Illinois Institute of Technology Department of Computer Science

Solutions to Second Examination

CS 430 Introduction to Algorithms Spring, 2016

Wednesday, February 15, 2016 10am-11:15am & 11:25am-12:40pm, 111 Life Sciences

Exam Statistics

114 students took the exam. The range of scores was 3–76, with a mean of 30.99, a median of 25, and a standard deviation of 18.18. The exam was clearly too hard, but not as hard as these statistics indicate: about dozen students got no credit on some problems because they brought in forbidden materials to the exam. Very roughly speaking, if I had to assign final grades on the basis of this exam only, 49 and above would be an A (22), 30–48 a B (28), 15–29 a C (44), 10–14 a D (11), below 10 an E (9).

Problem Solutions

1. (a) Using the Principle of Optimality, we can express L_i recursively as

$$L_j = \begin{cases} 1 & \text{if } j = 1, \\ \max\{L_{j-1}, \max_{\substack{1 \le i < j \\ a_{A_i} < a_j}} \{1 + L_i\}\} & \text{otherwise.} \end{cases}$$

Of course the values of A_j and B_j must be maintained in this recursive definition (that is, which of the various possibilities in the max was maximum with the lowest value of A_j and what the corresponding value of B_j is).

Without the hint, we might be tempted to express the recurrence looking at all increasing subsequences that have a_i as a member, recursively obtain longest increasing subsequences on the left and on the right, and the choose the combination with the overall maximum length. Writing the recurrence (and hence recursive code) is tricky because the element we insist on, a_i , must be larger than the largest element of the increasing subsequence on the left and smaller than the least element of the increasing subsequence on the right. This approach would mean adding two parameters to the recurrence so that it works out properly—a value that is an upper bound for a possible left increasing sequence and a a value that is a lower bound for a possible right increasing sequence.

(b) Let c_j be the rate of growth of the cost of evaluating L_j using the recurrence in (a). We

have

$$c_j = \begin{cases} 1 & \text{if } j = 1, \\ 1 + \sum_{1 \le i < j} c_i & \text{otherwise.} \end{cases}$$

Looking at the difference between successive values (as we did in the Quicksort recurrence to eliminate the summation), we find that $c_j - c_{j-1} = c_{j-1}$ implying $c_j = 2c_{j-1}$ and hence $c_j = 2^{j-1}$; we could also guess at this solution by computing the first few values and then verifying the guess by induction. Thus computing L_n would cost $\Theta(2^n)$.

(c) Following the hint, to memoize (a) we use an array L[j] to tell us the length of the LIS among a_1, a_2, \ldots, a_j , an array A[j] to tell us the index of the smallest possible largest (last) element of the LIS among a_1, a_2, \ldots, a_j , and an array B[j] to tell us the index of the smallest possible second largest element of the LIS among a_1, a_2, \ldots, a_j . Define a dummy element $a_0 = -\infty$; then L[1] = 1, A[1] = 1, and B[1] = 0. We have:

```
1: a_0 = -\infty
2: L[1] = 1
3: A[1] = 1
4: B[1] = 0
5: for j = 2 to n do
       L[j] \leftarrow L[j-1]
       A[j] \leftarrow A[j-1]
       B[j] \leftarrow B[j-1]
8:
       for i = 1 to j - 1 do
9:
         if L[j] = L[i] and a_{A[j]} > a_{A[i]} > a_{B[j]} then
10:
            // a_{A[i]} gives us a lower largest value for L[j]
11:
            A[j] \leftarrow A[i]
12:
         else if L[j] = L[i] and a_{A[j]} > a_j > a_{B[j]} then
13:
            // a_i gives us a lower largest value for L[j]
14:
15:
            A[j] \leftarrow j
         else if L[j] \leq L[i] and a_i > a_{A[i]} then
16:
            // we can add a_j to end of L[i]
17:
            L[j] \leftarrow L[i] + 1
18:
19:
             B[j] \leftarrow A[j]
20:
         end if
21:
22:
       end for
23: end for
```

- (d) The cost of this double-nested loop computation is $\sum_{1 < j \le n} \sum_{1 \le i < j} 1 = \sum_{1 < j \le n} (j-1) = n(n-1)/2 = \Theta(n^2)$.
- (e) We recover the LIS itself from the A[j] and B[j] values as follows:

```
1: procedure WriteLIS(i)
2: if i > 0 then
3: if A[i] = i then
```

```
WriteLIS(B[i])
4:
        SystemPrint(a_i)
5:
6:
     else
        WriteLIS(A[i])
7:
     end if
8:
9: end if
```

