

## 论题 2-1 作业

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### 1 [TC] Problem 2-1

- a.* For every sublist of length  $k$ , insertion sort can sort it in  $\Theta(k^2)$  worst-case time, and there are  $n/k$  sublists, so these sublists can be sorted by insertion sort in  $\Theta(k^2)n/k = \Theta(nk)$  worst-case time.
- b.* Apply the divide-and-conquer approach. Divide these sublists into two groups, each containing  $n/(2k)$  sublists, and merge these sublists recursively, and finally merge the two groups. Let  $m$  denote the number of the sublists, i.e.  $n/k$ , and  $T(m)$  denote the total running time of merging  $m$  sublists. The “divide”, “conquer” and “combine” steps take a running time of  $\Theta(1)$ ,  $2T(m/2)$ ,  $\Theta(km)$ , so the recurrence is

$$T(m) = \begin{cases} \Theta(1) & m = 1 \\ 2T(m/2) + \Theta(km) & m > 1 \end{cases}.$$

Solve this recurrence, we obtain  $T(m) = \Theta(km \lg(m)) = \Theta(n \lg(n/k))$ .

- c.* A standard merge sort takes a running time of  $\Theta(n \lg n)$ . When  $k = \Theta(\lg(n))$ ,  $\Theta(nk + n \lg(n/k)) = \Theta(n \lg n)$ . For every  $k = \omega(\lg n)$ ,  $\Theta(nk + n \lg(n/k)) = \Theta(nk) = \omega(n \lg n)$ . Therefore, the largest value of  $k$  is  $\Theta(\lg n)$ .
- d.* It mainly depends on the constant factors of merge sort and insertion sort when  $n$  is sufficiently large. Let  $c_1$  be the constant factor of merge sort,  $c_2$  be the constant factor of insertion sort. We can rewrite the total running time as  $T = c_1nk + c_2n \lg(n/k)$ . We should now minimize  $T$  with respect to  $k$ . Since  $T'_k = nc_1 - (nc_2)/(k \ln 2)$ ,  $k = c_2/(c_1 \ln 2)$  is a minimum point of  $T$ . So we can choose  $k = c_2/(c_1 \ln 2)$  in practice.

### 2 [TC] Problem 2-2

- a.*  $\langle A'[1], A'[2], \dots, A'[n] \rangle$  is a permutation of  $\langle A[1], A[2], \dots, A[n] \rangle$ .
- b.* At the start of each iteration, the subarray  $A[j..A.length]$  consists of the elements originally in  $A[j..A.length]$ , and  $A[j]$  is the smallest item of  $A[j..A.length]$ .

**Initialization** Prior to the first iteration of this loop, we have  $j = A.length$ . Therefore, the subarray  $A[j..A.length]$  consists only one element, and it is the original element in  $A[j..A.length]$ , and, of course, it is the smallest item of  $A[j..A.length]$ .

**Maintenance** In each iteration, we compare  $A[j]$  with  $A[j-1]$ . If  $A[j] \geq A[j-1]$ , we do nothing. Because  $A[j]$  is the smallest item of  $A[j..A.length]$ ,  $A[j-1]$  is the smallest item of  $A[j-1..A.length]$ . It  $A[j] >$

$A[j-1]$ , we exchange  $A[j]$  with  $A[j-1]$ . Because  $A[j]$  is the smallest item of  $A[j..A.length]$ ,  $A[j-1]$  is the smallest item of  $A[j-1..A.length]$  after exchanging, and  $A[j-1..A.length]$  consists of the elements originally from  $A[j-1..A.length]$ . Summarizing, the invariant still holds after an iteration.

**Termination** Finally, we get  $j = i$ . Therefore, the subarray  $A[i..A.length]$  consists of the elements originally in  $A[i..A.length]$ , and  $A[i]$  is the smallest item of  $A[i..A.length]$ , and that is what we want.

- c. At the start of each iteration,  $A[1..i]$  contains the  $i$  smallest elements of  $A[1..A.length]$ , in sorted order, and  $A[i+1..A.length]$  are the rest of the elements.

**Initialization** Prior to the first iteration, we have  $i = 1$ , and  $A[1..i]$  contains only one element, in sorted order, trivially. Moreover,  $A[i+1..A.length]$  are the rest of the elements obviously.

**Maintenance** After each iteration,  $A[i+1..A.length]$  consists of the elements originally in  $A[i+1..A.length]$ , and  $A[i+1]$  is the smallest item of  $A[i+1..A.length]$ . Since  $A[1..i]$  are the  $i$  smallest elements of  $A[1..A.length]$ ,  $A[i+1]$  is greater than or equal to every element in  $A[1..i]$ , but less than or equal to every element in  $A[i+2..A.length]$ . Therefore,  $A[1..i+1]$  contains the  $i+1$  smallest elements of  $A[1..A.length]$  in sorted order, and  $A[i+2..A.length]$  is the rest of the elements, i.e. the loop invariant still holds.

