# 算法设计与分析 **Algorithms Design & Analysis**

第一讲: 概述

## 参考教材

- Thomas H. Cormen, 《算法导论,第二 版》, 高等教育出版社(影印, 英文)。
- 吴伟昶, 《算法设计技巧与分析》, 电子工业 出版社,2005。





提醒: 不看教材很难会有好成绩 光看教材不保证有好成绩

不看教材很难有好基础 光看教材很难有好发展

## 课程训练内容

- 相关基本概念
- 算法设计方法
- 算法分析技术
- 解决实际问题的能力

注意: 不包含算法在某种具体语言下的实现

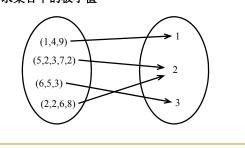
不包含算法调试是一门理论基础课程

编程实现是你自己的事情

(Program implementations are on your own)

## 从简单的问题开始

• 求集合中的极小值



# 解决问题 → Output Input · 三要素: 输入、输出、算法 算法在解决问题层面上具有"黒盒"特征

## 数据结构(Data Structure)

- 输入、输出和数据结构直接相关
- ■数据之间的逻辑关系、数据在计算机上的存储 方式、数据的操作。
- 数据结构与算法设计是密切相关的

# 算法 (Algorithm)

- Informally, an <u>algorithm</u> is any <u>well-defined</u> <u>computational procedure</u> that takes some value, or set of values, as <u>input</u> and produces some value, or set of values, as <u>output</u>.
- An algorithm is a step-by-step description of a procedure which, if followed closely, produces a well-defined result.
- 算法是为了求解问题而给出的指令序列。

程序只是算法的一种实现

#### 算法表述

- 自然语言 (ENGLISH)
- 算法描述语言 (Pseudo-code)
- 计算机程序语言(C++, Java)
- 硬件设计 (DSP)

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#### 算法一般特性

- 正确性:对于符合输入类型的任意输入数据,都产生正确的输出。
- Example: Given an election of 100,000,000 votes, determine the winner. (选举统计)

Approach 1: Count all 100,000,000 votes!

Approach 2: Select a sample of 1000,000 votes and declare the winner of the sample as the winner of the election.

==> with high probability, approach 2 will determine the winner correctly! But not always.(但是可以获得很高的正确性概率)

#### 算法一般特性

- <mark>有效性</mark>:每一步指令能够被有效的执行, 并且规定了指令的执行效果,结果应该具 有的数据类型,而且是可以预期的。
- <mark>确定性</mark>:每一步之后都要有确定的下一步 指令。
- 有穷性: 有限步内结束

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## 算法效率 ( Efficiency)

■ 有限资源

#### Computational resources :

- · CPU time (running time)计算时间
- · memory usage (space)存储空间
- messages sent along the network网络带宽
- These resources should be used wisely, and algorithms that are efficient in terms of time or space will help you do so。(效率分析是为了有效的利 用资源)

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# 资源开销与输入

 Resource consumption differs depending on the size of the input(资源开销与输入大小相关)

length of text(文本长度) number of records to be sorted(排序记录条目数量)

Resource consumption may even differ greatly for inputs
of the same size, depending on their structure(资源开销与输
入的组织结构有关)

highly unsorted input(无序) almost sorted input (有序)

Specify resource consumption as a function of the inputs size. 随输入规模变化的资源开销函数

#### 时空资源折中原理

对于同一个求解问题,一般会存在多种算法, 这些算法在时空开销上的优劣往往表现出"时 空折中"的性质,即为了改善一个算法的时间 开销,往往可以通过增大空间开销为代价,设 计出一个新算法来。

AB=Constant

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# 算法设计与分析 Algorithms Design & Analysis

第二讲: 算法渐进分析

华中科技大学软件学院 邱德红 主讲

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#### 渐进分析 (Asymptotic Analysis)

- 回答→ How does algorithm behave as the problem size gets very large?
  - Running time(运行时间)
  - Memory/storage requirements(存储需求)
  - □ Remember that we use the RAM model (RAM模式):
    - All memory equally expensive to access (存储空间 访问开销均等)
    - No concurrent operations(顺序无并发)
    - All reasonable instructions take unit time
      - Except, of course, function calls(指令执行周期相同,除 了程序调用之外)
    - Constant word size(字节长度一致)
      - Unless we are explicitly manipulating bits

