

# การอบรมคอมพิวเตอร์โอลิมปิก สอวน. ค่ายที่ 2 ปีการศึกษา 2566

ศูนย์คอมพิวเตอร์โอลิมปิก สอวน.

โรงเรียนสามเสนวิทยาลัย – มหาวิทยาลัยธรรมศาสตร์

(ศูนย์คอมพิวเตอร์โอลิมปิก สอวน. ส.ส. – มธ.)



ภาควิชาวิทยาการคอมพิวเตอร์

คณะวิทยาศาสตร์และเทคโนโลยี ม.ธรรมศาสตร์

# ช่วงที่ 3: อัลกอริทึม (Algorithms)

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ผศ. ดร.วนิดา พฤทธิวิทยา

อ. ดร.พิสิษฐ์ มรรคไพสิษฐ์

ผศ. ดร.ฐาปนา บุญชู

อ. ดร.ศาทนาฏ กิจศิริานุวัตร

# วัตถุประสงค์การเรียนรู้ (Our Goals)

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- learn several **algorithm design techniques** by studying a standard set of important algorithms from different areas of computing (different problem types)
- discuss the general framework for efficiency analysis
- so that, students
  - can develop algorithms on their own when given a computational problem
  - can formally argue the correctness of their algorithms
  - can analyze the efficiency of their algorithms

- บทนำ (Introduction)
- การวิเคราะห์ความซับซ้อนของอัลกอริทึม (Algorithmic Complexity)
- กลวิธีทางอัลกอริทึม
  - Brute Force
  - Divide-and-Conquer
  - Decrease-and-Conquer
  - Transform-and-Conquer
  - String Matching Algorithms
  - Backtracking & Branch-and-Bound
  - Greedy Algorithms
  - Graph Algorithms
  - Dynamic Programming

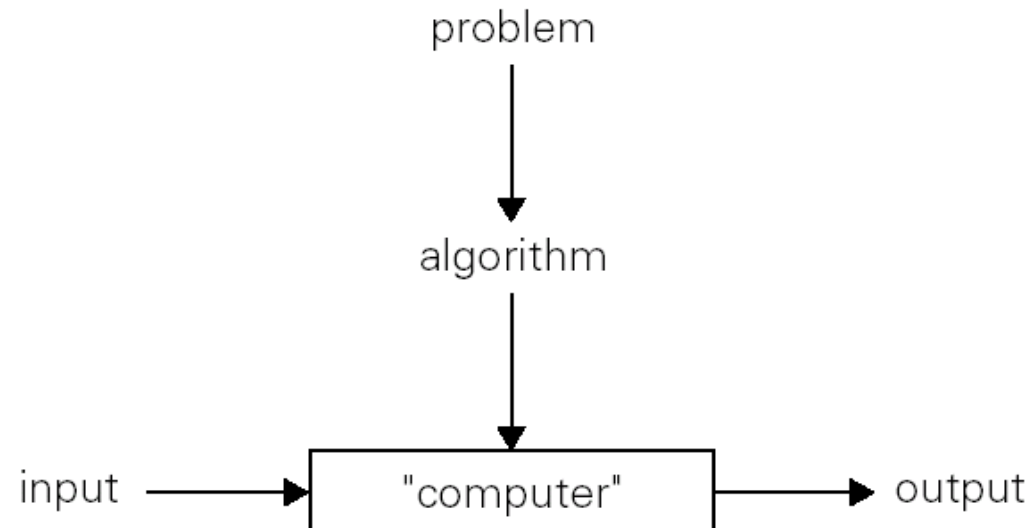
- บทนำ (Introduction)
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  - Dynamic Programming

# บทนำ (Introduction)

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# Introduction: What Is an Algorithm?

- an **algorithm** is a sequence of unambiguous instructions for solving a *well-specified computational problem*
- an **algorithm** is a sequence of computational steps that transforms the *legitimate input* into the *desired output* in a *finite amount of time*



# Introduction: Computational Problems

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- In theoretical computer science, a *computational problem* is a problem that can be solved by *an algorithm* (a step-by-step procedure that can be executed by a computer).
- These *computational problems* have well-defined inputs and outputs, and there exists a *deterministic algorithm* that can produce the correct output for any given input.



# Introduction: Computational Problems (2)

- For example,

## The Problem of Prime Factoring

Given a positive integer  $n$ , find a nontrivial prime factor of  $n$ .

## (decision problem) The Problem of Primality Testing

Given a positive integer  $n$ , determine if  $n$  is prime.

- More examples:
  - Computing the greatest common divisor of a pair of integers
  - Finding the shortest path between a pair of nodes in a finite graph

# Introduction: Computational Problems :

## Important Problem Types

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- **Sorting**: re-arrange a given set of orderable items
- **Searching**: Solution consists of one or more values that satisfies a given condition.
  - **String Processing**: E.g., **string matching** - searching for a given word in a text. Search patterns in DNA sequences database.
  - **Combinatorial problems**: consist in finding, among a finite set of objects, one that satisfies a set of constraints. **Job scheduling** problem. **Assignment** problem. **Knapsack** problem. **Graph Coloring** problem. **Closest-pair** problem.
  - **Graph problems**: E.g., find a path from one geometric location to another on a map
- **Geometric** and **Numerical problems**: E.g., is a given number  $n$  a prime?

# Introduction: Computational Problems (5)

- **Advantages:**
  - **Efficiency:** many computable problems can be solved efficiently using algorithms.
  - **Predictability:** since computable problems have well-defined inputs and outputs, the behavior of their algorithmic solutions is predictable. This makes it easier to reason about their properties and analyze their performance
- **Disadvantages:**
  - **Limitations:** not all problems are computable.
  - **Complexity:** although computable problems can be solved, *the efficiency of the algorithms can vary significantly*. Some problems may have solutions that are computationally expensive, requiring a large amount of time or resources to compute the result.

# Introduction: Computational Problems (6)

- A *computational problem* can be viewed as *a set of instances* or cases together with a, possibly empty, *set of solutions* for every instance/case.

For example, *the problem of Prime Factoring* can be viewed as follows:

{  $[n, p]$ , where  $n$  is an integer  $> 1$  and  $p$  is a set prime factors of  $n$  }

## The Problem of Prime Factoring

Given a positive integer  $n$ , find a nontrivial prime factor of  $n$ .

{  $[n=4, p = 2 \times 2]$  ,  $[n=6, p = 2 \times 3]$  ,  $[n = 15, p = 3 \times 5]$  ,  $[n = 60, p = 3 \times 4 \times 5]$  , ... }

# Introduction: ตัวอย่าง Computational Problems :

$\text{gcd}(m, n)$

Finding the greatest common divisor (gcd) of two non-negative, not both zero integers

- Problem Statement:

- *Input:* two non-negative, not both zero integers called  **$m$**  and  **$n$** ,  $m \geq n$
- *Output:* the greatest common divisor of  **$m$**  and  **$n$**
- a problem Instance: (18, 12), (60, 24), (12, 0)
- *Algorithms:* the intuitive algorithm, Euclid's Algorithm

# Introduction: ตัวอย่าง Computational Problems: $\gcd(m, n)$ : the Consecutive Integers Checking algorithm

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- Intuitively based on the definition of  $\gcd$

Algorithm  $\gcd(\mathbf{m}, \mathbf{n})$

If  $\mathbf{n} = 0$  Return  $\mathbf{m}$  and stop.

While  $\mathbf{n} > 1$  do

Begin

    If  $\mathbf{m}$  is divisible by  $\mathbf{n}$  Return  $\mathbf{n}$  and stop.

    Otherwise, decrease  $\mathbf{n}$  by one.

End

Return  $\mathbf{n}$ .

# Introduction: ตัวอย่าง Computational Problems:

## gcd(m, n): Euclid's algorithm

- $\text{gcd}(\mathbf{m}, \mathbf{n}) = \text{gcd}(\mathbf{n}, \mathbf{m} \bmod \mathbf{n})$

$$\text{gcd}(\mathbf{m}, \mathbf{0}) = \mathbf{m}$$

- Ex.  $\text{gcd}(\mathbf{18}, \mathbf{12})$

$$= \text{gcd}(\mathbf{12}, \mathbf{6})$$

$$= \text{gcd}(\mathbf{6}, \mathbf{0})$$

$$= \mathbf{6}$$

Algorithm  $\text{gcd}(\mathbf{m}, \mathbf{n})$

While  $\mathbf{n} \neq 0$  do

Begin

$$\mathbf{r} = \mathbf{m} \bmod \mathbf{n}$$

$$\mathbf{m} = \mathbf{n}$$

$$\mathbf{n} = \mathbf{r}$$

End.

Return  $\mathbf{m}$ .

# Introduction: ตัวอย่าง Computational Problems: The Sorting Problem (ปัญหาการเรียงลำดับข้อมูล)

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- **Sorting** problem has a variety of **interesting algorithmic solutions** that embody many Computer Science ideas:
  - Comparison vs. non-comparison based strategies,
  - Iterative vs. Recursive implementation,
  - Divide-and-Conquer paradigm (e.g., Merge Sort or Quick Sort),
  - Best / Worst / Average-case Time Complexity analysis,
  - Randomized Algorithms, etc.
- various **Sorting Algorithms**: Insertion Sort, Selection Sort, Bubble Sort, Merge Sort, Quick Sort



# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem (2)

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### *Sorting* Problem (การเรียงลำดับข้อมูล)

- Problem Statement:

- *Input*: a sequence of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$
- *Output*: a reordering of the input sequence  $\langle a_1, a_2, \dots, a_n \rangle$   
such that  $a_i \leq a_j$  whenever  $i < j$
- a *problem Instance*: the sequence  $\langle 31, 41, 59, 26, 41, 58 \rangle$
- given a problem instance: the *input* sequence  $\langle 31, 41, 59, 26, 41, 58 \rangle$
- any sorting algorithm that works correctly would yield the desired *output*, that is, the sequence  $\langle 26, 31, 41, 41, 58, 59 \rangle$

# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: -- in action

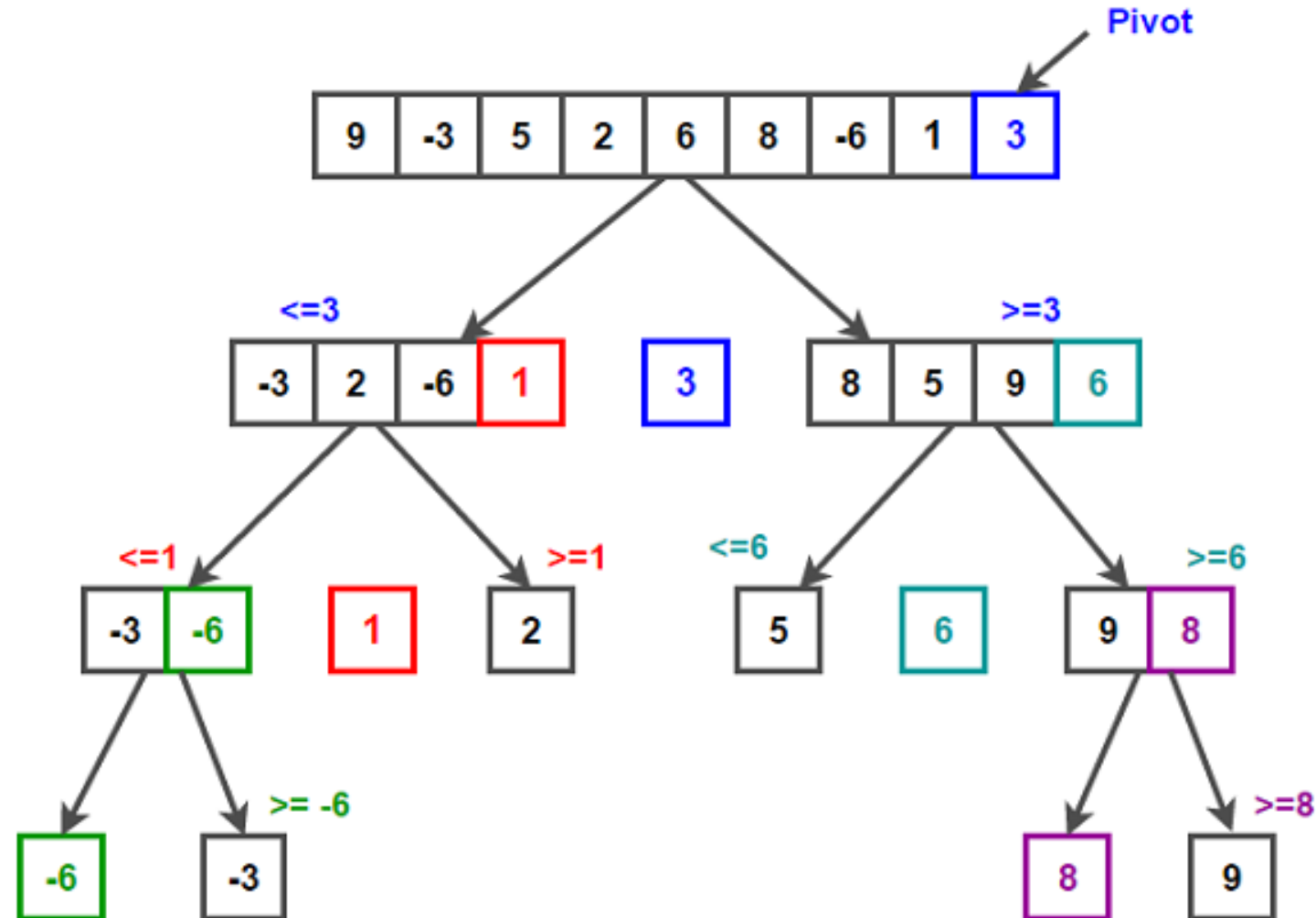
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- Bubble Sort
- Selection Sort
- Insertion Sort

