

Algorithm Design and Analysis

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名

姓

算法设计与分析

詹博华，杨大卫

This week's content

- Today Wednesday:
 - Introduction of teachers
 - Grading rules
 - Why study algorithms?
Chapter 3: Growth of functions
 - Exercises
- Tomorrow Thursday:
 - Exercise solutions
 - Chapter 4: Divide and Conquer
Introduction of the remaining content

这周的内容

- 今天周三：
 - 老师介绍
 - 课程评分规则
 - 为什么学习算法?
第三章：函数的增长
 - 函数的增长的练习
- 明天周四：
 - 练习题解答
 - 第四章：分治策略
以后内容的介绍

David N. Jansen

名

姓

杨大卫

- Swiss and Dutch
- Mathematics Master in Switzerland
- Computer Science Ph. D. in the Netherlands
- worked later in Germany, Netherlands, China (ISCAS)
- interested in formal verification. also interested in philosophy of science

- 瑞士人和同时荷兰人
- 在瑞士获得数学硕士学位
- 在荷兰获得计算机科学博士学位

prove correctness
of algorithms with
mathematical proof

Can
science answer all
questions?

"Will you
marry me?"

- 以后工作在德国，荷兰，中国（中科院软件研究所）
- 研究方向：形式化验证，也对科学的哲学感兴趣

用数学证明
算法的正确性

科学有能力解决
所有的问题吗？

你愿意嫁
给我吗？

詹博华的介绍

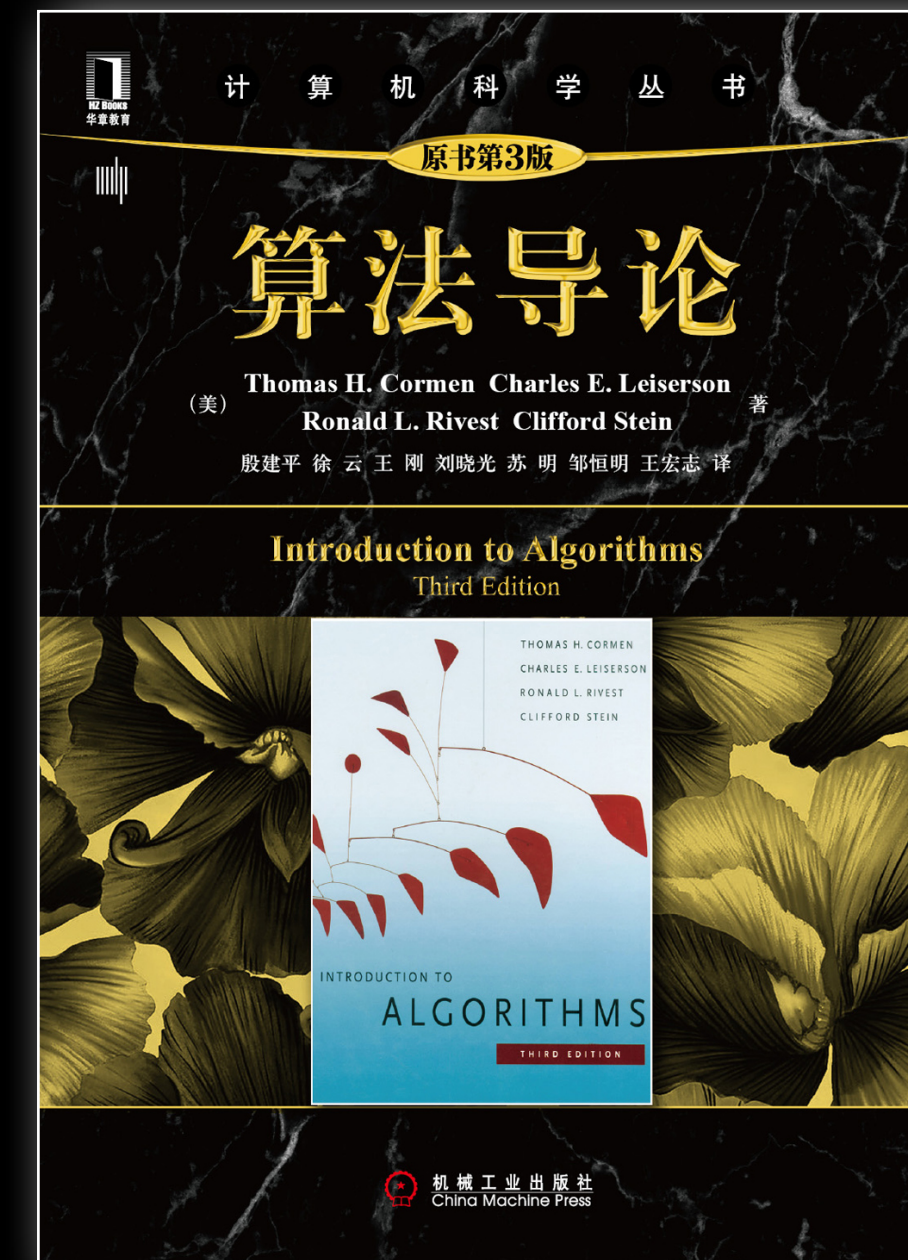
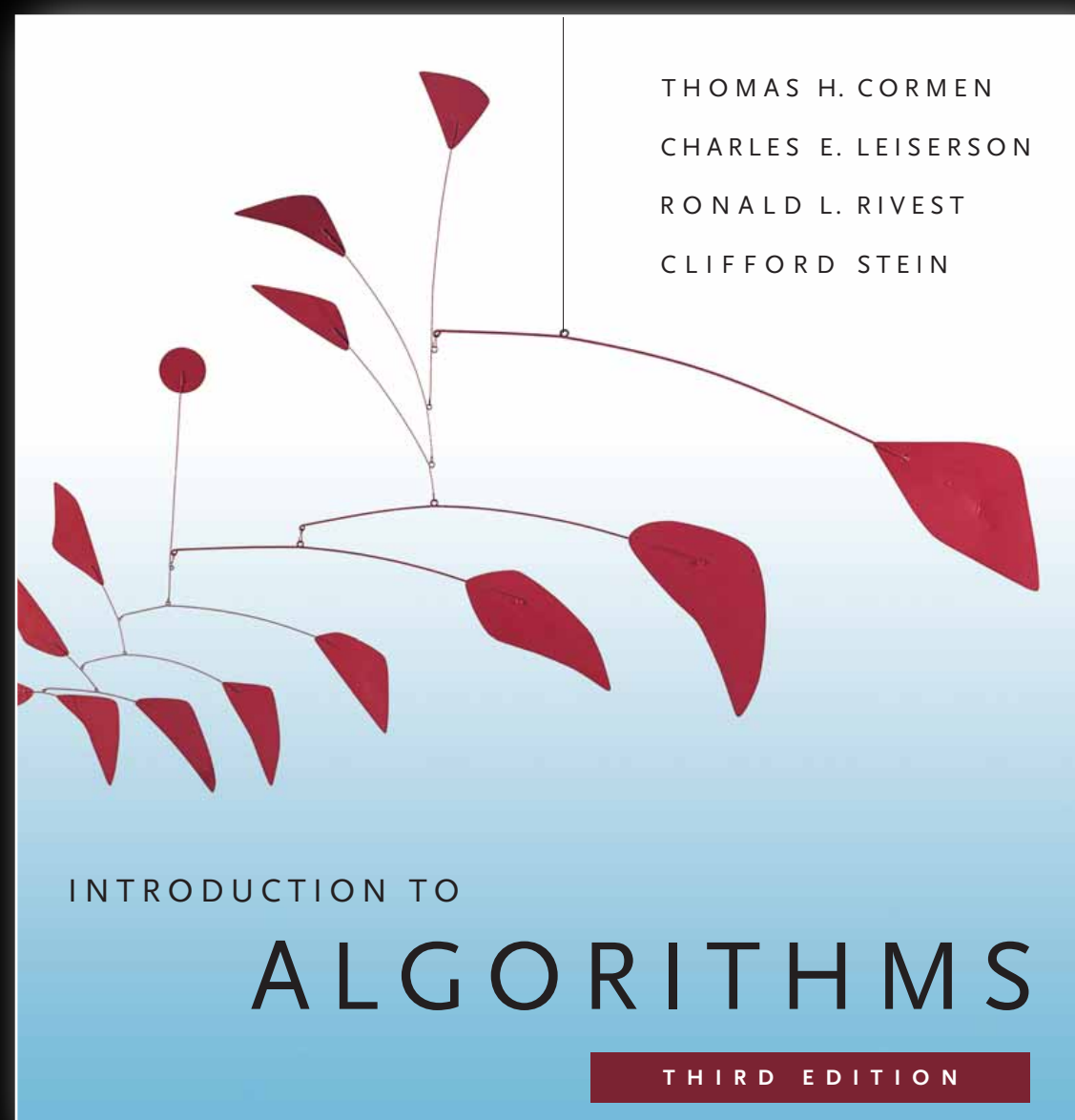
- 在美国获得数学学士和博士学位
- 2018年回国，加入中科院软件所，现任副研究员
- 研究方向：形式化方法、交互式定理证明、嵌入式系统的建模与验证

