Design and Analysis of Algorithms Lecture 1: Introduction



Ke Xu and Yongxin Tong (许可 与 童咏昕)

School of CSE, Beihang University

Outline

- About Us
- Course Details
- A.M. Turing Award Winners for Algorithms
- What Is This Course About
- What Are Algorithms
- What Does It Mean to Analyze An Algorithm
- Comparing Time Complexity

Instructors



Ke Xu Professor

State Key Lab. of Software Development Environment

Research Interests: Algorithms, Phase Transitions in NPC, Logic and Complexity, Data Mining



Yongxin Tong
Professor

State Key Lab. of Software Development Environment

Research Interests: Big Data, Crowd Intelligence, Data Mining, Federated/Reinforcement Learning

Contact and TAs

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TAs

- Yexuan Shi (Ph.D. Student)
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- Qiaoyang Liu(Master Student)
 - Email: qiaoyangliu@buaa.edu.cn
- Ruisheng Zhang (Master Student)
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- Xiaomin Pengmao (Senior Undergraduate Student)
 - Email: prime21@buaa.edu.cn

Instructors: Yongxin Tong

- Beihang University (2015.4 Current)
 - "Zhuoyue Program" Distinguished Researcher
 - State Key Lab. of Software Development Environment
 - Research Interests: Big Data and Crowd Intelligence

- HKUST (2010.8 2015.3)
 - Research Assistant Professor (2014.2 2015.3)
 - CSE Department, focused on big data and crowdsourcing
 - Ph.D. Student and Candidate (2010.8 2014.1)
 - CSE Department, focused on data mining and crowdsourcing

Instructors: Yongxin Tong



Yongxin Tong 童 咏 昕

Professor
Big Data Analysis Group
State Key Laboratory of Software Development Environment
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[Short Bio] [Research] [Publications] [Awards] [Experiences] [Professional Services] [Misc.]

Short Biography

Yongxin Tong is a Professor in the <u>State Key Laboratory of Software Development Environment</u> (SKLSDE) of the <u>School of Computer Science and Engineering</u> at <u>Beihang University (BUAA)</u>. He received a Ph.D. degree in Computing Science and Engineering from the <u>Department of Computer Science and Engineering</u>, <u>The Hong Kong University of Science and Technology (HKUST)</u>, under <u>Prof. Lei Chen</u>'s supervision. He also received a Master degree in Software Engineering at <u>Beihang University</u> and a Double Bachelor degree in Economics from <u>China Centre for Economic Research (CCER)</u> at <u>Peking University</u>.

Research Interests

- Crowdsourcing
- · Spatio-temporal Big Data
- · Federated Learning and Data Federation Services
- Differential Privacy and Secure Multi-Party Computation
- Uncertain Data Mining and Management
- · Social Network Analysis

Our Recent Tutorials and Surveys

- Nami Yongxin Tong, Zimu Zhou, Yuxiang Zeng, Lei Chen, Cyrus Shahabi. "Spatial Crowdsourcing: A Survey", The VLDB Journal (VLDBJ), 29(1): 217–250, January 2020. [VLDB 2017 Tutorial Slides]
- Nam Qiang Yang, Yang Liu, Tianjian Chen, Yongxin Tong. "Federated Machine Learning: Concept and Applications", ACM Transactions on Intelligent Systems and Technology (TIST), 10(2): No.12, February 2019.

Homepage: http://sites.nlsde.buaa.edu.cn/~yxtong/

Faculty Members in SKLSDE















李未教授

马殿富教授

吕卫锋教授 尹宝林教授 蔡维德教授 马世龙教授

张玉平教授

许可教授







郎波教授



杨钦教授





吴文峻教授 朱皞罡教授





童咏昕教授 杜博文教授刘祥龙副教授

















刘瑞副教授 丁嵘副教授吕江花副教授 诸彤宇副教授 罗杰副教授 王德庆副教授 孟宪海博士 李吉刚博士

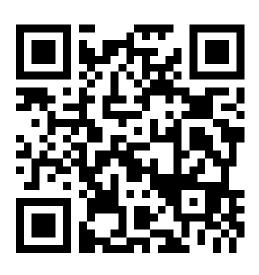
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MOOC