# 渐进分析(续1)

■ 渐进分析的目的是得到一个开销函数的渐进表 达式,比如大*○*渐进表达式

$$rateT(n) = O(f(n))$$

n: 问题规模

T (n): 资源开销函数

f(n): 问题规模的整函数表达式

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# 确定情况:矩阵求和(例)

```
void matrix_addition(double *M1, double *M2, int k)
                                       cost
                                                 times
  for(int i=0; i< k; i++)
                                        c1
                                                  k+1
     for(int j=0; j< k; j++)
                                        c2
                                                  k(K+1)
        M1[i][j]=M1[i][j]+M2[i][j];
                                        c3
          问题规模: n=k2
          T(n)=c1(k+1)+c2k(k+1)+c3kk
              =ak^2+bk+c=an+bn^{1/2}+c
                rateT(n) = O(f(n))
          f(n)=n
```

#### 条件情况: 插入排序(Insertion Sort)

```
InsertionSort(A, n) {
  for i = 2 to n {
     key = A[i]
     j = i - 1
     while (j > 0) and (A[j] > key) {
           A[j+1] = A[j]
           j = j - 1
     }
     A[j+1] = key
}
```

```
An Example: Insertion Sort
                            i = 2 j = 0 key = 10
            40
                 20
  30
       30
                            A[j] = \emptyset
                                             A[j+1] = 30
             3
           InsertionSort(A, n) {
             for i = 2 to n {
   key = A[i]
                  j = i - 1;
while (j > 0) and (A[j] > key) {
                         A[j+1] = A[j]
j = j - 1
                  A[j+1] = key
             }
          }
```

```
An Example: Insertion Sort

| 30 | 30 | 40 | 20 | | i = 2 | j = 0 | key = 10 |
| 1 | 2 | 3 | 4 | 3 | 4 | [j] = \emptyset | A[j+1] = 30

| InsertionSort(A, n) {
| for i = 2 to n {
| key = A[i] |
| j = i - 1; |
| while (j > 0) and (A[j] > key) {
| A[j+1] = A[j] |
| j = j - 1 |
| A[j+1] = key
| }
| A[j+1] = key
| }
```

```
An Example: Insertion Sort

| 10 | 30 | 40 | 20 | | i = 3 | j = 2 | key = 40 | | A[j] = 30 | A[j+1] = 40 |

InsertionSort(A, n) {
    for i = 2 to n {
        key = A[i] | | j = i - 1;
        while (j > 0) and (A[j] > key) {
            A[j+1] = A[j] | | j = j - 1 |
            A[j+1] = key
        }
        }
        A[j+1] = key
        }
}
```

```
An Example: Insertion Sort
                                          key = 20
                           i = 4 j = 3
            40
                20
  10
      30
                           A[j] = 40
                                           A[j+1] = 20
          InsertionSort(A, n) {
             for i = 2 to n {
   key = A[i]
                 j = i - 1;
while (j > 0) and (A[j] > key) {
                        A[j+1] = A[j]
j = j - 1
                 A[j+1] = key
            }
          }
```

```
An Example: Insertion Sort

| 10 | 30 | 40 | 40 | | i = 4 | j = 3 | key = 20 | A[j] = 40 | A[j+1] = 40 |

| InsertionSort(A, n) {
| for i = 2 to n {
| key = A[i] |
| j = i - 1;
| while (j > 0) and (A[j] > key) {
| A[j+1] = A[j] |
| j = j - 1 |
| A[j+1] = key
| }
| }
```

```
An Example: Insertion Sort
                           i = 4 j = 2 key = 20
            40
                 40
  10
      30
                           A[j] = 30
                                           A[j+1] = 40
            3
           InsertionSort(A, n) {
             for i = 2 to n {
   key = A[i]
                 j = i - 1;
while (j > 0) and (A[j] > key) {
                        A[j+1] = A[j]
j = j - 1
                 A[j+1] = key
             }
          }
```

```
An Example: Insertion Sort

| 10 | 30 | 30 | 40 | | i = 4 | j = 2 | key = 20 | A[j] = 30 | A[j+1] = 30 |

InsertionSort(A, n) {
    for i = 2 to n {
        key = A[i] | j = i - 1;
        while (j > 0) and (A[j] > key) {
            A[j+1] = A[j] | j = j - 1 |
            A[j+1] = key
        }
    }
}
```

```
An Example: Insertion Sort
                       i = 4 j = 1
                                     key = 20
  10
      30
          30
               40
                       A[j] = 10
                                     A[j+1] = 30
         InsertionSort(A, n) {
           for i = 2 to n {
    key = A[i]
    j = i - 1;
               \Rightarrow
               A[j+1] = key
           }
         }
```

```
An Example: Insertion Sort

| 10 | 30 | 30 | 40 | | i = 4 | j = 1 | key = 20 | A[j] = 10 | A[j+1] = 30 |

| InsertionSort(A, n) {
| for i = 2 to n {
| key = A[i] |
| j = i - 1;
| while (j > 0) and (A[j] > key) {
| A[j+1] = A[j] |
| j = j - 1 |
| A[j+1] = key
| }

| A[j+1] = key
```

```
An Example: Insertion Sort
                           i = 4 j = 1 key = 20
           30
                 40
 10
      20
                           A[j] = 10
                                           A[j+1] = 20
          InsertionSort(A, n) {
             for i = 2 to n {
   key = A[i]
                 j = i - 1;
while (j > 0) and (A[j] > key) {
                        A[j+1] = A[j]
j = j - 1
                 A[j+1] = key
             }
          }
                         Done!
```

```
Insertion Sort
InsertionSort(A, n)
                                                               Cost
                                                                           Repetitions
C_1
            \textbf{do } \textit{key} \leftarrow A[i]
                                                                 C_2
                                                                               n-1
      j \leftarrow i - 1
                                                                 C_4
                                                                               n-1
                                                                               \sum_{i=2}^{n} t_i
         while j > 0 and A[j] > key
                                                                 C_5
                                                                               \sum_{i=2}^{n} (t_i - 1)
                \operatorname{do} A[j+1] \leftarrow A[j]
                                                                 C_6
                                                                               \sum_{i=2}^{n} (t_i - 1)
                  j \leftarrow j - 1
                                                                 C_7
7 A[j+1] \leftarrow key
```

