

Homework 1

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Problem 1 (Textbook 1.18). 对以下函数, 按照他们的阶从高到低排序; 如果两个函数的阶相同, 表示为 $f(n) = \Theta(g(n))$.

$$2^{\sqrt{2 \log n}}, n \log n, \sum_{k=1}^n \frac{1}{k}, n2^n, (\log n)^{\log n}, 2^{2n}, 2^{\log \sqrt{n}}$$

$$n^3, \log n!, \log n, \log \log n, n^{\log \log n}, n!, n, \log 10^n$$

Answer. 我们用 $>$ 表示阶的大小关系:

$$n! > 2^{2n} > n2^n > (\log n)^{\log n} = n^{\log \log n} > n^3 > n \log n = \Theta(\log n!) >$$

$$n = \Theta(\log 10^n) > 2^{\log \sqrt{n}} > 2^{\sqrt{2 \log n}} > \log n = \Theta\left(\sum_{k=1}^n \frac{1}{k}\right) > \log \log n$$

Problem 2 (Textbook 1.19). 求解下列递推方程:

$$(1) \begin{cases} T(n) = T(n-1) + n^2 \\ T(1) = 1 \end{cases}$$

$$(2) \begin{cases} T(n) = T(n/2) + T(n/4) + cn \\ T(1) = 1 \end{cases}$$

$$(3) \begin{cases} T(n) = 5T(n/2) + (n \log n)^2 \\ T(1) = 1 \end{cases}$$

$$(4) \begin{cases} T(n) = T(n-1) + \frac{1}{n} \\ T(1) = 1 \end{cases}$$

Answer. (1) 用迭代法:

$$T(n) = n^2 + (n-1)^2 + \cdots + 4 + 1 = \frac{n(n+1)(2n+1)}{6}$$

(2) 使用递归树法: 设数深为 k , 那么 $n/2^k \geq 1 \implies k \leq \log n$:

$$T(n) = \sum_{k=0}^{\log n} \left(\frac{3}{4}\right)^k cn = 4cn \left(1 - \left(\frac{3}{4}\right)^{\log n}\right) = \Theta(n)$$

(3) 由主定理, $f(n) = (n \log n)^2 = O(n^{\log_2 5 - \epsilon})$, $T(n) = \Theta(n^{\log_2 5})$

(4) 用迭代法:

$$T(n) = \sum_{k=1}^n \frac{1}{k} = \Theta(\log n)$$

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Problem 3 (Textbook 1.21). 设原问题规模为 n , 从下述算法中选择一个最坏情况下时间复杂度最低的算法, 简述理由:

1. 算法 A: 将原问题划分为规模减半的 5 个子问题, 在线性时间内合并结果.
2. 算法 B: 将原问题划分为 2 个规模为 $n-1$ 的子问题, 在常量时间内合并结果.
3. 算法 C: 将原问题划分规模为 $n/3$ 的 9 个子问题, 在 $O(n^3)$ 内合并结果.

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Answer. 选择算法 A, 理由如下:

1. 算法 A: $T(n) = 5T(n/2) + O(n) \implies T(n) = \Theta(n^{\log_2 5})$.
2. 算法 B: $T(n) = 2T(n-1) + O(1) \implies T(n) = \Theta(2^n)$.
3. 算法 C: $T(n) = 9T(n/3) + O(n^3) \implies T(n) = \Theta(n^3)$.

算法 A 在最坏情况下复杂度最低.

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Problem 4 (Complexity Bounds). For each blank, indicate whether A_i is in O , Ω , or Θ of B_i . More than one space per row can be valid.

A	B	O	Ω	Θ
$10n$	n	✓	✓	✓
10	n	✓		
n^2	$2n$		✓	
n^{2021}	2^n	✓		
$n^{\log 9}$	$9^{\log n}$	✓	✓	✓
$\log(n!)$	$\log(n^n)$	✓	✓	✓
$(3/2)^n$	$(2/3)^n$		✓	
3^n	2^n		✓	
$n^{1/\log n}$	1	✓	✓	✓
$\log^5 n$	$n^{0.5}$	✓		
n^2	$4^{\log n}$	✓	✓	✓
$n^{0.2}$	$(0.2)^n$		✓	
$\log \log n$	$\sqrt{\log n}$	✓		
$\log(\sqrt{n})$	$\sqrt{\log n}$		✓	

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Problem 5 (Gauging Complexity from Code). Refer to the algorithms below for solving these problems. Give the **worst-case run-time complexity** in Big-Oh notation for each of the algorithms below.

