

## Data Structure and Algorithm

### Homework #1 Solution

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#### **Problem 1.** Basic Matrix Operation (30%)

Sample code is shown on the course website.

#### **Problem 2.** Asymptotic Notation (20%)

##### 2.1. (10%)

- (a) for  $i \in \{0, 1, 2, \dots, d\}$ ,  $a_i \cdot n^i \leq a_d \cdot n^k$  when  $n \geq \max\{1, \frac{a_i}{a_d}\}$

By summing the  $d+1$  inequalities, we can obtain that  $p(n) \leq (d+1) \cdot a_d \cdot n^k$  when

$$n \geq \frac{\max\{a_0, a_1, a_2, a_3, \dots, a_d\}}{a_d}.$$

$$\rightarrow p(n) = O(n^k)$$

- (b) The key idea is making  $|a_i| \cdot n^i \leq \frac{1}{d+1} \cdot a_d \cdot n^d$  for  $i \in \{0, 1, 2, \dots, d-1\}$

We know that the equality holds when  $n \geq \max\{1, (d+1) \cdot \frac{a_i}{a_d}\}$

$$\begin{aligned} p(n) &= \sum_{i=0}^d a_i \cdot n^i = a_d \cdot n^d + \sum_{i=0}^{d-1} a_i \cdot n^i \\ &\geq a_d \cdot n^d - \sum_{i=0}^{d-1} |a_i| \cdot n^i \\ &\geq a_d \cdot n^d - d \left( \frac{1}{d+1} \cdot a_d \cdot n^d \right) = \frac{1}{d+1} \cdot a_d \cdot n^d \\ &\geq \frac{1}{d+1} \cdot a_d \cdot n^k \text{ when } n \geq \max\{1, (d+1) \cdot \frac{a_0}{a_d}, (d+1) \cdot \frac{a_1}{a_d}, (d+1) \cdot \frac{a_2}{a_d}, \dots, (d+1) \cdot \frac{a_d}{a_d}\} \\ &\rightarrow p(n) = \Omega(n^k) \end{aligned}$$

- (c) We know that  $f(n) = \Theta(g(n))$  if and only if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$

(The proof is provided in 2.2(b))

From (a) and (b), we can obtain that  $p(n) = \Theta(n^k)$ .

- (d)  $\because k > d \therefore \lim_{n \rightarrow \infty} \frac{p(n)}{n^k} = 0$  (approach 0 from the right side)

by L'Hôpital's rule, for any  $\epsilon > 0$ , we can find an  $n_0$  s.t.  $0 \leq \frac{p(n)}{n^k} < \epsilon$  when  $n \geq n_0$ .

$\rightarrow$  for any  $c > 0$  we can find an  $n_0$  s.t.  $p(n) < c \cdot n^k$  when  $n \geq n_0$

$$\rightarrow p(n) = o(n^k)$$

- (e)  $\because k < d \therefore \lim_{n \rightarrow \infty} \frac{n^k}{p(n)} = 0$  (approach 0 from the right side)

by L'Hôpital's rule, for any  $\epsilon > 0$ , we can find an  $n_0$  s.t.  $0 \leq \frac{n^k}{p(n)} < \epsilon$  when  $n \geq n_0$ .

$\rightarrow$  for any  $c > 0$  we can find an  $n_0$  s.t.  $n^k < c \cdot p(n)$  when  $n \geq n_0$

$$\rightarrow p(n) = \omega(n^k)$$

##### 2.2. (10%)

- (a)  $f(n) = O(g(n))$

$\rightarrow$  We can find a  $c > 0$  and an  $n_0$  s.t.  $f(n) \leq c \cdot g(n)$  when  $n \geq n_0$

$\Rightarrow$  We can find a  $c' > 0$  and an  $n_1$  s.t.  $g(n) \geq c' \cdot f(n)$  when  $n \geq n_1$  ( $c' = \frac{1}{c}$ )

$\Rightarrow g(n) = \Omega(f(n))$

(b)  $f(n) = \Theta(g(n))$

$\Rightarrow$  We can find a  $c_1 > 0$ , a  $c_2 > 0$  and an  $n_0$  s.t.  $c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$  when  $n \geq n_0$

$\Rightarrow$  We can find a  $c_1 > 0$ , a  $c_2 > 0$  and an  $n_0$  s.t.  $f(n) \geq c_1 \cdot g(n)$  when  $n \geq n_0$  and  $f(n) \leq c_2 \cdot g(n)$  when  $n \geq n_0$

$\Rightarrow f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$

(c)  $\because g(n)$  is an asymptotically positive function.

$\therefore$  We can find an  $n_0$  s.t.  $g(n) > 0$  when  $n \geq n_0$

$\because f(n) = O(g(n)) \therefore$  We can find a  $c$  and an  $n_1$  s.t.  $f(n) \leq c \cdot g(n)$  when  $n \geq n_1$

$\rightarrow f(n) \cdot g(n) \leq c \cdot g(n) \cdot g(n)$  when  $n \geq \max\{n_0, n_1\}$

$\rightarrow f(n) \cdot g(n) = O((g(n))^2)$

(d)  $\because f(n)$  is an asymptotically positive function.