课程评分规则

- 40%大作业、60%期末考试
- 定期布置（不计分的）作业，用于熟悉考试内容和题目形式

Book

- We will follow this book:
Cormen / Leiserson / Rivest / Stein: Introduction to Algorithms.
3rd edition. MIT Press 2009.



Open office

- I want to offer a time to ask questions every week.
Suggested time: Thursday after the lecture.

开放时间

- 每周有时间可以问我问题。
建议：周四下课后

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**Why study
Algorithms?**

为什么学习算法？

What is an algorithm?

- a precise sequence of instructions that transform an input into an output
- satisfies some specification / goal
- The word “algorithm” is derived from the Persian mathematician Al Khwārizmī
- can be communicated in natural language, but often written as pseudocode

算法是什么？

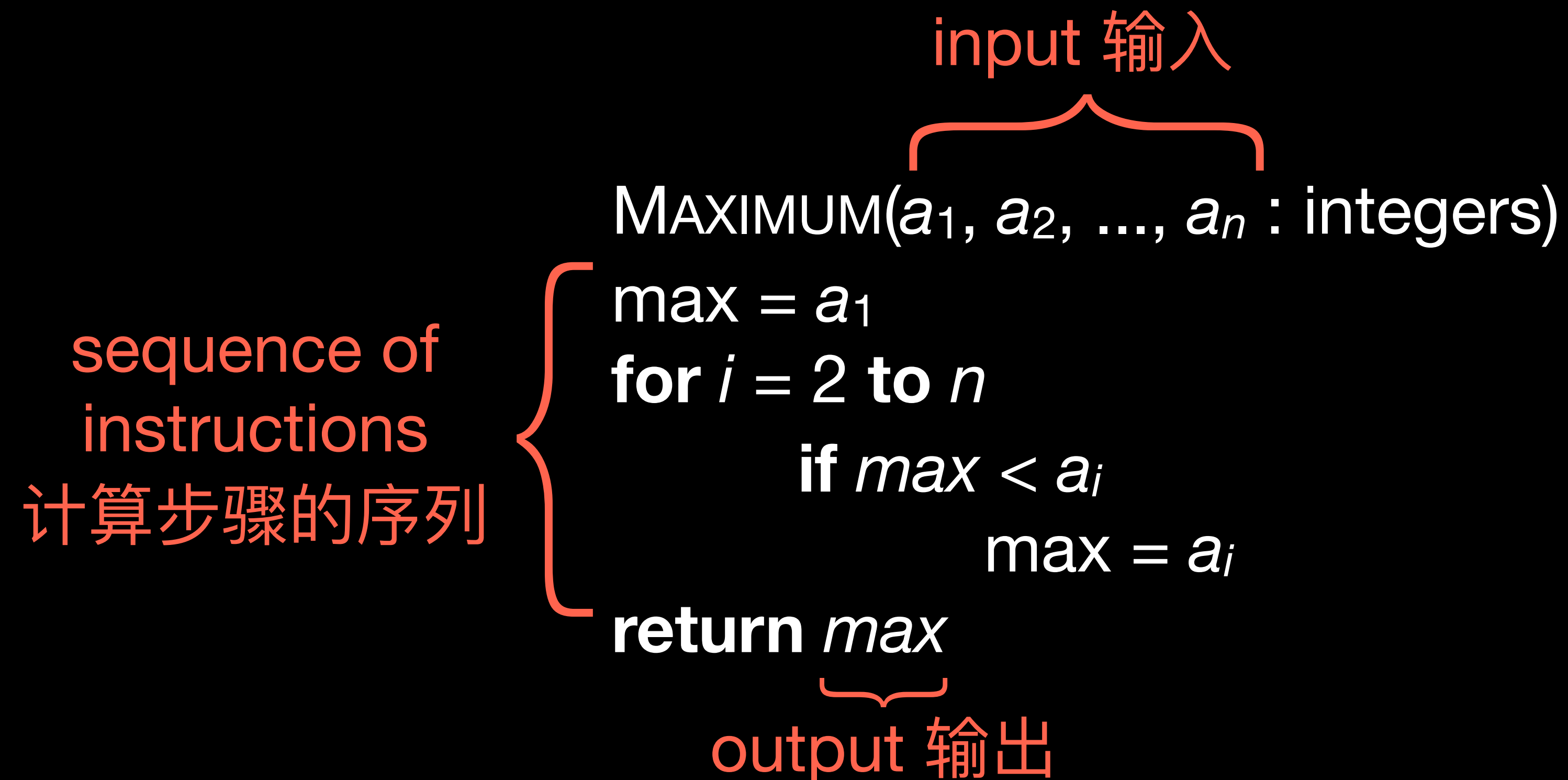
- 把输入转换成输出的计算步骤的序列
- 满足需求/目标
- “algorithm”词从波斯的数学家Al Khwārizmī
- 可以使用自然语言沟通算法，但通常也使用伪代码