**Termination** Finally we get  $i = A.length$ , therefore  $A[1..A.length]$  contains the  $A.length$  smallest elements of  $A[1..A.length]$  in sorted order. Hence, the algorithm is correct.

- d. Whatever the original sequence is, lines 3-4 will always be executed  $n(n-1)/2$  times, where  $n$  is the number of the elements of the original sequence. Therefore the worst-case running time of bubble sort is  $\Theta(n^2)$ , as much as the worst-case running time of insertion sort.

### 3 [TC] Problem 2-3

- a.  $\Theta(n)$ .

- b. Given the coefficients  $a_0, a_1, \dots, a_n$  and a value for  $x$ :

```

1  y = 0
2  for i = 0 to n
3      t = 1
4      for j = 1 to i
5          t = t * x
6      y = y + a_i * t
```

The running time of this algorithm is  $\theta(n^2)$ , worse than Horner's rule.

- c. **Initialization** Prior to the first iteration, we have  $y = 0$  and  $i = n$ , so the loop invariant trivially holds.

**Maintenance** Assume, prior the  $t$ th iteration, we have  $i = i_t$  and  $y = y_t$ . After the iteration and incrementing  $i$ , we get  $i_{t+1} = i_t - 1$  and

$$y_{t+1} = a_{i_t} + x * y_t = a_{i_t} + x * \sum_{k=0}^{n-(i_t+1)} a_{k+i_t+1} x^k$$

$$\begin{aligned}
&= a_{i_t} + \sum_{k=0}^{n-(i_{t+1}+2)} a_{k+i_{t+1}+2} x^{k+1} = a_{i_{t+1}+1} + \sum_{k=1}^{n-(i_{t+1}+1)} a_{k+i_{t+1}+1} x^k \\
&= \sum_{k=0}^{n-(i_{t+1}+1)} a_{k+i_{t+1}+1} x^k,
\end{aligned}$$

therefore the invariant still holds.

**Termination** At termination, we have  $i = -1$ . Substituting  $i$  for  $-1$  in invariant, we obtain

$$y = \sum_{k=0}^n a_k x^k.$$

- d.* We have proved that the algorithm is partially correct. Note that the algorithm will be terminated after  $n + 1$  loops, so the algorithm is totally correct.

## 4 [TC] Problem 2-3

- a.*  $(2, 1), (3, 1), (8, 6), (8, 1), (6, 1)$ .

- b.*  $\langle n, n-1, \dots, 1 \rangle$ .  
 $n(n-1)/2$ .

- c.* Assume there are  $I(n)$  inversions in the input array, then the running time of insertion sort is  $\Theta(n + I(n))$ .  
 Proof: In page 26,  $t_j$  stands for the number of times the **while** loop test is executed for that value of  $j$ . Every time we execute the **while** loop, we insert  $A[j]$  into the correct position, and exactly  $t_j - 1$  inversions are eliminated. Finally, all the inversions are eliminated, and the array is sorted in order. Substituting  $\sum_{j=2}^n (t_j - 1)$  for  $I(n)$  and rewriting the formula, we get

$$T(n) = an + bI(n) + c,$$

therefore the running time of insertion sort is  $\Theta(n + I(n))$ .

- d.* Let  $c$  be an integer representing the number of inversions.

1  $c = 0$

2 MODIFIED-MERGE-SORT( $A, 1, A.length$ )

MODIFIED-MERGE-SORT( $A, p, r$ )

1 **if**  $p < r$

2      $q = \lfloor (p+r)/2 \rfloor$

3     MODIFIED-MERGE-SORT( $A, p, q$ )

4     MODIFIED-MERGE-SORT( $A, q+1, r$ )

5     MODIFIED-MERGE( $A, p, q, r$ )

MODIFIED-MERGE( $A, p, q, r$ )

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1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  Let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else
17          $A[k] = R[j]$ 
18         // Add the number of inversions  $(A[i], R[j]), (A[i + 1], A[j]) \dots (A[q], A[j])$  to  $c$ 
19          $c = c + (q - i + 1)$ 
20      $j = j + 1$ 