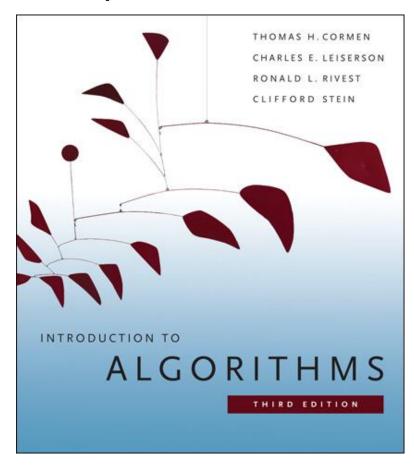
• 辅助线上MOOC课程: 算法设计与分析(中文)

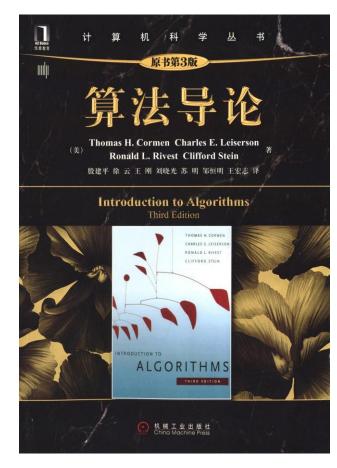




Textbook

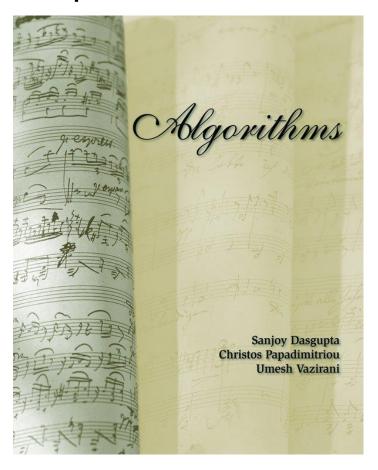
- Textbook: Introduction to Algorithms (3rd ed.)
 - by Cormen, Leiserson, Rivest and Stein (CLRS)
 - Prepublication version available online

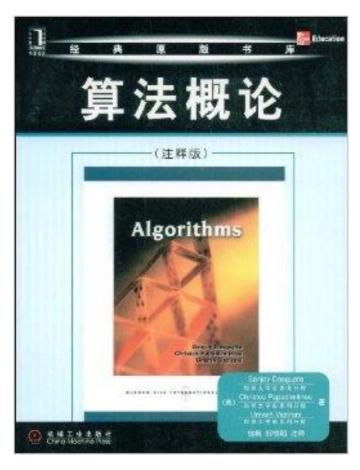




References (1)

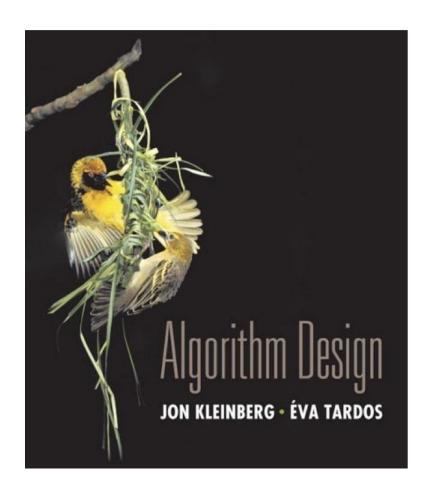
- Reference: *Algorithms*
 - by Dasgupta, Papadimitriou, and Vazirani (DPV)
 - Prepublication version available online

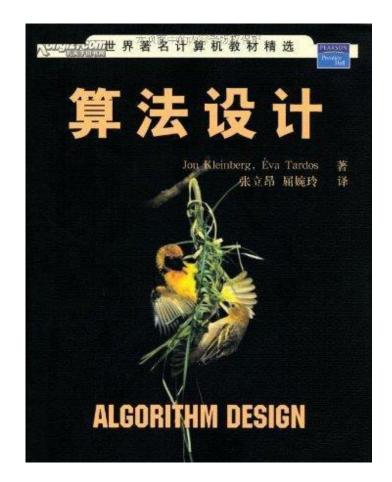




References (2)