See <https://visualgo.net/en/sorting>

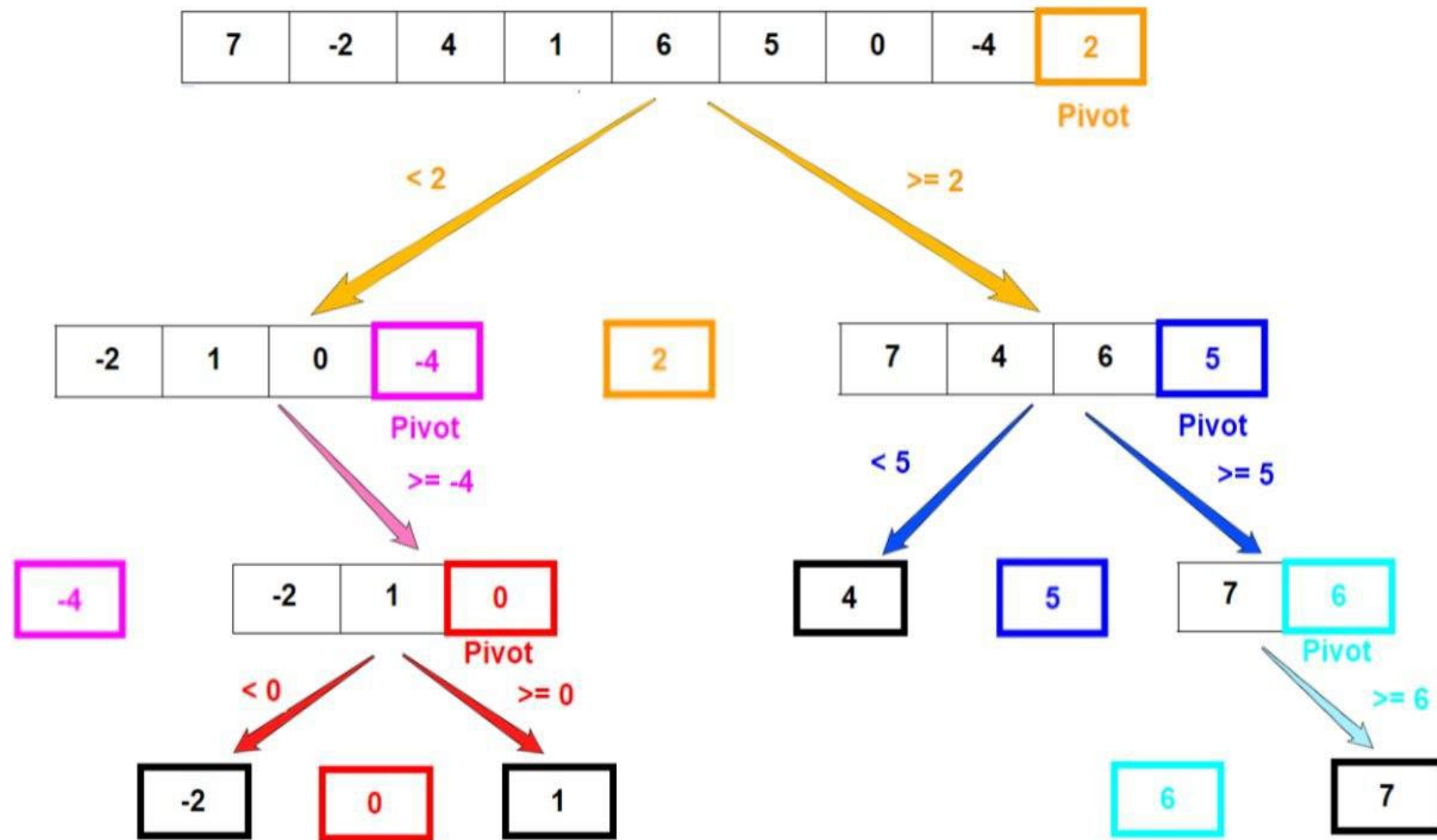
# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: Quick Sort -- in action



# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: Quick Sort -- in action (2)



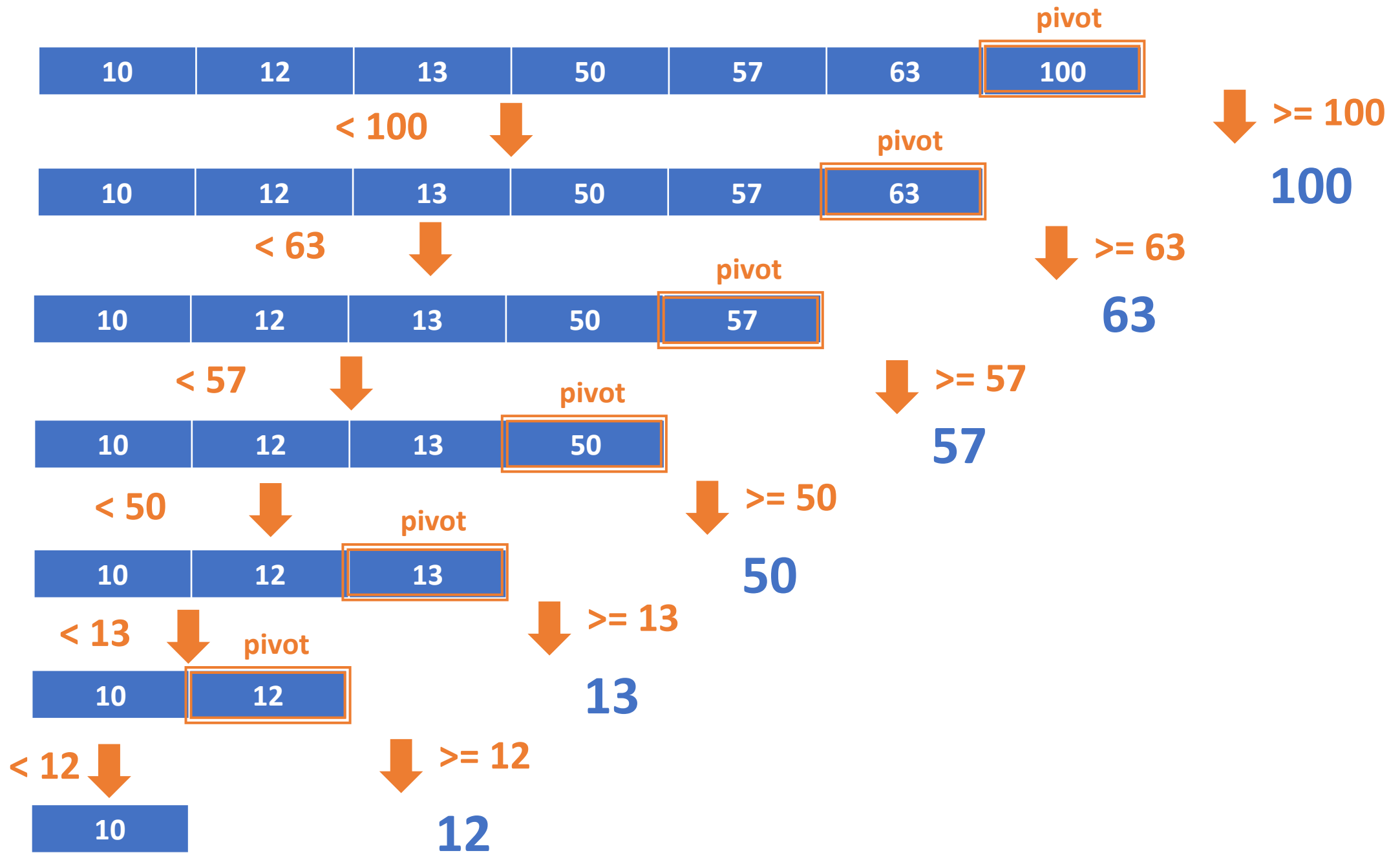
# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: Quick Sort -- in action (3)

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- Worst-Case Scenario: ลิสต์มีข้อมูลจำนวน 7 ตัว

<10, 12, 13, 50, 57, 63, 100>



# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: Quick Sort -- in action (5)