- (a) Take $r_1 = 2$, $r_2 = 3$, and $r_3 = 4$. Buying the equipment in increasing order has a total $\cos t \ 100(2+3^2+4^3) = 7500$, but in reverse order the total cost is $100(4+3^2+2^3) = 2100$.
 - (b) Following the hint, suppose there is an optimal solution O that differs from our highestcost-first solution S. Then O must contain an "inversion" in which two successive months, t and t+1, in which the increase rate of the equipment bought in month t, r_t is less than the rate for the equipment bought in month t+1, r_{t+1} ; that is, $r_t < r_{t+1}$. Swapping these two purchases saves us

$$100(r_t^t + r_{t+1}^{t+1}) - 100(r_{t+1}^t + r_t^{t+1}).$$

We want to show this is a positive amount; that is, $r_t^t+r_{t+1}^{t+1}>r_{t+1}^t+r_t^{t+1}$ Rearranging, we need $r_{t+1}^{t+1}-r_{t+1}^t>r_t^{t+1}-r_t^t$

$$r_t^t + r_{t+1}^{t+1} > r_{t+1}^t + r_t^{t+1}$$

$$r_{t+1}^{t+1} - r_{t+1}^{t} > r_t^{t+1} - r$$

or

$$r_{t+1}^t(r_{t+1}-1) > r_t^t(r_t-1),$$

which is true because $r_i > 1$ for all i and $r_t < r_{t+1}$.

- (c) The running time of the algorithm is dominated by $\Theta(n \log n)$, the cost of sorting the equipment into decreasing order by rate; the rest of the algorithm is just outputting the result.
- 3. Let n be the number of elements on the stack before an operation. The calculations are nearly identical to those done in chapter 17 of CLRS (lecture notes of February 29):

$$\begin{array}{lll} {\rm AMORT_{POP}} & = & {\rm ACTUAL_{POP}} + |{\rm stack}|_{after}^2 - |{\rm stack}|_{before}^2 \\ & = & 1 + (n-1)^2 - n^2 = -2(n-1) \leq 0 = O(n). \\ {\rm AMORT_{PUSH}} & = & {\rm ACTUAL_{PUSH}} + |{\rm stack}|_{after}^2 - |{\rm stack}|_{before}^2 \\ & = & 1 + (n+1)^2 - n^2 = 2(n+1) = O(n). \\ {\rm AMORT_{MULTIPOP}}(k) & = & {\rm ACTUAL_{MULTIPOP}}(k) + |{\rm stack}|_{after}^2 - |{\rm stack}|_{before}^2 \\ & = & k(k-1)/2 + (n-k)^2 - n^2 \\ & = & k^2/2 - k/2 - 2kn \\ & = & k(k/2 - 2n - 1/2) < 0 = O(n), \end{array}$$

because $k \leq n$.

4. We want a complexity of $\Omega(m \log n)$, therefore a reasonable guess is that the disjoint-set tree generated with union by rank has a depth $\Omega(\log n)$. Such a tree can be easily achieved. Suppose that, without the loss of generality, n is a power of 2 and we have $\{x_1, x_2, \ldots, x_n\}$ since there are n Make-Set operations. Then, because our Union is a union by rank, we need to union small trees as soon as possible so that we won't have a tree where all the small trees are linked to one single tree. In an extreme case, we may end up with a tree of depth 1, where n-1 nodes are linked to one node with is the root. Therefore, after preparing the set by running n Make-Set operations on $\{x_1, x_2, \cdots, x_n\}$, we run Union $\{x_1, x_2\}$, Union $\{x_3, x_4\}$, Union $\{x_5, x_6\}$ and so forth in the first round. After all individual nodes are unioned, we union the trees by running Union $\{x_1, x_3\}$, Union $\{x_5, x_7\}$ and so forth by unioning each pair of consecutive trees in the second round. This is repeated until there is only a single tree (set) left, at which time there were $\log n$ rounds of unions. It is easy to see we have performed n-1 Union operations $(\frac{n}{2} + \frac{n}{4} + \cdots + 2 + 1)$, and the tree has a depth $O(\log n)$ because at each round the depth increase by 1. So far, we have n Make-Set operations and n-1 Union operations.

Suppose we have k FIND-SET operations. If we perform FIND-SET on the node of greatest depth, each FIND-SET will be $\Theta(\log n)$. Then, the total complexity of m = 2n - 1 + k operations will be $\Omega(n + n - 1 + k \log n)$. When $k \ge n$, $k = \Theta(m)$, and we have $\Omega(2n - 1 + k \log n) = \Omega(m \log n)$.