

# ADVANCED ANALYSIS OF ALGORITHMS

## CPS 5440

# UNIT I: THE ROLE OF ALGORITHMS IN COMPUTING AND FOUNDATIONS REVIEW

*The theoretical study of computer-program performance and resource usage.*

What's more important than performance?

- modularity
- correctness
- maintainability
- functionality
- robustness
- user-friendliness
- programmer time
- simplicity
- extensibility
- reliability

# WHY STUDY ALGORITHMS AND PERFORMANCE?

- Algorithms help us to understand ***scalability***.
- Performance often draws the line between what is feasible and what is impossible.
- Algorithmic mathematics provides a ***language*** for talking about program behavior.
- Performance is the ***currency*** of computing.
- The lessons of program performance generalize to other computing resources.
- Speed is fun!

# 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

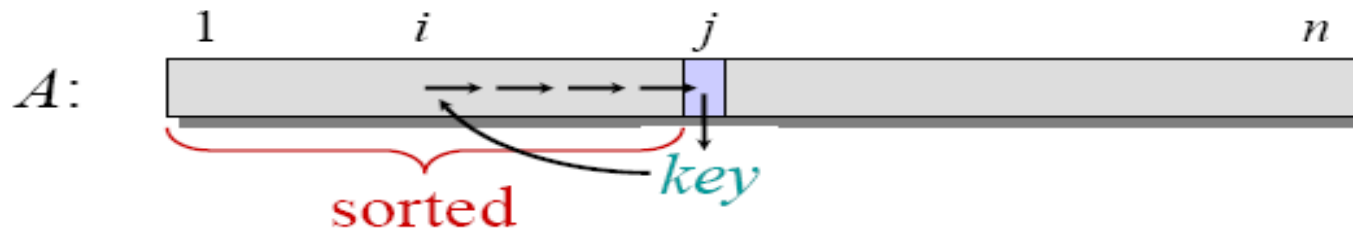
# INSERTION SORT

“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

“pseudocode” {

```
INSERTION-SORT ( $A, n$ )  $\triangleleft A[1 \dots n]$   
  
  for  $j \leftarrow 2$  to  $n$   
    do  $key \leftarrow A[j]$   
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```