- **Worst-Case Scenario:** จากตัวอย่าง ลิสต์มีข้อมูลจำนวน 7 ตัว จะพบว่า

รอบที่ 1: ทำการเปรียบเทียบ 6 ครั้ง

รอบที่ 2: ทำการเปรียบเทียบ 5 ครั้ง

รอบที่ 3: ทำการเปรียบเทียบ 4 ครั้ง

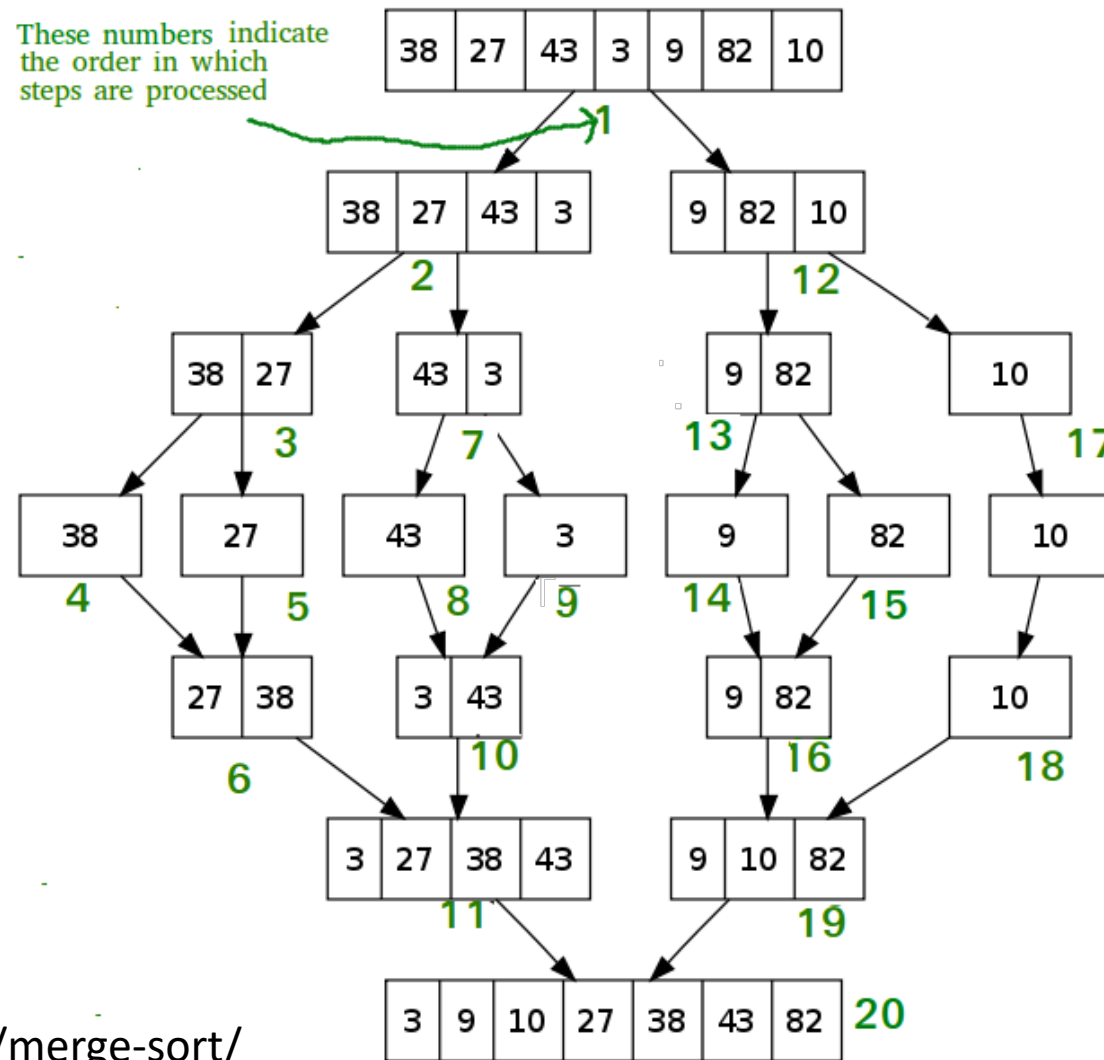
...

รอบที่ 6: ทำการเปรียบเทียบ 1 ครั้ง

$$\begin{aligned}\text{เมื่อ } n = 7, \sum_{i=1}^{7-1} (7 - i) &= \sum_{i=1}^{n-1} (n - i) = \sum_{i=1}^{n-1} n - \sum_{i=1}^{n-1} i \\ &= \underbrace{(n + n + \cdots + n)}_{n-1 \text{ ครั้ง}} - \frac{(n-1) \times (n)}{2} \\ &= (n \times (n - 1)) - \frac{n^2 - n}{2} = \frac{n^2 - n}{2}\end{aligned}$$

# Introduction: ตัวอย่าง Computational Problems:

## The Sorting Problem: Merge Sort -- in action





# Introduction:

## Fundamentals of Algorithmic Problem Solving

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1. Understand the problem
2. Choose between exact and approximate problem solving
3. Decide on algorithm design techniques and appropriate data structures
4. Specify the algorithm: natural language, pseudocode, flowcharts
5. Prove the algorithm's correctness
  1. Exact algorithms: the algorithm yields a required result for every legitimate input
  2. Approximate algorithms: the error produced by the algorithm does not exceed a pre-defined limit
6. Analyze the algorithm
7. Coding the algorithm

# Introduction (19)

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- several algorithms for solving the same problem may exist
  - different people have different ideas when solving a problem
  - different *algorithm design techniques*
- one algorithm can be implemented in several different ways using different *data structures*
- as a result, *different algorithms* that are based on *different ideas* and employ *different data structures* could produce the desired result with largely *different speeds* (different *execution time* or *run time*)

# Introduction (20)

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- given an algorithm implemented with one algorithm design technique using one type of data structure
- there are many factors that still play a role in the *running time* of the algorithm
- Let's look at the Sorting problem as our example
  - the number of items to be sorted (input size)
  - the extent to which the items are already somewhat sorted
  - the architecture of the computer
  - the kind of storage devices to be used: M/M, disks (magnetic or SSD), or tapes

# พื้นฐานการวิเคราะห์ความซับซ้อนของอัลกอริทึม

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# What is an Analysis of Algorithms?

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- “**Analysis of algorithms**” means an investigation of **an algorithm’s efficiency** with respect to two resources: running time and memory space
  - การประเมินประสิทธิภาพของอัลกอริทึม
- **time efficiency ประสิทธิภาพเชิงเวลา** (time complexity) indicates how fast an algorithm in question runs
- **space efficiency ประสิทธิภาพเชิงพื้นที่** (space complexity) refers to the amount of memory units the algorithm requires

# Analysis of Algorithms: the Analysis Framework

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- it is logical to investigate an algorithm's efficiency as “a function of some parameter  $n$  indicating **the algorithm's input size**”

$T(n)$  : time taken by the algorithm to run when input size =  $n$

# Analysis of Algorithms: the Analysis Framework (2)

- Example: the problem of sorting elements in a list
  - the **algorithm's input size** is indicated by **the list's size**
- Example: the problem of searching for the maximum element in a list
  - the **algorithm's input size** is indicated by **the list's size**
- Example: problems of matrix addition, matrix subtraction
  - the **algorithm's input size** is indicated by **the matrix's size**
- Example: problem of evaluating a polynomial degree  $k$ 
$$p(x) = a_k x^k + a_{k-1} x^{k-1} + \dots + a_1 x + a_0$$
  - the **algorithm's input size** is indicated by **the degree of polynomial**

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency

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- **Time efficiency** is analyzed by counting the number of times an algorithm's **basic operation** is executed on input of size  $n$ 
  - **basic operation**: the operation contributing the most to the total running time

**Rule for finding algorithm's basic operation: it is the most time-consuming operation in the algorithm's innermost loop**



# Analysis of Algorithms: the Analysis Framework

## Time Efficiency (2)

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Example: the problem of searching for the maximum element in a list

Algorithm FindMax( $A[0..n-1]$ )

max =  $A[0]$ ;

For ( $i = 1; i < n; i++$ ) do

    if ( $\text{max} < A[i]$ ) max =  $A[i]$ ;

- **basic operation**: key comparison inside loop 'for'

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency (3)

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Example: the problem of sorting all elements in a list

Algorithm BubbleSort( $A[0..n-1]$ )

For ( $i = 0; i < n-1; i++$ )

    For ( $j = 0; j < n - i - 1; j++$ )

        if ( $A[j] > A[j+1]$ )

            Swap( $A[j], A[j+1]$ );

- **basic operations** : key comparison inside the inner loop for  
swap operation inside the inner loop for

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency (4)

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- let  $n$  be the algorithm's input size
- let  $c_{op}$  be the execution time of an algorithm's **basic operation** on a particular computer
- let  $C(n)$  be the number of times this basic operation needs to be executed for this algorithm Ex.  $C(n) = n$
- then, the **running time**  $T(n)$  of a program implementing this algorithm on this particular computer can be estimated by the formula

$$T(n) \approx c_{op} \times C(n)$$

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency (5)

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- We usually focus on how much the running time changes as the input size increases, not the running time for a particular input size

**How much longer will the algorithm run if we double its input size?**

$$\frac{T(2n)}{T(n)} = \frac{c_{op} \times C(2n)}{c_{op} \times C(n)}$$

note that we were able to answer the question w/o actually knowing the value of  $c_{op}$ : the value was neatly **cancelled out** in the ratio

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency (6)

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- it is for these reasons that the time efficiency analysis framework ignores multiplicative constants and concentrates on the count's order of growth to within a constant multiple for large-size inputs

$$T(n) \approx C(n)$$

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency (7)

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- When discussing **the count's order of growth** to within a constant multiple for **large-size inputs**, we do not need to be exact
- We usually use **asymptotic notations** to express the time efficiency of an algorithm (**the count's order of growth** )

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Asymptotic Notations

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- to **compare** and **rank** the **orders of growth** of an algorithm's **basic operation counts**, computer scientists use three notations:
  1.  $O$  (big Oh)
  2.  $\Omega$  (big Omega)
  3.  $\Theta$  (big Theta)

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Asymptotic Notations: big-Oh

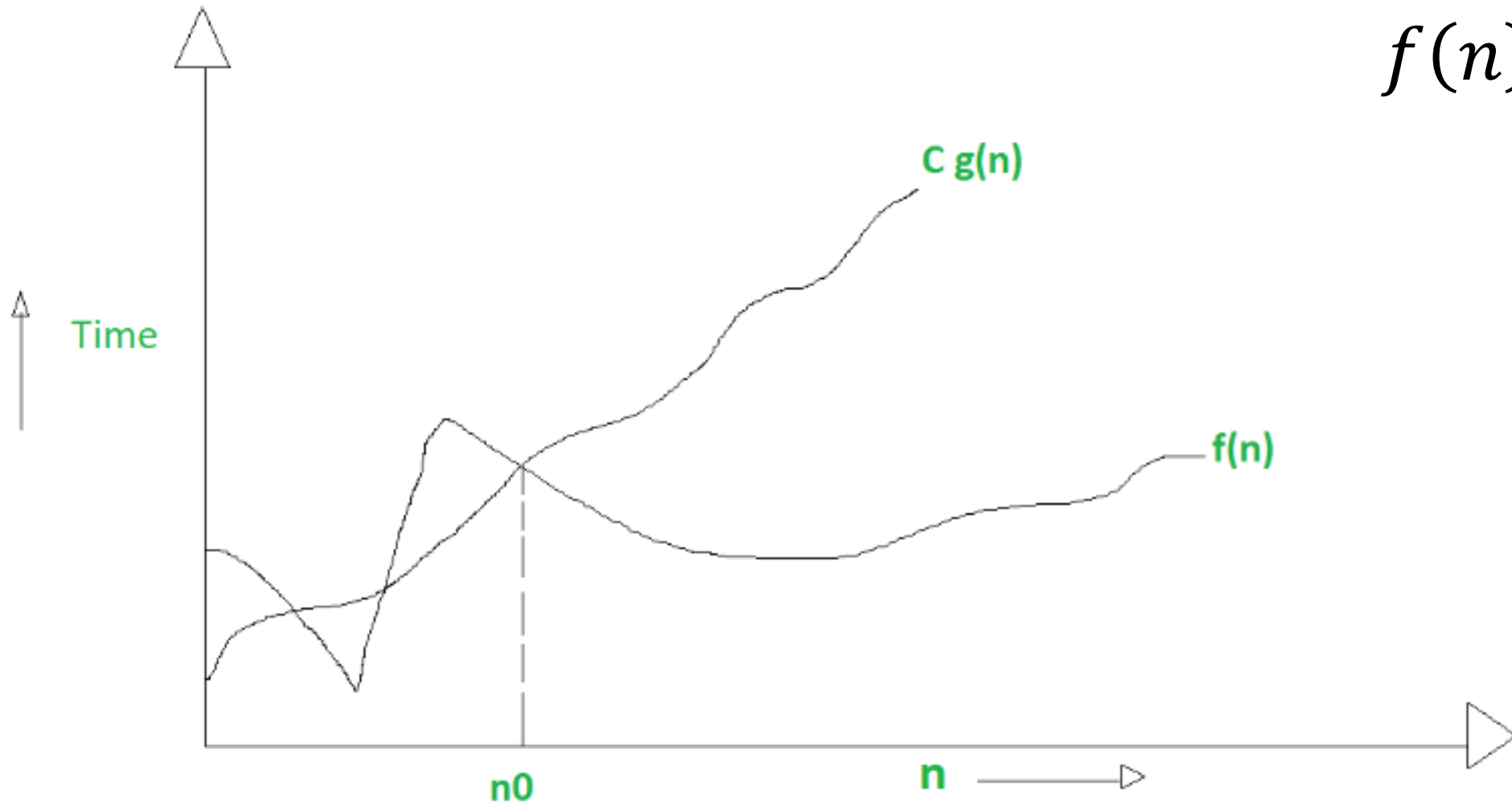


- $O(g(n))$  is the set of all functions with a **smaller or same order of growth** as  $g(n)$  (to within a constant multiple, as  $n$  goes to infinity).
- $f(n) \in O(g(n))$  as there exists  $C > 0$  and  $n_0$  such that
$$f(n) \leq C g(n), \text{ whenever } n \geq n_0$$
- $g(n)$  is an “upper-bound”.
- for example,

