

Design and Analysis of Algorithms

CSE 5311

Lecture 1 Administration & Introduction

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Department of Computer Science and Engineering

Administration

- **Course CSE 5311**

- What: Design and Analysis of Algorithms
- When: M/W 1:00pm – 2:20pm
- Where: WH 311
- Who: Song Jiang (song.jiang@uta.edu)
- Office Hour: Mon. & Wed. 10:30 ~ 11:30am at SIER319
- or by appointments
- TA: Mr. Zongyao Lyu, ERB413, office hour is Tu/Th 3:00pm-4:00pm
- Homepage: <http://ranger.uta.edu/~sjiang/CSE5311-fall-19/index.htm>
(Please visit this website regularly)

- **About your instructor**

- Research areas: file and storage system, operating system, parallel and distributed computing, and high performance computing,

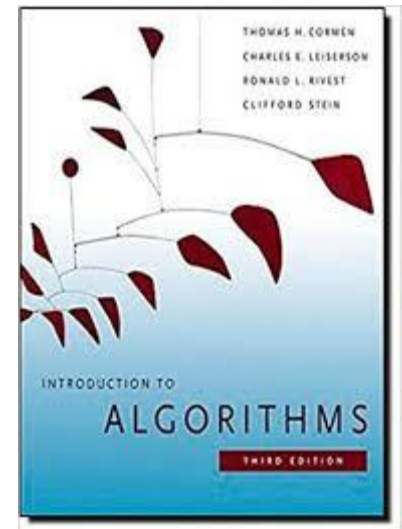
Study Materials

- **Prerequisites**

- CSE 2320 Algorithms and Data Structures or its equivalents
- Programming skills on a high-level language, such as C and Java.
- Mathematical background on summations, sets, relations, probability, and matrix computation.

- **Textbook**

Thomas Cormen, Charles Leiserson, Ronald Rivest, and Clifford Stein. [Introduction to Algorithms](#). 3rd ed. MIT Press, 2009.



Grading

- **Distribution**

- 5% Class attendance
- 30% Homework Assignments
- 20% Quizzes
- 20% Midterm Exam
- 25% Final Exam

100%

- Bonus credits may be offered for voluntarily and correctly answering in-class questions.

- **Attention**

- Homework is as important as any other aspects of your grade.
- Attendance is strongly encouraged.
- The university makeup policy will be strictly adhered to. Generally, no make-up exams/quizzes except for university sanctioned reasons.
- When missing an exam/quiz due to unavoidable circumstances, PLEASE notify the instructor and **request a makeup approval ahead of time.**

Grading

- **Late Assignments**

Late assignments will be accepted with a 20% penalty applied for each day late up to 2 days. Assignments submitted later than 2 days after the original due time will not be accepted.

- **Collaboration Policy**

Students are allowed and encouraged to collaborate on homework assignments. However, **You must write up each problem solution by yourself without assistance**, even if you collaborate with others to solve the problem. If you obtain a solution through research (e.g., on the Web), acknowledge your source, and write up the solution in your own words. **It is a violation of this policy to submit a problem solution that you cannot orally explain to the instructor or GTA.**

Final Grade

- **Final Letter Grade**

- [90 100] --- A
- [80 90) --- B
- [70 80) --- C
- [60 70) --- D
- [00 60) --- F

- **Note**

- [] denotes inclusion and () denotes exclusion.
- Your final weighted scores may be curved before assignment of your letter grade.

What's the Course About?

- **The theoretical study of analysis and design of computer algorithms**
 - **Analysis:** predict the cost of an algorithm in terms of resources and performance
 - **Design:** design algorithms which minimize the cost
- **Basic goals for an algorithm**
 - Always correct
 - Always terminates
- **Our class: performance**

Algorithms

- An **algorithm** is any well-defined computational procedure that takes some value, or set of values, as **input** and produces some value, or set of values, as **output**.

- An example problem: sorting

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

Output: A permutation (reordering) $\langle a'_1, a'_2, \dots, a'_n \rangle$ of the input sequence such that $a'_1 \leq a'_2 \leq \dots \leq a'_n$.

An instance of a problem: $\langle 31; 41; 59; 26; 41; 58 \rangle$

- Design algorithms for a problem:
 - Find a longest common subsequence of
 $X = \langle x_1, x_2, \dots, x_m \rangle$ and $Y = \langle y_1, y_2, \dots, y_n \rangle$

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!

Machine Model

- **Generic Random Access Machine (RAM)**
 - Executes operations sequentially
 - Set of primitive operations: Arithmetic, Logical, Comparisons, Function calls
- **Simplifying assumption**
 - All operations cost one unit
 - Eliminates dependence on the speed of our computer
 - Otherwise impossible to verify and to compare



Analysis of algorithms

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



The problem of sorting

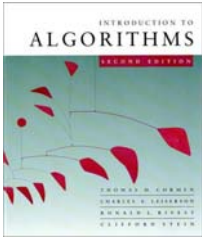
Input: sequence $\langle a_1, a_2, \dots, a_n \rangle$ of numbers.

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

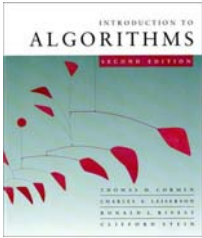
Output: 2 3 4 6 8 9



Insertion sort

“pseudocode”

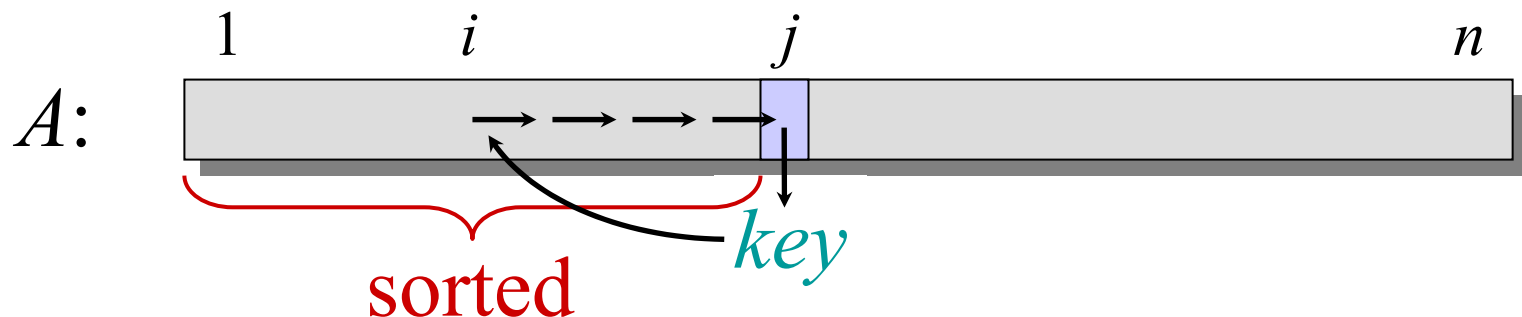
```
INSERTION-SORT ( $A, n$ )     $\triangleright A[1 \dots n]$ 
  for  $j \leftarrow 2$  to  $n$ 
    do  $key \leftarrow A[j]$ 
       $i \leftarrow j - 1$ 
      while  $i > 0$  and  $A[i] > key$ 
        do  $A[i+1] \leftarrow A[i]$ 
           $i \leftarrow i - 1$ 
       $A[i+1] = key$ 
```



Insertion sort

“pseudocode”