## EXAMPLE OF INSERTION SORT

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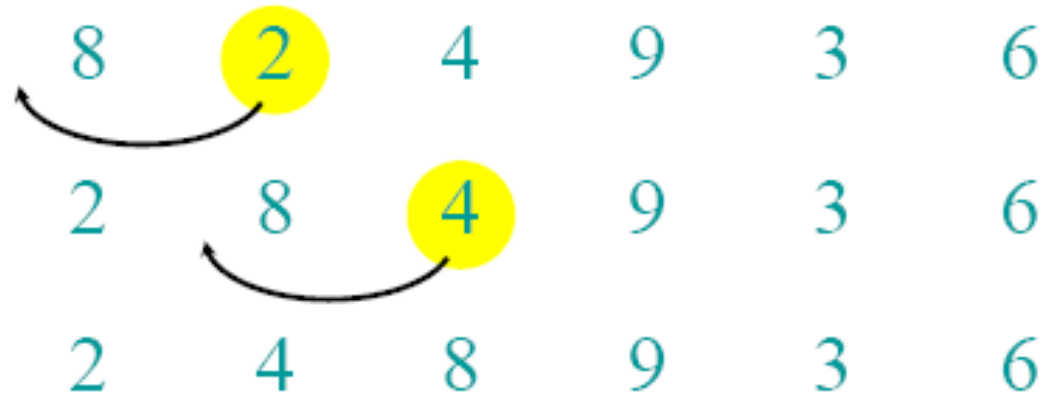
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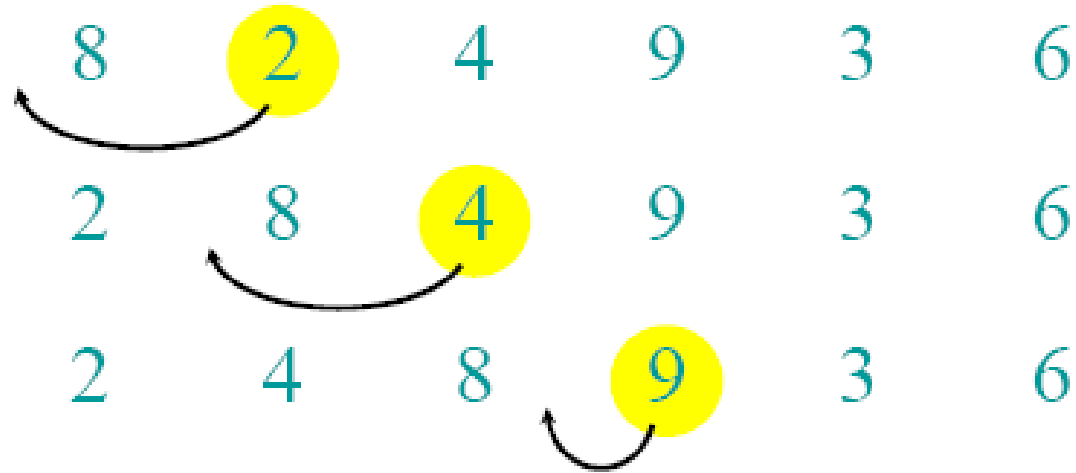
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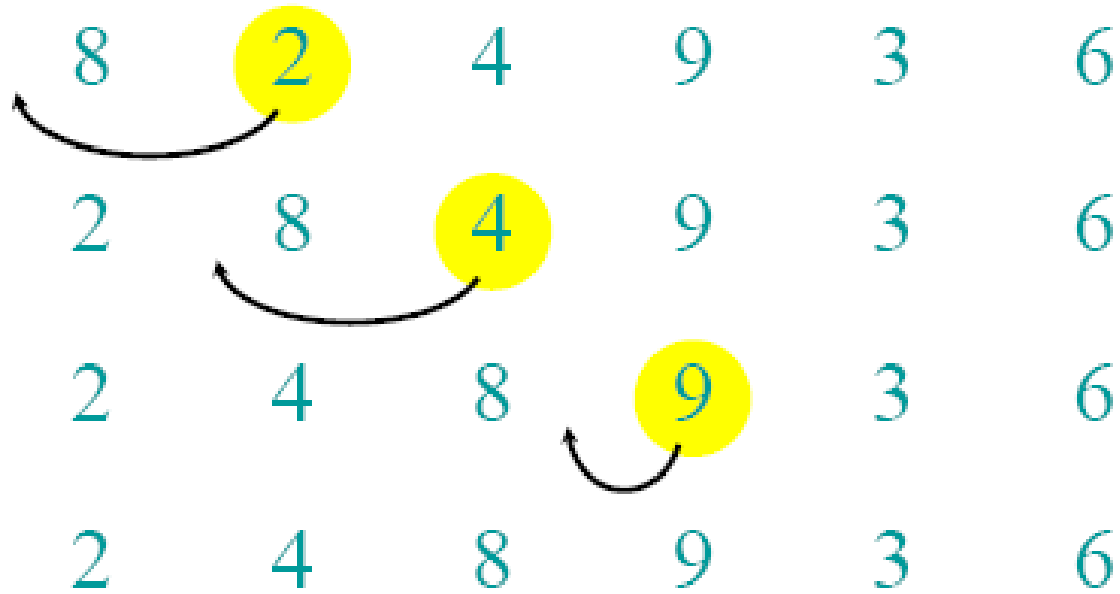
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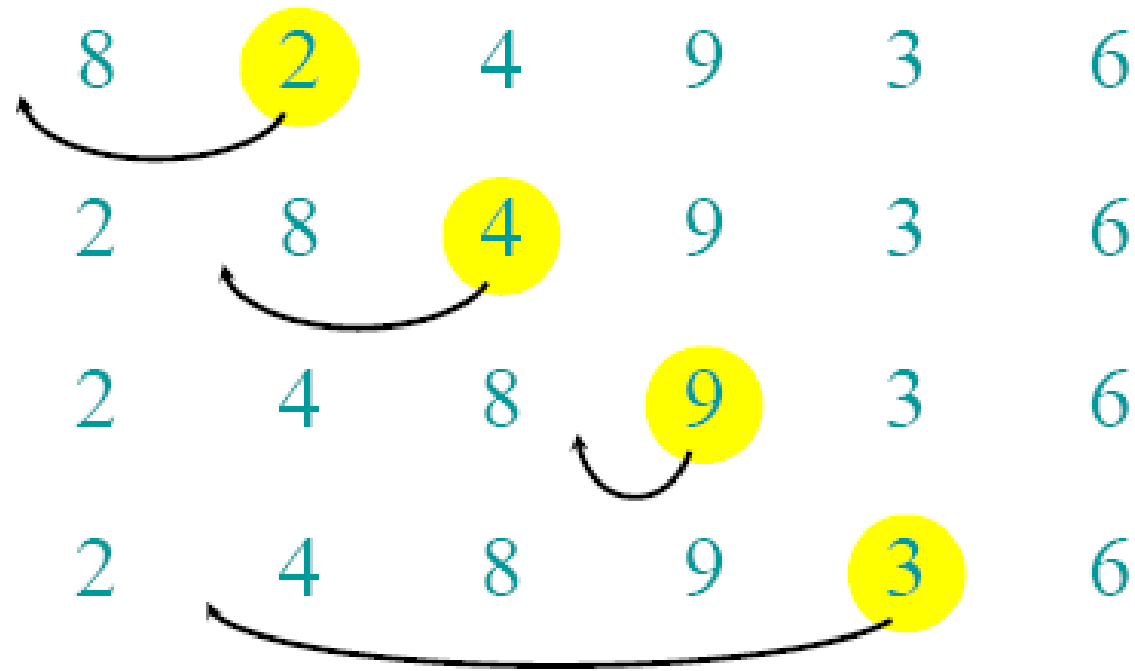
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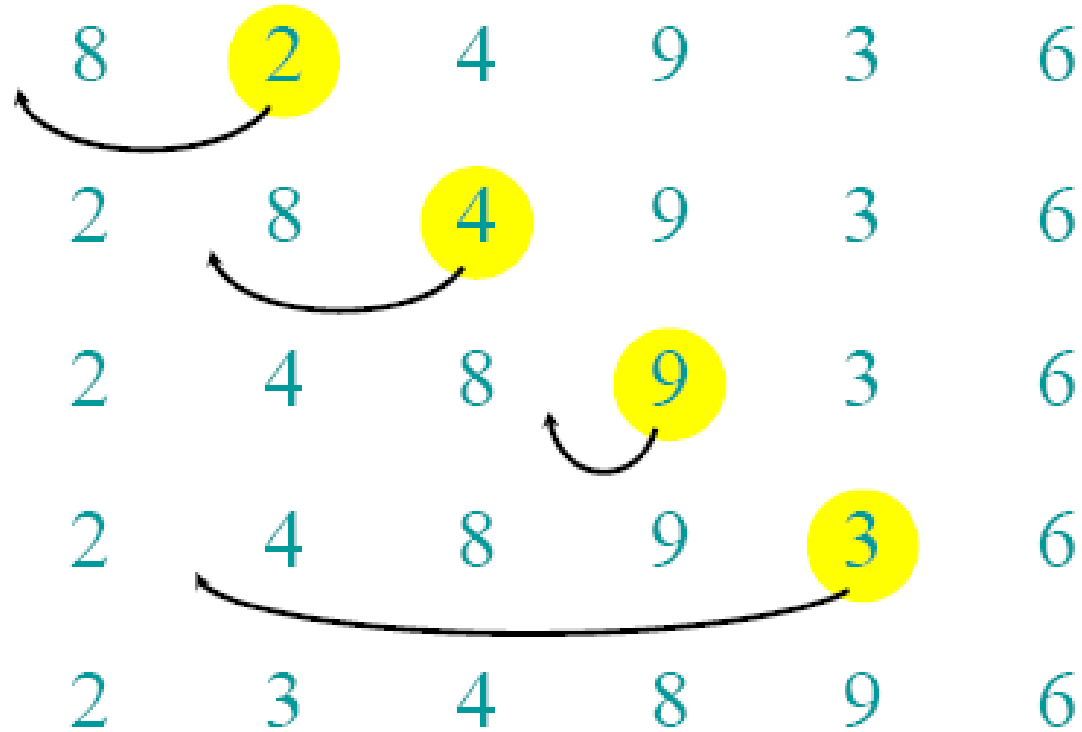
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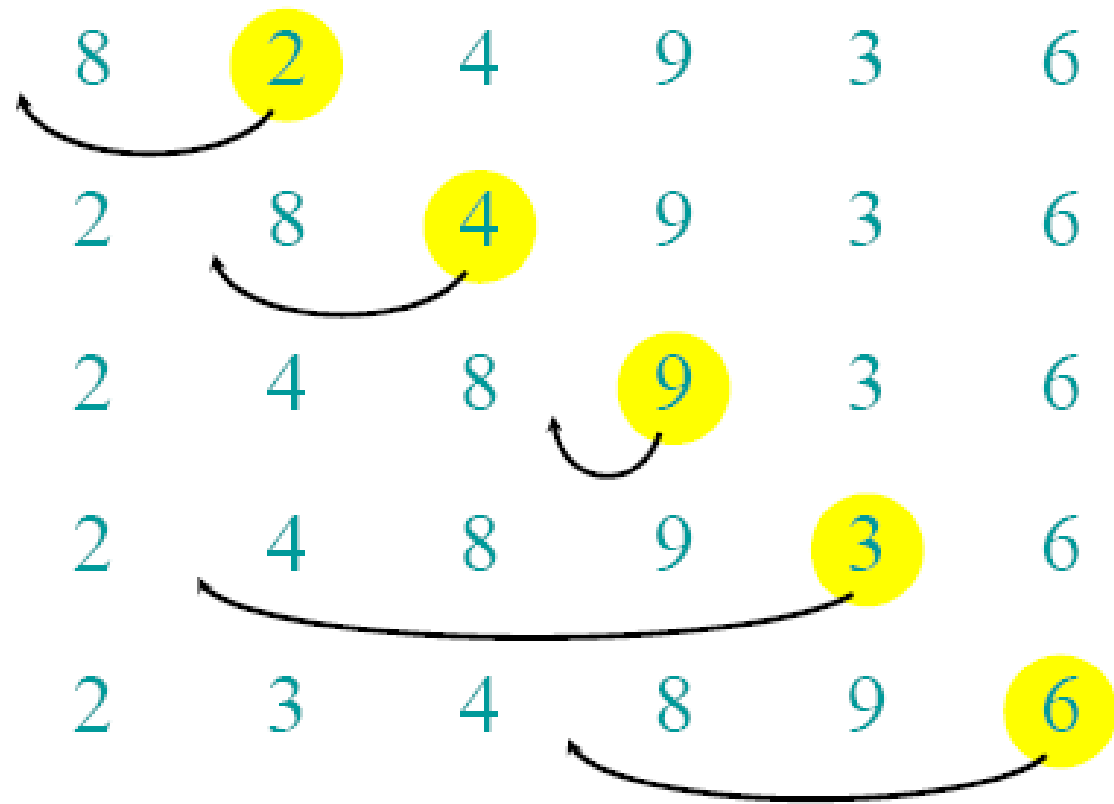
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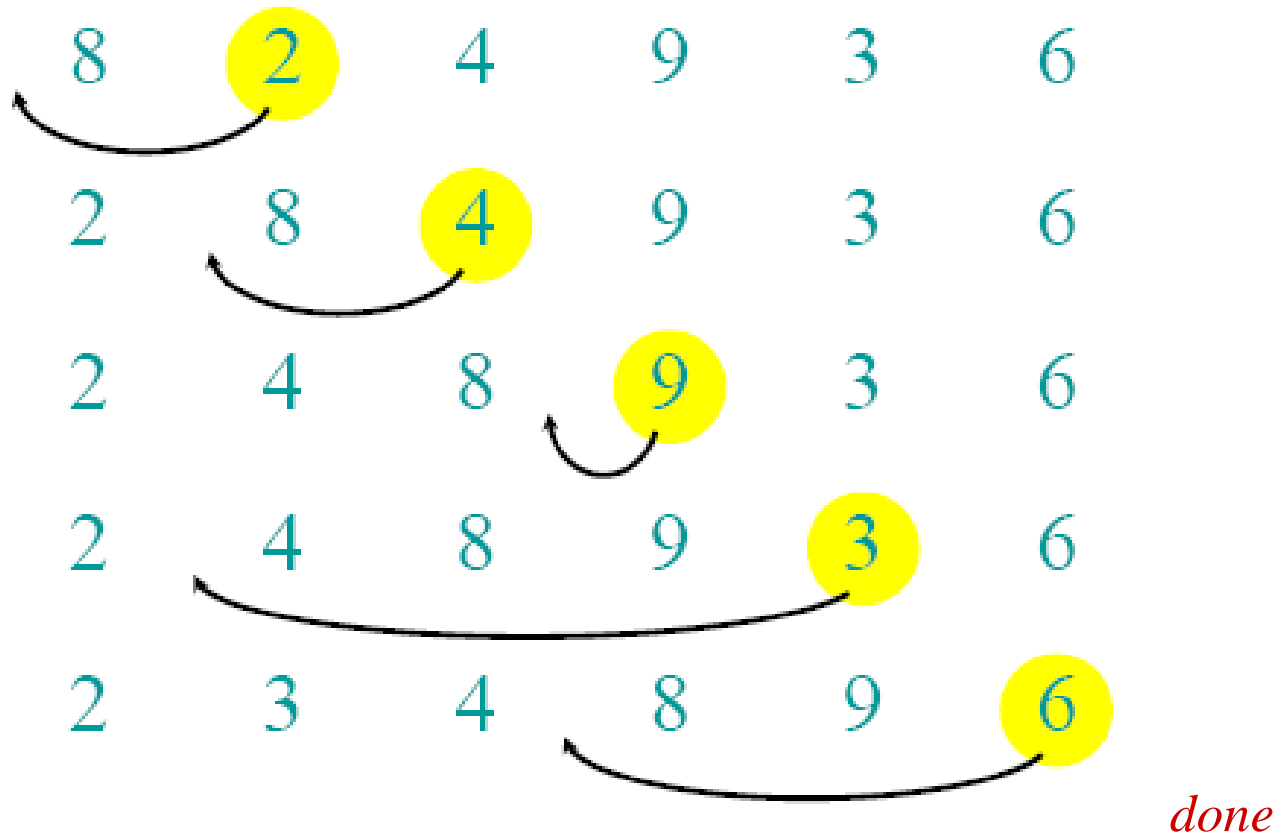


## EXAMPLE OF INSERTION SORT





## EXAMPLE OF INSERTION SORT



- The running time depends on the input:  
an already sorted sequence is easier to sort.
- Parameterize 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  $n$ .
- Need assumption of statistical distribution of inputs.

**Best-case:** (bogus)

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

*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  $n \rightarrow \infty$ .