$$\begin{array}{lll} n \in O(n^2), & 100n + 5 \in O(n^2), & \frac{1}{2}n(n-1) \in O(n^2) \\ n^3 \notin O(n^2), & 0.00001n^3 \notin O(n^2), & n^4 + n + 1 \notin O(n^2) \end{array}$$



$$f(n) \in O(g(n))$$



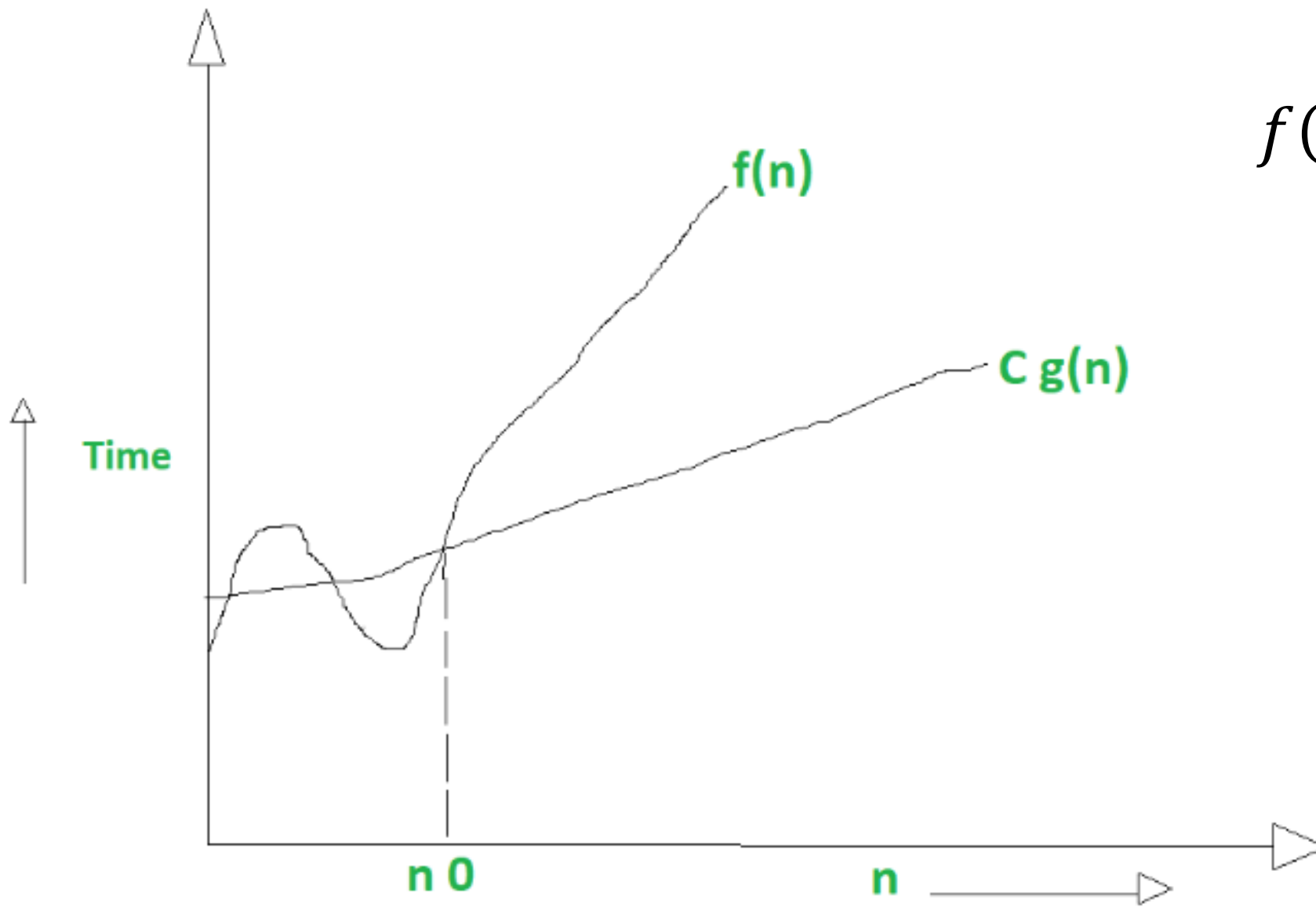
# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Asymptotic Notations: big-Omega



- $\Omega(g(n))$ , stands for the set of all functions with a **larger or same order of growth** as  $g(n)$  (to within a constant multiple, as  $n$  goes to infinity).
- $f(n) \in \Omega(g(n))$  as there exists  $C > 0$  and  $n_0$  such that
$$f(n) \geq C g(n), \text{ whenever } n \geq n_0$$
- $g(n)$  is a “lower-bound”.
- for example,

$$n^3 \in \Omega(n^2), \quad \frac{1}{2}n(n-1) \in \Omega(n^2), \quad \text{but } 100n + 5 \notin \Omega(n^2).$$



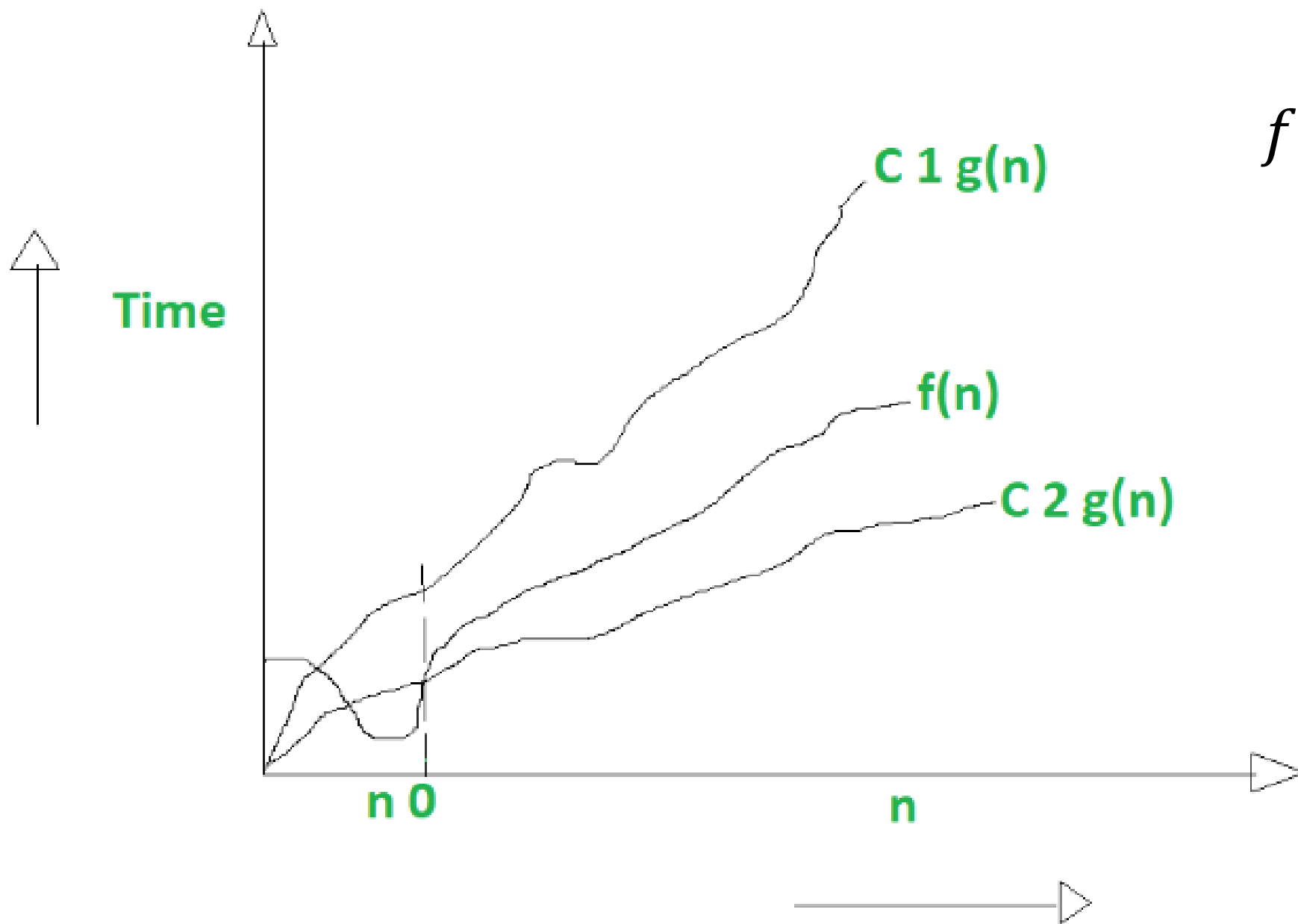
$$f(n) \in \Omega(g(n))$$

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Asymptotic Notations: big-Theta



- $\Theta(g(n))$  is the set of all functions that have the **same order of growth** as  $g(n)$  (to within a constant multiple, as  $n$  goes to infinity).
- $f(n) \in \Theta(g(n))$  as there exists  $C_1$  and  $C_2 > 0$  and  $n_0$  such that
$$C_1 g(n) \leq f(n) \leq C_2 g(n), \text{ whenever } n \geq n_0$$



$$f(n) \in \Theta(g(n))$$

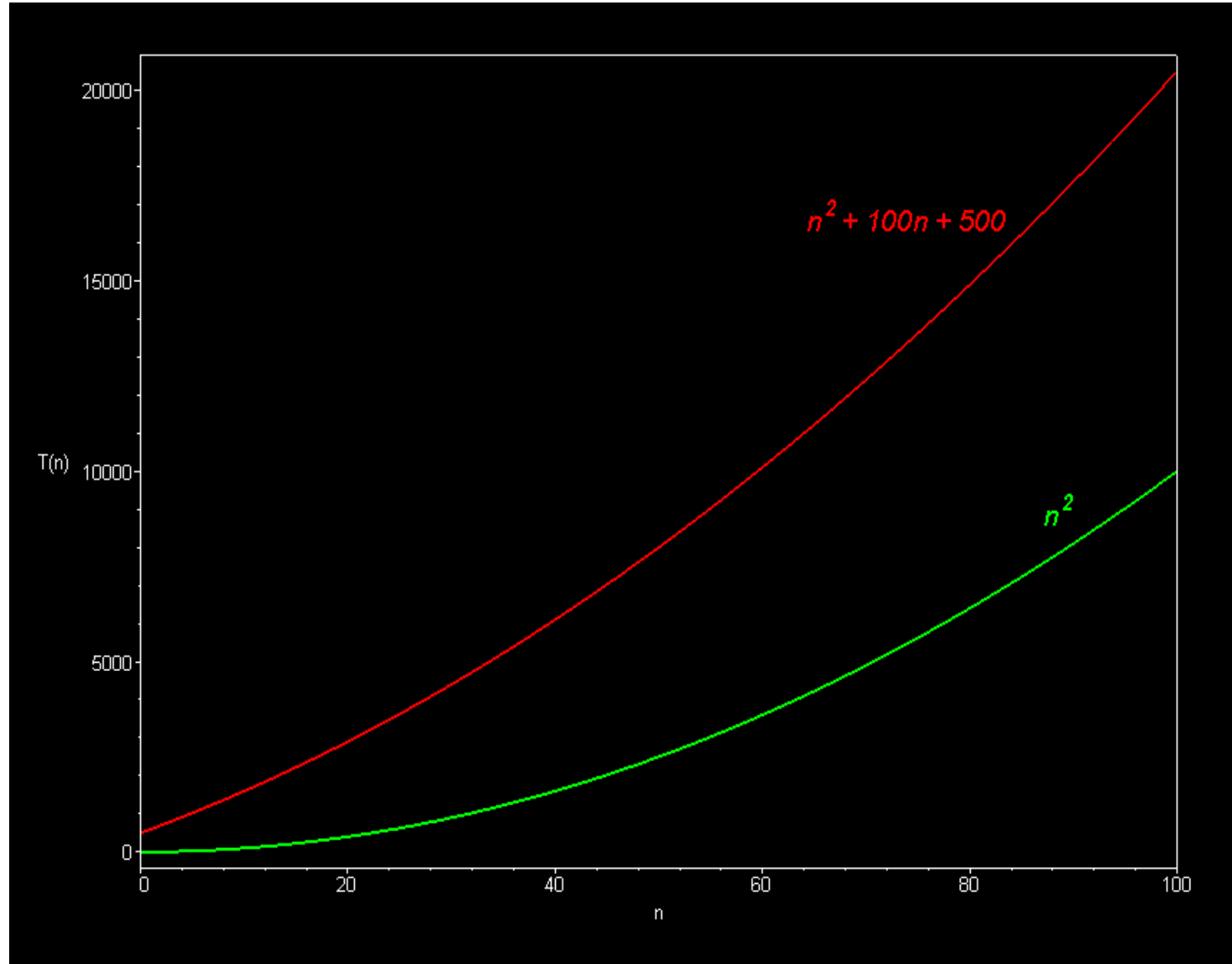
# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Asymptotic Notations: big-Theta