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Example of insertion sort

8 2 4 9 3 6



Example of insertion sort



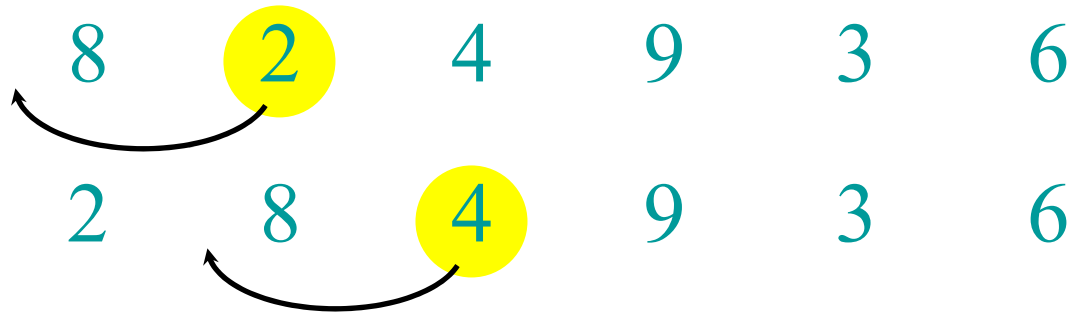


Example of insertion sort



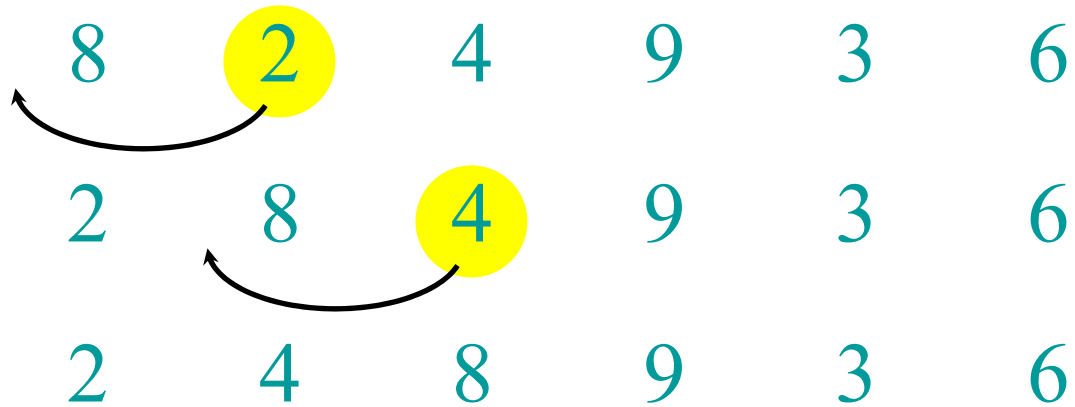


Example of insertion sort



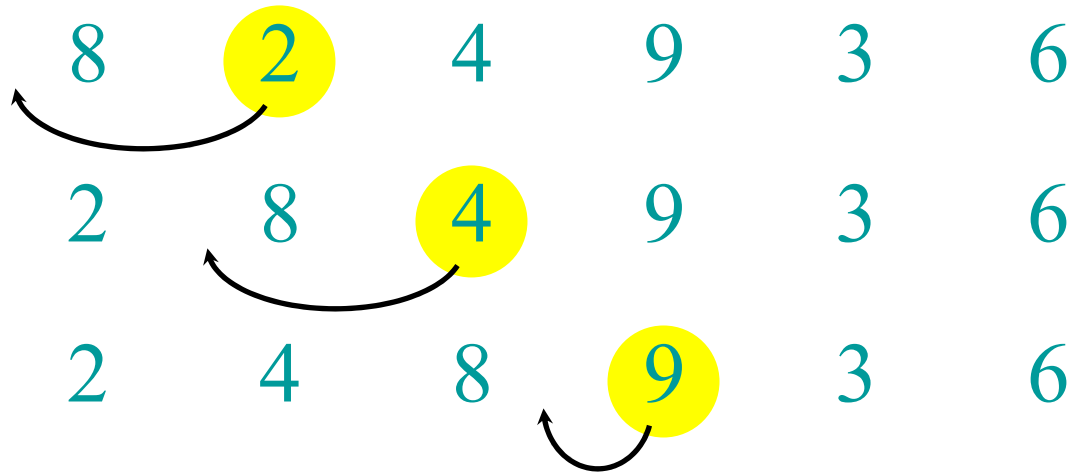


Example of insertion sort



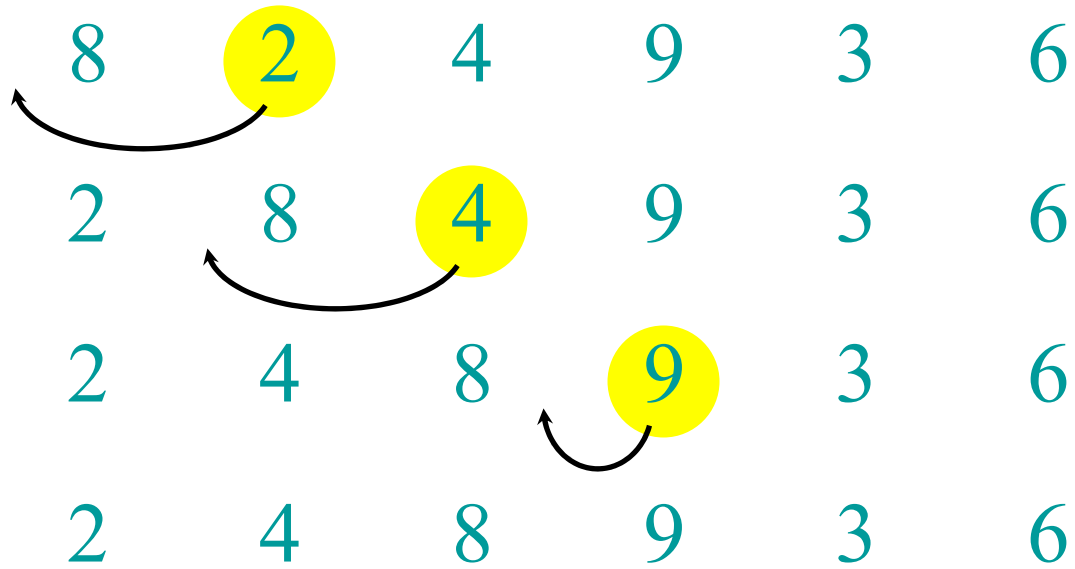


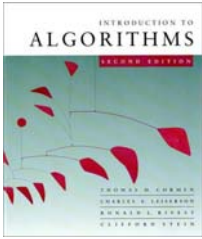
Example of insertion sort



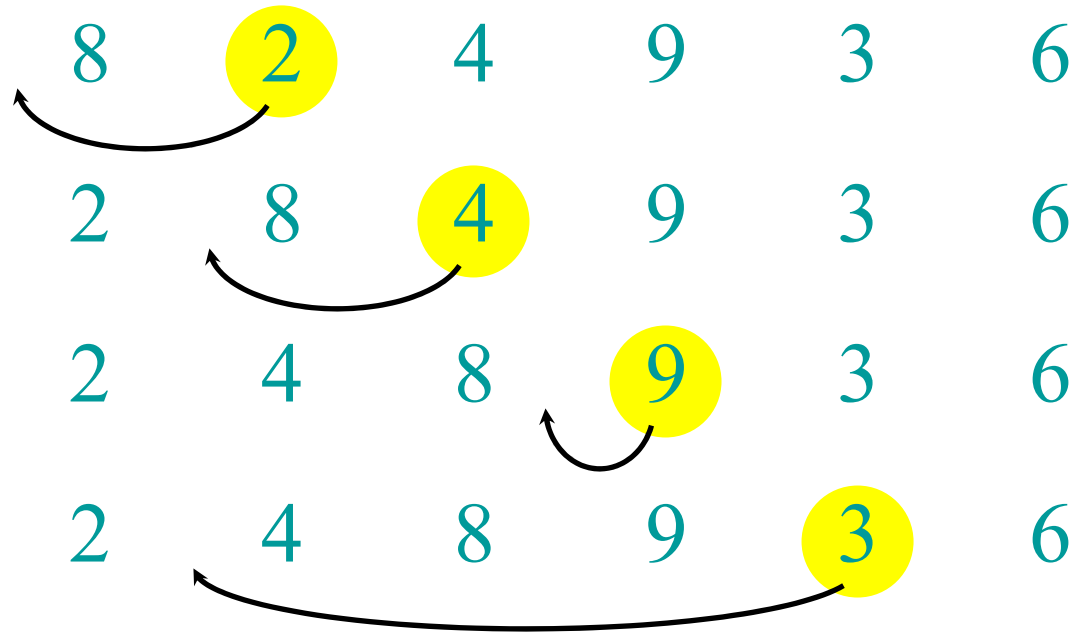


Example of insertion sort



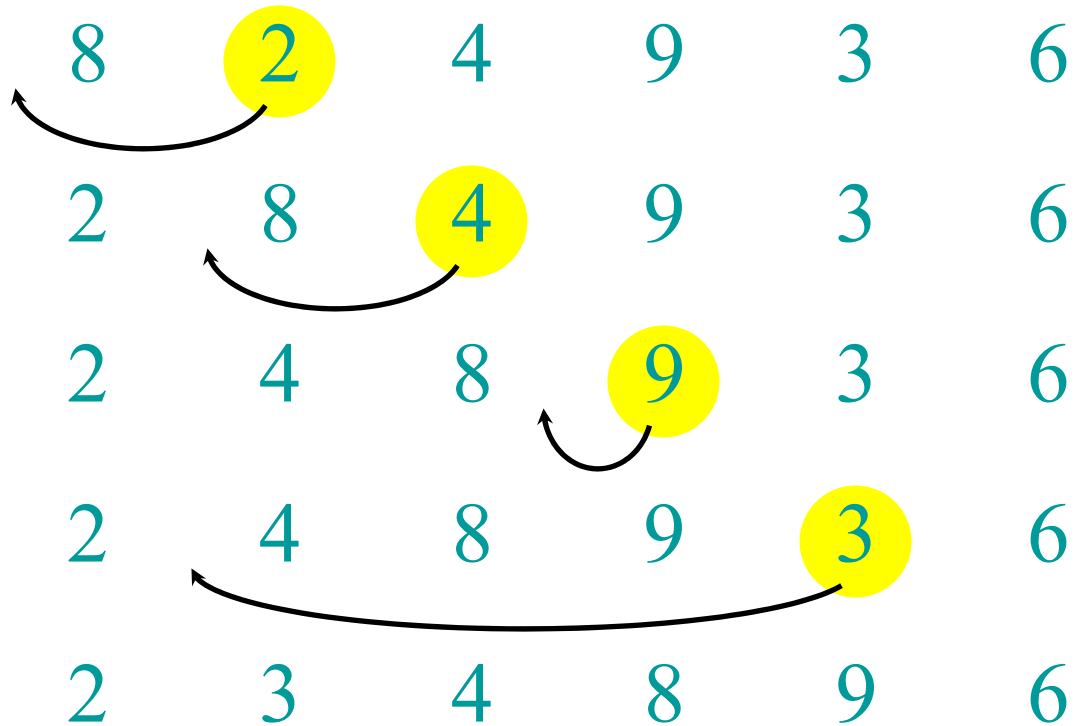


Example of insertion sort



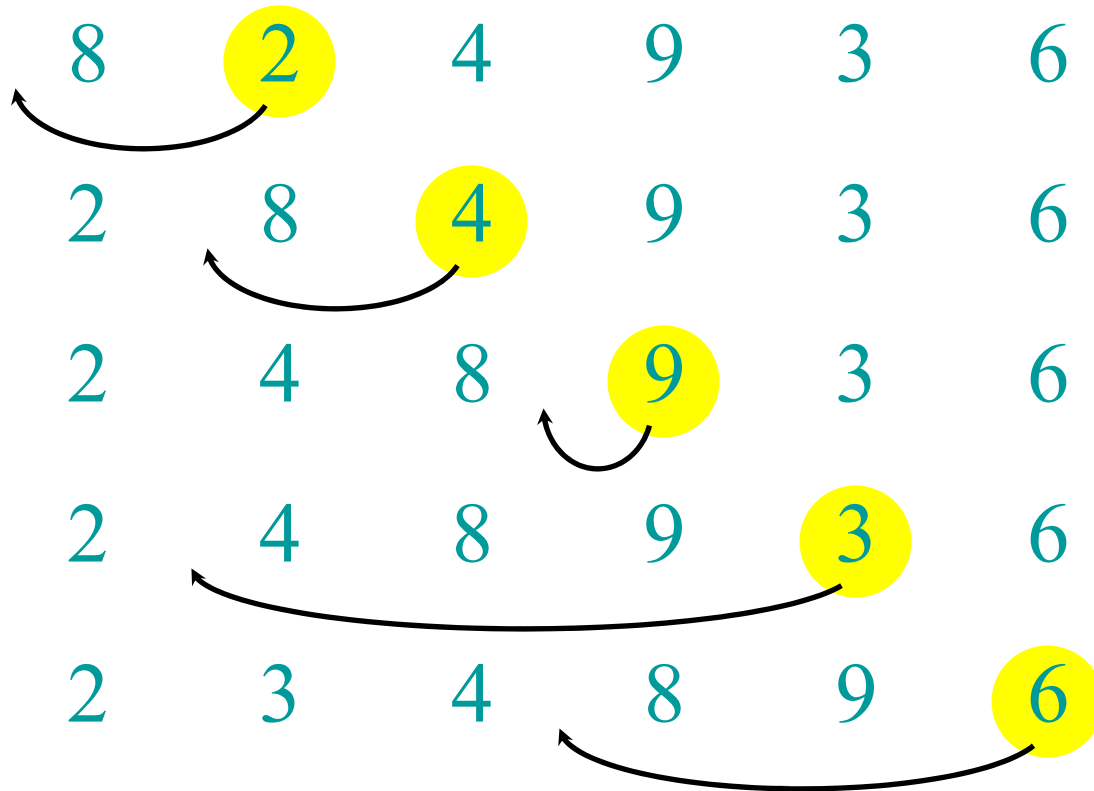


Example of insertion sort



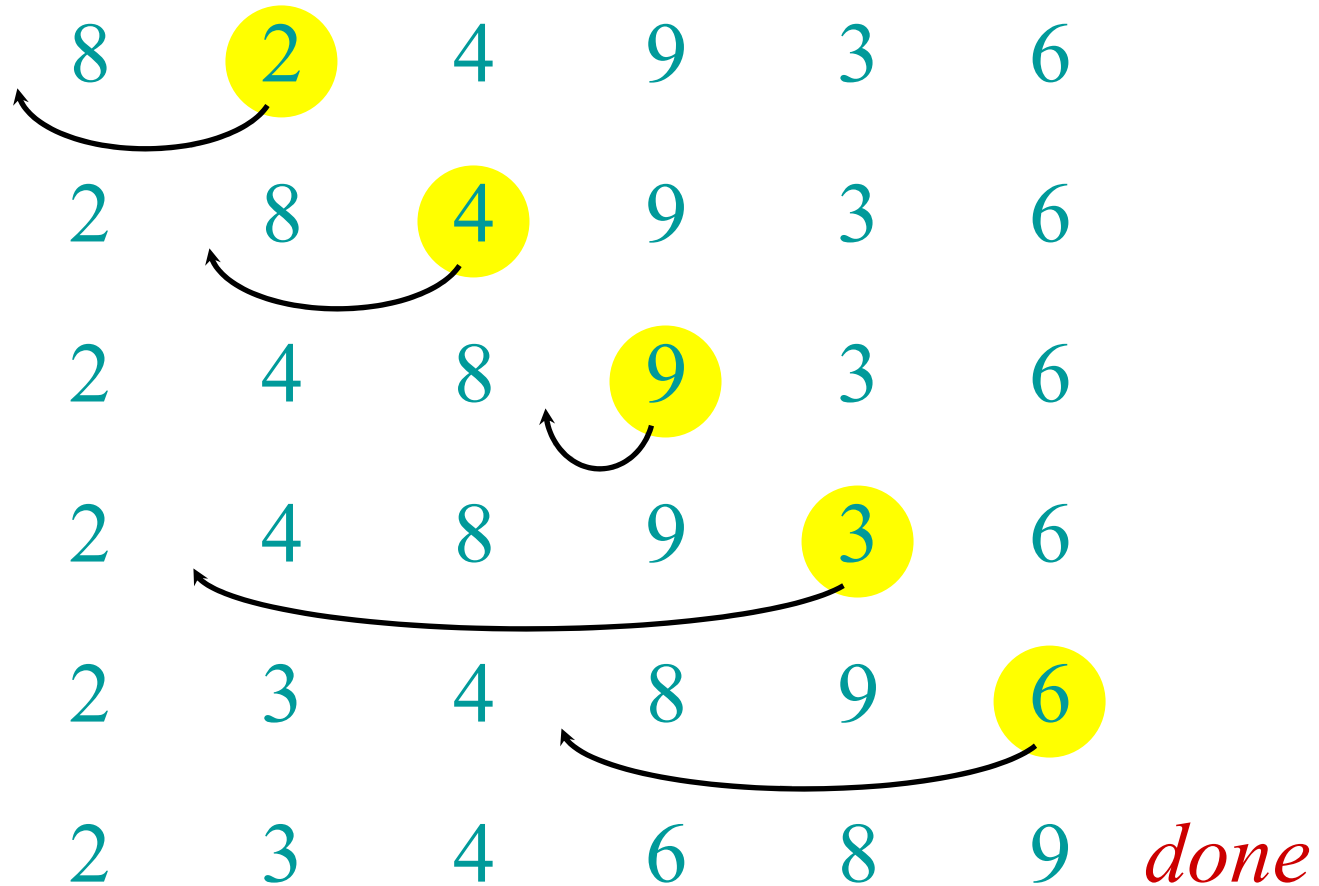


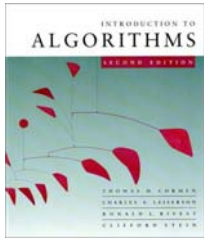
Example of insertion sort





Example of insertion sort





Running time

