tree

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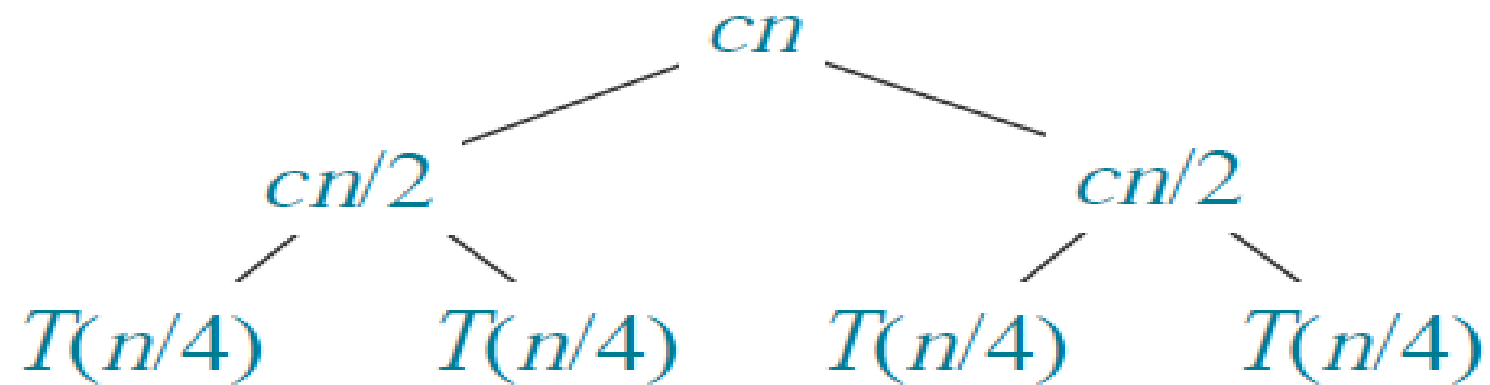