Algorithm 1: Search an element in an array of length n

Input: An array A of size n and an element e

Output: Index of the element e

```

1  $i \leftarrow 0$ ;
2 while  $i < n$  do
3   if  $A[i] = e$  then
4     return  $i$ ;
5   end
6    $i \leftarrow i + 1$ ;
7 end
8 return  $-1$                                      // If element  $e$  is not found

```

Algorithm 2: Replace an element in an array of length n with a given element

Input: An array A of size n , an element e present in the array, and a new element E

Output: Updated array

```

1  $i \leftarrow 0$ ;
2 while  $i < n$  do
3    $\text{currElement} \leftarrow A[i]$ ;
4   if  $\text{currElement} = e$  then
5      $A[i] \leftarrow E$ ;
6     return  $A$ ;
7   end
8   else
9      $A[i] \leftarrow \text{currElement}$ ;
10  end
11   $i \leftarrow i + 1$ ;
12 end
13 return  $A$                                      // If element  $e$  is not found, return the original array

```

Algorithm 3: Sort an array of length n using bubble sort

Input: An array A of size n

Output: Sorted array

```

1 for  $i \in [0, n - 1)$  do
2   /* Declare  $i$  as the index having the current smallest element */
3   for  $j \in [i + 1, n)$  do
4     /* Find the smallest element in the range  $[i, n)$  */
5     if  $A[j] > A[j + 1]$  then
6        $\text{smallest} \leftarrow j$ 
7     end
8   end
9   swap ( $A[i], A[\text{smallest}]$ );
10 end
11 return  $A$ ;

```



Answer. (1) **Algorithm 1:** 在最坏情况下, 需要完整遍历一遍数组, 时间复杂度为 $\Theta(n)$.

(2) **Algorithm 2:** 在最坏情况下, 替换的值在数组的最后一位, 需要完整遍历一遍数组, 时间复杂度为 $\Theta(n)$.

(3) **Algorithm 3:** 设最坏情况下所需操作数为 $T(n)$, 则 (c_1, c_2 为常数):

$$T(n) = \sum_{i=0}^{n-2} (c_1(n-i-1) + c_2) = \Theta(n^2)$$

故时间复杂度为 $\Theta(n^2)$.

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Problem 6 (Partial Sum of a 1D Array). Given an array A consisting of n integers $A[1], A[2], \dots, A[n]$. You want to obtain a 2D $n \times n$ array B where $B[i][j]$ (for $i < j$) contains the sum of array entries $A[i]$ through $A[j]$, i.e., $A[i] + A[i+1] + \dots + A[j]$. The value of array entry $B[i][j]$ is left unspecified whenever $i > j$.

Here's a basic algorithm to solve this problem:

Algorithm 4: Basic Algorithm

```

1 for  $i = 1$  to  $n$  do
2   for  $j = i + 1$  to  $n$  do
3     Add up array entries  $A[i]$  through  $A[j]$ ;
4     store the result in  $B[i][j]$ 
5   end
6 end
```

- (1) For some function f you may choose, give a bound of $O(f(n))$ on the runtime of this algorithm.
- (2) For this same function f , prove that the runtime of the algorithm is also $\Omega(f(n))$ (This shows an asymptotically tight bound of $\Theta(f(n))$ on the runtime).
- (3) Although the algorithm you analysis in (1)(2) is the most intuitive way to solve the problem, after all, it contains some highly unnecessary sources of inefficiency. Give a different algorithm to solve the problem, with an asymptotically better runtime. In other words, you should develop an algorithm with runtime $O(g(n))$, where $\lim_{n \rightarrow \infty} g(n)/f(n) = 0$.

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Answer. (1) 设 $T_1(n)$ 为 *Basic Algorithm* 的时间复杂度, 那么:

$$T_1(n) = \sum_{i=1}^n \left(1 + \sum_{j=i+1}^n (j-i+2) \right) = \sum_{i=1}^n \left(\frac{n^2}{2} + \frac{i^2}{2} + \frac{5n}{2} - in - \frac{5i}{2} + 1 \right) = \Theta(n^3)$$

故取 $f(n) = cn^3$ (c 为常数), 则 $T_1(n) = O(n^3)$.

(2) 由于 $T_1(n) = \Theta(n^3)$, 故 $T_1(n) = \Omega(n^3)$.

(3) 注意到对于 $B[i][j]$ ($j > i$), $\sum_{k=i}^{j-1} A[k]$ 已经在 $B[i][j-1]$ 中计算过, 无需重复计算, 可以用动态规划的思想优化算法:

Algorithm 5: Optimized Algorithm

```
1 for  $i = 1$  to  $n$  do
2    $B[i][i] = A[i];$ 
3 end
4 for  $i = 1$  to  $n$  do
5   for  $j = i + 1$  to  $n$  do
6      $B[i][j] = B[i][j - 1] + A[j];$ 
7   end
8 end
```

设 $T_2(n)$ 为 *Optimized Algorithm* 的时间复杂度, 那么:

$$T_2(n) = \sum_{i=1}^n \left(1 + \sum_{j=i+1}^n 1 \right) = \sum_{i=1}^n (n - i + 1) = \Theta(n^2)$$

令 $g(n) = cn^2$ (c 为常数), 则 $T_2(n) = O(n^2)$.

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