- 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 analyses

Worst-case: (usually)

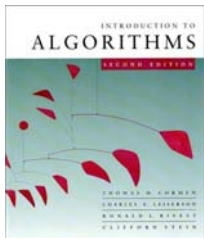
- $T(n)$ = maximum time of algorithm on any input of size n .

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.



Machine-independent time

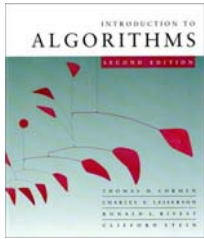
What is insertion sort's worst-case time?

- 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.
- Look at *growth* of $T(n)$ as $n \rightarrow \infty$.

“Asymptotic Analysis”



Θ -notation

Math:

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

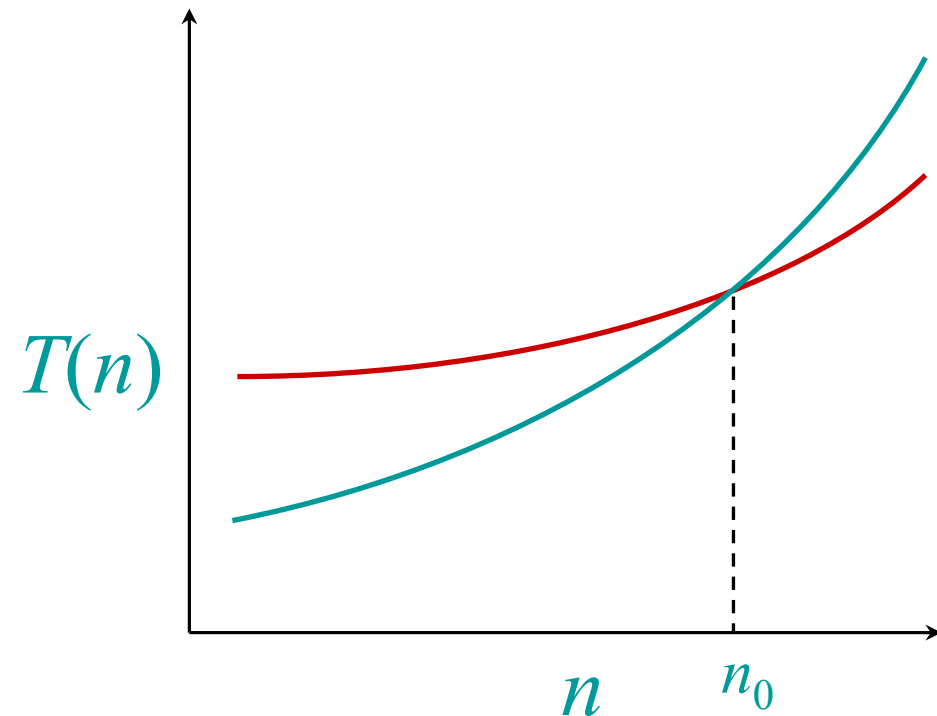
Engineering:

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

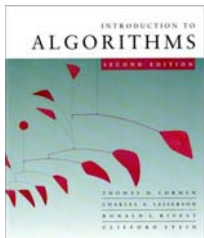


Asymptotic performance

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.



Insertion sort analysis

Worst case: Input reverse sorted.