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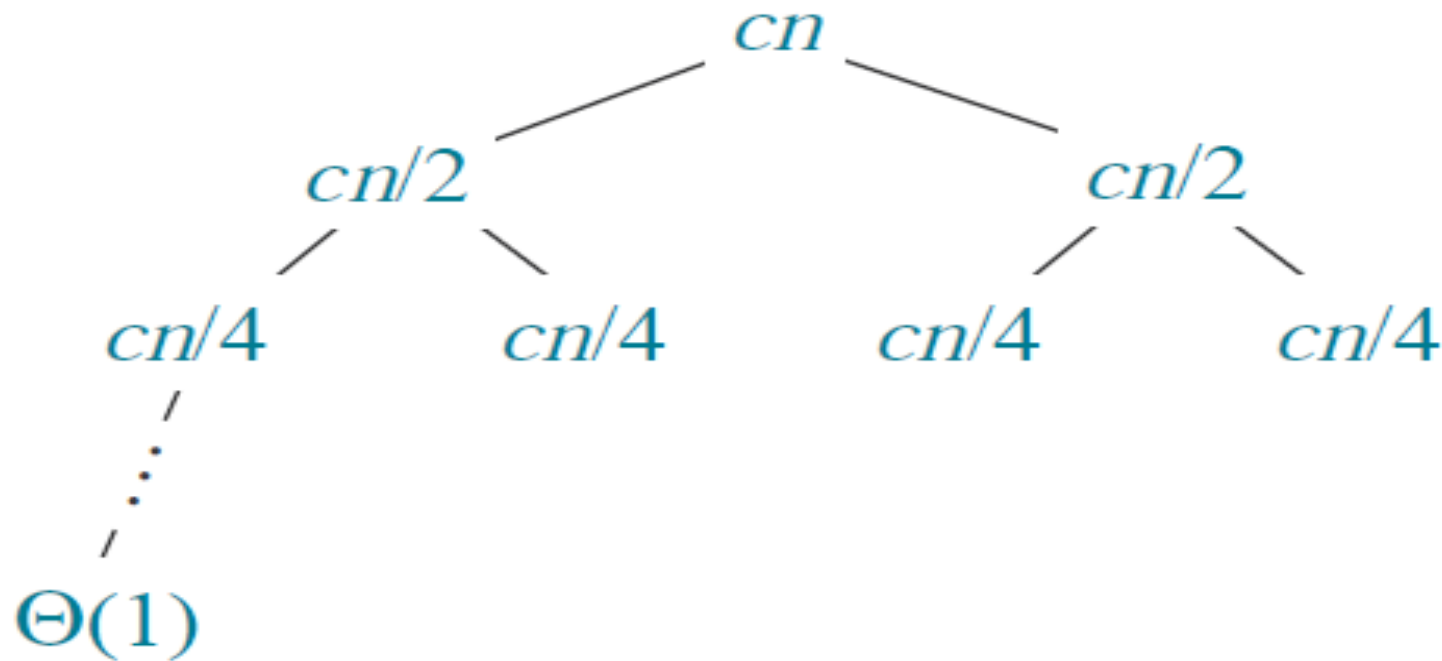
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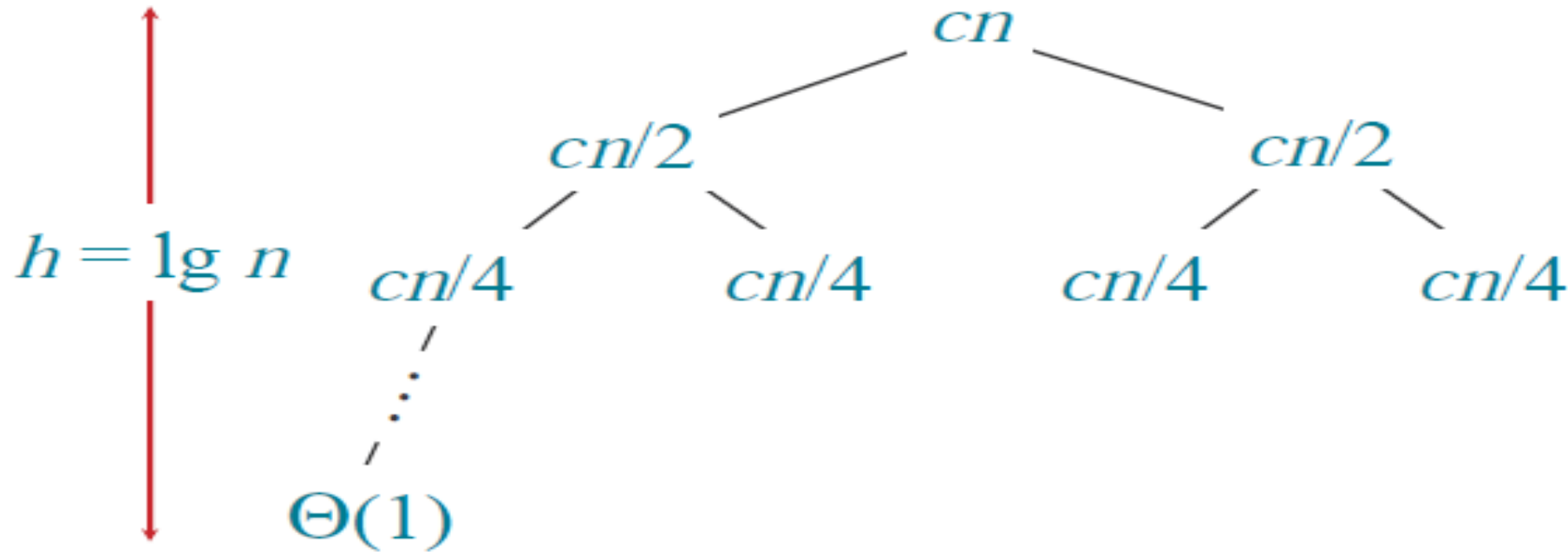