Example: maximum

例子：最大元素

- specification: find the largest number in a sequence of integers

- 需求：
找到一个整数的序列的最大的元素



Properties of Algorithms

- definite: every step of the algorithm is defined precisely
 - correct: produces the correct output for every input
 - general: applicable to all problems of a certain class
 - finite: for each input, it terminates in a finite number of steps
- 确定的
 - 正确的
 - 一般的
 - 有限的

Why study algorithms?

- Need to solve problems correctly:
good algorithm has proof of correctness
- Need to solve problems efficiently:
good algorithm is fast /
uses little memory

A little history

- algorithms existed before computers
 - Babylonian algorithm:
2500 BC, division
 - Euclid's algorithm:
300 BC, greatest common divisor
- Many new algorithms with computers
 - bubble sort → quick sort
- 算法计算机前已有
 - 巴比伦：2500 BC，出发
 - 欧几里得：300BC，最大公因子
- 冒泡排序 → 快速排序

Growth of Functions (Big-O Notation)

Chapter 3

函数的增长 (大O记号)

第三章

Insertion Sort 插入排序

- animation with numbered cards

Insertion Sort

INSERTION-SORT(array A)

for $j := 2$ **to** $A.length$

$key = A[j]$

$i = j - 1$

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

$A[i+1] = A[i]$

$i = i - 1$

$A[i + 1] = key$

sequence $A[1 \dots j-1]$
is sorted

select an element
to be sorted

Insert the element
into the sorted sequence
 $A[1 \dots j-1]$

sequence $A[1 \dots j]$
is sorted

Insertion Sort: Loop Invariant

- Correctness proof of an algorithm with loops:
uses a loop invariant 环不变式.
- **Initialisation:** When the loop starts, the loop invariant holds.
- **Maintenance:** If the loop invariant holds at the beginning of the loop body, then it holds at the end of the loop body.
- **Termination:** When the loop terminates, the property required after the loop holds.