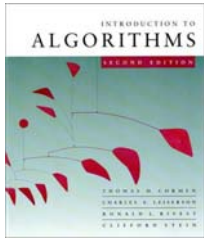
$$T(n) = \sum_{j=2}^n \Theta(j) = \Theta(n^2) \quad [\text{arithmetic series}]$$

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

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

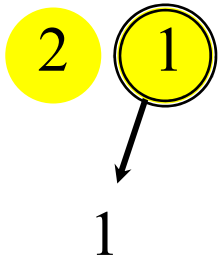


Merging two sorted arrays

20 12

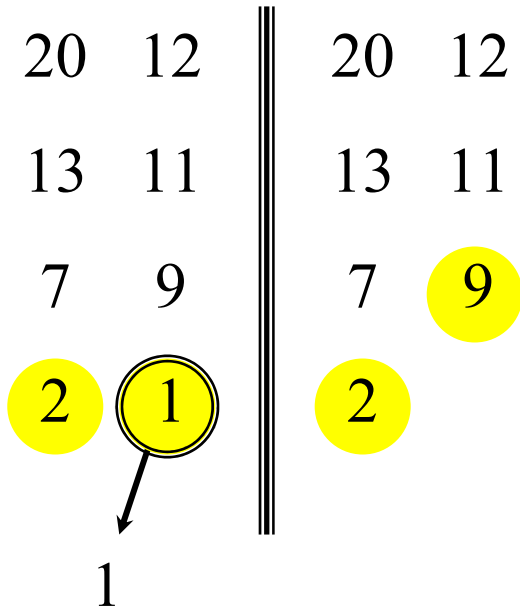
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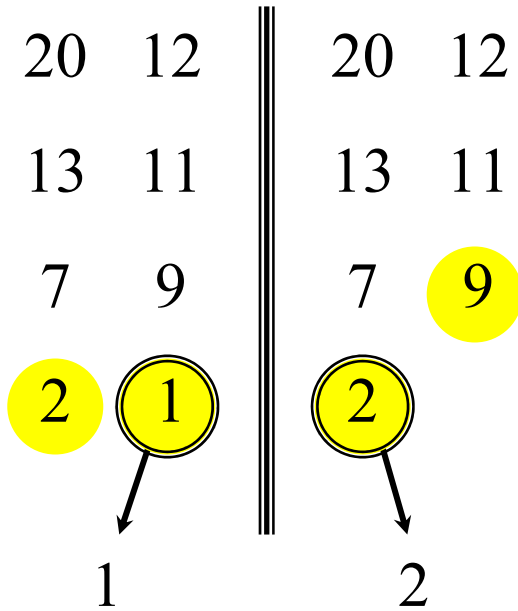


Merging two sorted arrays



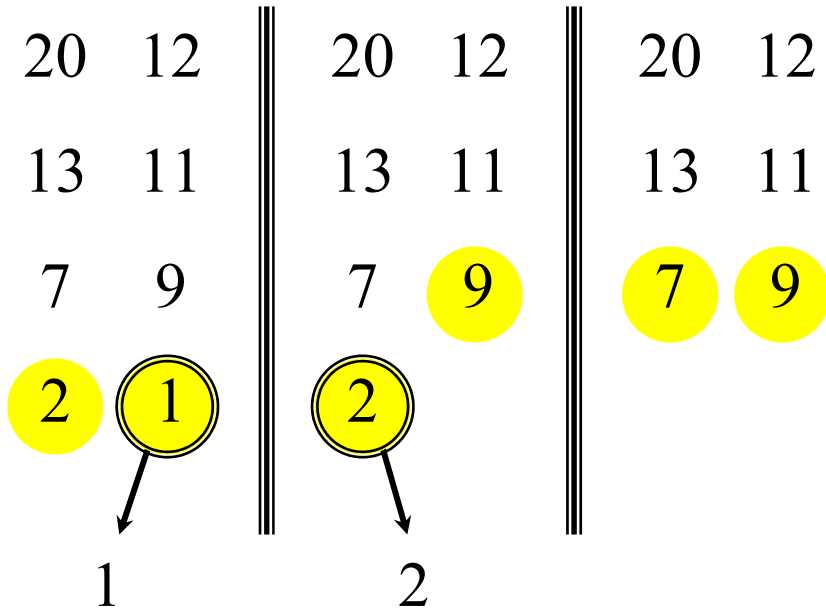


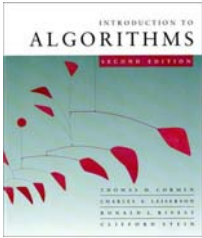
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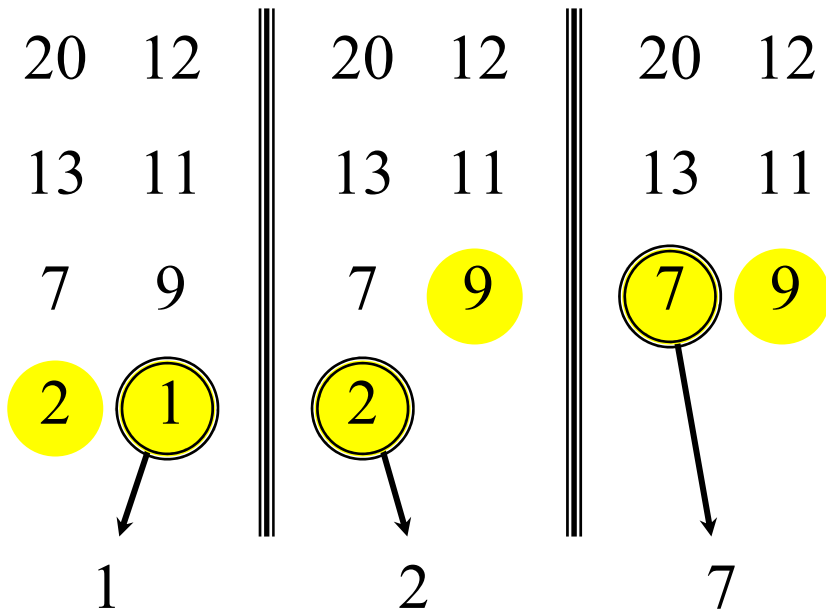


Merging two sorted arrays



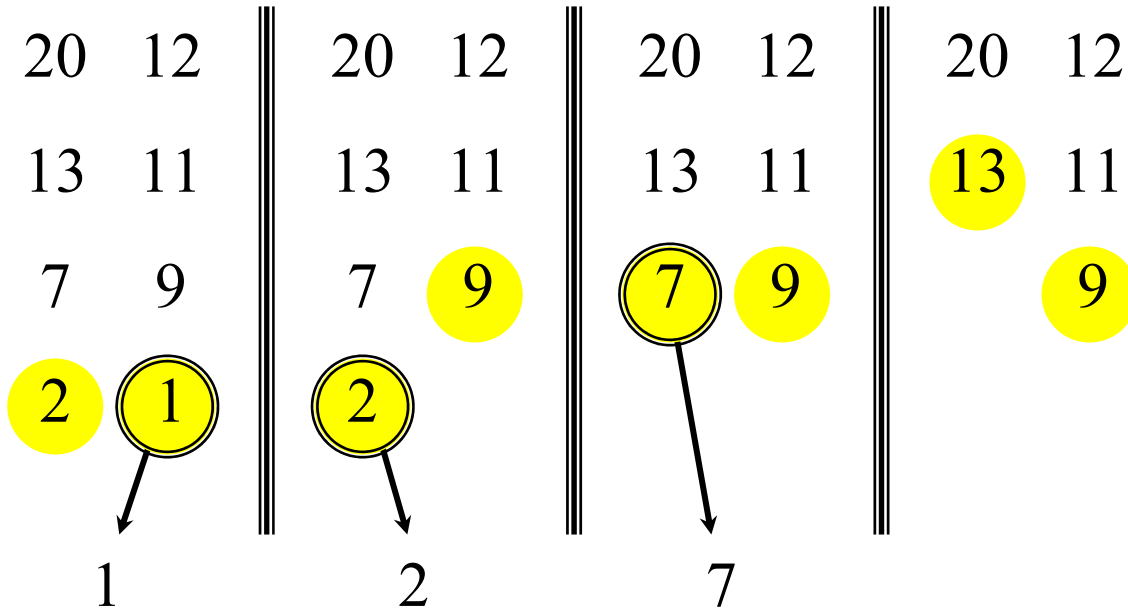


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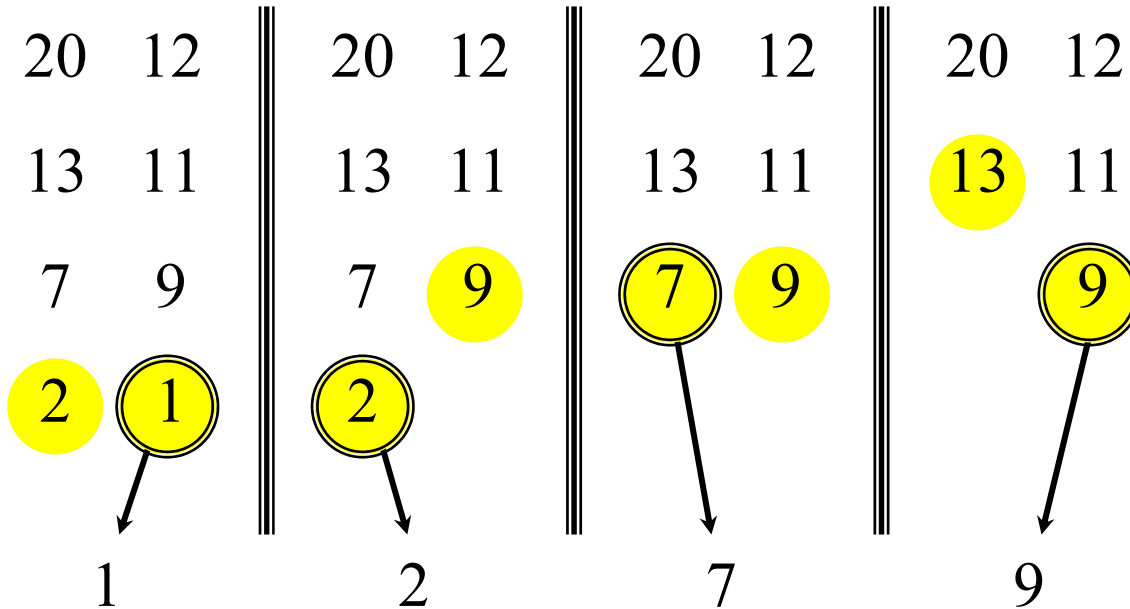