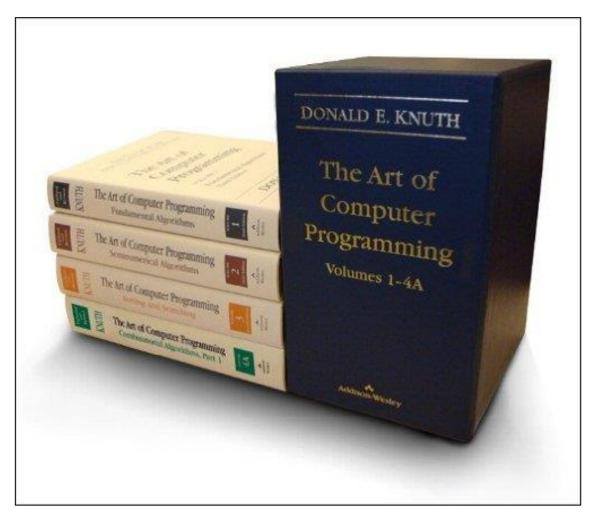
- Reference: Algorithm Design
 - by Kleinberg and Tardos (KT)





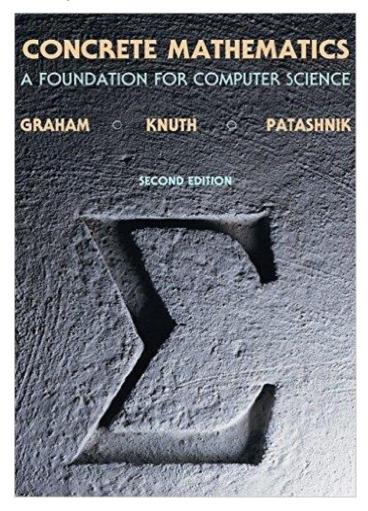
References (3)

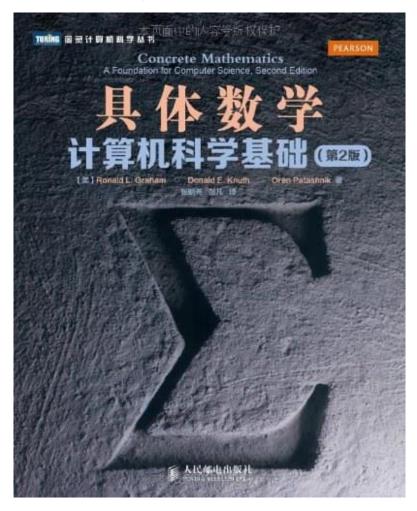
- Reference: The Art of Computer Programming
 - by Donald E. Knuth



References (4)

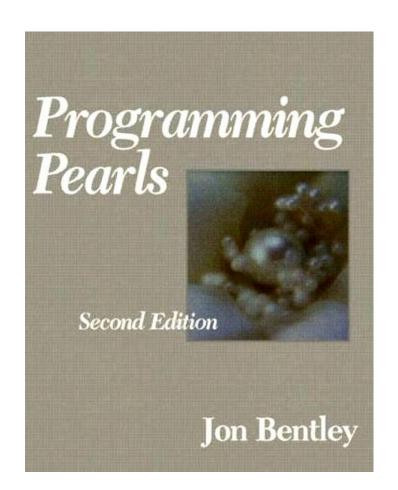
- Reference: Concrete Mathematics (2nd ed.)
 - by Graham, Knuth, Patashnik (GKP)