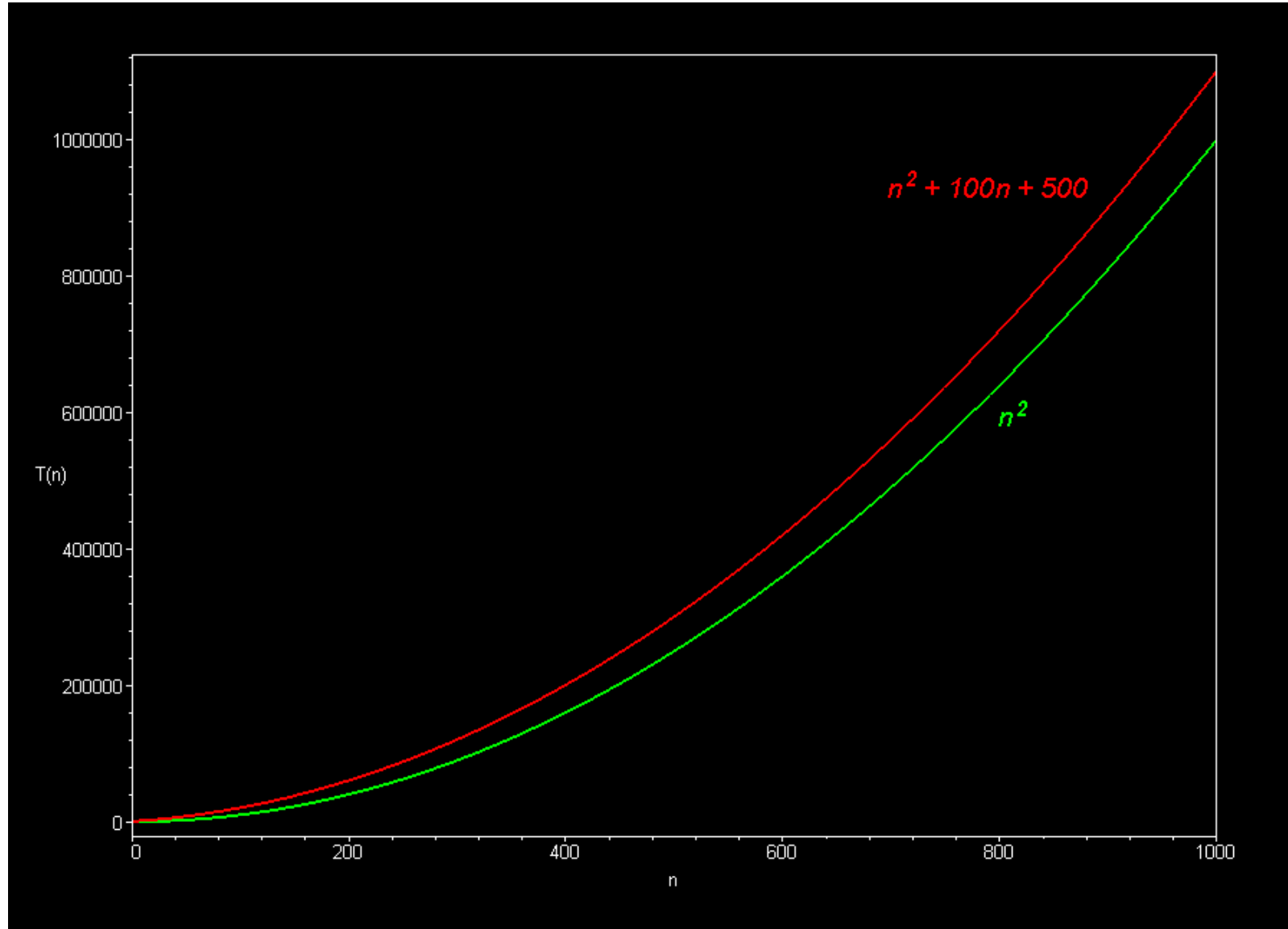
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- thus, every quadratic function  $an^2 + bn + c$  with  $a > 0$  is in  $\Theta(n^2)$

## Asymptotic Notations: Informal Introduction: big-Theta (2)

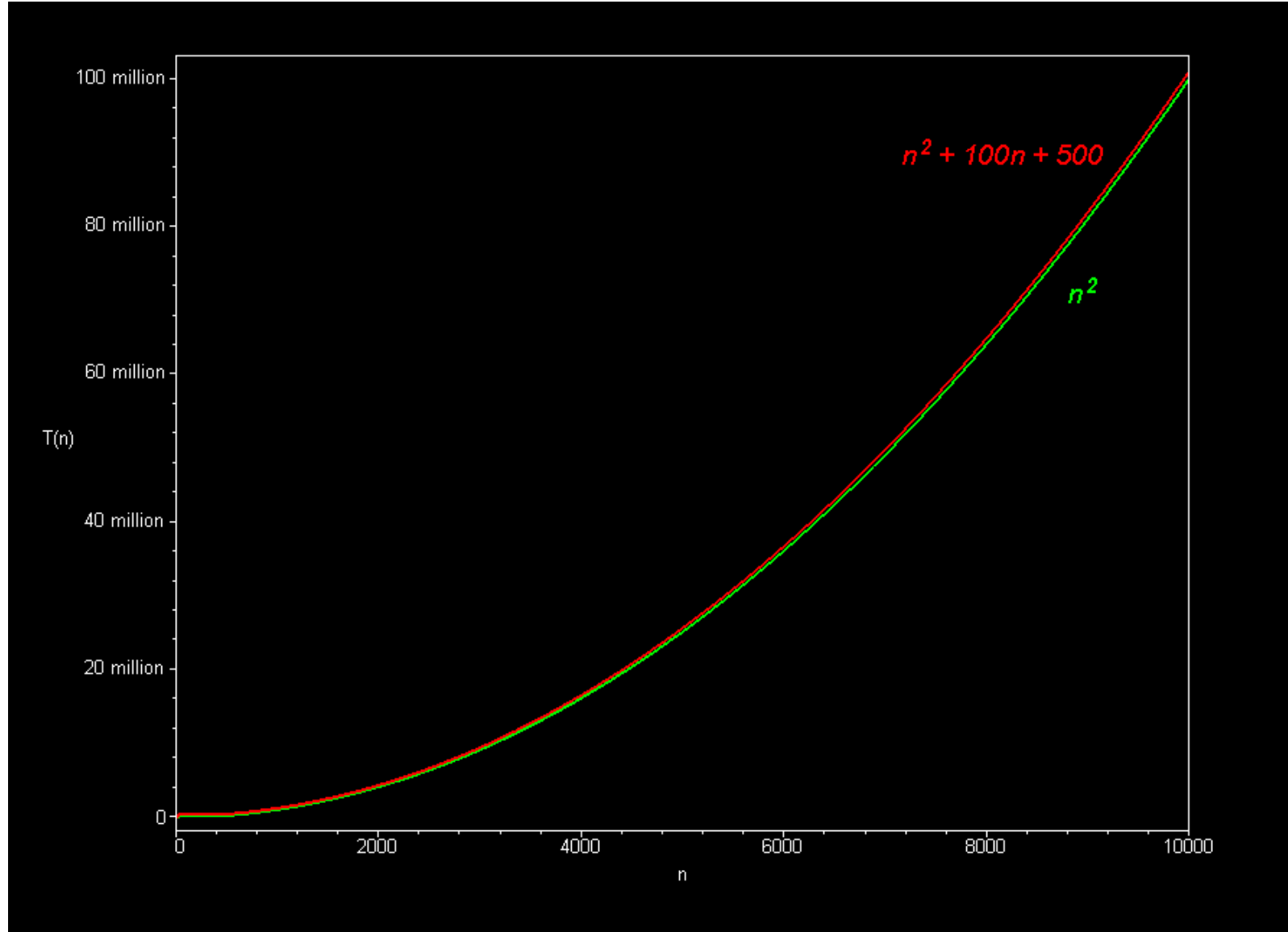


## Asymptotic Notations: Informal Introduction: big-Theta (3)





## Asymptotic Notations: Informal Introduction: big-Theta (4)



# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Basic Asymptotic Efficiency Classes

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Class	Name
1	Constant
$\log n$	Logarithmic
$n$	linear
$n \log n$	“n-log-n”
$n^2$	Quadratic
$n^3$	Cubic
$2^n$	exponential
$n!$	factorial

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Order of Growth

**TABLE 2.1** Values (some approximate) of several functions important for analysis of algorithms

$n$	$\log_2 n$	$n$	$n \log_2 n$	$n^2$	$n^3$	$2^n$	$n!$
10	3.3	$10^1$	$3.3 \cdot 10^1$	$10^2$	$10^3$	$10^3$	$3.6 \cdot 10^6$
$10^2$	6.6	$10^2$	$6.6 \cdot 10^2$	$10^4$	$10^6$	$1.3 \cdot 10^{30}$	$9.3 \cdot 10^{157}$
$10^3$	10	$10^3$	$1.0 \cdot 10^4$	$10^6$	$10^9$		
$10^4$	13	$10^4$	$1.3 \cdot 10^5$	$10^8$	$10^{12}$		
$10^5$	17	$10^5$	$1.7 \cdot 10^6$	$10^{10}$	$10^{15}$		
$10^6$	20	$10^6$	$2.0 \cdot 10^7$	$10^{12}$	$10^{18}$		
	Logarithmic	Linear	Linearithmic	Quadratic	Cubic	Exponential	Factorial

- the function *growing the slowest* among these is the **logarithmic function**
- the function *growing the fastest* among these is the **factorial function**
- a program implementing an algorithm with a **logarithmic basic-operation count** runs practically on inputs of all realistic sizes
- a program implementing an algorithm that requires an **exponential number of basic-operations** are practical for only problems of very small sizes

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Order of Growth (3)