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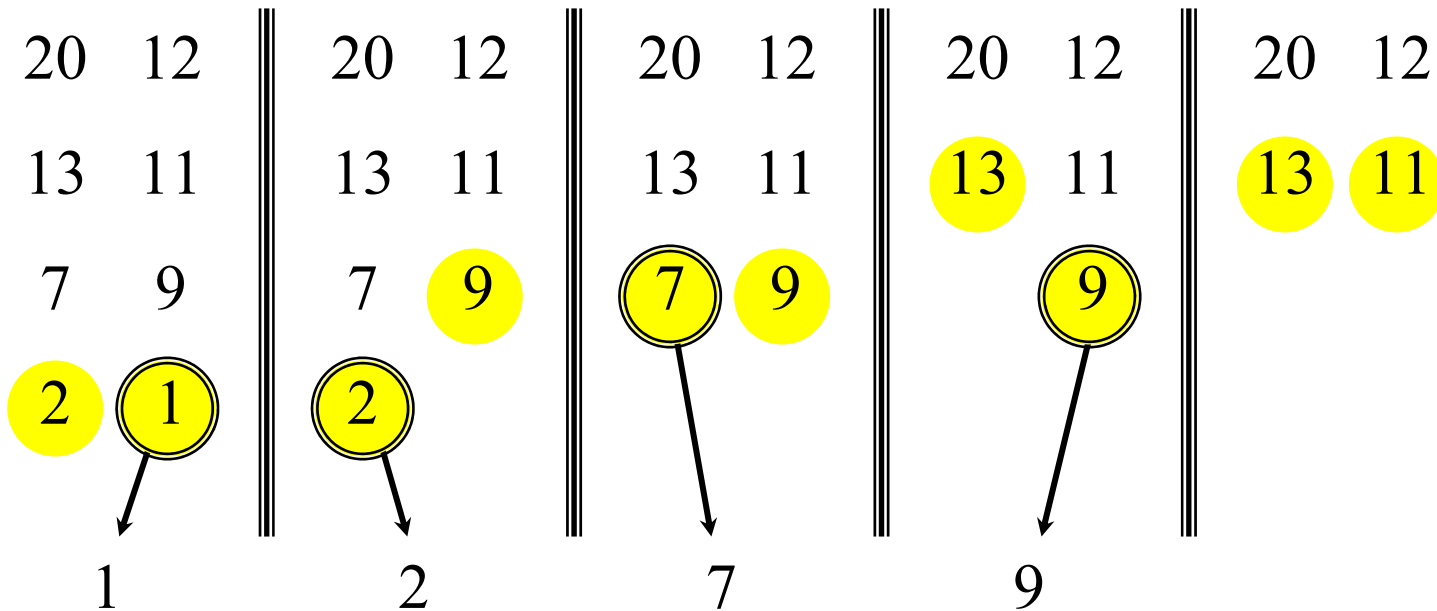


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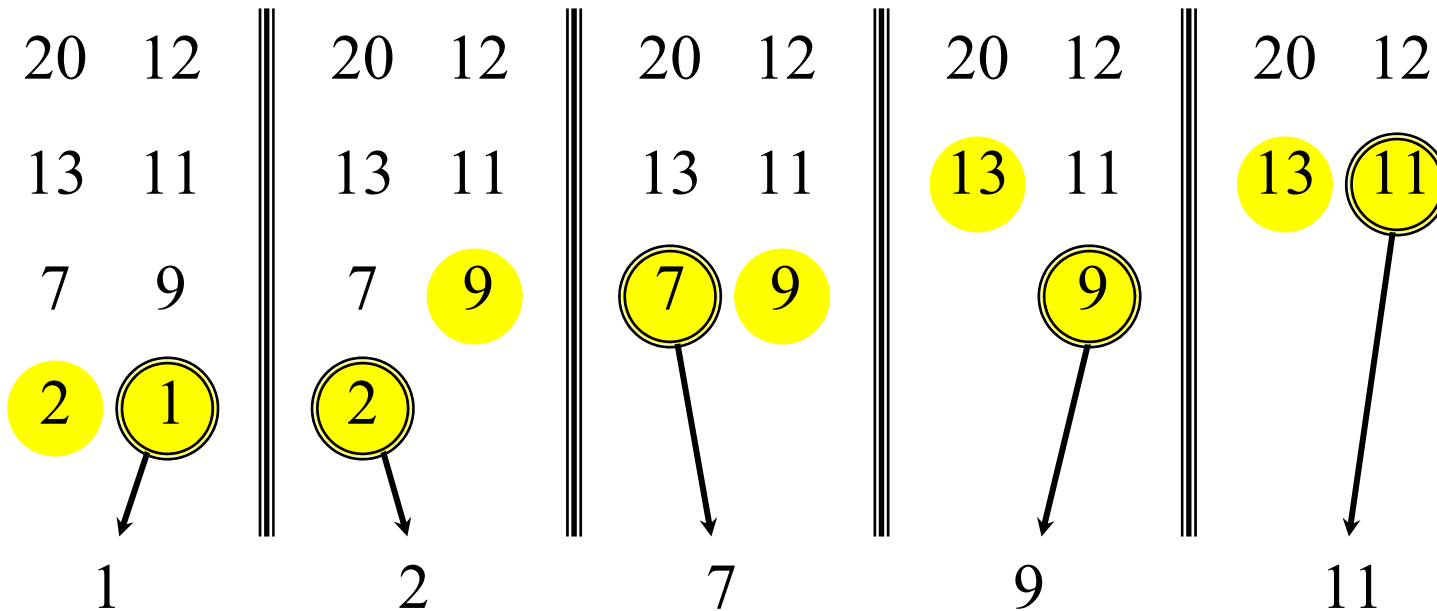


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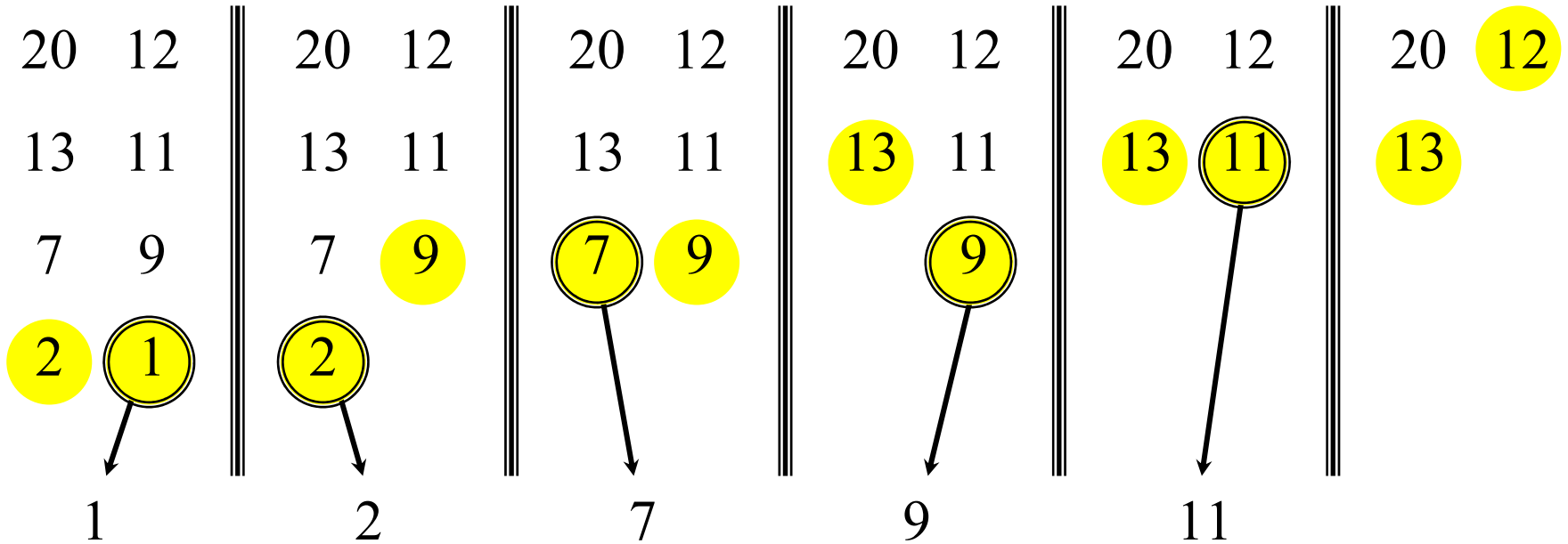