“Asymptotic Analysis”

## *Math:*

$\Theta(g(n)) = \{ f(n) : \text{there exist positive constants } c_1, c_2, \text{ and}$

$n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n)$

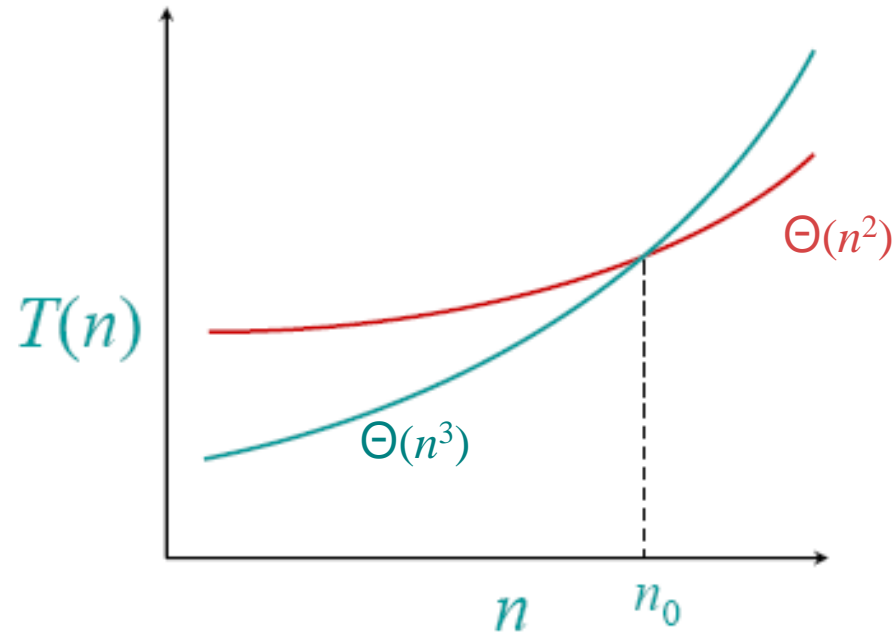
$\text{for all } n \geq n_0 \}$

## *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.

**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  $n$ .

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

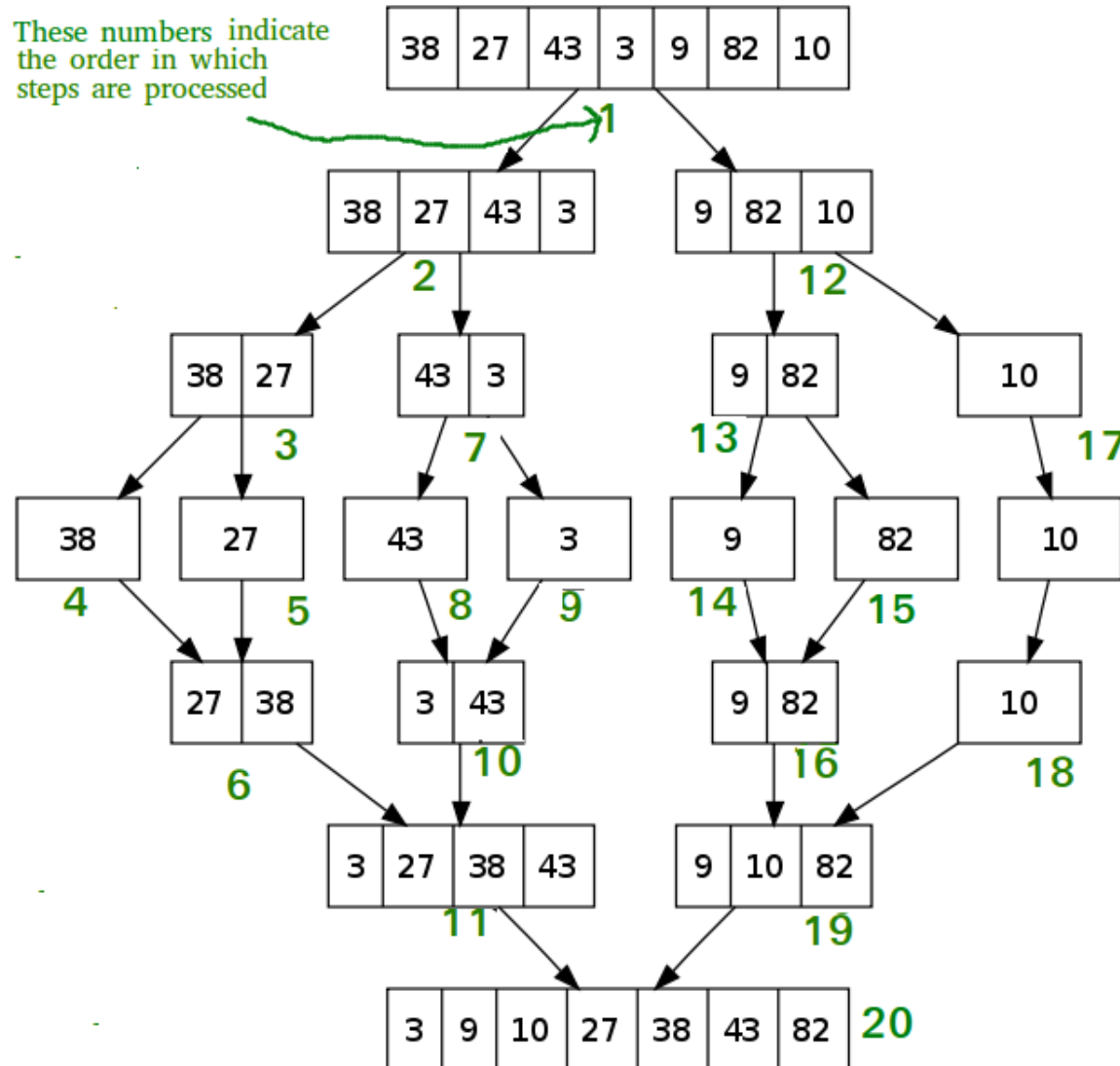
1. If  $n = 1$ , done.
2. Recursively sort  $A[1 \dots \lfloor n/2 \rfloor]$   
and  $A[\lfloor n/2 \rfloor + 1 \dots n]$ .
3. “*Merge*” the 2 sorted lists.

*Key subroutine:* MERGE



# MERGE SORT EXAMPLE

These numbers indicate the order in which steps are processed



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

