

# Algorithm Design and Analysis

## Assignment 1

**Deadline: March 23, 2025**

1. (24 points) Asymptotic notations.

(a) (24 points) In each of the following situations, indicate whether  $f = O(g)$ , or  $f = \Omega(g)$ , or both (in which case  $f = \Theta(g)$ ). Justify your answer.

1.  $f(n) = (n+1)!$  and  $g(n) = n!$

$f = \Omega(g)$  because  $\frac{f(n)}{g(n)} = (n+1)$  is unbounded.

2.  $f(n) = 2^{n+1}$  and  $g(n) = 2^n$

$f = \Theta(g)$  because  $\frac{f(n)}{g(n)} = 2$  is a constant.

3.  $f(n) = 2^n$  and  $g(n) = 3^n$

$f = O(g)$  because  $\frac{f(n)}{g(n)} = (\frac{2}{3})^n$  converges to 0.

4.  $f(n) = n^{1/2}$  and  $g(n) = 5^{\log_2 n}$

$f = O(g)$  because  $\log_2 5 > 1/2$ , so that  $\frac{f(n)}{g(n)} = \frac{n^{1/2}}{n^{\log_2 5}}$  converges to 0.

5.  $f(n) = 100n + \log n$  and  $g(n) = n + (\log n)^2$

$f = \Theta(g)$  because  $\frac{f(n)}{g(n)} = 100$  is a constant.

6.  $f(n) = (\log n)^{\log n}$  and  $g(n) = n/\log n$

$f = \Omega(g)$  because  $\frac{f(n)}{g(n)} = (n+1)$  is unbounded.

7.  $f(n) = (\log n)^{\log n}$  and  $g(n) = 2^{(\log_2 n)^2}$

$f = O(g)$  because  $\frac{f(n)}{g(n)} = \frac{(\log n)^{\log n}}{2^{(\log_2 n)^2}}$  converges to 0.

8.  $f(n) = \sum_{i=1}^n i^k$  and  $g(n) = n^{k+1}$

- if  $k > -1$ ,  $f = \Theta(g)$  because  $\frac{f(n)}{g(n)} = \frac{\sum_{i=1}^n i^k}{n^{k+1}}$  converges to  $\frac{1}{k+1}$ .
- if  $k \leq -1$ ,  $f = O(g)$  because  $\frac{f(n)}{g(n)} = \frac{\sum_{i=1}^n i^k}{n^{k+1}}$  is unbounded.

(b) (Not for credit, just for fun: 0 points) Suppose  $f : \mathbb{Z}^+ \rightarrow \mathbb{R}^+$  and  $g : \mathbb{Z}^+ \rightarrow \mathbb{R}^+$  are increasing functions. Is it always true that we have either  $f(n) = O(g(n))$  or  $f(n) = \Omega(g(n))$ ?

**I think the answer is no. Suppose  $f(n) = 2^n$ , while  $n=2k$  and  $f(n) = 2^{n-1} + 1$ , while  $n=2k+1$ ;  $g(n) = 2^{n-1} + 1$ , while  $n=2k$  and  $g(n) = 2^n$ , while  $n=2k+1$ . Apparently,  $f(n) = O(g(n))$ ,  $n=2k$ , while  $f(n) = \Omega(g(n))$ ,  $n=2k+1$ , which means we have neither at  $\mathbb{Z}^+$ .**

2. (25 points) Given an array  $A[1 \cdots n]$  of integers, a pair of indices  $(i, j)$  is an *inversion* if  $i < j$  and  $A[i] > A[j]$ . Design an algorithm that counts the number of inversions in  $O(n \log n)$  time.

**Suppose  $p$  is the number of inversions.**

**Devide the array  $A$  into two halves (let's say  $M, N$  with length  $m, n$ ) with indices  $(i, j)$ .**

**For  $i$  from 0 to  $m-1$ ,  $j$  from 0 to  $n-1$ , repeat the following steps:**

**1: Record  $\min(M_i, N_j)$  in  $A'$**

**2: If  $M_i < N_j$ , then  $p = p + j$  and  $i = i + 1$ ; If  $M_i > N_j$ , then  $j = j + 1$  (a bit counterintuitive but practical)**

**3: if  $i > m$ , break; or if  $j > n$ , then  $p = p + (m - i) * j$  and break.**

**4: Record the rest to  $A'$ .**

**Apparently,  $T(n) = 2T(n/2) + O(n)$ , which means  $T(n) = O(n \log n)$ .**

3. (25 points) Given an array of  $n$  integers  $x_1, x_2, \dots, x_n$ , there are queries of the following form: given an integer  $1 \leq k \leq n$ , you need to return the  $k$ -th smallest integers in the array. Obviously, if we use  $O(n \log n)$  time to preprocess the array by sorting it, we can answer each query in  $O(1)$  time. In the class, we see that each query can be answered in  $O(n)$  time without any preprocessing (the Median-of-the-Medians algorithm). Now, design an  $O(n)$  preprocessing algorithm so that you can answer each query in  $O(k)$  time. Suppose  $f$  is the preprocessing function, defined as follows:

**1:** Use the Median-of-the-Medians algorithm to find the median, denoted as  $m$ .

**2:** Divide the array into three parts:

- **L:** those smaller than  $m$ ;
- **M:** those equal to  $m$ ;
- **R:** those larger than  $m$ .