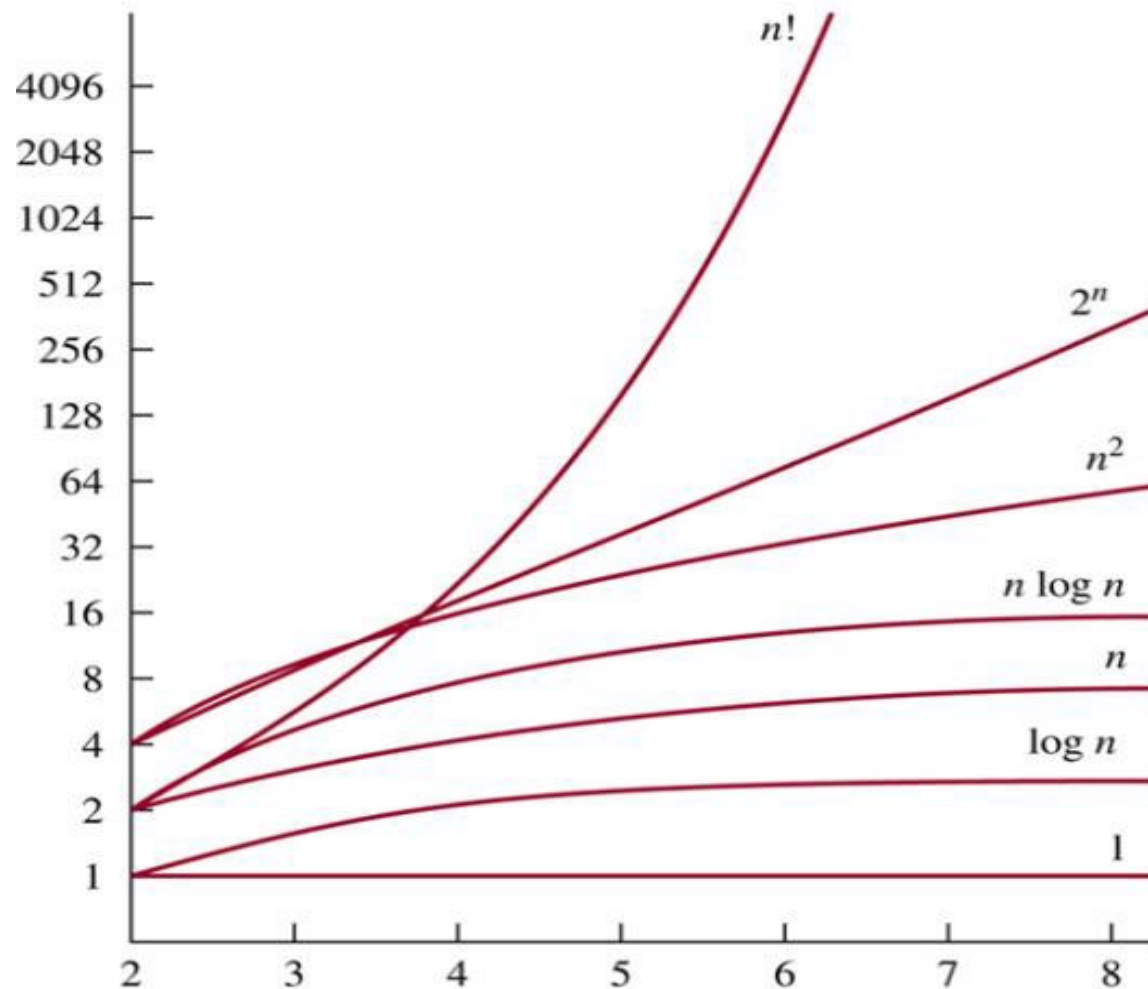
- Example. See how functions  $C(n)$  with different growths react to a twofold increase in the value of their arguments ( $n$ )

	$C(n)$	$C(2n)$
1. Logarithmic	$\log_2 n$	$\log_2 2n = \log_2 2 + \log_2 n = 1 + \log_2 n$ [i.e. <b>increase by just 1</b> ]
2. Linear	$n$	$2n$ [i.e. <b>twofold</b> ]
3. Linearithmic	$n \log_2 n$	$2n(1 + \log_2 n) = 2n + 2n \log_2 n$ [i.e. <b>slightly more than twofold</b> ]
4. Quadratic	$n^2$	$(2n)^2 = 4n^2$ [i.e. <b>fourfold</b> ]
5. Cubic	$n^3$	$(2n)^3 = 8n^3$ [i.e. <b>eightfold</b> ]
6. Exponential	$2^n$	$2^{2n} = (2^n)^2 = 2^n \times 2^n$ [i.e. <b><math>2^n</math>-fold</b> ]
7. Factorial	$n!$	$(2n)!$

# Analysis of Algorithms: the Analysis Framework:

## Time Efficiency: Order of Growth (4)

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# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Example (1)

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For an algorithm with  $C(n) = 0.5n(n-1)$

- $C(n) = 0.5n^2 - 0.5n$
- big-Oh for the time efficiency of the algorithm is  $O(n^2)$

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Example (2)

---

Example: the problem of searching for the maximum element in a list

Algorithm FindMax(A[0..n-1])

max = A[0];

For (i = 1; i < n; i++) do

    if (max < A[i]) max = A[i];

- **basic operation**: key comparison inside loop for
- $C(n) = n - 1$
- big-Oh for the time efficiency of the algorithm is  $O(n)$



# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Example (3)

Example: the problem of sorting all elements in a list

Algorithm BubbleSort(A[0..n-1])

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=0}^{n-i-2} 2 = \sum_{i=0}^{n-2} 2(n-i-1)$$

For (i = 0; i < n-1; i++)

For (j=0; j < n - i -1; j++)

$$= 2[\sum_{i=0}^{n-2} (n-i-1)] = (n-1)(n)$$

if (A[j] > A[j+1])

Swap(A[j], A[j+1]);

big-Oh for the time efficiency

of the algorithm =  $O(n^2)$

- **basic operation**: key comparison inside the inner loop for  
swap operation inside the inner loop for

$$\begin{aligned}
C(n) &= \sum_{i=0}^{n-2} \sum_{j=0}^{n-i-2} 2 = \sum_{i=0}^{n-2} 2(n-i-1) \\
&= 2 \left[ \sum_{i=0}^{n-2} (n-i-1) \right] \\
&= 2 \left[ (n-1) + (n-2) + \cdots + 1 \right] \\
&= 2 \left[ \sum_{i=1}^{n-1} i \right] \\
&= 2 \times \frac{(n-1)(n-1+1)}{2} \\
&= (n-1)(n)
\end{aligned}$$

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=0}^{n-i-2} 2 = \sum_{i=0}^{n-2} 2(n-i-1)$$

$$= 2 \left[ \sum_{i=0}^{n-2} n - \sum_{i=0}^{n-2} i - \sum_{i=0}^{n-2} 1 \right]$$

$$= 2 \left[ n(n-2-0+1) - \frac{(n-2)(n-1)}{2} - (n-2-0+1) \right]$$

$$= 2 \left[ n(n-1) - \frac{(n-2)(n-1)}{2} - (n-1) \right] = n(n-1)$$

## Time Efficiency: Worst-Case, Best-Case, Average-Case

---

- There are many algorithms for which running time depends *not only* on an **input size** but also on **the specifics of a particular input**:
  - worst-case, best-case, and average-case running time

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Worst-Case

---

- the **worst-case efficiency** of an algorithm is its efficiency for **the worst-case input** of size  **$n$** , which is an input of size  **$n$**  for which the algorithm **runs the longest** among all possible inputs of that size

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Best-Case

---

- the **best-case efficiency** of an algorithm is its efficiency for **the best-case input** of size  **$n$** , which is an input of size  **$n$**  for which the algorithm runs **the fastest** among all possible inputs of that size.
- **best case does not mean** the smallest input

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Average-Case

---

- **the average-case efficiency** seeks to provide the algorithm's behavior on a “typical” or “**random**” input.
- to analyze the algorithm's **average-case efficiency**, we must make some **assumptions** about possible inputs **of size  $n$** .

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Worst-Case, Best-Case, Average-Case:

### Example



- Example: Sequential Search Algorithm

**ALGORITHM** **SequentialSearch** ( **A** [ 0 . . ( **n** - 1 ) ], **K** )

// Search for a given value in a given array by sequential search

// **Input**: an array **A** [ 0 . . **n** - 1 ] and a search key **K**

// **Output**: return the index of the first element of **A** that matches **K**

// or -1 if there are no matching elements

**i**  $\leftarrow$  0

**while** **i** < **n** **and** **A**[**i**]  $\neq$  **K** **do**

**i**  $\leftarrow$  **i** + 1

**if** **i** < **n** **return** **i**

**else return** -1



# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Worst-Case, Best-Case, Average-Case:

### Example (2)



- Sequential Search Algorithm: Worst Case:
  - when there are no matching elements or
  - the first matching element happens to be the last one on the list

$$C_{\text{worst}}(n) = n$$

- the **worst-case analysis** works as an **upper-bound** for the running time of an algorithm: it guarantees that for any instance of size **n**, the running time **will not exceed**  $C_{\text{worst}}(n)$

# Analysis of Algorithms: the Analysis Framework

Time Efficiency: Worst-Case, Best-Case, Average-Case:

## Example (2)



- Sequential Search Algorithm: Best Case:
  - when the first element in the list equals to the search key

$$C_{\text{best}}(n) = 1$$

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Worst-Case, Best-Case, Average-Case:

### Example (3)



- Sequential Search Algorithm: Average Case:
- **Standard assumptions**
  - (1) the probability of a successful search is equal to  $p$   
( $0 \leq p \leq 1$ )
  - and
  - (2) the probability of the first match occurring  
in the  $i^{\text{th}}$  position of the list is the same for every  $i$

# Analysis of Algorithms: the Analysis Framework

## Time Efficiency: Worst-Case, Best-Case, Average-Case:

### Example (4)



- in the case of a **successful search**, the **probability of the first match occurring in the  $i^{\text{th}}$  position** of the list is  **$p/n$  for every  $i$** , and the **number of comparisons** made by the algorithm in such a situation is obviously  **$i$**
- in the case of an **unsuccessful search**, the **number of comparisons** is  **$n$**  with the **probability of such a search being  $(1-p)$** .

- Therefore, 
$$\begin{aligned} C_{avg}(n) &= \left[ 1 \cdot \frac{p}{n} + 2 \cdot \frac{p}{n} + \cdots + i \cdot \frac{p}{n} + \cdots + n \cdot \frac{p}{n} \right] + n \cdot (1 - p) \\ &= \frac{p}{n} [1 + 2 + \cdots + i + \cdots + n] + n(1 - p) \\ &= \frac{p}{n} \frac{n(n+1)}{2} + n(1 - p) = \frac{p(n+1)}{2} + n(1 - p). \end{aligned}$$

# Brute Force

---

- ***Brute force*** is a straightforward approach to solving a problem, usually directly based on the problem statement and definitions of the concepts involved.
- ***Brute force*** = “just do it”
- Often, the brute-force strategy is indeed the one that is easiest to apply.

# Brute-Force Algorithm: Exponentiation

## ALGORITHM Exponentiation ( b, n )

// Compute  $b^n = b \times b \times \dots \times b$  (total n times)

// Input: a base b which is an integer and an exponent n which is a positive integer

// Output: return the result of b to the power n,  $b^n$

result  $\leftarrow$  1

for (i = 1; i <= n; i++) do

    result  $\leftarrow$  result \* b

return result

consider the multiplication to be  
the algorithm's **basic operation**

ลูปนี้ทำงาน n รอบ

แต่ละรอบทำงาน 'คูณ' 1 ครั้ง

$$C(n) = \sum_{i=1}^n 1 = n$$

$$\therefore \text{big-Oh} = O(n)$$

# Brute-Force Algorithm: gcd(m, n) : the Consecutive Integers Checking algorithm

ALGORITHM gcd ( m, n )

// Compute the Greatest Common Divisor (GCD) for two non-negative, not both zero

// integers **m** and **n**

// Input: two non-negative, not both zero integers called m and n AND  $m \geq n$

// Output: return gcd(m, n)

If **n** = 0 Return **m** and stop.

While **n** > 1 do

Begin

If **m** is divisible by **n** Return **n** and stop.

Otherwise, decrease **n** by one.

End

Return **n**.

consider the division to be the  
algorithm's basic operation

ลูปนี้ทำงาน n รอบ (worst-case)

แต่ละรอบทำงาน 'หาร' 1 ครั้ง

$$C(n) = \sum_{i=1}^n 1 = n$$

$$\therefore \text{big-Oh} = O(n)$$



# Brute-Force Algorithm: Matrix Multiplication

$$\begin{array}{ccc} A & B & C \\ \text{row } i \left[ \begin{array}{|c|c|c|c|c|} \hline \square & \square & \square & \square & \square \\ \hline \end{array} \right] * \left[ \begin{array}{|c|} \hline \square \\ \square \\ \square \\ \square \\ \square \\ \hline \end{array} \right] \text{col. } j & = & \left[ \begin{array}{|c|} \hline C[i,j] \\ \hline \end{array} \right] \end{array}$$

$$C[i, j] = A[i, 0]B[0, j] + \dots + A[i, k]B[k, j] + \dots + A[i, n-1]B[n-1, j]$$

where  $0 \leq i, j \leq n-1$ .