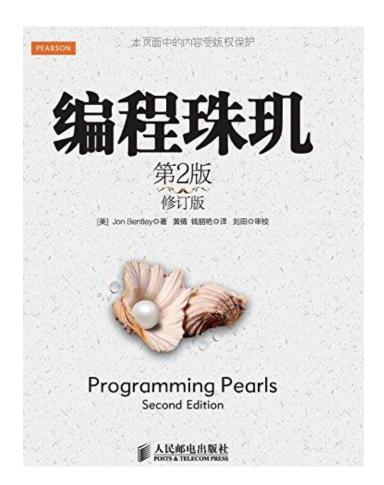




References (5)

- Reference: Programming Pearls (2nd ed.)
 - by Jon Bentley





Prerequisites

- We assume you know:
 - Linked Lists, Stacks, Queues
 - Binary Search Trees
 - Traversals
 - Searching (but not analysis)
- What have you learnt previously?
 - Graph algorithms
 - Breadth-first search (BFS)
 - Depth-first search (DFS)
 - Topological sort (TS)
 - Minimum Spanning Trees (MST)
 - Dijkstra's shortest path algorithm (SP)

Tentative Syllabus

- Basics
 - Asymptotic Notations and Recurrences
- Divide and Conquer Algorithms
 - MCS Problem, PM Problem, and Quicksort
- Dynamic Programming Algorithms
 - 0-1 Knapsack, Rod-Cutting, CMM, LCS, and MDE
- Greedy Algorithms
 - Huffman Coding and Fractional Knapsack
- Graph Algorithms
 - BFS, DFS, SP, MST, Max Flow and Matching
- Dealing with Hard Problems
 - Problem Classes (P, NP, NPC) and Approximation Alg.

Lectures and Tutorials

- Lectures
 - Slides will be available on course web page.

- Tutorials (补充练习)
 - There will be 12 tutorials in this semester.
 - The tutorials will provide more examples to illustrate the material you learnt in class.
 - The first tutorial will be released on next week.

Grading Scheme

- (40%) Four Assignments
 - Each assignment has 10% scores.
 - Each takes 14 days. The first one will be released on Sep. 21.
 - After each submission due, we will post the solution and WON'T accept any assignment.
 - Failing to do any of these will be considered PLAGIARISM, and may result in a failing grade in the course.

- (60%) Final Exam
 - It covers the materials of the entire semester.

Classroom Etiquette

No roll-call in our class!

- Turn off cell phone ringers.
 - No phone conversations in room.
- Latecomers should enter quietly.

No LOUD talking among selves during lectures.

WeChat Group



算法设计-6系-2020秋季

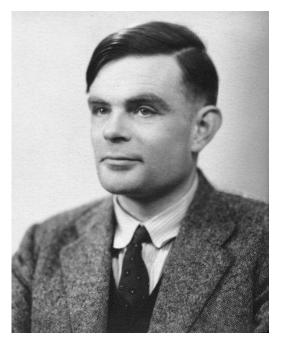


该二维码7天内(9月13日前)有效,重新进入将更新

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A.M. Turing Award



Alan M. Turing

From 2007 to 2013, the award was accompanied by a prize of US \$250,000 by Intel and Google. Since 2014, the award has been accompanied by a prize of US \$1 million by Google.



Nobel Prize of Computing

Since 1966, there have been 70 recipients of A.M. Turing Award! This year is the 53rd anniversary of A.M. Turing Award!