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***Key subroutine:* MERGE**

# MERGING TWO SORTED ARRAYS

20 12

13 11

7 9

2 1

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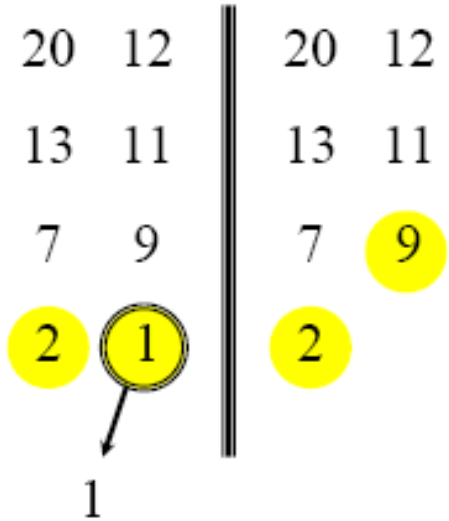
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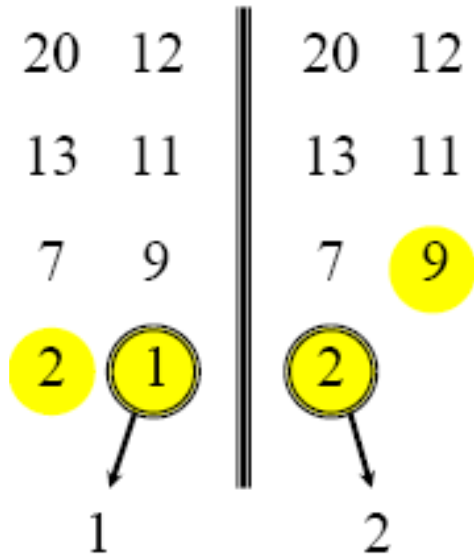
1



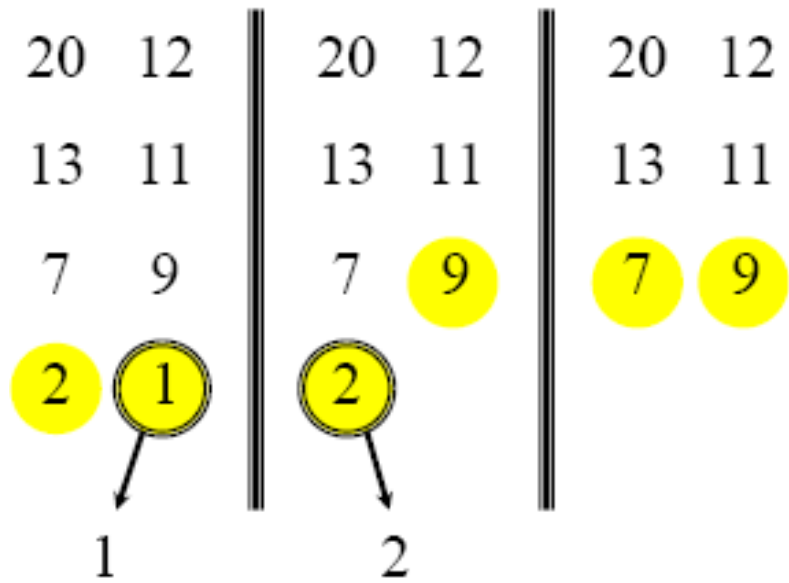
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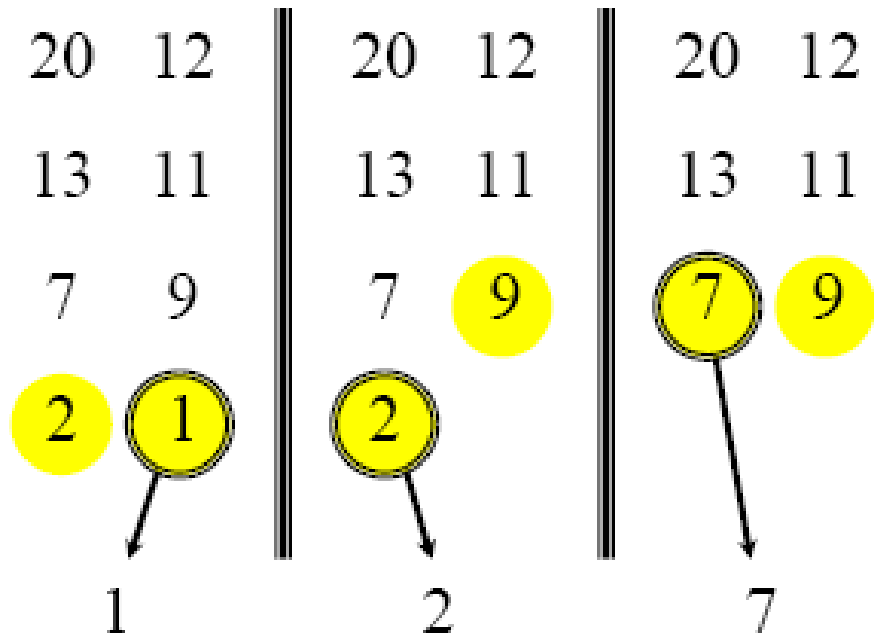
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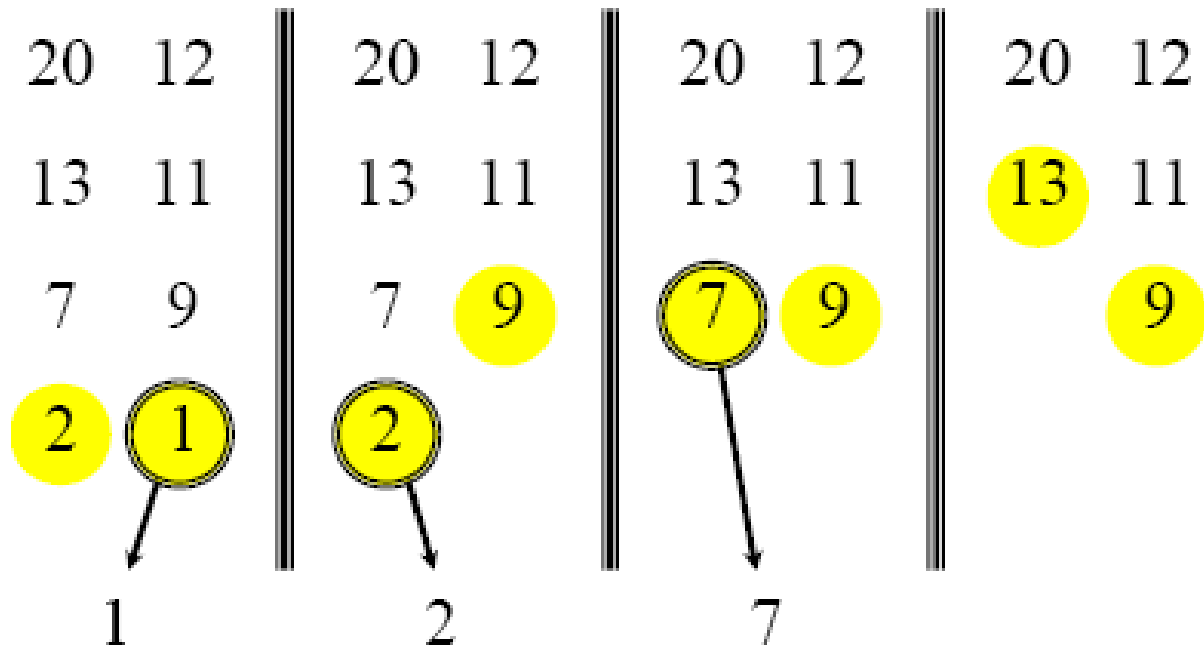
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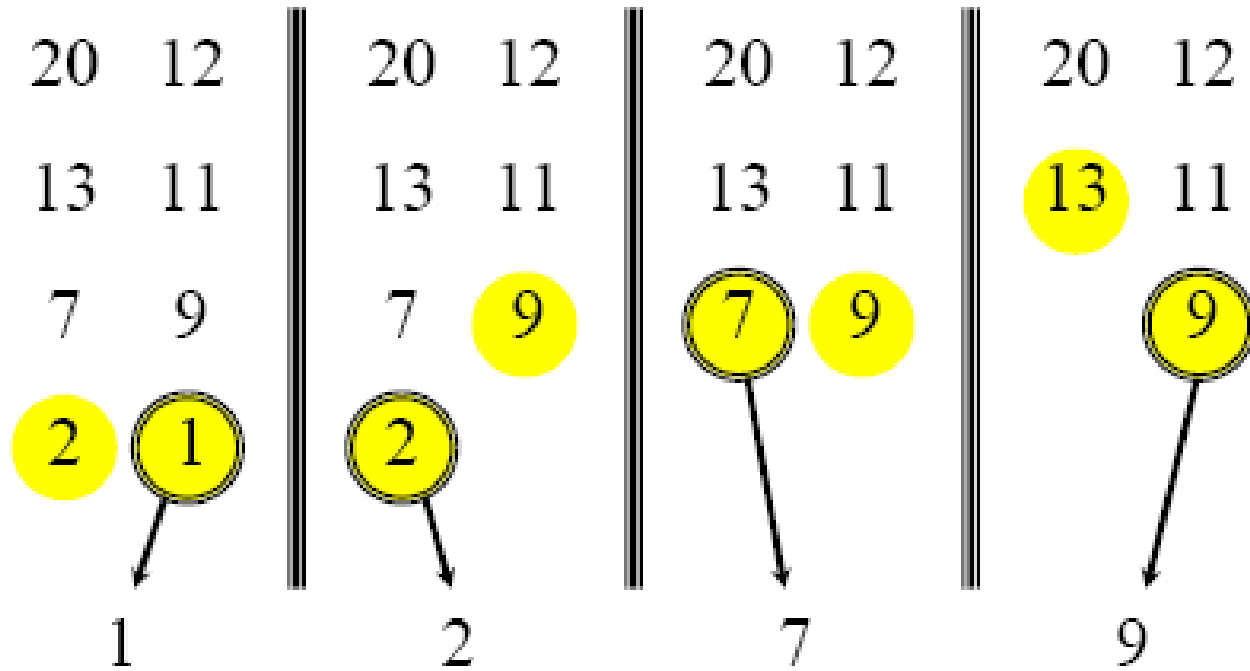


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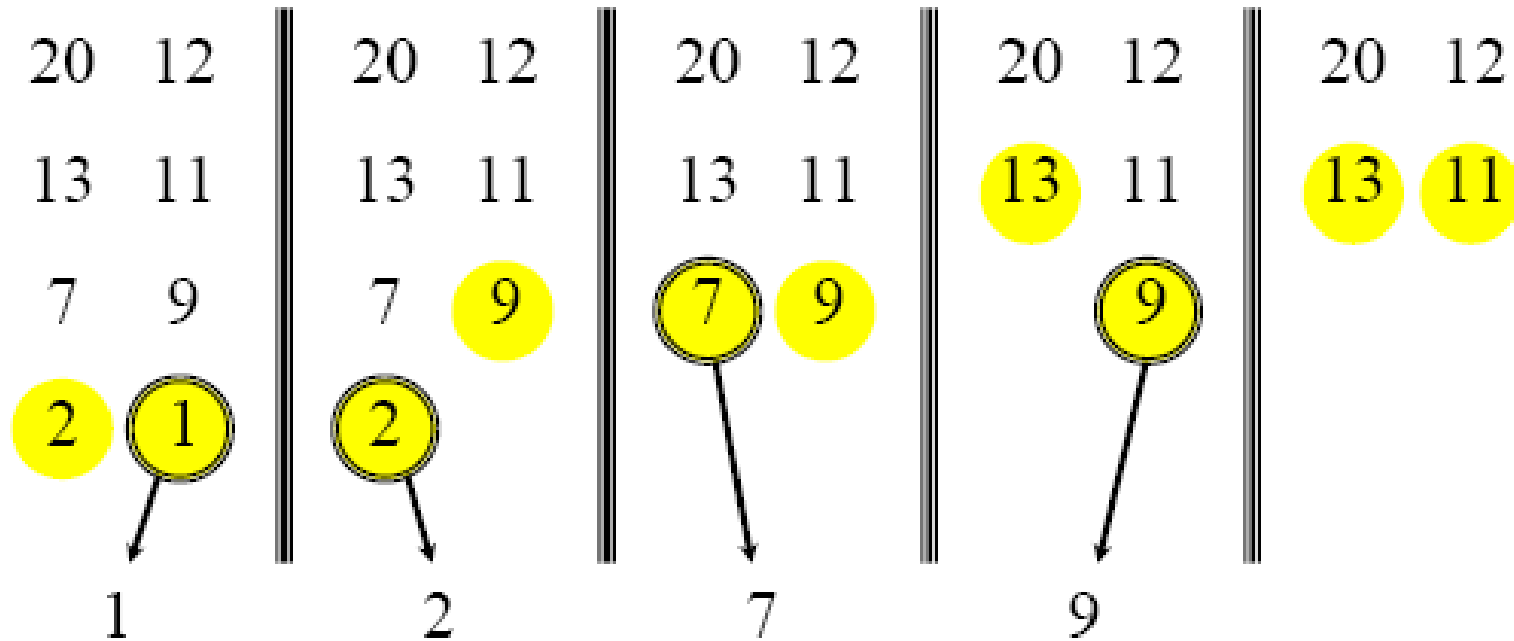