# Brute-Force Algorithm: Matrix Multiplication (2)

**ALGORITHM** *MatrixMultiplication*( $A[0..n-1, 0..n-1]$ ,  $B[0..n-1, 0..n-1]$ )  
//Multiplies two square matrices of order  $n$  by the definition-based algorithm  
//Input: Two  $n \times n$  matrices  $A$  and  $B$   
//Output: Matrix  $C = AB$   
**for**  $i \leftarrow 0$  **to**  $n - 1$  **do**  
    **for**  $j \leftarrow 0$  **to**  $n - 1$  **do**  
         $C[i, j] \leftarrow 0.0$   
        **for**  $k \leftarrow 0$  **to**  $n - 1$  **do**  
             $C[i, j] \leftarrow C[i, j] + A[i, k] * B[k, j]$   
**return**  $C$

consider the multiplication to be the algorithm's **basic operation**



The measurement of the input size is 'n' (the matrix order)

## Brute-Force Algorithm: Matrix Multiplication (3)

- just **one multiplication** is made **for each repetition of the innermost loop**, i.e., for each value of the loop's variable **k** between its limits **0** and **n-1**
- that is, the number of **multiplications** made for every specific pair of variables **i** and **j** is

$$\sum_{k=0}^{n-1} 1$$

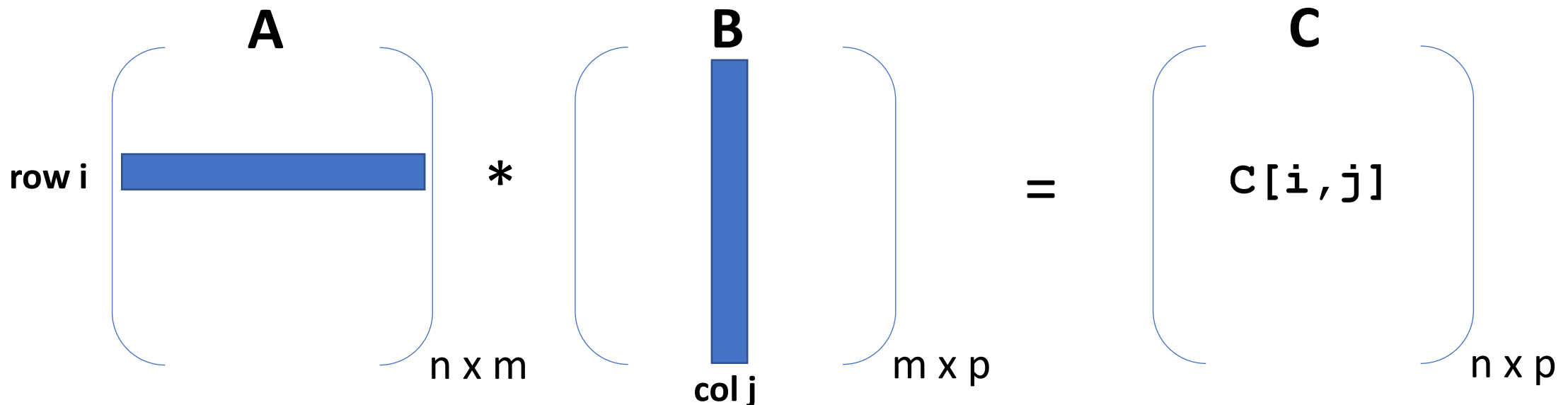
- the total number of **multiplications** **M(n)** is expressed by the following:

$$M(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \sum_{k=0}^{n-1} 1 = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} n = \sum_{i=0}^{n-1} n^2 = n^3 \in \Theta(n^3)$$

# Brute-Force Algorithm: Exercise 1

**Matrix Multiplication:** given one  $n \times m$  matrix (called  $A$ ) and one  $m \times p$  matrix (called  $B$ ),

- write a brute-force version of MatrixMultiplication algorithm for computing their product  $C = A \times B$  and
- find the time efficiency of such definition-based algorithm



# Brute-Force and The Sorting Problem

---

the application of the brute-force approach to the problem of sorting

The Sorting Problem: given a list of  $n$  orderable items, re-arrange them in non-decreasing order

- What would be the most straight-forward method for solving the sorting problem?

# Brute-Force and The Sorting Problem:

## Sorting using Exhaustive Search

---

- This approach generates all permutations of the  $n$  orderable items
- Then, check for the one satisfying the non-decreasing condition

# Brute-Force and The Sorting Problem:

## Selection Sort

- The **selection sort algorithm** sorts an array by repeatedly finding the minimum element from unsorted part and putting it at the beginning.
- The algorithm maintains two subarrays in the given array.
  - 1) The subarray which is already sorted.
  - 2) Remaining subarray which is unsorted.
- In every iteration of selection sort, the minimum element from the unsorted subarray is picked and moved to the end of sorted subarray.
- After  $n-1$  passes, the list is sorted

$$\begin{array}{ccc} & & \begin{array}{c} \text{-----} \\ \downarrow \qquad \downarrow \end{array} \\ A_0 \leq A_1 \leq \dots \leq A_{i-1} & | & A_i, \dots, A_{min}, \dots, A_{n-1} \\ \text{in their final positions} & & \text{the last } n - i \text{ elements} \end{array}$$

# Brute-Force and The Sorting Problem:

## Selection Sort : Example

	89	45	68	90	29	34	<b>17</b>
17		45	68	90	<b>29</b>	34	89
17	29		68	90	45	<b>34</b>	89
17	29	34		90	<b>45</b>	68	89
17	29	34	45		90	<b>68</b>	89
17	29	34	45	68		90	<b>89</b>
17	29	34	45	68	89		90



# Brute-Force and The Sorting Problem:

## Selection Sort (3)

---

**ALGORITHM** *SelectionSort*( $A[0..n - 1]$ )

*//Sorts a given array by selection sort*

*//Input: An array  $A[0..n - 1]$  of orderable elements*

*//Output: Array  $A[0..n - 1]$  sorted in nondecreasing order*

**for**  $i \leftarrow 0$  **to**  $n - 2$  **do**

$min \leftarrow i$

**for**  $j \leftarrow i + 1$  **to**  $n - 1$  **do**

**if**  $A[j] < A[min]$   $min \leftarrow j$

    swap  $A[i]$  and  $A[min]$

# Brute-Force and The Sorting Problem:

## Selection Sort (4)

- the input size is given by the number of elements in the list  $n$
- the basic operation is the key comparison  $A[j] < A[min]$

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] = \sum_{i=0}^{n-2} (n-1-i).$$

$$= (n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 = \sum_{i=1}^{n-1} i = \frac{(n-1)n}{2}$$

$$\in O(n^2)$$

# Brute-Force and The Sorting Problem: Bubble Sort

- Bubble Sort is the simplest sorting algorithm that works by repeatedly swapping the adjacent elements if they are in wrong order
- Pass  $i$  ( $0 \leq i \leq n - 2$ ) of bubble sort can be represented by the following diagram:

$$A_0, \dots, A_j \overset{?}{\leftrightarrow} A_{j+1}, \dots, A_{n-i-1} \mid A_{n-i} \leq \dots \leq A_{n-1}$$

in their final positions

# Brute-Force and The Sorting Problem:

## Bubble Sort (2)

- Example: sort the list (5, 1, 4, 2, 8) in non-decreasing order  
(เรียงจากน้อยไปมาก)

### First Pass:

( **5** 1 4 2 8 )  $\rightarrow$  ( 1 **5** 4 2 8 ),  
( 1 **5** 4 2 8 )  $\rightarrow$  ( 1 4 **5** 2 8 ),  
( 1 4 **5** 2 8 )  $\rightarrow$  ( 1 4 2 **5** 8 ),  
( 1 4 2 **5** 8 )  $\rightarrow$  ( 1 4 2 5 **8** ),

### Second Pass:

( 1 **4** 2 5 8 )  $\rightarrow$  ( 1 4 2 5 8 )  
( 1 **4** 2 5 8 )  $\rightarrow$  ( 1 2 **4** 5 8 ),  
( 1 2 **4** 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 4 **5** 8 )  $\rightarrow$  ( 1 2 4 5 8 )

### Third Pass:

( 1 2 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )

# Brute-Force and The Sorting Problem:

## Bubble Sort (3)

**ALGORITHM** *BubbleSort*( $A[0..n - 1]$ )

*//Sorts a given array by bubble sort*

*//Input: An array  $A[0..n - 1]$  of orderable elements*

*//Output: Array  $A[0..n - 1]$  sorted in nondecreasing order*

**for**  $i \leftarrow 0$  **to**  $n - 2$  **do**

**for**  $j \leftarrow 0$  **to**  $n - 2 - i$  **do**

**if**  $A[j + 1] < A[j]$  swap  $A[j]$  and  $A[j + 1]$

- consider key comparison and key swap as the basic operations of the algorithm
- consider the worst-case where key swaps are done at every iteration

# Brute-Force and The Sorting Problem:

## Bubble Sort (4)

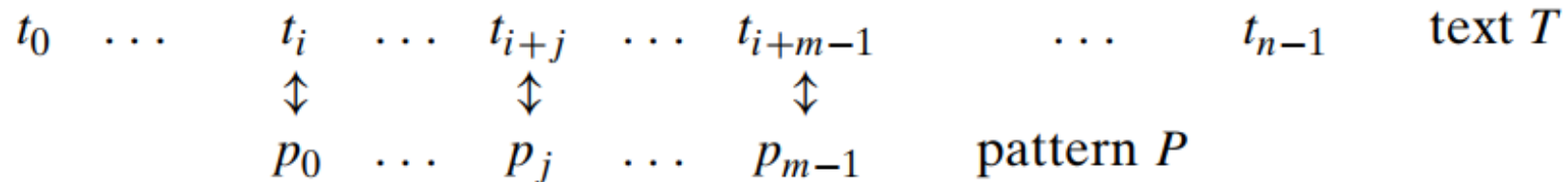
$$\begin{aligned} C(n) &= \sum_{i=0}^{n-2} \sum_{j=0}^{n-2-i} 1 = \sum_{i=0}^{n-2} (n-1-i) \\ &= (n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 = \sum_{i=1}^{n-1} i = \frac{(n-1)n}{2} \in O(n^2) \end{aligned}$$

$$S_{worst}(n) = C(n) \in O(n^2)$$

The overall time complexity of the algorithm is  $O(n^2)$

# Brute-Force String Matching

- given a string of  $n$  characters called the ***text*** and a string of  $m$  characters ( $m \leq n$ ) called the ***pattern***, find a substring of the text that matches the pattern.



## Brute-Force String Matching (2)

[illegible]

- Pattern = “NOT”



# Brute-Force String Matching (3)

**ALGORITHM** *BruteForceStringMatch*( $T[0..n - 1]$ ,  $P[0..m - 1]$ )

//Implements brute-force string matching

//Input: An array  $T[0..n - 1]$  of  $n$  characters representing a text and

// an array  $P[0..m - 1]$  of  $m$  characters representing a pattern

//Output: The index of the first character in the text that starts a

// matching substring or  $-1$  if the search is unsuccessful

**for**  $i \leftarrow 0$  **to**  $n - m$  **do**

$j \leftarrow 0$

**while**  $j < m$  **and**  $P[j] = T[i + j]$  **do**

$j \leftarrow j + 1$

**if**  $j = m$  **return**  $i$

**return**  $-1$

} comparisons between corresponding  
characters in Pattern and current Text window

- The basic operation is the comparison between each character in Pattern and those in the current text window.
- worst-case scenario: we need to make all  $m$  comparisons prior to realizing that this is not a match for every text window

# Brute-Force String Matching (4)

---

- $C(n) = \sum_{i=0}^{n-m} (m) = m (n - m - 0 + 1)$   
 $= nm - m^2 + m$
- Since  $m \leq n$ ,  $C(n) \in O(n m)$

## Exercise 1: Insertion Sort

---

- Given a list of  $N$  integers, sort the list in non-decreasing order.