#### 不同情况:

■ 最好情况: 输入为10、20、30、40

$$t_i = 1$$

$$T(n) = (C_1 + C_2 + C_4 + C_5 + C_8)n - (C_2 + C_4 + C_5 + C_8)$$

$$= an + b$$

不同情况:

■ 最坏情况: 输入为40、30、20、10

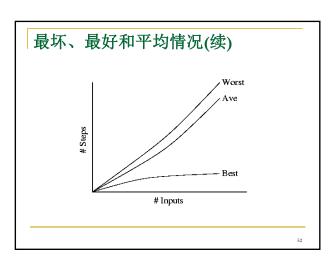
$$t_i = i$$

$$T(n) = an^2 + bn + c$$

# 最坏、最好和平均情况

- The worst case running time of an algorithm is the function defined by the maximum number of steps taken on any instance of size.(输入量为n时的最大运行步骤数目)
- The best case running time of an algorithm is the function defined by the minimum number of steps taken on any instance of size n.(输入量为n 时的最小运行步骤数目)
- The average-case running time of an algorithm is the function defined by an average number of steps taken on any instance of size n.(输入量为n时的平均运行步骤数目)

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#### Average case analysis

- Drawbacks(平均分析缺点)
  - Based on a probability distribution of input instances(与输入的分布有关)
  - How do we know if distribution is correct or not? (问题: 如何确定输入的分布?)
- Usually more complicated to compute than worst case running time(通常比最坏情况计算 困难)
  - Often worst case running time is comparable to average case running time(最坏情况通常比较接近于 平均情况)

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#### Worst case analysis

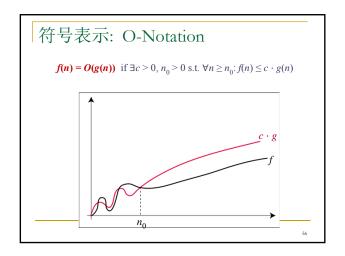
- Typically much simpler to compute as we do not need to "average" performance on many inputs(简单)
  - Instead, we need to find and understand an input that causes worst case performance(需要确定导致 最坏情况下的输入)
- Provides guarantee that is independent of any assumptions about the input(独立于有关 输入的假设)
- Often reasonably close to average case running time(通常接近平均情况)