Insertion Sort: Loop Invariant

- Loop invariant for insertion sort: At the beginning of an iteration, the sequence $A[1 \dots j-1]$ is sorted.
- **Initialisation:** When the loop starts ($j = 2$), $A[1 \dots 1]$ is sorted.
- **Maintenance:** If at the beginning of the loop body $A[1 \dots j-1]$ is sorted, then $A[1 \dots j]$ is sorted at the end of the loop body.
- **Termination:** When the loop terminates ($j = n$), $A[1 \dots n]$ is sorted.



property at the end of the loop body

Insertion Sort: Running Time 运行时间

- We want to know how long INSERTION-SORT takes, depending on $A.length$.
- Problem: operations on different computers take different time.
- Simplification: only count operations (we will see later that it's ok.)

Insertion Sort

```
INSERTION-SORT(array A)
for  $j := 2$  to  $A.length$ 
     $key = A[j]$ 
     $i = j - 1$ 
    while  $i > 0$  and  $A[i] > key$ 
         $A[i+1] = A[i]$ 
         $i = i - 1$ 
     $A[i + 1] = key$ 
```

This line is executed $\leq \dots$ times

$n = A.length$

$n-1$

$n-1$

$2 + 3 + \dots + n$

$1 + 2 + \dots + (n-1)$

$1 + 2 + \dots + (n-1)$

$n-1$

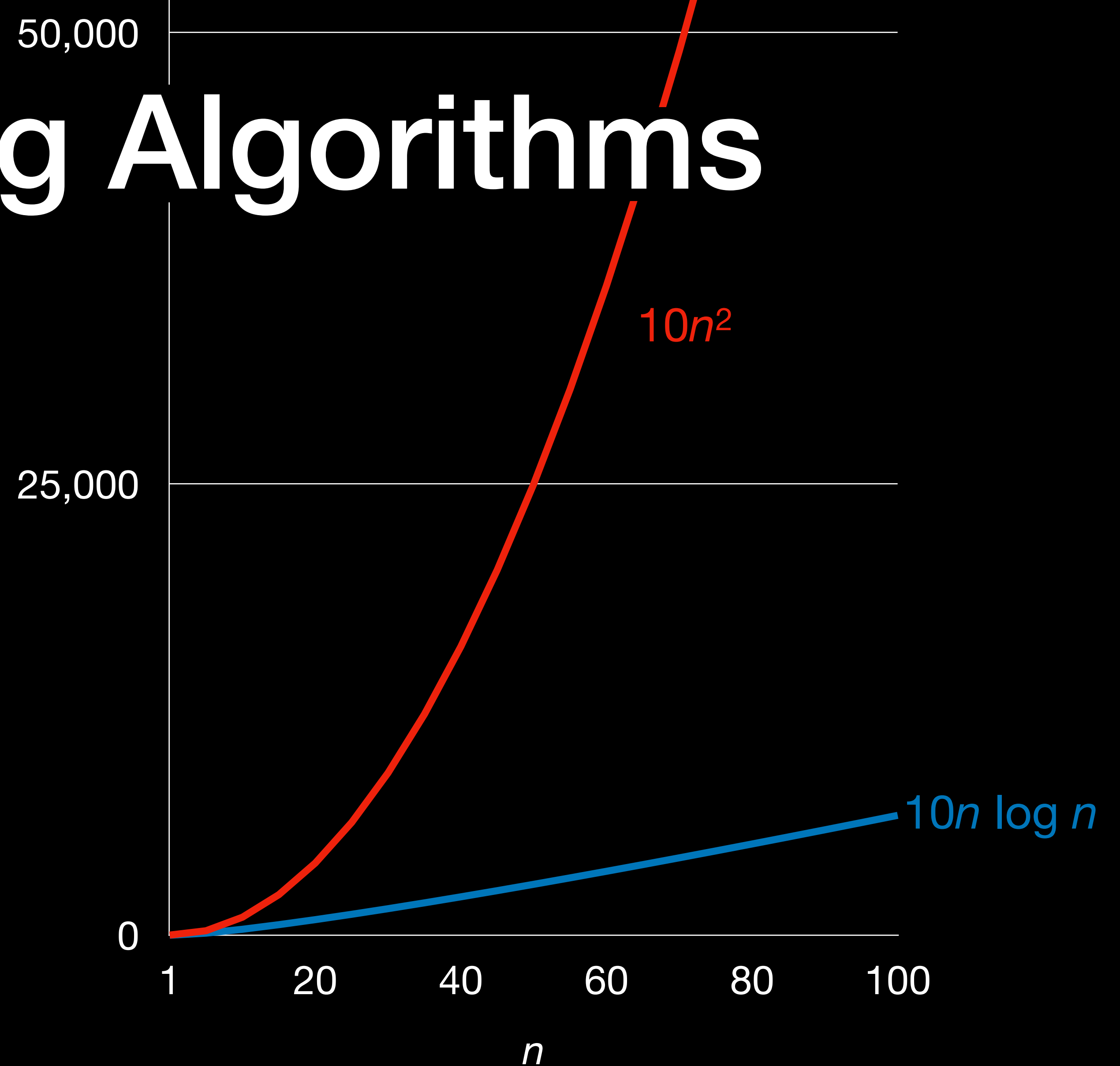
Running time is important if $n = A.length$ is large, but then $n \ll n^2$