Merging two sorted arrays



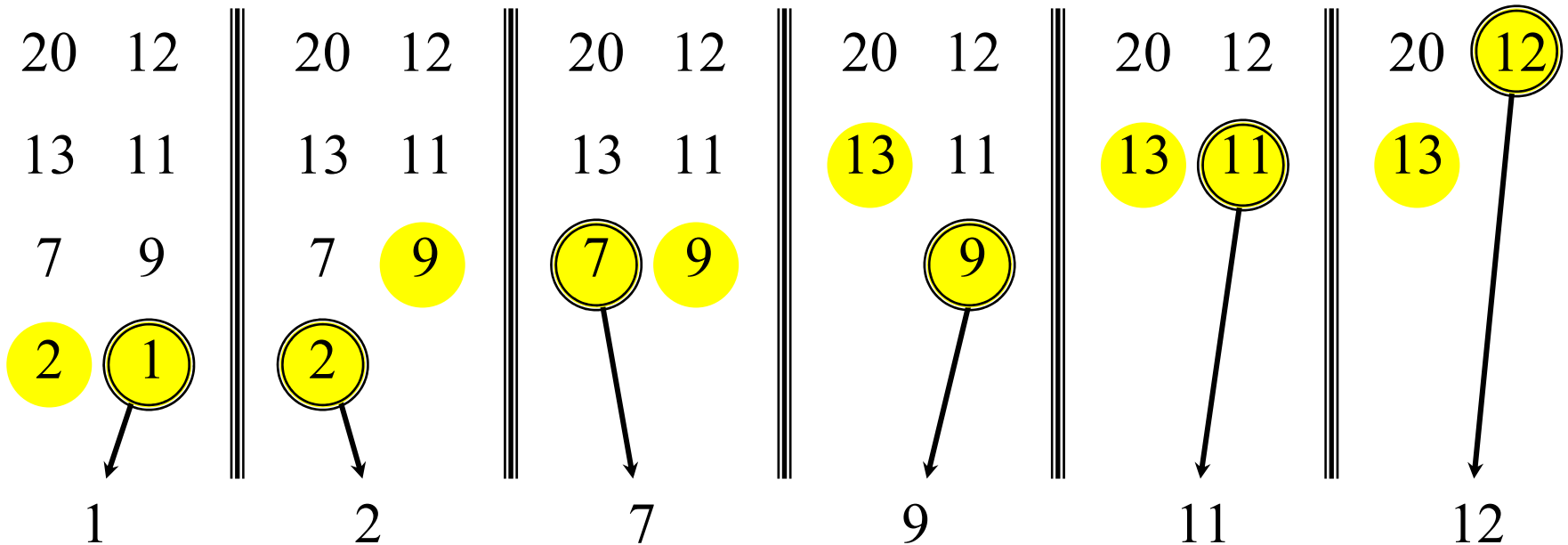


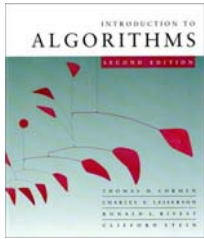
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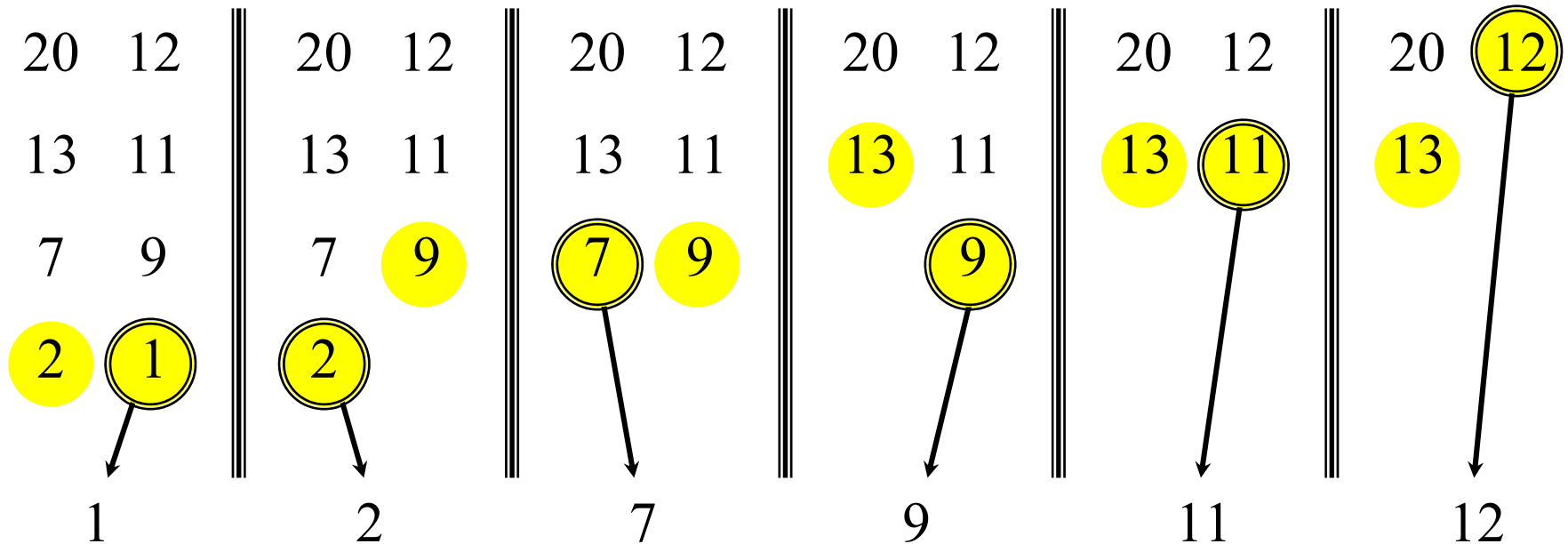


Merging two sorted arrays

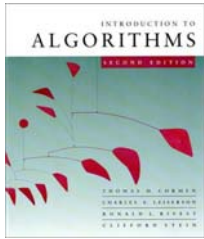




Merging two sorted arrays



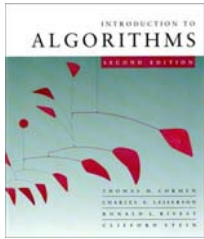
Time = $\Theta(n)$ to merge a total of n elements (linear time).



Analyzing merge sort

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

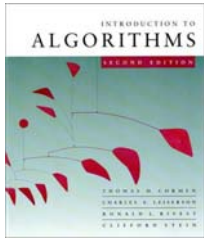
Sloppiness: Should be $T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor)$,
but it turns out not to matter asymptotically.



Recurrence for merge sort

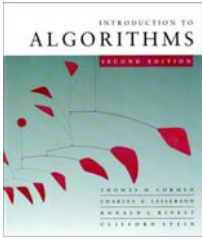
$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

- 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

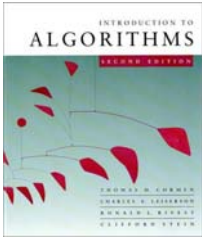
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.



Recursion tree

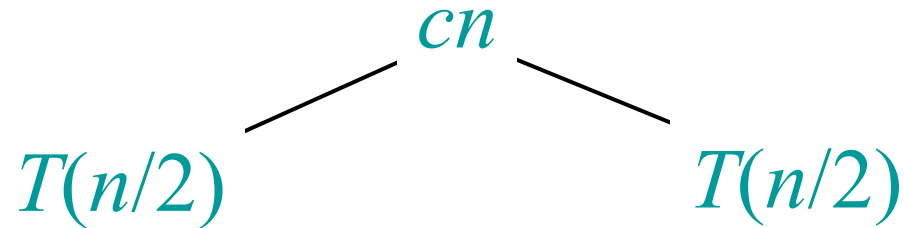
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.

$$T(n)$$



Recursion tree

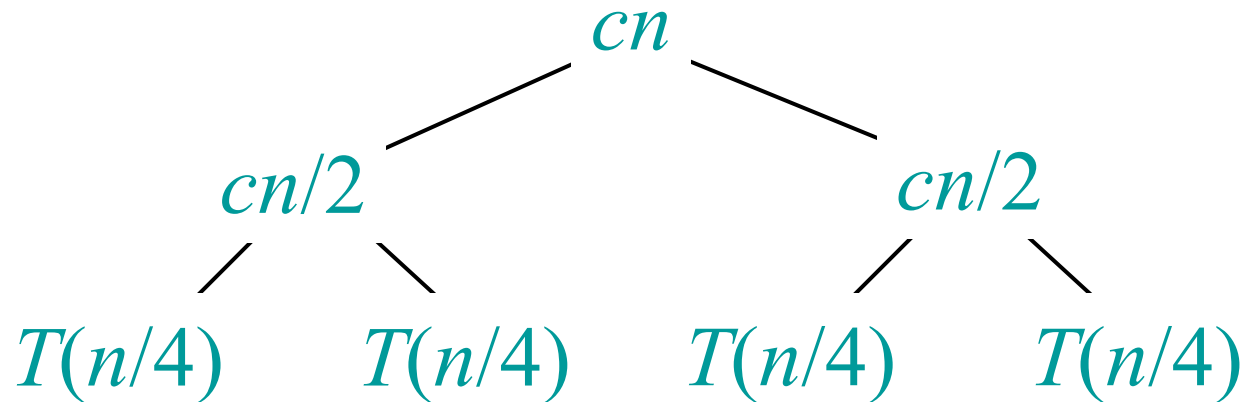
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Recursion tree

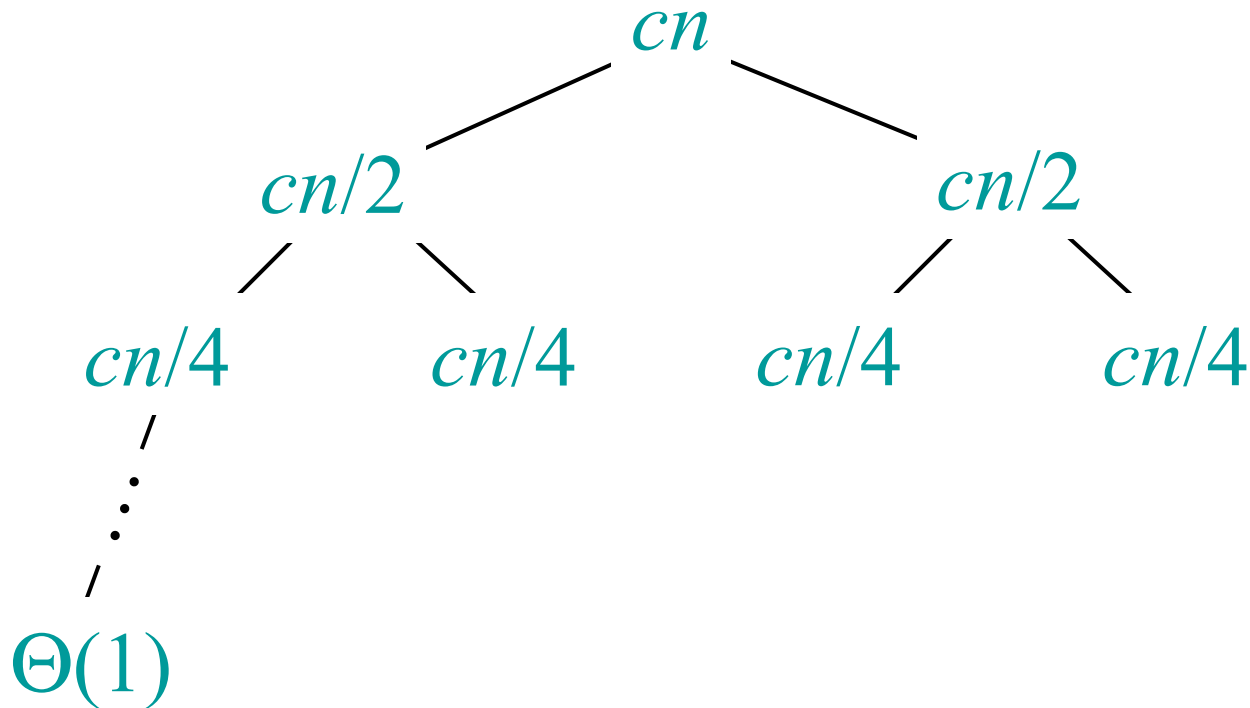
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Recursion tree