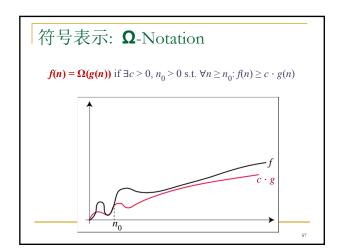
# 渐进分析与阶的增长

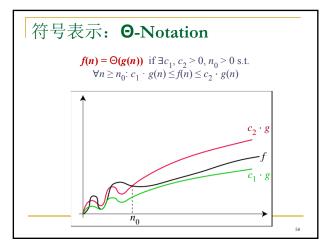
- 开销函数的估计是相对的而不是绝对的
- 独立于机器的算法开销估计
- 独立于实现技术的算法自身测度表示
- 关心的是大规模输入的情况

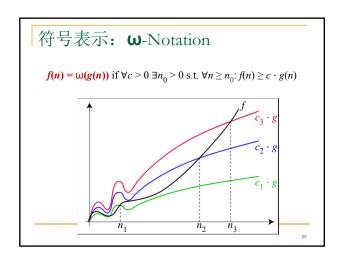
忽略了开销函数的低阶项和常数项,度量的是算 法渐进的开销,此时,开销函数的阶的增长决定了 开销的大小。

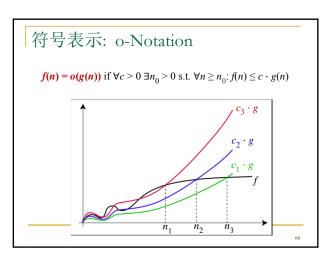
$$T(n) = an^2 + bn + c$$











# A Few Simple Facts

- $f(n) = \Omega(f(n))$ f(n) = O(f(n)) $f(n) = \Theta(f(n))$
- $f(n) = O(g(n)) \text{ and } g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$   $f(n) = \Omega(g(n)) \text{ and } g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$
- $f(n) = \Theta(g(n))$  and  $g(n) = \Theta(h(n)) \Rightarrow f(n) = \Theta(h(n))$
- $f(n) = O(g(n)) \Longleftrightarrow g(n) = \Omega(f(n))$   $f(n) = o(g(n)) \Longleftrightarrow g(n) = \omega(f(n))$
- f(n) = O(g(n)) and  $f(n) = \Omega(g(n)) \Rightarrow f(n) = \Theta(g(n))$
- $f_1(n) = O(g_1(n))$  and  $f_2(n) = O(g_2(n)) \Rightarrow f_1(n) + f_2(n) = O(g_1(n) + g_2(n))$
- $f(n) = O(g(n)) \Rightarrow f(n) + g(n) = O(g(n))$

# 渐进分析和极限(Asymptotic Analysis and

If 
$$\lim_{n\to\infty} \frac{f(n)}{g(n)} = 0$$
, then  $f(n) = o(g(n))$ .

If 
$$\lim_{n\to\infty} \frac{f(n)}{g(n)} = c$$
, for some constant  $c > 0$ , then  $f(n) = \Theta(g(n))$ .

If 
$$\lim_{n\to\infty} \frac{f(n)}{g(n)} = 0$$
, then  $a^{f(n)} = o(a^{g(n)})$ , for any  $a > 1$ .

$$f(n) = o(g(n)) \Longrightarrow a^{f(n)} = o(a^{g(n)})$$
, for any  $a > 1$ .

$$f(n) = \Theta(g(n)) \implies a^{f(n)} = \Theta(a^{g(n)})$$

常见的规模的整函数表达式的f(n)

- f(n)=1, 常数函数,不依赖于n
- f(n)=log n, 对数函数, 比线性函数增长慢
- f(n)=n
- $f(n)=n^2$
- f(n)=nlog(n)
- f(n)=a<sup>n</sup>, 指数增长

算法效率比较

■ We consider algorithm A better than algorithm B if (算法A好于算法B,如果满足以下公式)

 $T_A(n) = o(T_B(n))$ 

合并排序(Merge Sort)