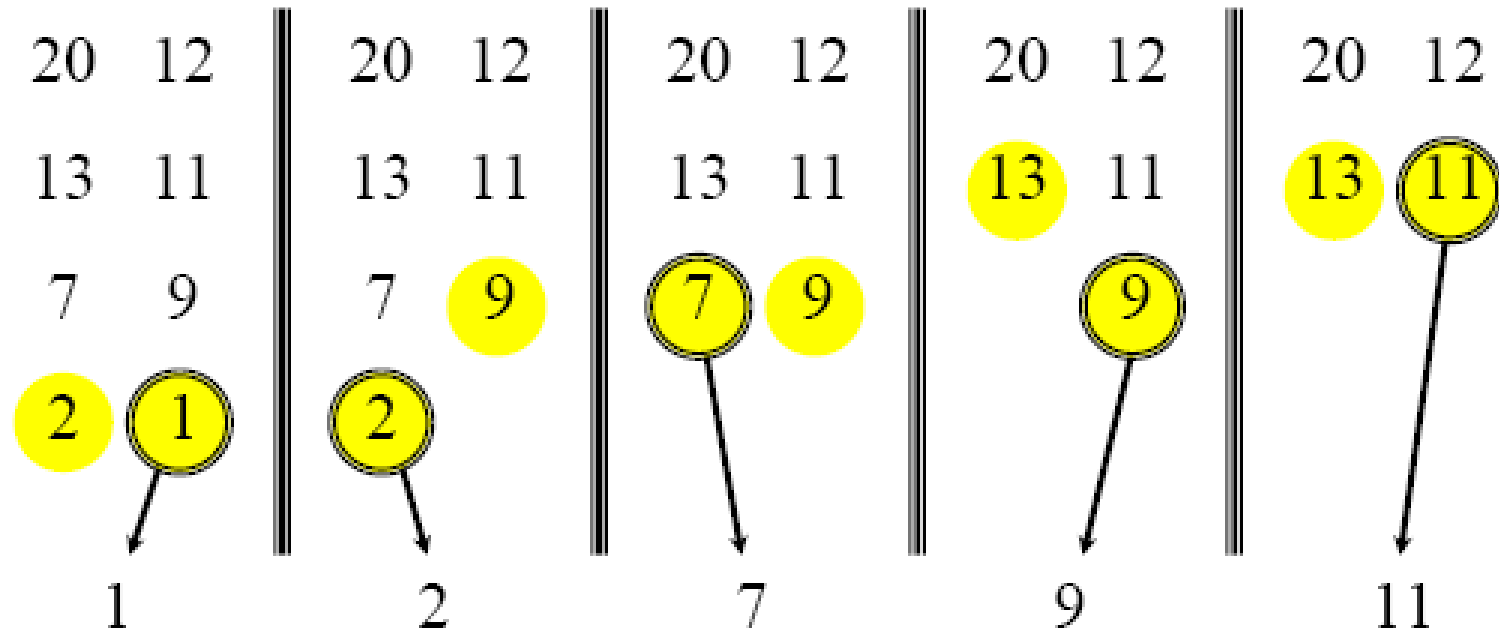
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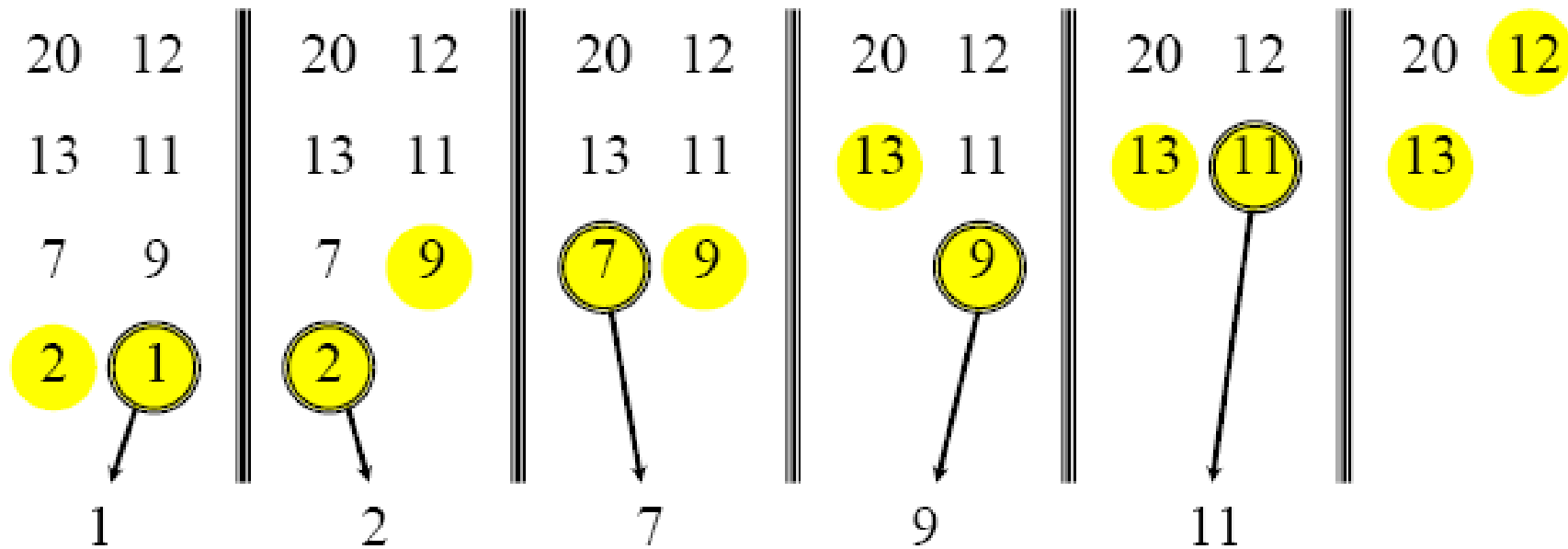
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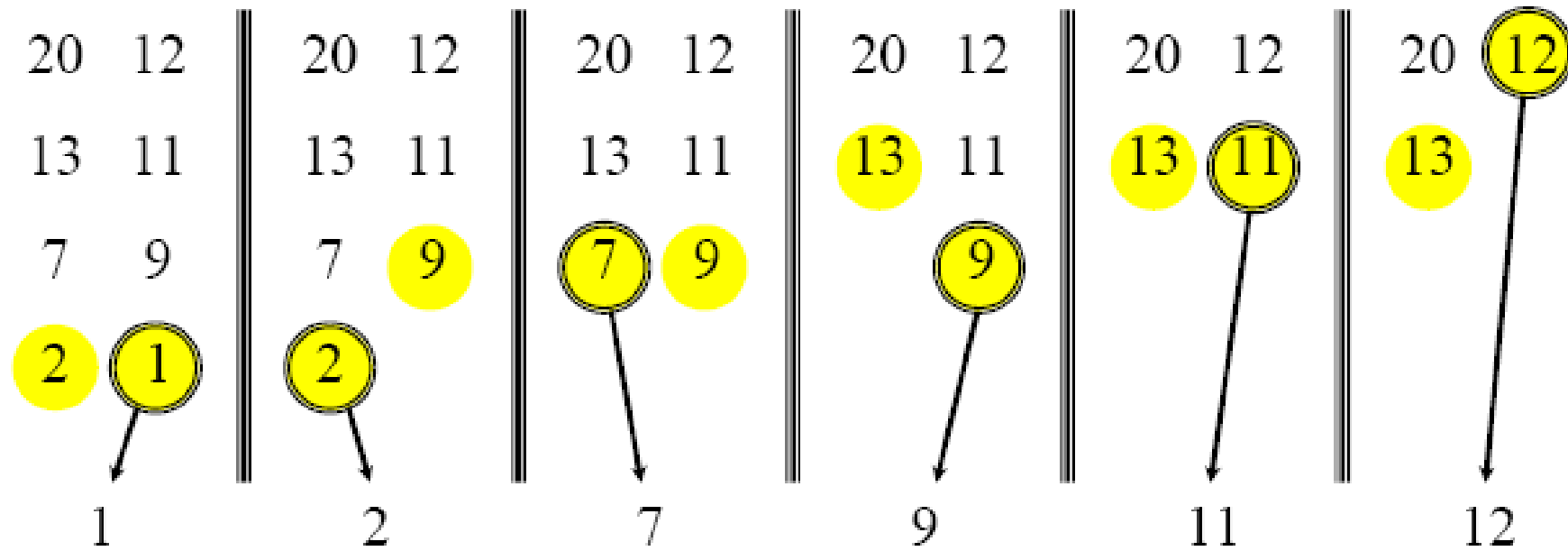
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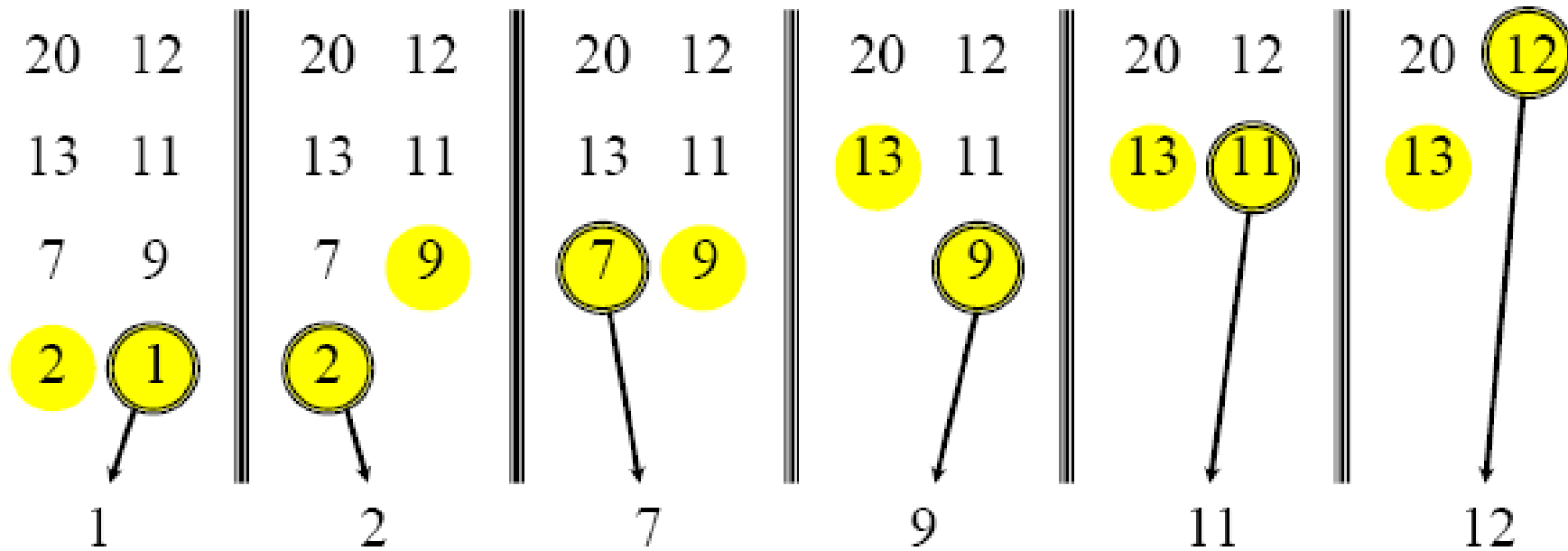
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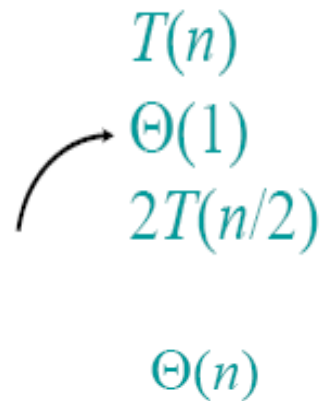


## MERGING TWO SORTED ARRAYS



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

# ANALYZING MERGE SORT



A diagram illustrating the recurrence relation for Merge Sort. It shows the equation  $T(n) = \Theta(1) + 2T(n/2)$  with a curved arrow pointing from the  $2T(n/2)$  term to the  $\Theta(1)$  term. Below this equation is the expression  $\Theta(n)$ .