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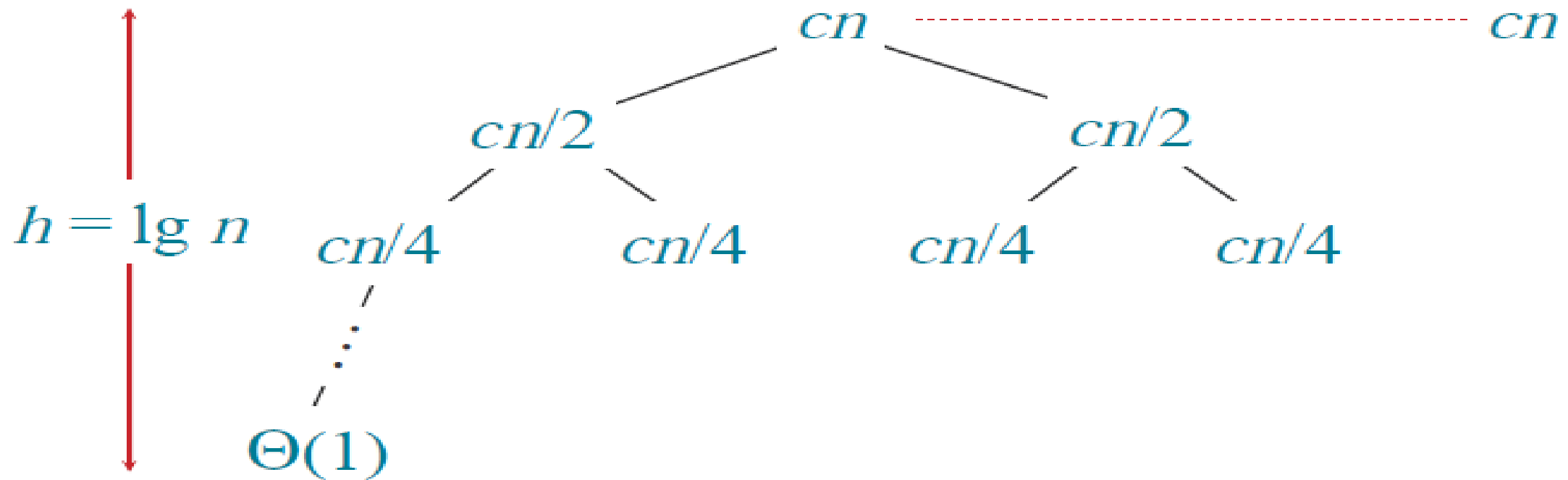
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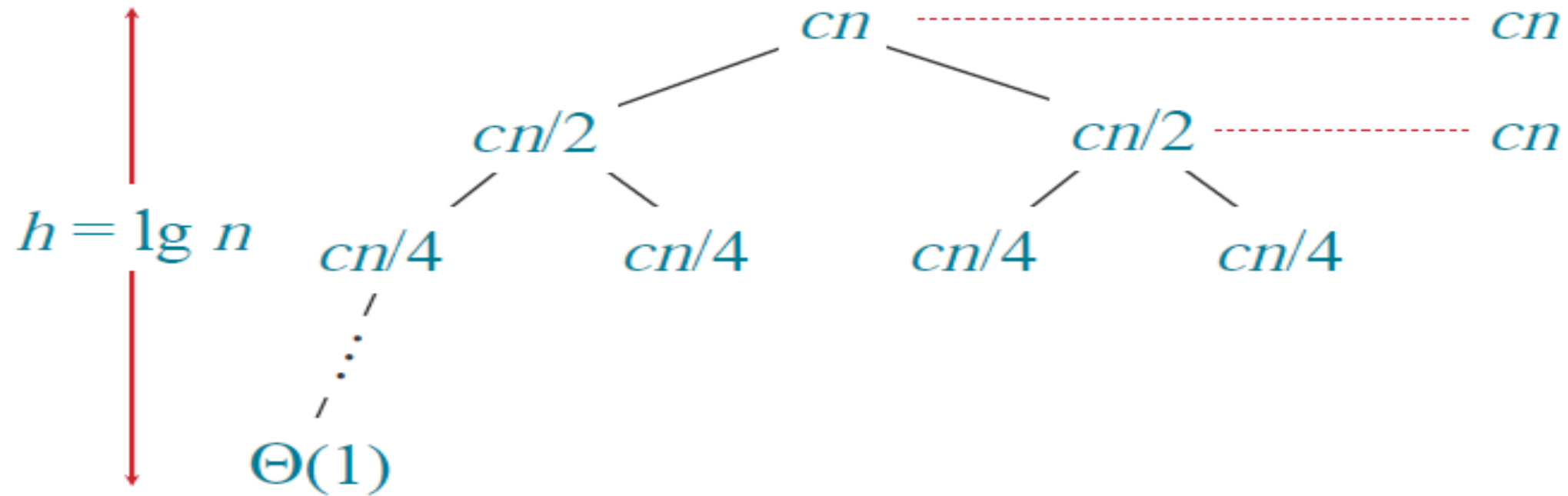
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