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## 5 [TC] Problem 3-2

$A$	$B$	$O$	$o$	$\Omega$	$\omega$	$\Theta$
$\lg^k n$	$n^\epsilon$	yes	yes	no	no	no
$n^k$	$c^n$	yes	yes	no	no	no
$\sqrt{n}$	$n^{\sin n}$	no	no	no	no	no
$2^n$	$2^{n/2}$	no	no	yes	yes	no
$n^{\lg c}$	$c^{\lg n}$	yes	no	yes	no	yes
$\lg(n!)$	$\lg(n^n)$	yes	no	yes	no	yes

## 6 [TC] Problem 3-3

a. List:

$2^{2^{n+1}}$	$2^{2^n}$	$(n+1)!$	$n!$	$n2^n$	$e^n$
$2^n$	$(\frac{3}{2})^n$	$n^{\lg \lg n}$	$(\lg n)^{\lg n}$	$(\lg n)!$	$\sqrt{2}^{\lg n}$
$n^3$	$4^{\lg n}$	$n^2$	$n \lg n$	$2^{\lg n}$	$n$
$2^{\sqrt{2} \lg n}$	$\lg^2 n$	$\lg(n!)$	$\ln n$	$\sqrt{\lg n}$	$\ln \ln n$
$2^{\lg^* n}$	$\lg^* n$	$\lg^*(\lg n)$	$\lg(\lg^* n)$	$n^{1/\lg n}$	1

Equivalence classes:  $\{2^{2^{n+1}}\}$ ,  $\{2^{2^n}\}$ ,  $\{(n+1)!\}$ ,  $\{n!\}$ ,  $\{n2^n\}$ ,  $\{e^n\}$ ,  $\{2^n\}$ ,  $\{(\frac{3}{2})^n\}$ ,  $\{(\lg n)^{\lg n}, n^{\lg \lg n}\}$ ,  $\{(\lg n)!\}$ ,  $\{\sqrt{2}^{\lg n}\}$ ,  $n^3$ ,  $\{n^2, 4^{\lg n}\}$ ,  $\{n \lg n\}$ ,  $\{n, 2^{\lg n}\}$ ,  $\{2^{\sqrt{2 \lg n}}\}$ ,  $\{\lg^2 n\}$ ,  $\{\lg(n!)\}$ ,  $\{\sqrt{\lg n}\}$ ,  $\{\ln n\}$ ,  $\{\ln \ln n\}$ ,  $\{2^{\lg^* n}\}$ ,  $\{\lg^*(\lg n), \lg^* n\}$ ,  $\{\lg(\lg^* n)\}$ ,  $\{1, n^{1/\lg n}\}$ .

**b.**  $f(n) = (2^{2^{n+2}})^{\sin n}$

## 7 [TC] Problem 3-4

**a.** False. Take  $f(n) = n$ ,  $g(n) = n^2$ ,  $f(n) = O(g(n))$ , but  $g(n) \neq O(f(n))$ .

**b.** False. Take  $f(n) = n$ ,  $g(n) = n^2$ ,  $f(n) + g(n) = n + n^2 = \Theta(n^2) \neq \Theta(\min(f(n), g(n))) = \Theta(n)$ .

**c.** True. Since  $f(n) = O(g(n))$ , there exists positive constant  $c > 1$  such that for all sufficiently large  $n$ ,  $1 \leq f(n) \leq cg(n)$  holds. Therefore,  $\lg(f(n)) \leq \lg c + \lg(g(n))$ . Take  $c' = \lg c + 1$ , then  $0 \leq \lg(f(n)) \leq c' \lg(g(n))$  for all sufficiently large  $n$ . Hence,  $\lg(f(n)) = O(\lg(g(n)))$ .

**d.** False. Take  $f(n) = n \lg n$ ,  $g(n) = \lg(n!)$ ,  $f(n) = O(g(n))$ , however,  $2^{f(n)} = n^n \neq O(2^{g(n)}) = O(n!)$ .

**e.** False. Take  $f(n) = 1/n$ , then  $(f(n))^2 = 1/n^2$ , however,  $\lim_{n \rightarrow \infty} f(n)/(f(n))^2 = +\infty$ , that means,  $f(n)$  could not be asymptotically upper-bounded by  $(f(n))^2$ .

**f.** True. By transpose symmetry we know this is true.

**g.** False. Take  $f(n) = 4^n$ , then  $\Theta(f(n/2)) = \Theta(2^n)$ , however  $4^n \neq \Theta(2^n)$ .

**h.** True. By the definition of  $o$ -notation, for any positive constant  $c$ , there exists a positive integer  $n_0$ , for any integer  $n > n_0$ ,  $0 \leq o(f(n)) \leq cf(n)$  holds. For sufficiently large positive integer  $n$ ,  $f(n) \leq f(n) + o(f(n)) \leq (1+c)f(n)$ , therefore  $f(n) + o(f(n)) = \Theta(f(n))$ .