Total $\leq 1\frac{1}{2}n^2 + 3\frac{1}{2}n - 4$ times

See next slide

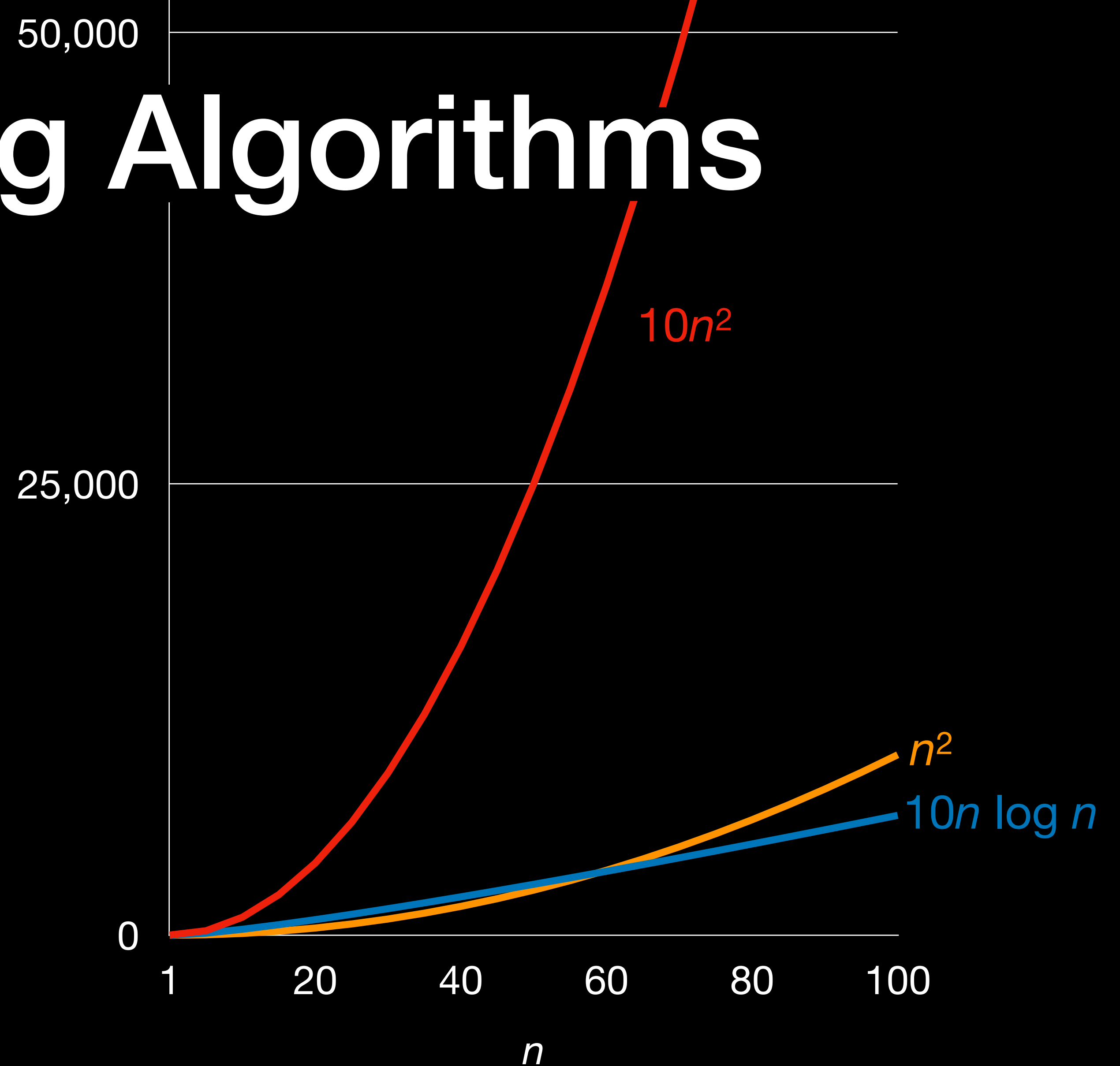
Comparing Algorithms

- Example:
 - **algorithm A** executes $\approx 10n^2$ instructions.
 - **algorithm B** executes $\approx 10n \log n$ instructions.
- Now let's make algorithm A 10× faster.



Comparing Algorithms

- Example:
 - **algorithm A** executes $\approx 10n^2$ instructions.
 - **algorithm B** executes $\approx 10n \log n$ instructions.
- Now let's make algorithm A 10× faster.
 - ➔ For $n \geq 59$, **algorithm B** is still faster than the **improved algorithm A**.



Comparing Algorithms

- For large inputs, growth rate of the function is most important.
(For small inputs, most algorithms are fast.)
- Big-O notation: brief notation to indicate the growth rate of a function.

Big-O Notation

- Notation to compare **asymptotic growth rate of functions**
(Which function $\mathbb{N} \rightarrow \mathbb{R}^+$ is larger, for large parameters?)
- used to describe order of growth of algorithms
(Which algorithm is faster for large input sizes?)

Big-O Notation

- $O(g(n))$ is a set of functions
all functions that do not grow faster than $g(n)$
 $\{ f(n) \mid \exists c, n_0: \forall n \geq n_0: 0 \leq f(n) \leq cg(n) \}$
- $O()$ gives an upper bound on a function.

Big-O Notation

- $n = O(n^2)$.
- $2n = O(n)$.
- $n = O(2n - 4)$.
- The worst-case running time of insertion sort: $1\frac{1}{2}n^2 + 3\frac{1}{2}n - 4$.