$$\begin{array}{l} T(n) \\ \Theta(1) \\ 2T(n/2) \\ \Theta(n) \end{array}$$

## 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

## 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.



# MERGE SORT COMPLEXITY ANALYSIS

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

**Iteration Method:**  
**Generate a Guess**

$$T(n) = 2T(n/2) + n \quad \rightarrow \text{Equation 1}$$

Substitute  $n/2$  in place of  $n$  in Eq 1:

$$T(n/2) = 2T(n/4) + n/2 \quad \rightarrow \text{Equation 2}$$

Substitute Eq 2 in Eq 1

$$\begin{aligned} T(n) &= 2(2T(n/4) + n/2) + n \\ &= 2^2 T(n/2^2) + 2n \end{aligned} \quad \rightarrow \text{Equation 3}$$

Substitute  $n/4$  in place of  $n$  in Eq 1

$$\begin{aligned} T(n/4) &= 2T(n/8) + n/4 \quad \rightarrow \text{Equation 4} \\ T(n) &= 2^2 T(n/8) + 2n \end{aligned}$$

# MERGE SORT COMPLEXITY ANALYSIS

$$T(n) = 2^3 T\left(\frac{n}{2^3}\right) + 3n$$

$$T(n) = 2^4 T\left(\frac{n}{2^4}\right) + 4n$$

$\vdots$   
 $\vdots$

$$T(n) = 2^i T\left(\frac{n}{2^i}\right) + i * n$$

$$T(1) = 1$$

$$\text{Let } \frac{n}{2^i} = 1 \rightarrow n = 2^i$$

$$\log_2 n = \log_2 2^i \rightarrow i = \log_2 n$$

$$T(n) = n T(1) + n \log_2 n$$

$$\mathbf{T(n) = n + n \log_2 n}$$

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

**Iteration Method:**

**Generate a Guess**

**But we need to prove it formally:**

**Substitution method!**

# RECURSION TREE

**Recursion Tree Method:**  
Another method to  
Generate a Good Guess  
for solving a recurrence.

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

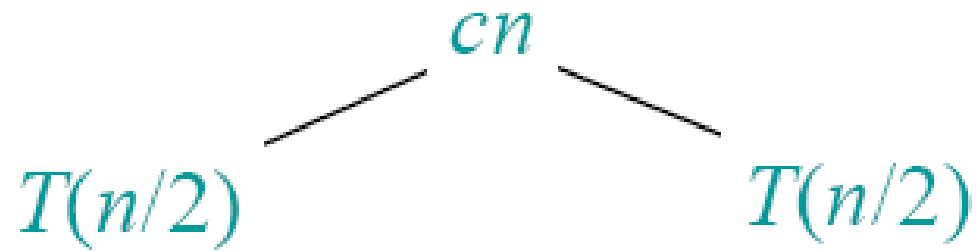
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$$T(n)$$

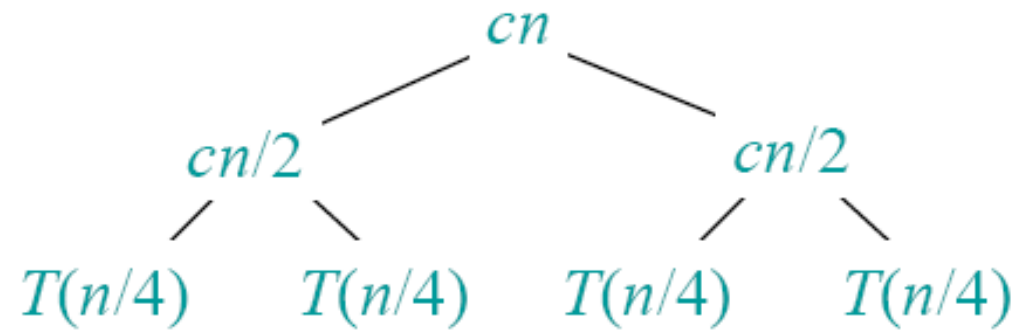
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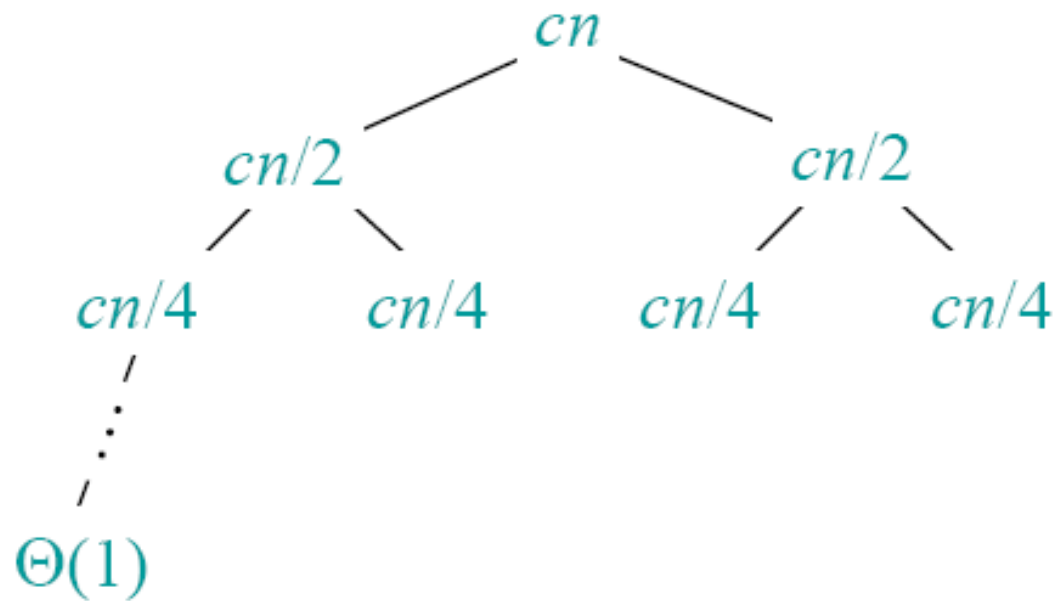