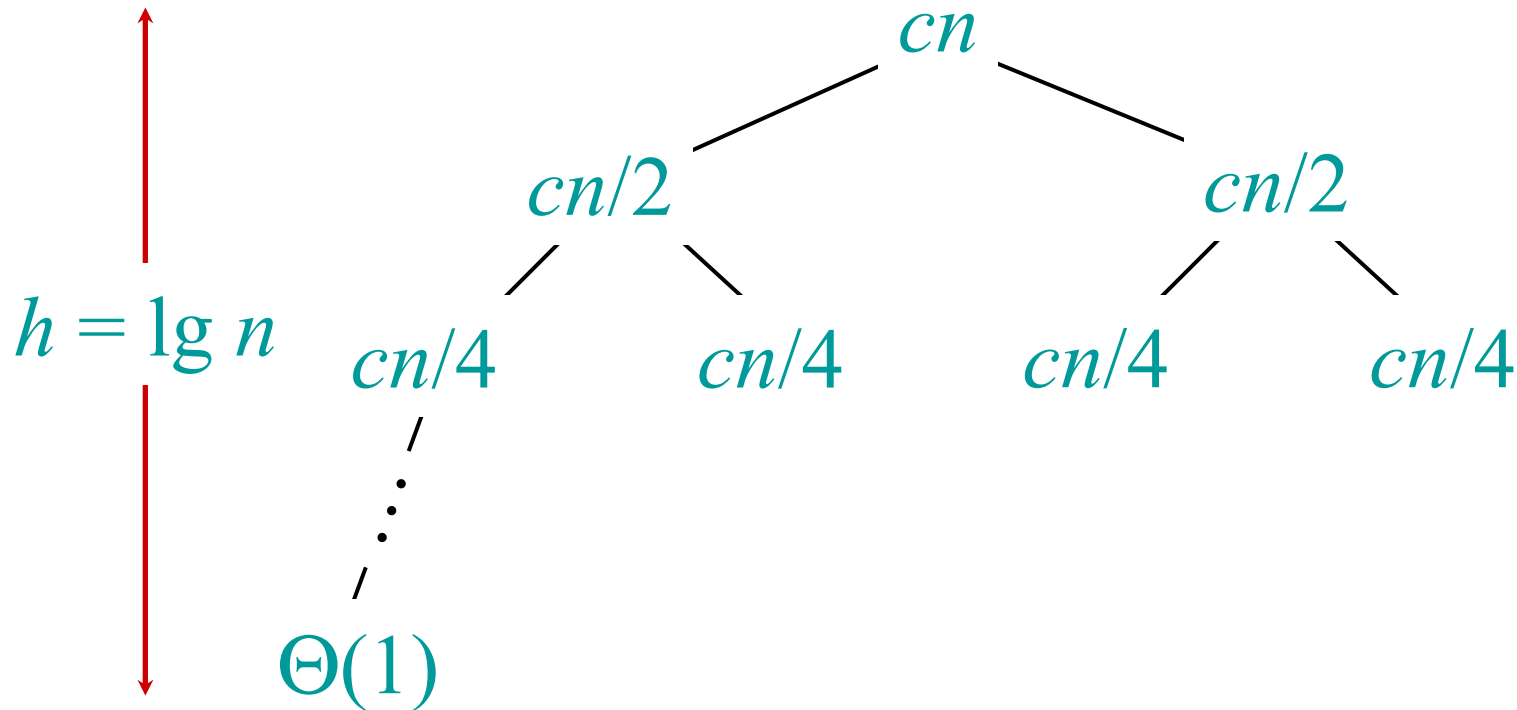
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Recursion tree

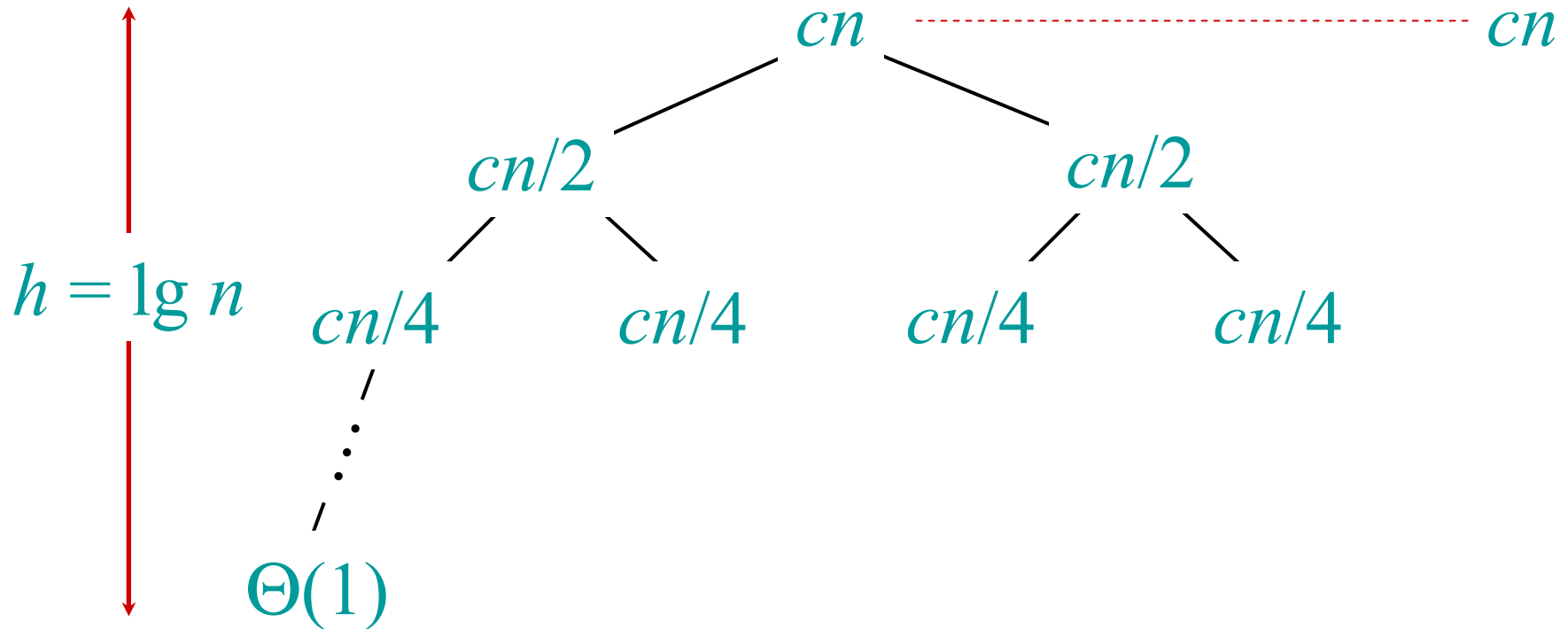
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Recursion tree

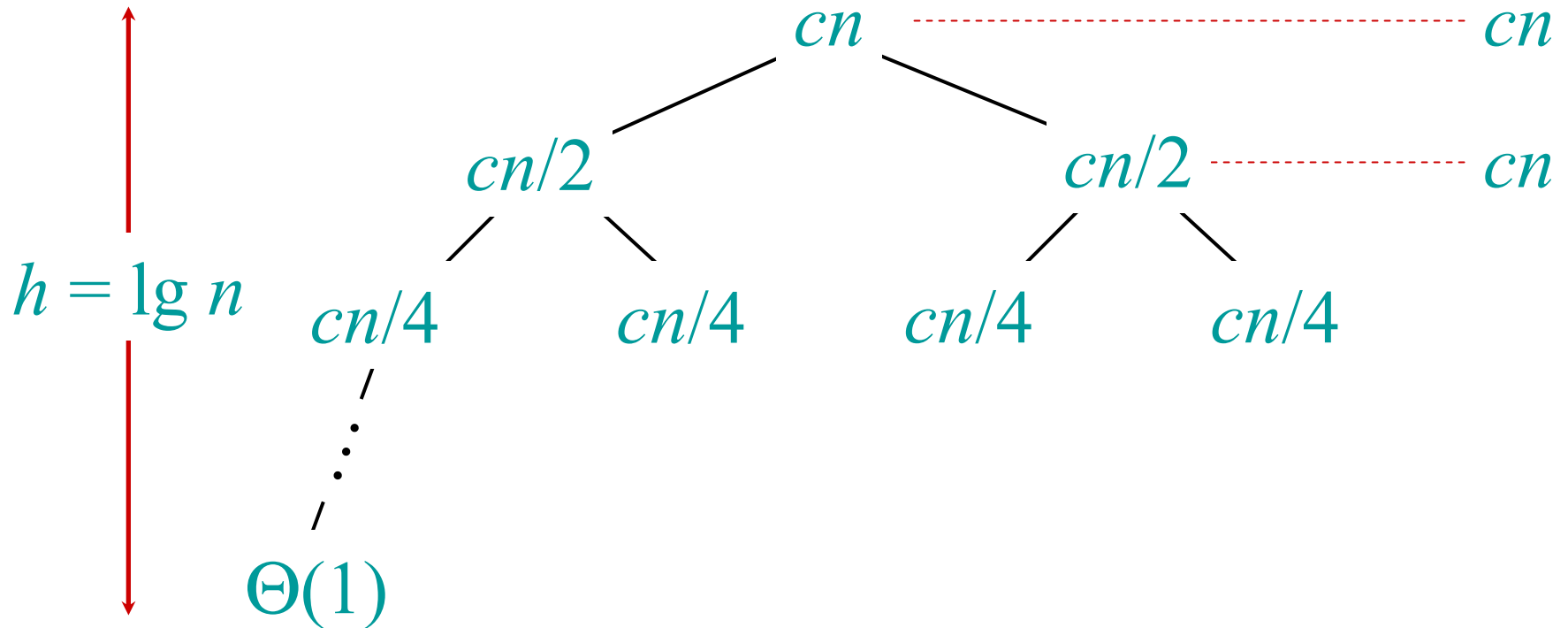
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.