This time is in $O(n^2)$.
It is also in $O(n^3)$.
It is not in $O(n \log n)$.

Big- Ω Notation

- $\Omega(g(n))$ is a set of functions
all functions that do not grow slower than $g(n)$
 $\{ f(n) \mid \exists c, n_0: \forall n \geq n_0: 0 \leq cg(n) \leq f(n) \}$
- $\Omega()$ gives a lower bound on a function.

Big- Ω Notation

- $n^2 = \Omega(n)$.
- $n = \Omega(2n + 5)$.
- The best-case running time of insertion sort is $5n - 4$.
It is in $\Omega(n)$.
It is not in $\Omega(n^2)$.

$O, \Omega, \Theta, o, \omega$

Notation	English Pronunciation	Meaning	Formula
$O(g(n))$	(big) Oh of $g(n)$	functions that don't grow faster than g	$\{ f(n) \mid \exists c, n_0: \forall n \geq n_0: 0 \leq f(n) \leq cg(n) \}$
$\Omega(g(n))$	(big) Omega of $g(n)$	functions that grow at least as fast as g	$\{ f(n) \mid \exists c, n_0: \forall n \geq n_0: 0 \leq cg(n) \leq f(n) \}$
$\Theta(g(n))$	Theta of $g(n)$	functions that grow equally fast as g	$O(g(n)) \cap \Omega(g(n))$
$o(g(n))$	small oh of $g(n)$	functions that grow slower than g	$\{ f(n) \mid \forall c > 0: \exists n_0: \forall n \geq n_0: 0 \leq f(n) < cg(n) \}$
$\omega(g(n))$	small omega of $g(n)$	functions that grow faster than g	$\{ f(n) \mid \forall c > 0: \exists n_0: \forall n \geq n_0: 0 \leq cg(n) < f(n) \}$

Further Examples

- $n = \Theta(2n)$.
- $4n^3 + 2n^2 + 5 = \Theta(n^3)$.
- $n = o(n^2)$. $n \notin o(2n + 5)$.
- $n = \omega(\log n)$.