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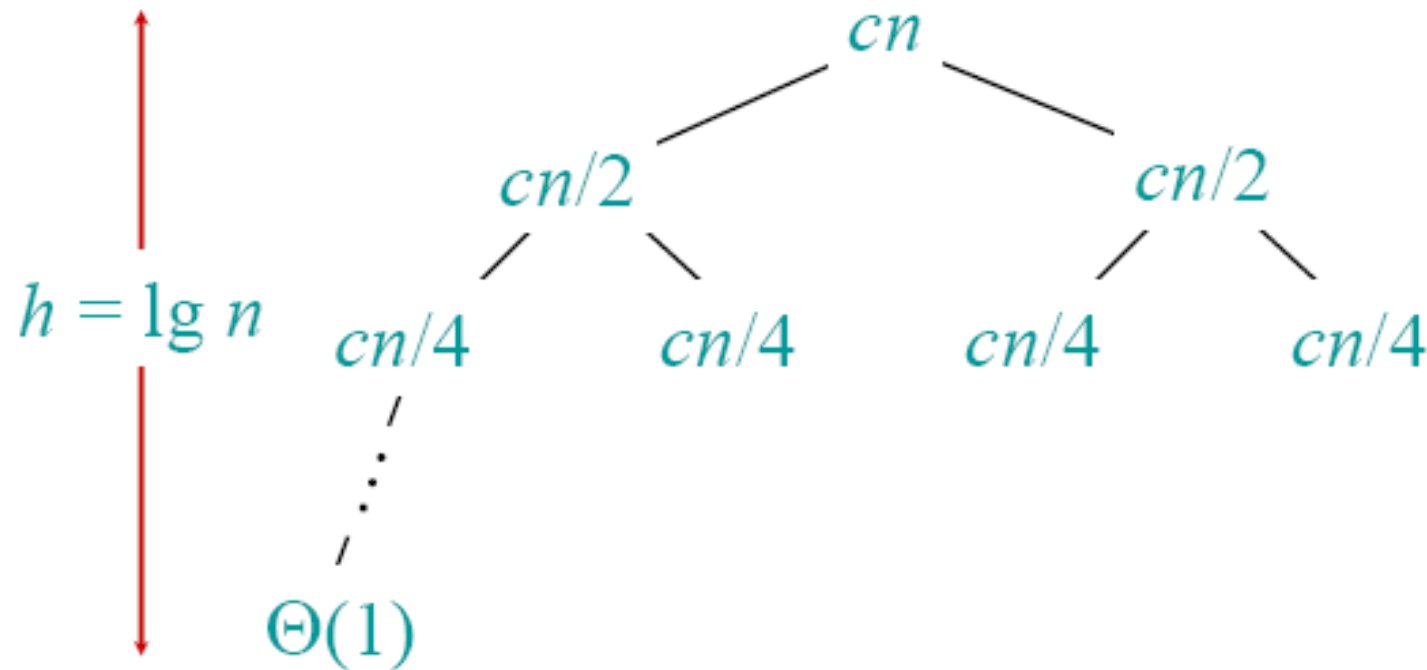
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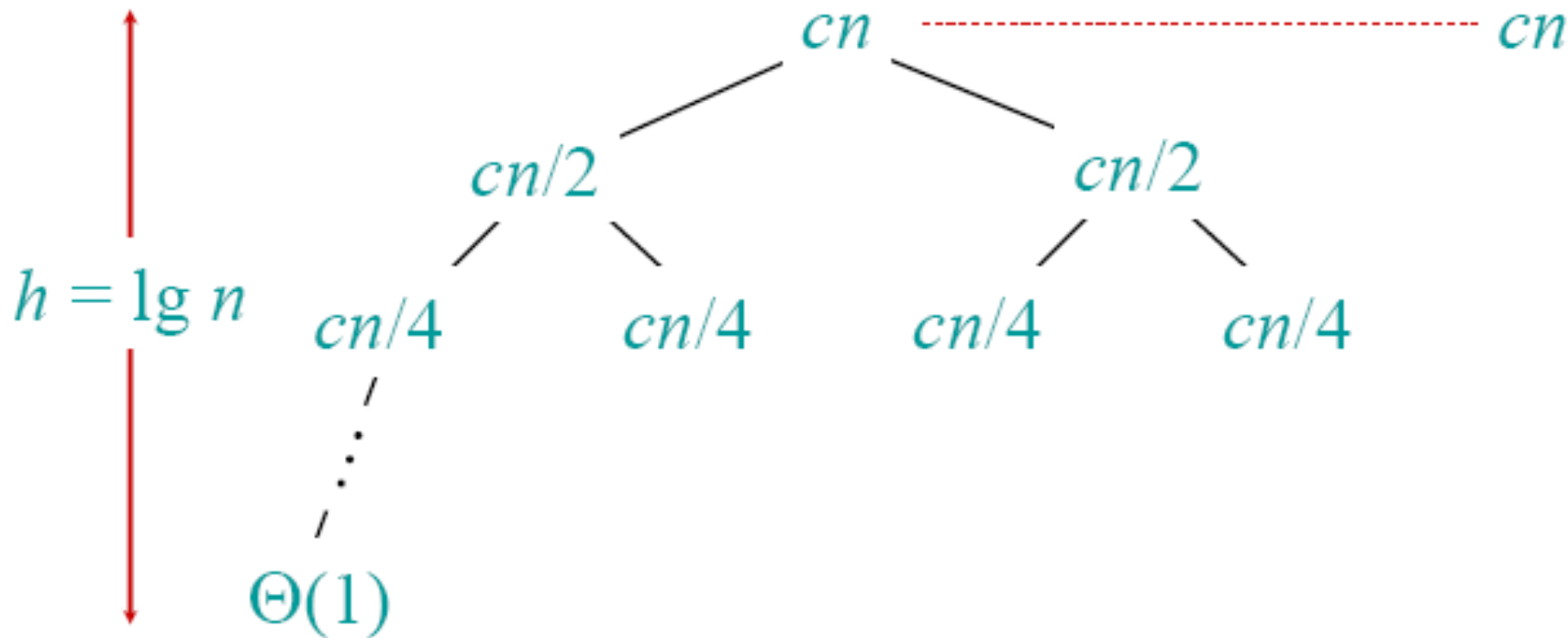
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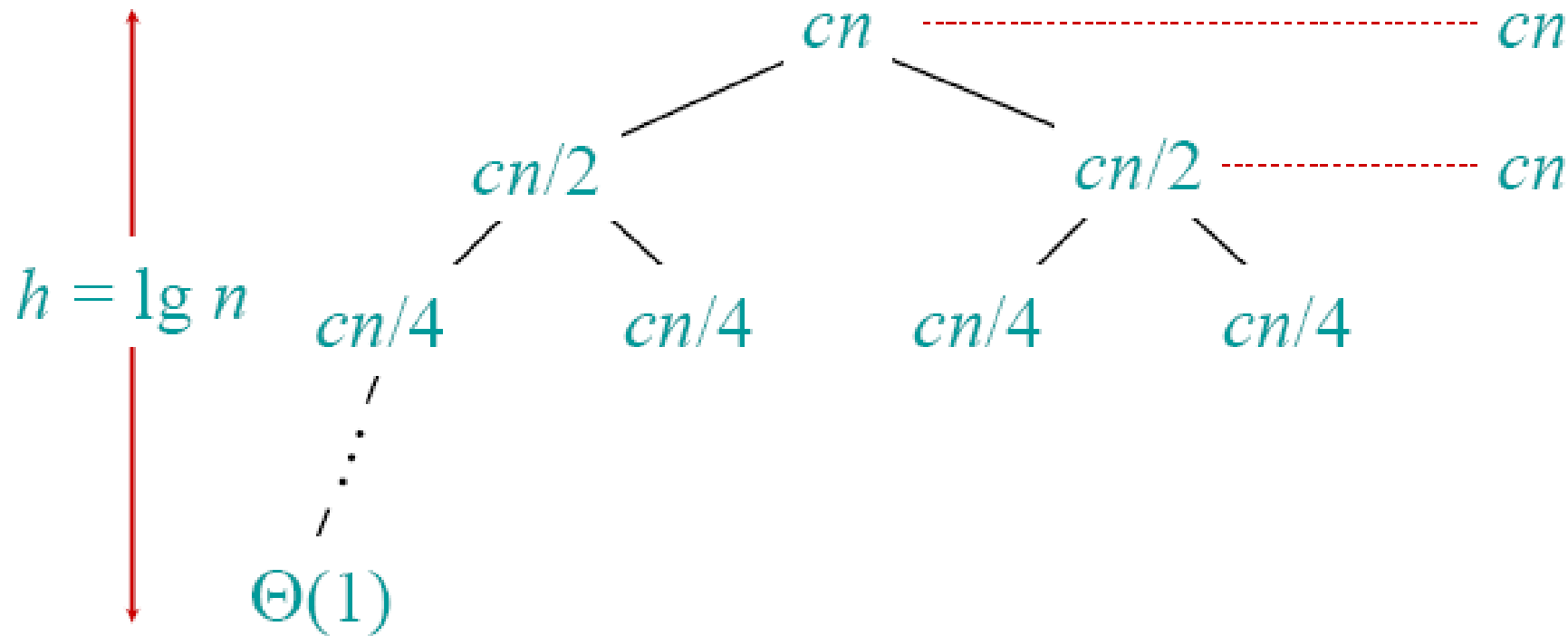
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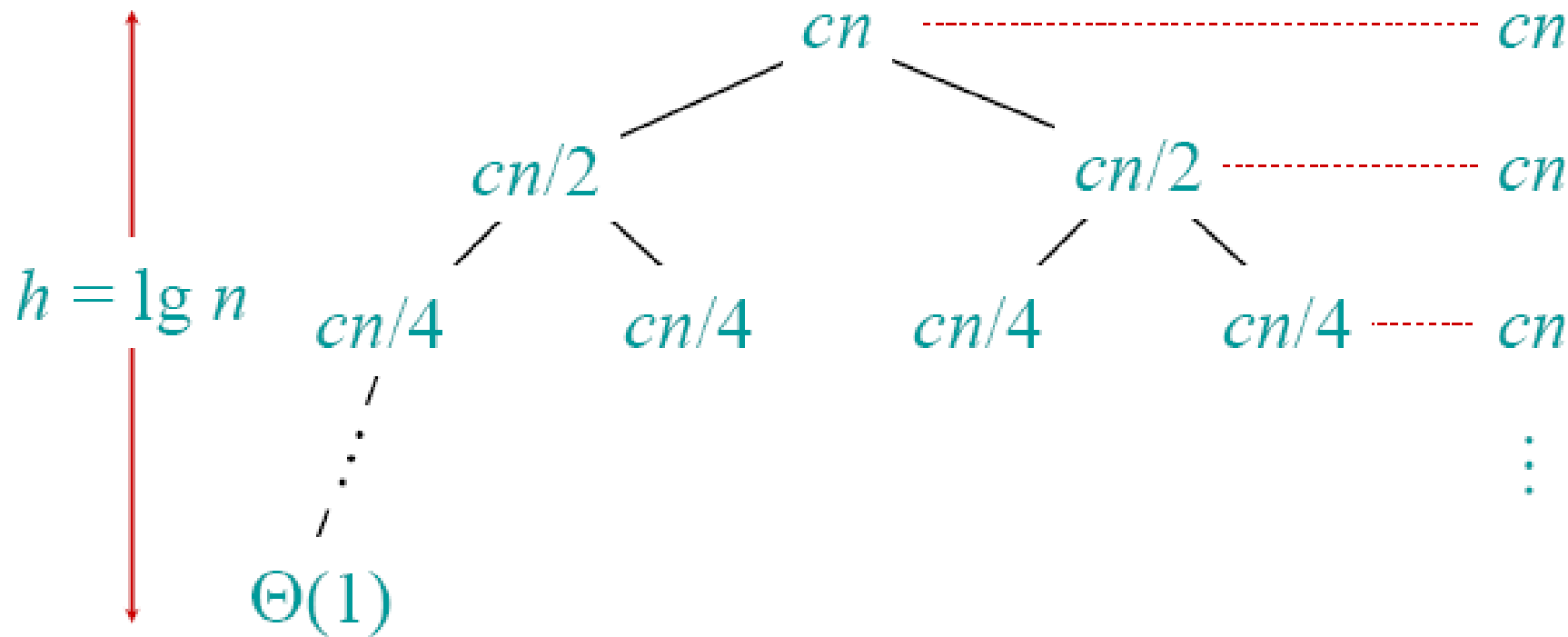
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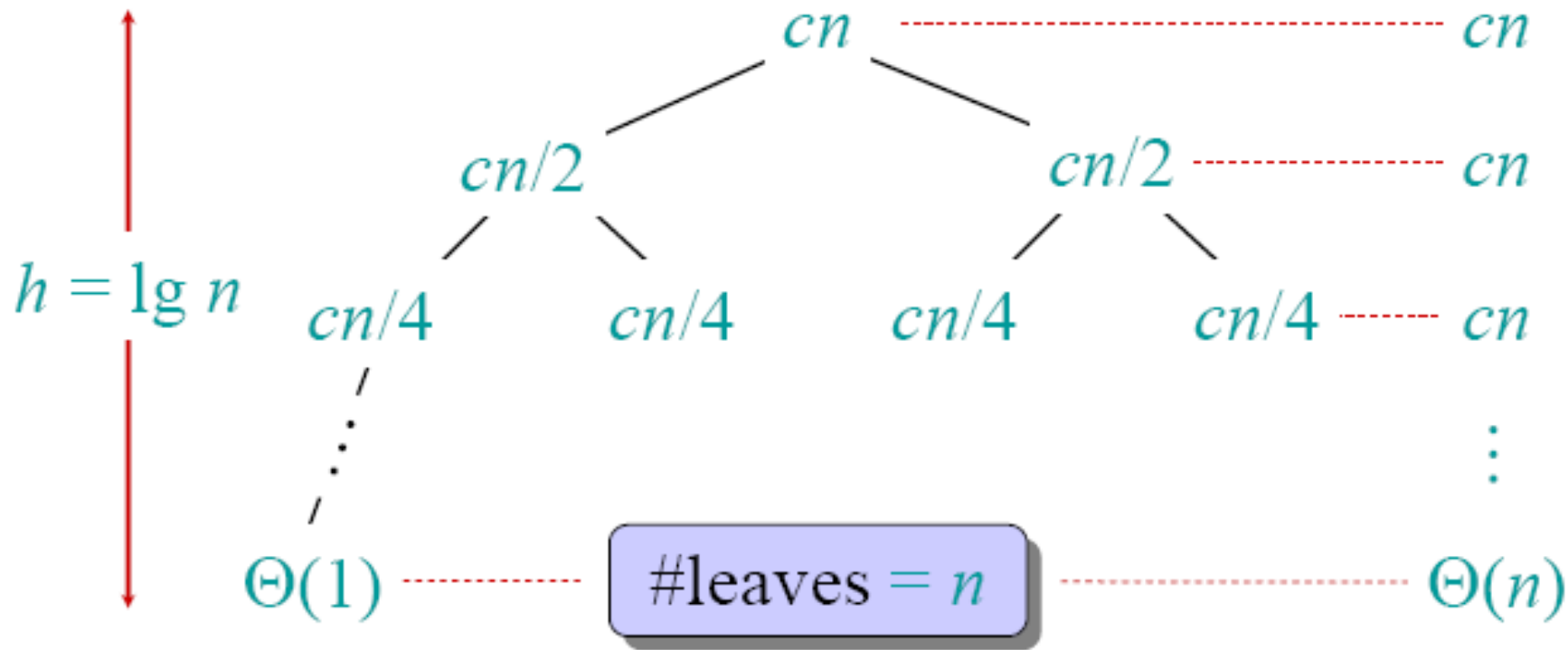
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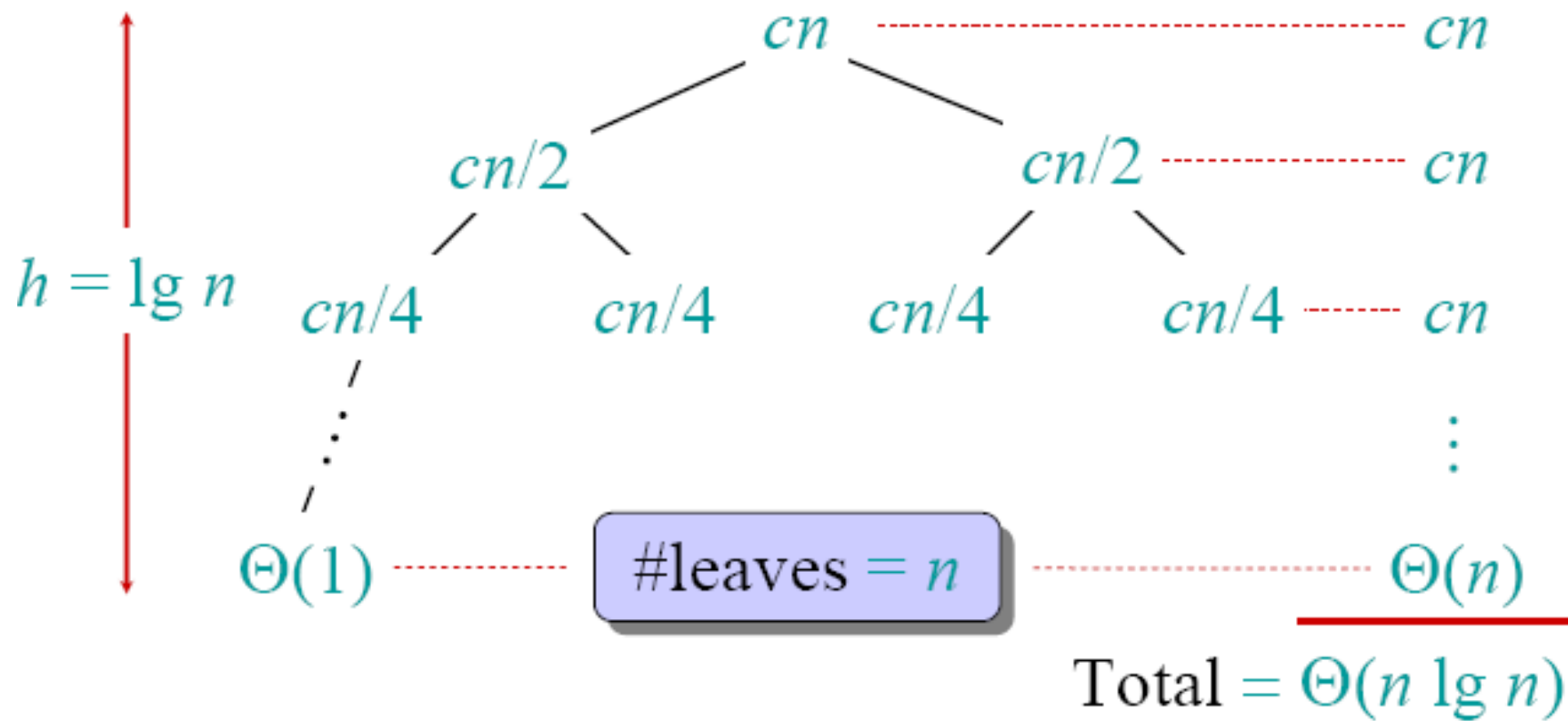
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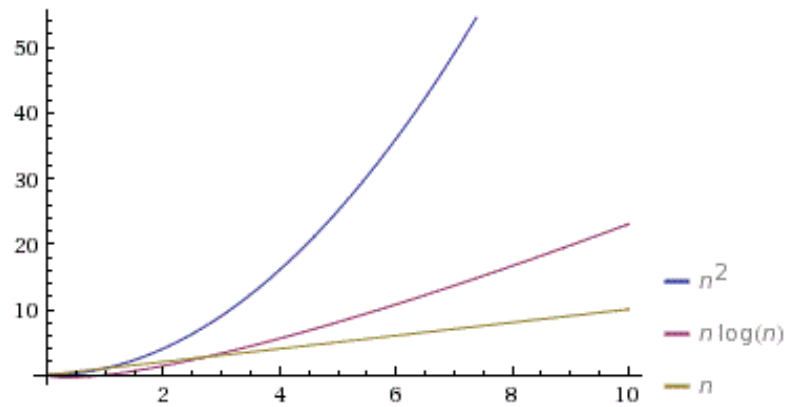
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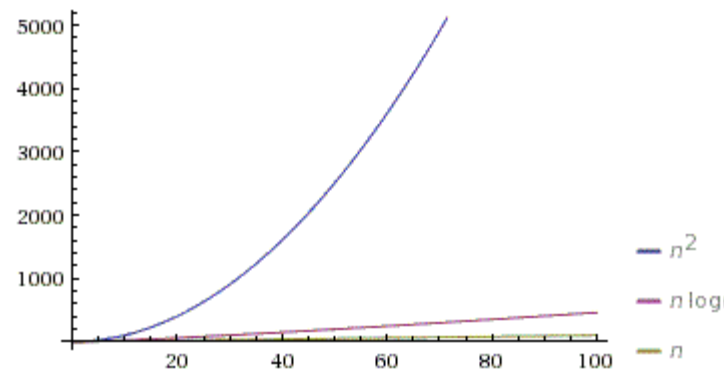
# RECURSION TREE

Plot:



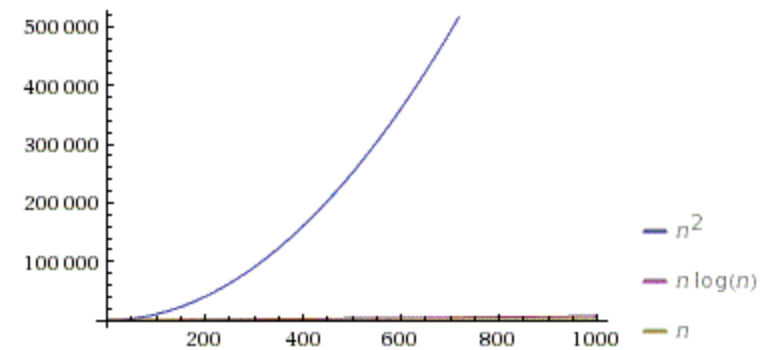
$n = [0; 10]$

Plot:



$n = [0; 100]$

Plot:



$n = [0; 1000]$

## 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!