A.M. Turing Award Winners for Algorithms



Donald E. Knuth 1974, USA



Robert W. Floyd 1978, USA



Stephen A. Cook 1982, USA



Richard M. Karp 1985, USA



John Hopcroft 1986, USA



Robert Tarjan 1986, USA



Juris Hartmanis 1993, Latvia



Richard E. Stearns 1993, USA



Manuel Blum 1995, Venezuela



Andrew Yao 2000, China



Leslie G. Valiant 2010, Hungarian



Silvio Micali 2012, Italy



Shafi Goldwasser 2012, USA



Martin Hellman 2015, USA



Whitfield Diffie 2015, USA

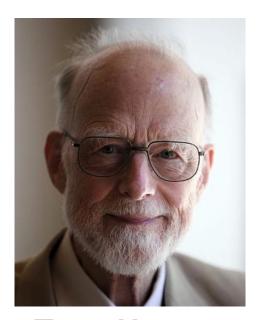
Other Related A.M. Turing Award Winners



Edsger W. Dijkstra
The Recipient in 1972,
Netherlands,

Contributions: ALGOL Father,

Related Work: Dijkstra Algorithm

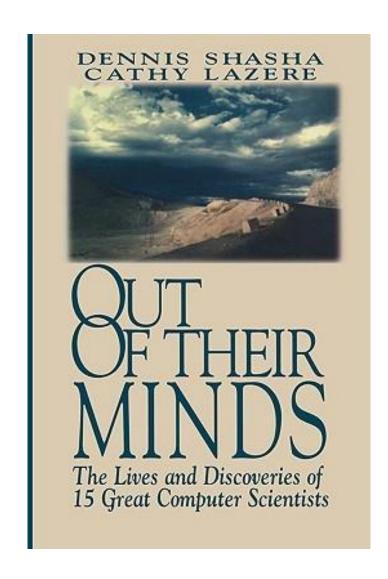


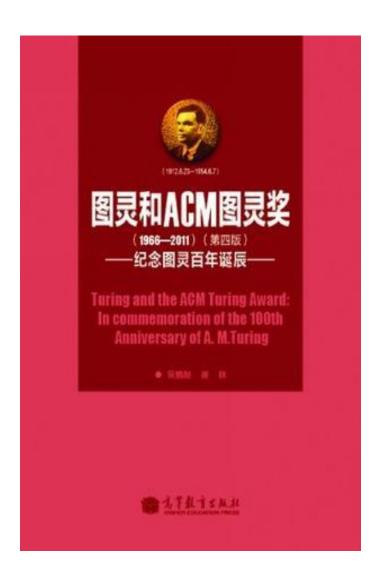
Tony Hoare The Recipient in 1980, UK,

Contributions: Hoare logic,

Related Work: QuickSort

Books of A.M. Turing Award Winners





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Example (Chain Matrix Multiplication)

Want: ABCD = ?

Method 1: (AB)(CD)

Method 2: A((BC)D)

Method 1 is much more efficient than Method 2. (Expand the expression on board)

- There is usually more than one algorithm for solving a problem.
- Some algorithms are more efficient than others.
- We want the most efficient algorithm.

- If we have a number of alternative algorithms for solving a problem, how do we know which is the most efficient?
- To do so, we need to analyze each of them to determine its efficiency.
- Of course, we must also make sure the algorithm is correct.

- In this course, we will discuss fundamental techniques for:
 - Designing efficient algorithms,
 - Proving the correctness of algorithms,
 - Analyzing the running times of algorithms

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Note:

Analysis and design go hand-in-hand:
 By analyzing the running times of algorithms, we will know how to design fast algorithms

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Computational Problem

Definition

A computational problem is a specification of the desired input-output relationship

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Example (Computational Problem)

Sorting

- Input: Sequence of *n* numbers $\langle a_1, \dots, a_n \rangle$
- Output: Permutation (reordering)

$$\langle a_1', a_2', \cdots, a_n' \rangle$$

such that $a_1' \leq a_2' \leq \cdots \leq a_n'$

Instance

Definition

A problem instance is any valid input to the problem.

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Example (Instance of the Sorting Problem)

(8, 3, 6, 7, 1, 2, 9)

Algorithm

Definition

An algorithm is a well defined computational procedure that transforms inputs into outputs, achieving the desired input-output relationship

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An algorithm is a well defined computational procedure that transforms inputs into outputs, achieving the desired input-output relationship

Definition

A correct algorithm halts with the correct output for every input instance. We can then say that the algorithm solves the problem

Example: Insertion Sort

Pseudocode:

```
Input: A[1 \dots n] is an array of numbers for j \leftarrow 2 to n do key \leftarrow A[j]; i \leftarrow j-1; while i \geq 1 and A[i] > key do A[i+1] \leftarrow A[i]; i \leftarrow i-1; end A[i+1] \leftarrow key; end
```

key

Sorted

Unsorted

Where in the sorted part to put "key"?