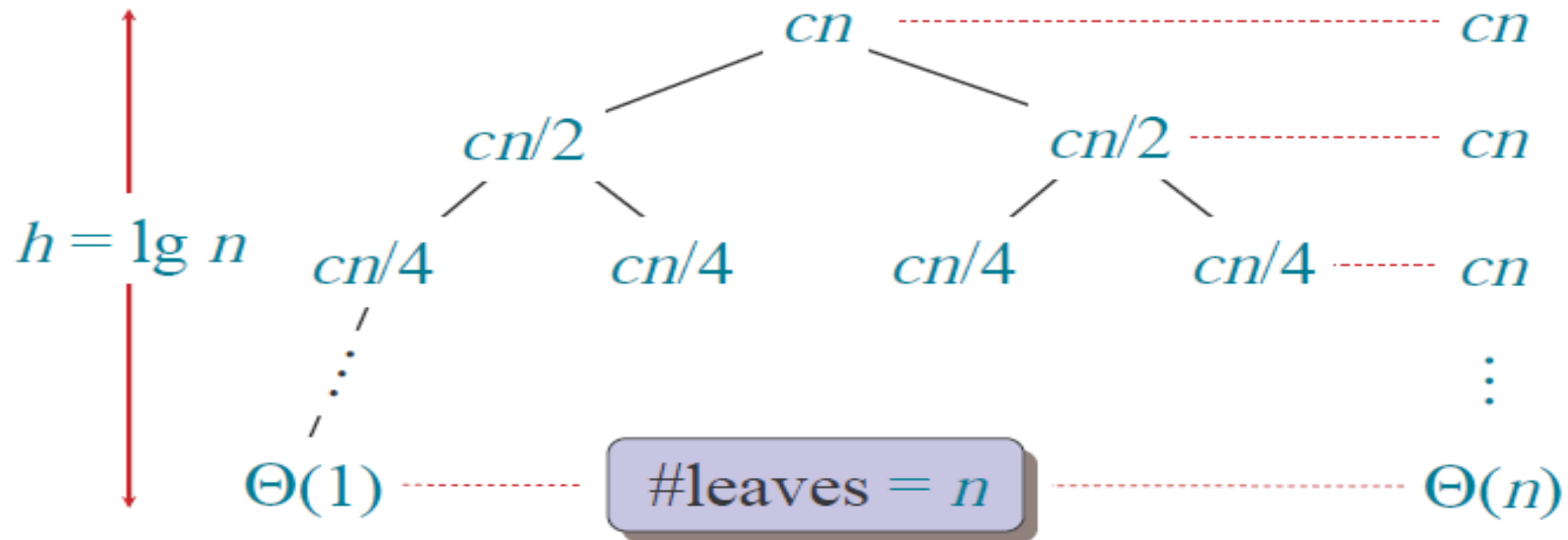
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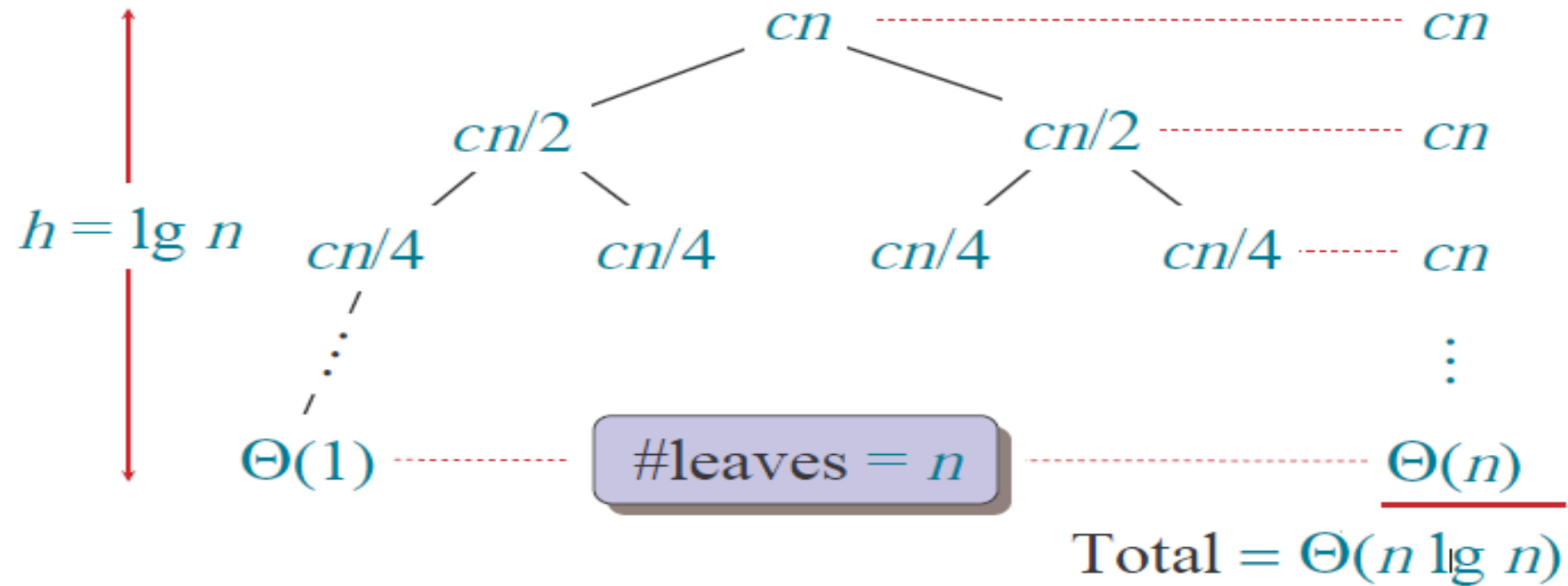
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# Conclusions