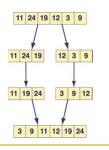
If |A| = 1, the data is already sorted; done. (一个元素的情况)

Otherwise:

Split the input into two pieces of equal size. (二等分)

Recursively sort these two pieces. (二部分递归排序)

Construct a sorted sequence from the two sorted subsequences. (整合排序好的两部分)



合并: Merge(A, p, q, r)

to temporary arrays L and R(左右临时数组)

 $1 \ n_1 \leftarrow q - p + r$  $2 n_2 \leftarrow r - q$ 

- 3 **for**  $i = 1..n_1$
- 4 **do**  $L[i] \leftarrow A[p+i-1]$ 5 **for**  $j = 1..n_2$ 6 **do**  $R[j] \leftarrow A[q+j]$
- Create two artificial end markers that are never copied to

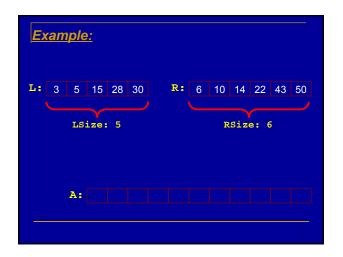
A: (结束标志)  $7 L[n_1 + 1] \leftarrow \infty$   $8 R[n_2 + 1] \leftarrow \infty$ 

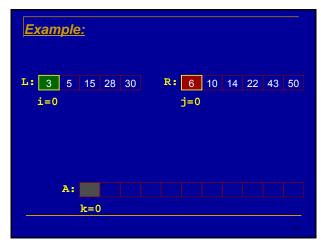
• Copy A[p..q] and A[q+1..r] • Repeatedly copy the smallest element from *L* and *R* to *A*:

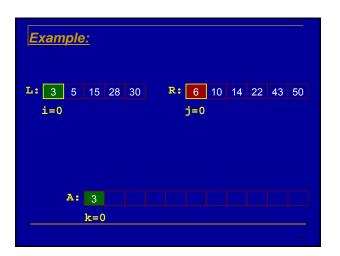
(重复的将左右零时数组中 最小的元素存储到A中)

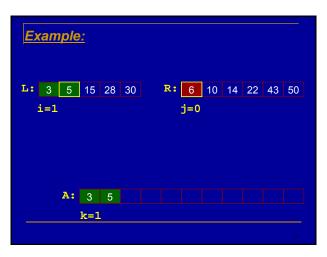
 $9i \leftarrow 1$  $10i \leftarrow 1$ 

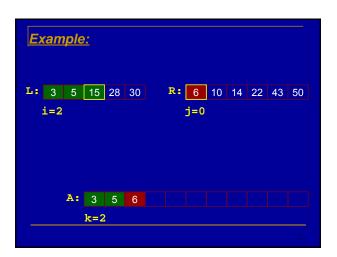
- 11 **for** k = p..r12 **do** if  $L[i] \leq R[j]$ then  $A[k] \leftarrow L[i]$  $i \leftarrow i + 1$ 13
  - else  $A[k] \leftarrow R[j]$