As is mentioned above, step 2 takes  $O(n)$  time.

**Repeat:** apply  $f$  on  $L$  till there is only one integer in  $L$ .

After the preprocess above, we can get an array of medians of  $L$ , which could be used as a mark in the searching step.

**Searching:**

- If  $k < |L|$ , in that  $L$  is an ordered array,  $T = O(k)$ ;
- If  $|L| < k < |L| + |M|$ , clearly  $k$  is the answer;
- If  $k > |L| + |M|$ , in that  $|R| < \frac{n}{2}$ ,  $T = O(\frac{n}{2}) < O(k)$ .

In conclusion, the searching step takes  $O(k)$  time, which satisfies the conditions.

4. (26 points) You are given a 2D discrete topographical map  $A[0, \dots, n-1; 0, \dots, m-1]$  representing a landscape. The number  $A[i, j]$  represents the altitude at position  $(i, j)$ . If it rains over the landscape, the water will form pools at each position where  $A[i, j]$  is less than each adjacent position, i.e., those  $A[i', j']$  for which  $|i - i'| + |j - j'| = 1$ . (You can assume all altitudes are distinct and that there is a “wall” at the edge of the map. For example, water pools at  $(0, 0)$  if  $A[0, 0]$  is less than  $A[0, 1]$  and  $A[1, 0]$ .)

- (a) Suppose  $m = 1$ , so  $A$  is a 1D array. Give a divide and conquer algorithm for finding *one* position where water pools. Write a recurrence for this algorithm. Analyze its running time.

**Algorithm:** Suppose  $\text{mid} = n//2$ , compare  $A[1, \text{mid}-1]$ ,  $A[1, \text{mid}]$  and  $A[1, \text{mid}+1]$ . Apparently, if  $A[1, \text{mid}-1] > A[1, \text{mid}]$  and  $A[1, \text{mid}+1] > A[1, \text{mid}]$ , then  $A[1, \text{mid}]$  is the pool.

Else, choose the smaller half and repeat the steps above.

**Recurrence:**  $T(n) = T(n/2) + O(1)$ .

Thus  $T(n) = O(\log n)$ .

- (b) Give another divide and conquer algorithm when  $m = n$ . Analyze its running time with a recurrence relation.

Because  $m=n$ , searching direction is still on a line, so actually (2) equals to (1).

**Algorithm:** Suppose  $\text{mid} = n//2 = m//2$ , compare  $A[\text{mid}, \text{mid}]$ ,  $A[\text{mid}+1, \text{mid}+1]$  and  $A[\text{mid}-1, \text{mid}-1]$ .

Apparently, if  $A[\text{mid}, \text{mid}] < A[\text{mid}+1, \text{mid}+1]$  and  $A[\text{mid}, \text{mid}] < A[\text{mid}-1, \text{mid}-1]$ , then  $A[\text{mid}, \text{mid}]$  is the pool.

Else, choose the smaller direction and repeat the steps above.

**Recurrence:**  $T(n) = T(n/2) + O(1)$ .

Thus  $T(n) = O(\log n)$ .

- (c) Generalize your algorithms from part (a) and (b) to work for any  $m$  and  $n$ . The running time should *smoothly* interpolate between the running times of (a) and (b). **Algorithm:** Suppose  $\text{mid}_m = m//2$ ,  $\text{mid}_n = n//2$ , compare  $A[\text{mid}_m, \text{mid}_n]$  with  $A[\text{mid}_m \pm 1, \text{mid}_n]$  and  $A[\text{mid}_m, \text{mid}_n \pm 1]$ .

If  $A[\text{mid}_m, \text{mid}_n]$  is minimum, then it's the pool.

Else, choose the smallest direction, and repeat the steps above.

**Recurrence:** for each step,  $T(m, n) = T(m/2, n) + O(n)$  or  $T(m, n/2) + O(m)$ .

Thus  $T(m, n) = O(m+n)$ .

5. How long does it take you to finish the assignment (including thinking and discussion)?  
Give a score (1,2,3,4,5) to the difficulty. Do you have any collaborators? Please write down their names here.

**About 7 hours.**

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**Ruifeng Shang.**