Recursion tree

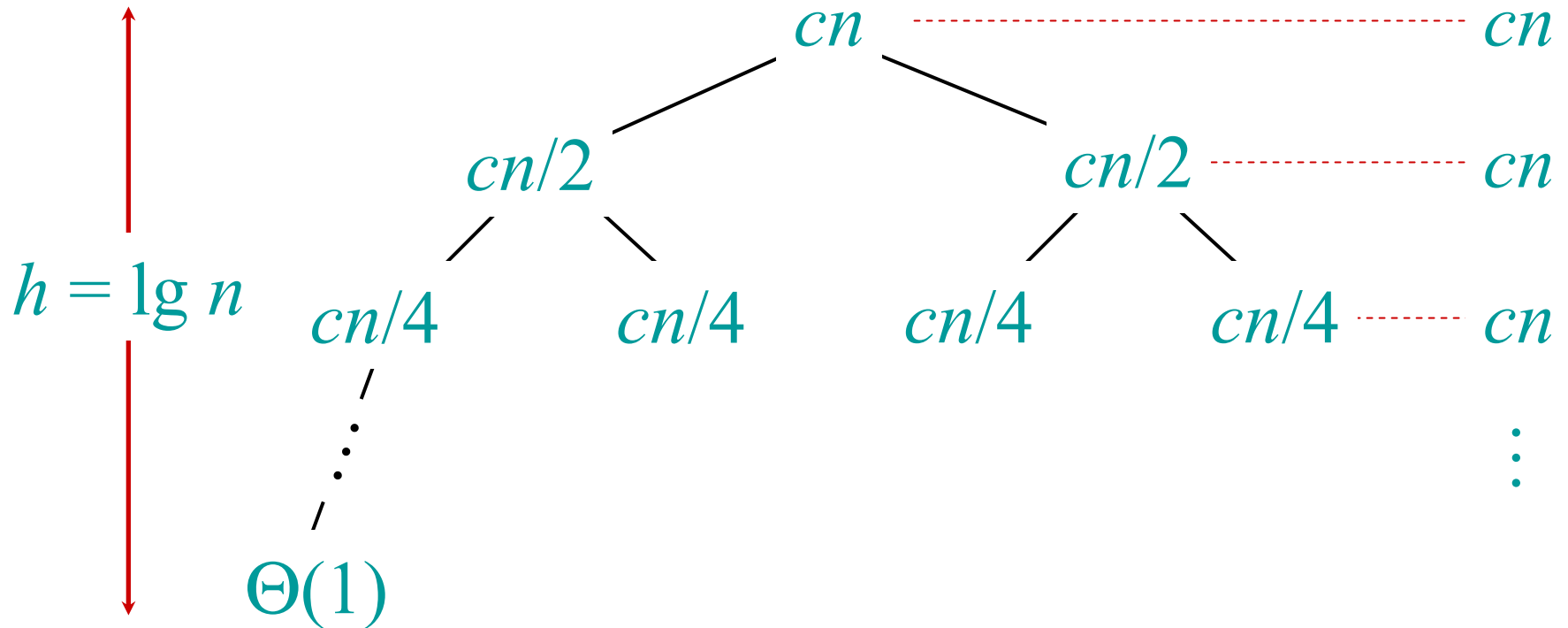
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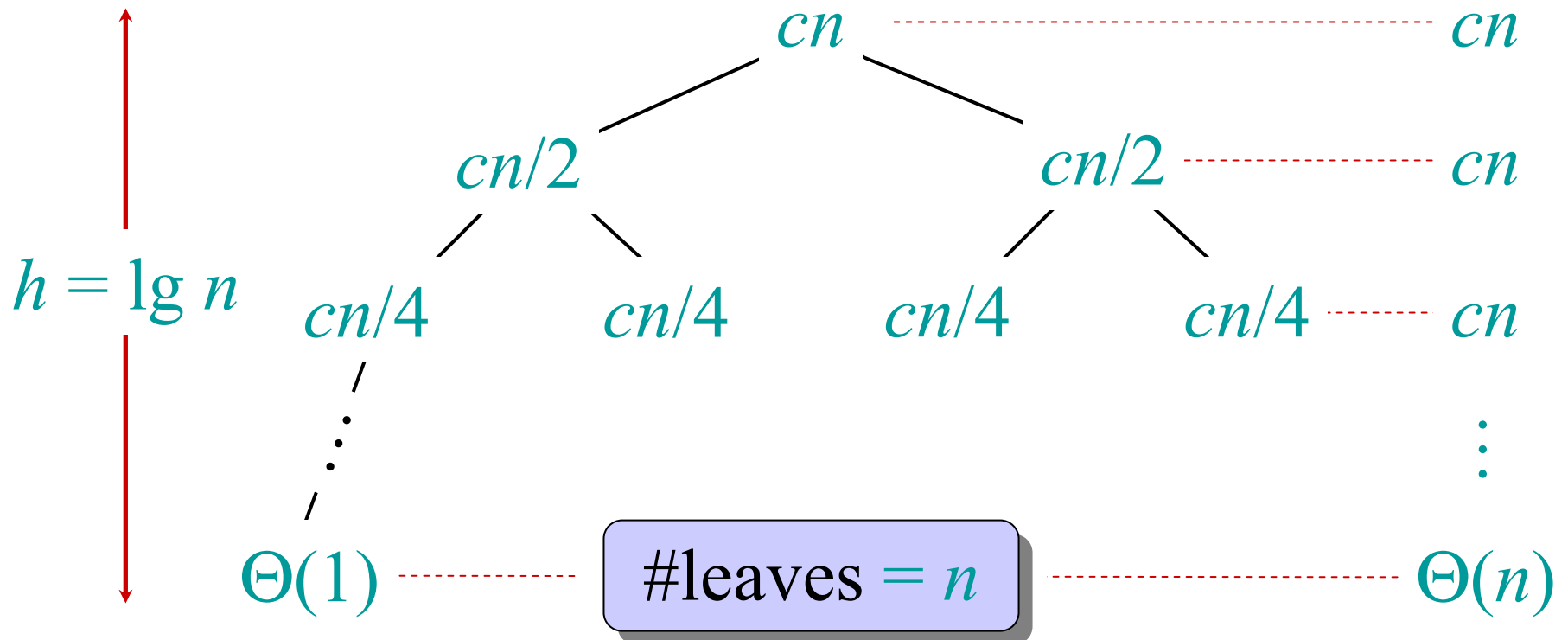
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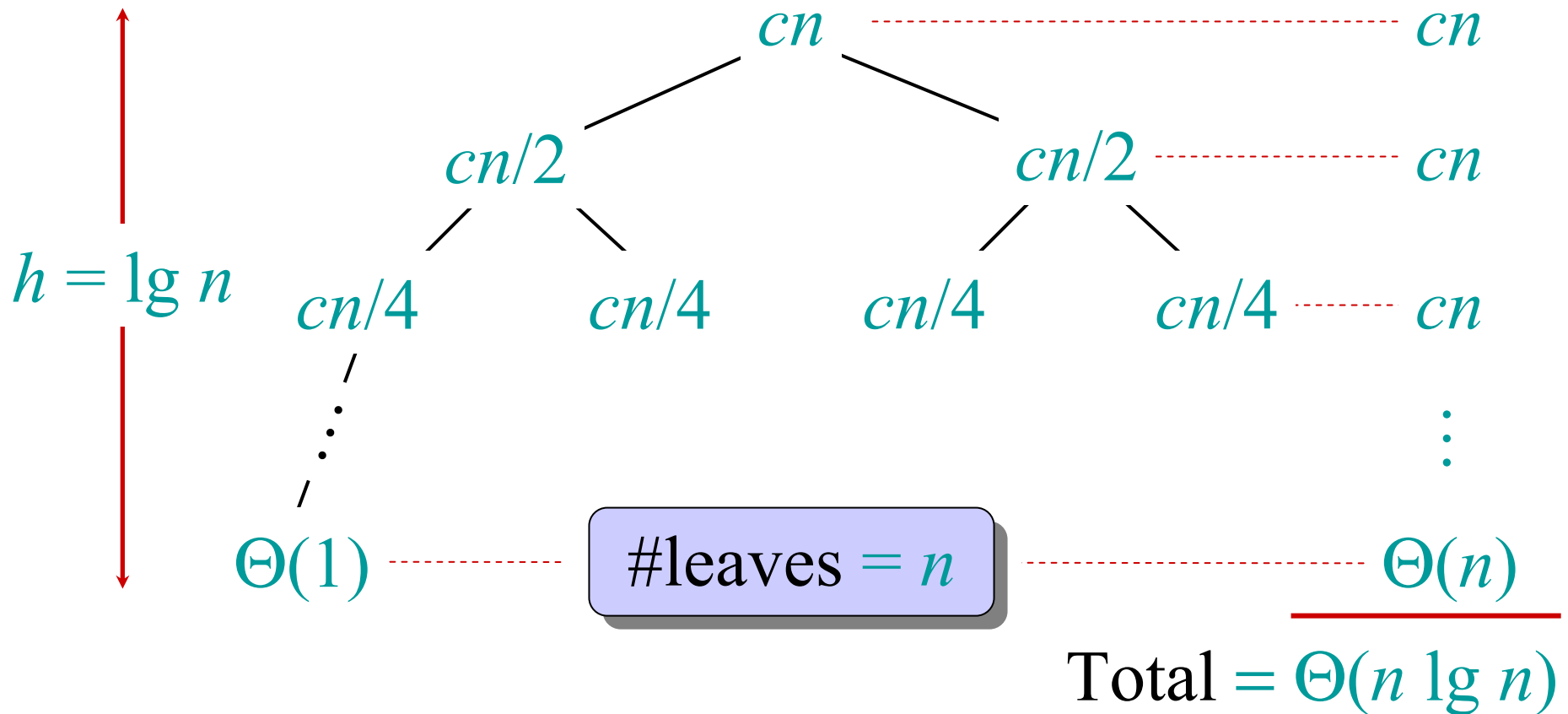
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Recursion tree

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!