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$\Theta(n \lg n)$  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.



# **Implementation**

# Implementation of merge sort

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**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**

# Implementation of merge sort

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*Key subroutine:* MERGE

```
[1] # Definition of merge sort
def mergeSort(A, p:int, r:int):
    if p < r:
        q = (p+r) // 2
        mergeSort(A, p, q)
        mergeSort(A, q+1, r)
        merge(A, p, q, r)

def merge(A, p:int, q:int, r:int):
    i = p; j = q+1; t = 0
    tmp = [0 for i in range(len(A))]
    while i <= q and j <= r:
        if A[i] <= A[j]:
            tmp[t] = A[i]; t += 1; i += 1
        else:
            tmp[t] = A[j]; t += 1; j += 1
    while i <= q:
        tmp[t] = A[i]; t += 1; i += 1
    while j <= r:
        tmp[t] = A[j]; t += 1; j += 1
    i = p; t = 0
    while i <= r:
        A[i] = tmp[t]; t+=1; i+=1
```

```
[2] # List
input_list1 = [8, 2, 4, 9, 3, 6]
print(input_list1)

[8, 2, 4, 9, 3, 6]
```

```
[3] # Sorting
mergeSort(input_list1, 0, len(input_list1)-1)
print(input_list1)

[2, 3, 4, 6, 8, 9]
```

```
[4] # random list
import random
input_list2 = random.sample(range(100),10)
print(input_list2)

[42, 72, 31, 30, 96, 35, 49, 3, 20, 12]
```

```
[5] # Sorting
mergeSort(input_list2, 0, len(input_list2) -1 )
print(input_list2)

[3, 12, 20, 30, 31, 35, 42, 49, 72, 96]
```

# Example code test

- Code test: <https://www.acmicpc.net/problem/2751>
- Solving the problem using merge sort
- Example result of submission

2751	aikiho	모든 언어 ▼	모든 결과 ▼	검색
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제출 번호	아이디	문제	결과	메모리	시간	언어	코드 길이	제출한 시간
48335930	aikiho	2751	맞았습니다!!	139560 KB	1444 ms	PyPy3 / 수정	806 B	2분 전

**THANK YOU**