How Does It Work?

An incremental approach: To sort a given array of length n, at the *i*th step it sorts the array of the first *i* items by making use of the sorted array of the first *i* - 1 items

Example

```
Sort A = \langle 6, 3, 2, 4, 5 \rangle with insertion sort Step 1: \langle 6, 3, 2, 4, 5 \rangle
Step 2: \langle 3, 6, 2, 4, 5 \rangle
Step 3: \langle 2, 3, 6, 4, 5 \rangle
Step 4: \langle 2, 3, 4, 6, 5 \rangle
Step 5: \langle 2, 3, 4, 5, 6 \rangle
```

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- Predict resource utilization
 - Memory (space complexity)
 - Running time (time complexity) -- focus of this course

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 - depends on the speed of the computer
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- In light of the above factors, how can we compare different algorithms in terms of their running times?
- We want to find a way of measuring running times that is mathematically elegant and machine-independent.

 We will measure the running time as the number of primitive operations (e.g., addition, multiplication, comparisons) used by the algorithm

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 - Sorting: number of items to be sorted

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- Input size n: rigorous definition given later
 - Sorting: number of items to be sorted
 - Graphs: number of vertices and edges

Best Case: An instance for a given size n that results in the fastest possible running time.

Best Case: An instance for a given size *n* that results in the fastest possible running time.

Example (Insertion sort)

$$A[1] \leq A[2] \leq A[3] \leq \cdots \leq A[n]$$

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Example (Insertion sort)

$$A[1] \le A[2] \le A[3] \le \cdots \le A[n]$$

The number of comparisons needed is equal to

$$\underbrace{1 + 1 + 1 + \dots + 1}_{n-1} = n - 1 = \Theta(n)$$

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I KCy	

Sorted Unsorted "key" is compared to only the element right before it.

Worst Case: An instance for a given size *n* that results in the slowest possible running time.

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Example (Insertion sort)

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Example (Insertion sort)

$$A[1] \ge A[2] \ge A[3] \ge \cdots \ge A[n]$$

The number of comparisons needed is equal to

$$1+2+\cdots+(n-1)=\frac{n(n-1)}{2}=\Theta(n^2)$$

key

Sorted Unsorted

"key" is compared to everything element before it.

Average Case: Running time averaged over all possible instances for the given size, assuming some probability distribution on the instances.

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Example (Insertion sort)

 $\Theta(n^2)$, assuming that each of the n! instances is equally likely (uniform distribution).

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Example (Insertion sort)

 $\Theta(n^2)$, assuming that each of the n! instances is equally likely (uniform distribution).

key

Sorted Unsorted

On average, "key" is compared to half of the elements before it.

Best case: Clearly useless

- Best case: Clearly useless
- Worst case: Commonly used, will also be used in this course
 - Gives a running time guarantee no matter what the input is
 - Fair comparison among different algorithms

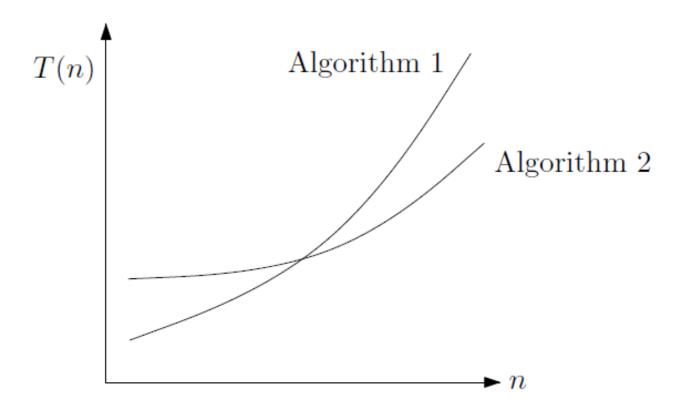
- Best case: Clearly useless
- Worst case: Commonly used, will also be used in this course
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- Average case: Used sometimes
 - Need to assume some distribution: real-world inputs are seldom uniformly random!
 - Analysis is complicated