## Exercise 2: Prime Numbers

---

- Given a positive integer  $N > 1$ , check whether  $N$  is a prime number.

### Definitions:

- I. A prime number is a whole number greater than 1 that **cannot** be made by multiplying other whole numbers.
- II. A prime number is a whole number greater than 1, that has only two factors , i.e. 1 and the number itself.
- III. A prime number is divisible only by the number 1 or itself.

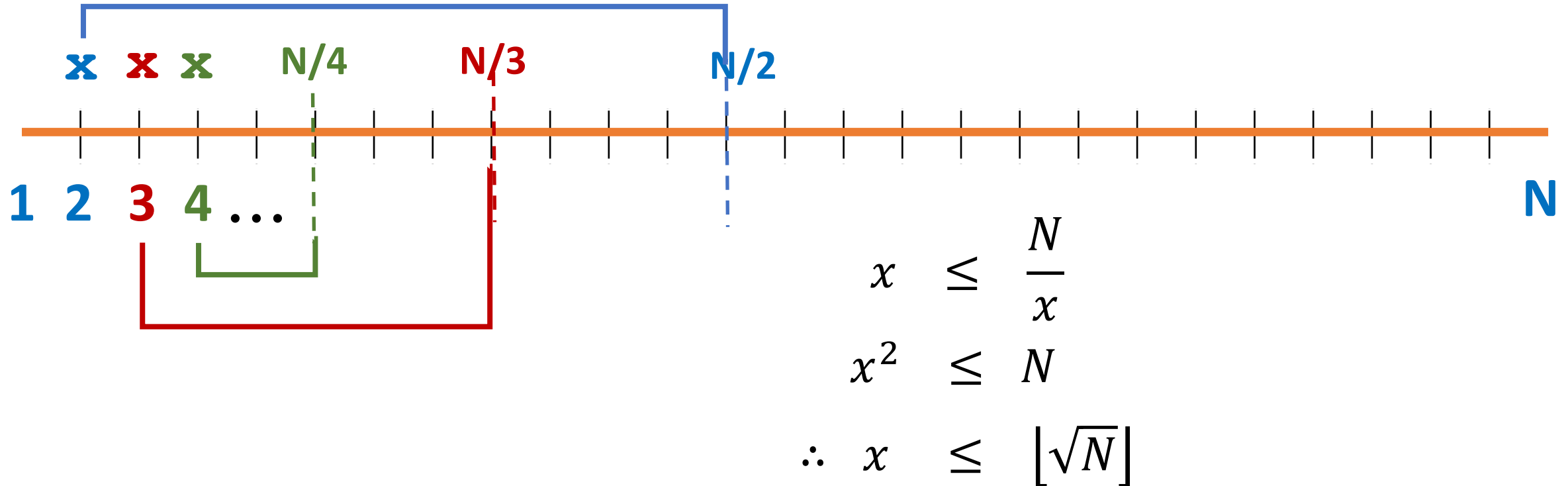
## Exercise 2: Prime Numbers: Method I

- from the definition of  $N$  being a prime number, we test whether  $N$  is divisible by another integer  $x$
- what do we know about the integer  $x$  ?
  - $x$  must be less than  $N$ .
  - ถ้าไม่นับรวม 1 ตัวประกอบที่เล็กที่สุดที่เป็นได้สำหรับ  $N$  คือ 2 (The smallest  $x$  is 2).
- so, we set  $x$  from 2 to  $N-1$ .
- Time complexity =  $O(N)$

## Exercise 2: Prime Numbers: Method II

- from the definition of  $N$  being a prime number, we test whether  $N$  is divisible by another integer  $x$
- what do we know about the integer  $x$  ?
  - $x$  must be less than  $N$ .
  - ถ้าไม่นับรวม 1 ตัวประกอบที่เล็กที่สุดที่เป็นไปได้สำหรับ  $N$  คือ 2 (The smallest  $x$  is 2).
  - **what else do we know?**
    - ดังนั้น ตัวประกอบที่ใหญ่ที่สุดที่เป็นไปได้สำหรับ  $N$  คือ  $N/2$
- so, we know that we can set  $x$  from 2 to  $N/2$
- Time complexity still =  $O(N)$

## Exercise 2: Prime Numbers: Method III



- So, we can set  $x$  from 2 to  $\lfloor \sqrt{N} \rfloor$
- Time complexity is now  $O(\sqrt{N})$

## Exercise 3: Nugget Numbers

- ร้านฟาสต์ฟู้ดแห่งหนึ่งขายนักเก็ตเป็นกล่อง โดยมีกล่องนักเก็ตอยู่ 3 ขนาด ได้แก่ ขนาดเล็ก ขนาดกลาง และขนาดใหญ่ บรรจุนักเก็ต จำนวน 6 ชิ้น, 9 ชิ้น, และ 20 ชิ้น ตามลำดับ
- เลขนักเก็ต คือ เลขจำนวนเต็มบวกที่เกิดจากผลรวมของจำนวนชิ้นนักเก็ตในกล่องขนาดต่าง ๆ ยกตัวอย่างเช่น
  - เลข 6 เป็นเลขนักเก็ต เพราะเป็นจำนวนนักเก็ตในกล่องขนาดเล็กหนึ่งกล่อง
  - เลข 9 เป็นเลขนักเก็ต เพราะเป็นจำนวนนักเก็ตในกล่องขนาดกลางหนึ่งกล่อง
  - เลข 12 เป็นเลขนักเก็ต เพราะเกิดจากการรวมกันของจำนวนนักเก็ตในกล่องขนาดเล็ก 2 กล่อง
  - เลข 15 เป็นเลขนักเก็ต เพราะเกิดจากการรวมกันของจำนวนนักเก็ตในกล่องขนาดเล็ก และ ขนาดกลาง อย่างละหนึ่งกล่อง
  - เลข 20 เป็นเลขนักเก็ต เพราะเป็นจำนวนนักเก็ตในกล่องขนาดใหญ่หนึ่งกล่อง
  - เลข 4 และ 10 ไม่เป็นเลขนักเก็ต
- ให้นักศึกษาเขียนโปรแกรมเพื่อรับเลขจำนวนเต็มบวก  $N \geq 6$  และคำนวณพร้อมแสดงผลลัพธ์เป็นเลขนักเก็ตทั้งหมดที่มีค่าน้อยกว่าหรือเท่ากับ  $N$  แบบเรียงตามลำดับจากน้อยไปมาก



## Exercise 3: Nugget Numbers (2)

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- **Challenge**: เขียนโปรแกรมเพื่อรับค่าเลขจำนวนเต็มบวก **N** และแสดงผลลัพท์ว่า **N** เป็นเลขนิกเก็ตหรือไม่?

# Brute-Force: Exhaustive Search

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- *Exhaustive search* is simply a brute-force approach to combinatorial problems.
- It suggests generating each and every element of the problem domain, selecting those of them that satisfy all the constraints, and then finding a desired element (e.g., the one that optimizes some objective function).
- Traveling Salesman Problem, Knapsack Problem, Assignment Problem

# Exhaustive Search:

## Traveling Salesman Problem

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- Given a set of  $n$  cities, find the shortest tour through the whole city set such that the salesman visits each city only once before returning to the city where the tour started

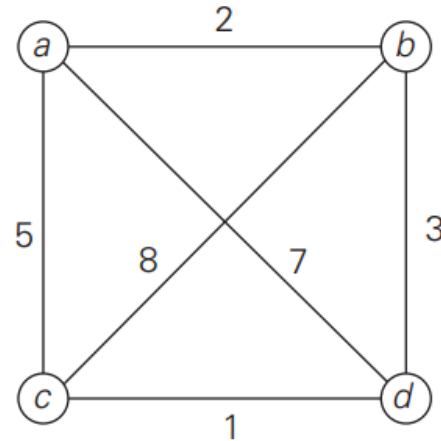
# Exhaustive Search:

## Traveling Salesman Problem (2)

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- The problem can be modeled a weighted undirected graph  $G=(V, E)$ :
  - $V$  represents the set of cities
  - $E$  maintains the road information between cities
  - Each edge in  $E$  has an associated weight specifying distance between two cities
- Then, the problem can be stated as the problem of finding the shortest *Hamiltonian circuit* of the graph.
- a Hamiltonian circuit is defined as a cycle that passes through all the vertices of the graph exactly once.

# Exhaustive Search: Traveling Salesman Problem (3)



Tour

Length

$a \rightarrow b \rightarrow c \rightarrow d \rightarrow a$	$l = 2 + 8 + 1 + 7 = 18$	
$a \rightarrow b \rightarrow d \rightarrow c \rightarrow a$	$l = 2 + 3 + 1 + 5 = 11$	optimal
$a \rightarrow c \rightarrow b \rightarrow d \rightarrow a$	$l = 5 + 8 + 3 + 7 = 23$	
$a \rightarrow c \rightarrow d \rightarrow b \rightarrow a$	$l = 5 + 1 + 3 + 2 = 11$	optimal
$a \rightarrow d \rightarrow b \rightarrow c \rightarrow a$	$l = 7 + 3 + 8 + 5 = 23$	
$a \rightarrow d \rightarrow c \rightarrow b \rightarrow a$	$l = 7 + 1 + 8 + 2 = 18$	

# Exhaustive Search: Knapsack Problem

- Given  $n$  items of known weights  $w_1, w_2, \dots, w_n$  and values  $v_1, v_2, \dots, v_n$  and a knapsack of capacity  $W$ , find the most valuable subset of the items that fit into the knapsack.
- Example:  $W = 10$

	weight	value
1	7	\$42
2	3	\$12
3	4	\$40
4	5	\$25

# Exhaustive Search: Knapsack Problem (2)

	weight	value
1	7	\$42
2	3	\$12
3	4	\$40
4	5	\$25

- list all combinations:  
 $C(n,0) + C(n,1) + C(n,2) + \dots + C(n,n)$
- Ex  $n = 4$   
all combinations =  $1 + 4 + 6 + 4 + 1$   
= 16 possible patterns

Subset	Total weight	Total value
$\emptyset$	0	\$ 0
{1}	7	\$42
{2}	3	\$12
{3}	4	\$40
{4}	5	\$25
{1, 2}	10	\$54
{1, 3}	11	not feasible
{1, 4}	12	not feasible
{2, 3}	7	\$52
{2, 4}	8	\$37
<b>{3, 4}</b>	<b>9</b>	<b>\$65</b>
{1, 2, 3}	14	not feasible
{1, 2, 4}	15	not feasible
{1, 3, 4}	16	not feasible
{2, 3, 4}	12	not feasible
{1, 2, 3, 4}	19	not feasible



# Exhaustive Search: Assignment Problem

- There are  $n$  people who need to be assigned to execute  $n$  jobs, one person per job. (That is, each person is assigned to exactly one job and each job is assigned to exactly one person.)
- The cost that would accrue if the  $i^{\text{th}}$  person is assigned to the  $j^{\text{th}}$  job is a known quantity  $C[i, j]$  for each pair  $i, j = 1, 2, \dots, n$ .
- The problem is to find an assignment with the minimum total cost.

# Exhaustive Search: Assignment Problem (2)

## Example $n = 4$

- the exhaustive-search approach to the assignment problem would require **generating all the permutations** of integers  $1, 2, \dots, n$ ,
- computing the total cost of each assignment by summing up the corresponding elements of the cost matrix,
- and finally selecting the one with the smallest sum.

	Job 1	Job 2	Job 3	Job 4
Person 1	9	2	7	8
Person 2	6	4	3	7
Person 3	5	8	1	8
Person 4	7	6	9	4