Facts about Big-O Notation

- Transitivity: $f(n) = O(g(n))$ and $g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$
 $f(n) = \Omega(g(n))$ and $g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$
also for Θ , o and ω
- Symmetry of Θ : $f(n) = \Theta(g(n)) \Leftrightarrow g(n) = \Theta(f(n))$
- Duality: $f(n) = O(g(n)) \Leftrightarrow g(n) = \Omega(f(n))$
- Be careful: $f(n) = O(g(n))$ and $f(n) \notin \Theta(g(n)) \not\Rightarrow f(n) = o(g(n))$

Example: $f(n) = n$ if n is odd 奇数 $g(n) = n^2$
 $= n^2$ if n is even 偶数

Big-O Notation for Functions

- When approximating a function, also use these notations.
- Example: number of comparisons in a certain sort algorithm
 $= 2n \log n + O(n)$.
 Meaning: $= 2n \log n + f(n)$, for some function $f(n) = O(n)$.
- Other uses: memory consumption of an algorithm
 general approximation of functions

Exercises 练习

- 1.2-2
Suppose we are comparing implementations of insertion sort and merge sort on the same machine. For inputs of size n , insertion sort runs in $8n^2$ steps, while merge sort runs in $64n \lg n$ steps. For which values of n does insertion sort beat merge sort?
- 假设我们正比较插入排序与合并排序在相同机器上的实现。对规模为 n 的输入，插入排序运行 $8n^2$ 步，而合并排序运行 $64n \lg n$ 步。问对哪些 n 值，插入排序优于合并排序？

Exercises 练习

Problem 1-1: Comparison of running times

For each function $f(n)$ and time t in the following table, determine the largest size n of a problem that can be solved in time t , assuming that the algorithm to solve the problem takes $f(n)$ microseconds (μsec).

思考题 1-1: 运行时间的比较

假设求解问题的算法需要 $f(n)$ 微妙，对下表中的每个函数 $f(n)$ 和时间 t ，确定可以在时间 t 内求解的问题的最大规模 n 。

$t = 1 \text{ sec}, 1 \text{ min}, 1 \text{ hour}, 1 \text{ day}, 1\text{month/月}, 1 \text{ year}, 100 \text{ years}$

$f(n) = \lg n, \sqrt{n}, n, n \lg n, n^2, n^3, 2^n, n! \quad (\lg n = \log_2 n)$

Exercises 练习

	1秒	1分	1小时	1天	1月	1年	1世纪
$\lg n$							
\sqrt{n}							
n							
$n \lg n$							
n^2							
n^3							
2^n							
$n!$							

Exercises 练习

- 2.1-3

Consider the searching problem:

Input: A sequence of n numbers $A = (a_1, a_2, \dots, a_n)$ and a value v .

Output: An index i such that $v = A[i]$, or the special value NIL if v does not appear in A .

Write pseudocode for linear search, which scans through the sequence, looking for v .

Using a loop invariant, prove that your algorithm is correct. Make sure that your loop invariant fulfills the three necessary properties.

- 考虑以下查找问题：

输入： n 个数的一个序列 $A = (a_1, a_2, \dots, a_n)$ 和一个值 v 。

输出： 下标 i 使得 $v = A[i]$ 或者当 v 不在 A 中出现时，特殊值 NIL。

写出线性查找的伪代码，它扫描整个序列来查找 v 。使用一个循环不变式来证明你的算法的正确性。确保你的循环不变式满足三条必要的性质。

Exercises 练习

- 3-2.3
Prove / 证明
 - $n! = \omega(2^n)$
 - $n! = o(n^n)$
 - $\log(n!) = \Theta(n \log n)$
- You may use Stirling's approximation / 可以使用斯特林近似公式:
$$n! = \sqrt{2\pi n} (n/e)^n (1 + \Theta(1/n))$$