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 - Gives a running time guarantee no matter what the input is
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- Average case: Used sometimes
 - Need to assume some distribution: real-world inputs are seldom uniformly random!
 - Analysis is complicated
 - Will not be used in this course

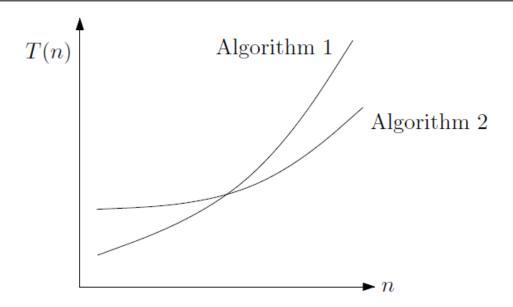
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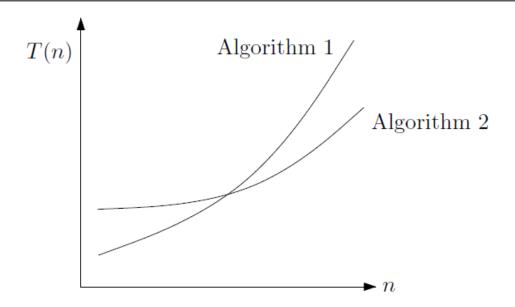
Comparing Time Complexity



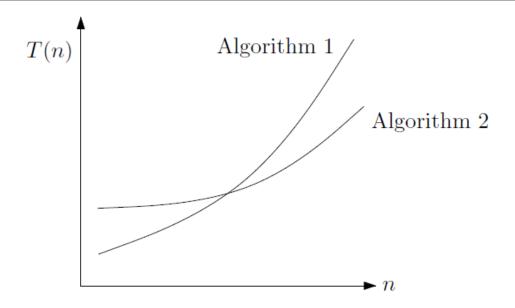
- Which algorithm is superior for large n?
 - T(n) for Algorithm 1 is $3n^3 + 6n^2 4n + 17$
 - T(n) for Algorithm 2 is $7n^2$ 8n + 20
- Clearly, Algorithm 2 is superior.



• T(n) for Algorithm 1 is $3n^3 + 6n^2 - 4n + 17 = \Theta(n^3)$



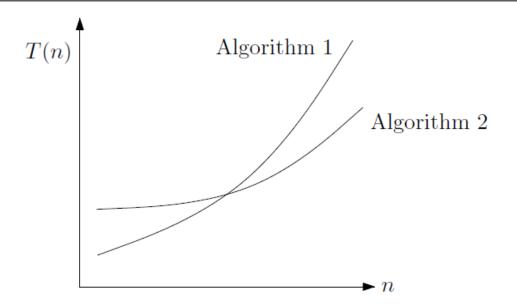
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Θ-notation

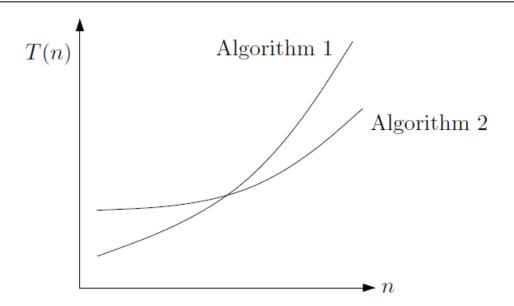
Drop low-order terms; ignore leading constants



- T(n) for Algorithm 1 is $3n^3 + 6n^2 4n + 17 = \Theta(n^3)$
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Θ-notation

- Drop low-order terms; ignore leading constants
- Look at growth of T(n) as n→∞



- T(n) for Algorithm 1 is $3n^3 + 6n^2 4n + 17 = \Theta(n^3)$
- T(n) for Algorithm 2 is $7n^2 8n + 20 = \Theta(n^2)$

Θ-notation

- Drop low-order terms; ignore leading constants
- Look at growth of T(n) as n→∞
- When n is large enough, a Θ(n²) algorithm always beats a Θ(n³) algorithm