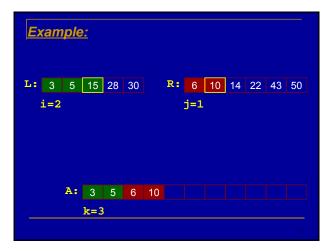


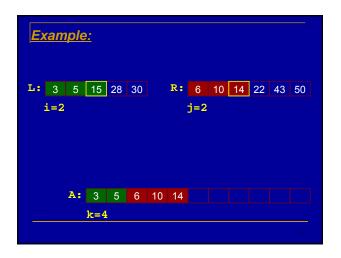


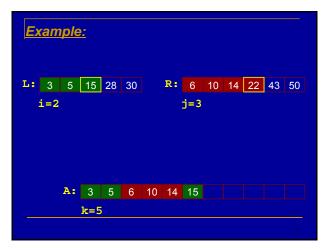


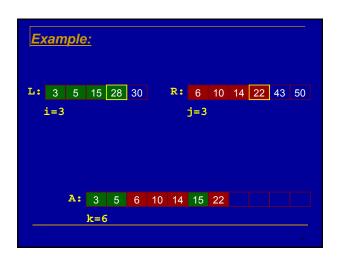


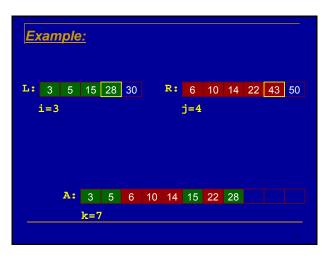


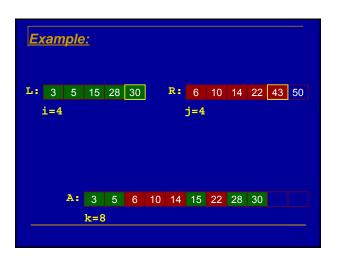


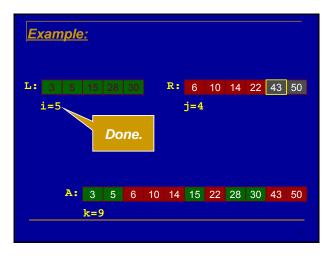


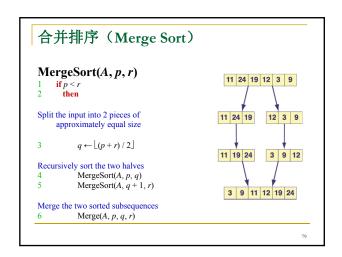


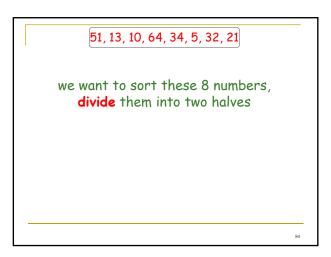


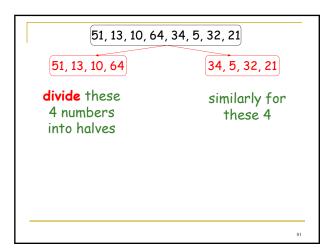


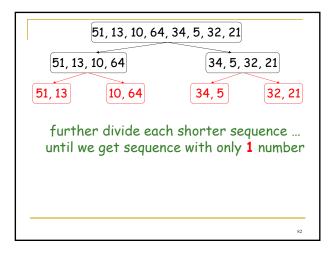


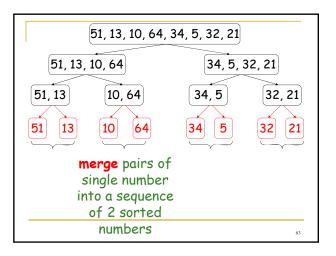


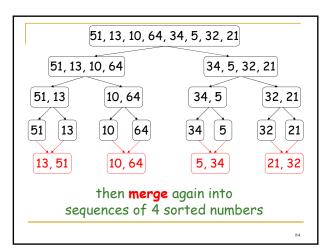


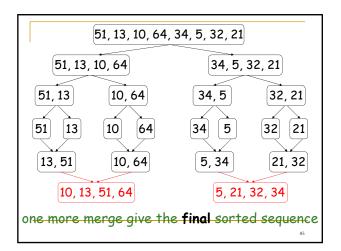


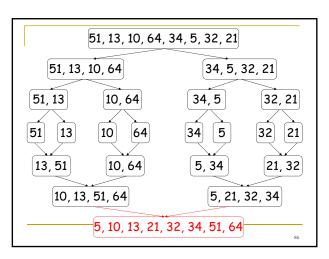












# 合并排序分析(Running Time of Merge Sort) • MergeSort(A, p, r) • 1 if p < r• 2 then • Split the input into 2 pieces of approximately equal size • 3 $q \leftarrow \lfloor (p+r)/2 \rfloor$ $c_1 +$ • Recursively sort the two halves • 4 MergeSort(A, p, q) T(n/2) +• 5 MergeSort(A, q+1, r) T(n/2) +• Merge the two sorted subsequences • 6 Merge(A, p, q, r) $c_2 \cdot n$

小结 (Summation)

- We have studied two sorting algorithms (两种排序算法):
  - $ext{ o}$  Insertion Sort takes  $c_1 \cdot n^2$  time. (插入排序)
- Even if  $c_2 \gg c_{1'}$  Merge Sort will ultimately outperform Insertion Sort (合并排序优于插入排序)