$\therefore$  We can find an  $n_0$  s.t.  $f(n) > 0$  when  $n \geq n_0$

$\because g(n)$  is an asymptotically positive function.

$\therefore$  We can find an  $n_1$  s.t.  $g(n) > 0$  when  $n \geq n_1$

$\because f(n) = O(g(n)) \therefore$  We can find a  $c$  and an  $n_2$  s.t.  $f(n) \leq c \cdot g(n)$  when  $n \geq n_2$

$\rightarrow f(n) \cdot f(n) \leq c \cdot f(n) \cdot g(n) \leq c \cdot (c \cdot g(n)) \cdot g(n)$  when  $n \geq \max\{n_0, n_1, n_2\}$

$\rightarrow (f(n))^2 = O((g(n))^2)$

(e) Let  $f(n) = 2 \cdot g(n)$ , then  $2^{f(n)} = 2^{2 \cdot g(n)} = 2^{g(n)} \cdot 2^{g(n)} = (2^{g(n)})^2$ .

It's easy to see that  $(2^{g(n)})^2 = O(2^{g(n)})$  doesn't necessarily hold for most cases.

$g(n) = \log_2 n$  is one of them.

**Problem 3.** Time and Space Complexities (15%)

3.1. (4%)  $O(m \log n)$ . Since the outer while-loop must run  $m$  times and the inner while-loop will run  $\log n$  times in worst case. All of other operations only runs in constant time.

3.2. (4%)  $O(n + m)$ . Since the for-loop must run  $n$  times and the while-loop must run  $m$  times. All of other operations only runs in constant time.

3.3. (3%)  $O(K)$ . Except the space cost of the input, the function uses three variables  $C, i, searchnum$ . The total size of them is  $K + 1 + 1 = O(K)$ .

3.4. (4%) *binary\_search*. Since  $m = O(1)$ , we can reduce the time complexity of each function. (  $T(\text{binary\_search}) = O(\log n), T(\text{count\_search}) = O(n)$  ) In result of comparison, *binary\_search* dominates *count\_search* in time and space both.

**Problem 4.** Stack and Queue (20%)

4.1. (5%) 5 4 3 1 2

4.2. (5%) 1 5 4 2 3

4.3. (6%) Set 1: stack. Set 2: queue. Set 3: neither.

4.4. (4%) Assume that the hidden data structure is a stack (queue) and simulate the operations.

If all the output from stack (queue) fit the output of sequence in problem description, then the data structure may be a stack (queue).

**Problem 5.** Scorched Toast (15%)

5.1. (5%)

```
1  Function brute-force{
2      maxArea = 0
3      for( upper = 1 ; upper ≤ M ; upper++ ){
4          for( lower = upper; lower ≤ M ; lower++ ){
5              for( left = 1 ; left ≤ N ; left++ ){
6                  for( right = left ; right ≤ N ; right++ ){
7                      flag = True
8                      for( i = upper ; i ≤ lower ; i++ ){
9                          for( j = left ; j ≤ right ; j++ ){
10                             if( Bij == 0 ){
11                                 flag = False
12                             }
13                         }
14                     }
15                     tmpArea = (lower - upper + 1) × (right - left + 1)
16                     if( flag == True and tmpArea > maxArea){
17                         maxArea = tmpArea
18                     }
19                 }
20             }
21         }
22     }
23     return maxArea
24 }
```

The first four loops, they cost  $O(M^2N^2)$  to enumerate the four boundaries of sub-rectangles. To check the status of each sub-rectangle, it needs  $O(MN)$  in the worst case. So the time complexity of this algorithm is  $O(M^2N^2) \times O(MN) = O(M^3N^3)$ .

5.2. (5%)

```

1  Function calculate-S{
2      for( i = 1 ; i ≤ M ; i++ ){
3          for( j = 1 ; j ≤ N ; j++ ){
4              if( i == 1 ){
5                  S(i,j) = Bij
6              }else{
7                  if( Bij == 1 ){
8                      S(i,j) = S(i-1,j) + 1
9                  }else{
10                     S(i,j) = 0
11                 }
12             }
13         }
14     }
15 }

```

5.3. (5%)

```

1  Function solve-by-row{
2      maxArea = 0
3      for( row = 1 ; row ≤ M ; row++ ){
4          for( left = 1 ; left ≤ N ; left++ ){
5              for( right = left ; right ≤ N ; right++ ){
6                  H = ∞
7                  for( j = left ; j ≤ right ; j++ ){
8                      if( S(row,j) < H ){
9                          H = S(row,j)
10                     }
11                 }
12                 tmpArea = H × (right - left + 1)
13                 if( flag == True and tmpArea > maxArea){
14                     maxArea = tmpArea
15                 }
16             }
17         }
18     }
19     return maxArea
20 }

```