## Illinois Institute of Technology Department of Computer Science

## Solutions to Second Examination

CS 430 Introduction to Algorithms Spring, 2018

Wednesday, March 7, 2018 10am–11:15am & 11:25am–12:40pm 111 Robert A. Pritzker Science Center

## **Exam Statistics**

XXX students took the exam; there were XX no-shows recorded as 0. The range of scores was XX–XX, with a mean of XX.XX (this excludes the no-shows), a median of XX, and a standard deviation of XX.XX. Very roughly speaking, if I had to assign final grades on the basis of this exam only, above XX would be an A (XX), XX–XX a B (XX), XX–XX a C (XX), XX–XX a D (XX), below XX an E (XX). Every student should have been able to get almost full credit on the first and fourth problems, plus a dozen or so points on the remaining problems; no score should have been below 50.

## **Problem Solutions**

1. We saw in class in Lecture 7 (January 31), that the external path length can equivalently be defined recursively as

$$EPL(\square) = 0$$
  
 $EPL(x) = EPL(x.LEFT) + EPL(x.RIGHT) + n + 1$ 

where x.LEFT and x.RIGHT are the left and right subtrees, respectively, of x, and n is the number of internal nodes in the subtree rooted at x. Thus EPL(x) does not satisfy the hypothesis of Theorem 14.1 (page 346 in CLRS), namely that EPL(x) depend only on EPL(x.LEFT) and EPL(x.RIGHT); hence we cannot conclude that it can be maintained in a red-black tree. But Theorem 14.1 only gives us sufficient conditions for maintainence, not necessary conditions.

If we knew the number of internal nodes in the subtree rooted at x, then EPL(x) would satisfy the hypothesis of Theorem 14.1. So, denote by SIZE(x) the number of internal nodes in the subtree rooted at x. SIZE(x) can be written recursively as

$$SIZE(\square) = 0$$
  
 $SIZE(x) = SIZE(x.LEFT) + SIZE(x.RIGHT) + 1,$ 

and hence SIZE(x) satisfies the hypothesis of Theorem 14.1 and can be maintained in a red-black tree. Thus by maintaining SIZE(x) we can also maintain EPL(x) in a red-black tree.

2. (a) Let  $S_{ij}$  be the largest sum that can be reached from row i, diagonal j. Then,

$$S_{ij} = \begin{cases} T_{ij} & \text{if } i = N \\ T_{ij} + \max(S_{i+1,j}, S_{i+1,j+1}), & \text{if } i < N \end{cases}$$

The largest sum on a path from the apex to the bottom is  $S_{00}$ .

(b) Let  $t_{ij}$  be the time to compute  $S_{ij}$ . Then,

$$t_{ij} = \begin{cases} O(1) & \text{if } i = N \\ O(1) + t_{i+1,j} + t_{i+1,j+1}, & \text{if } i < N \end{cases}$$

Induction on i proves that  $t_{N-i,k} = \Theta(2^i)$  for all i.

- (c) Memoizing the computed values of  $S_{ij}$  means that O(1) time is required for each  $t_{ij}$ , assuming the row below has already been computed. There are about  $N^2/2$  elements in the triangular array, so the time with memoization is  $O(N^2)$ .
- (d) To keep track of the path giving the largest sum, we add a direction to the memo:  $d_{ij}$  is "left" or "right" according to which element is the max in the second line of the equation for  $S_{ij}$ .
- 3. (a) Here are three jobs for which the stated greedy method schedules 1 job, and clearly that is the best possible because the three jobs all conflict:

- (b) The greedy algorithm always gives the optimum schedule: You can either argue parallel to the proof in the text (or lecture) or just transform the jobs  $\{(s_i, f_i)\}$  to the mirror image set of jobs  $\{(-f_i, -s_i)\}$ . Then by the proof for the finishing-time-first version, the greedy algorithm always gives the optimum schedule.
- 4. Take the potential function to be  $\Phi(D_i) = 2b_i$ , twice the number of 1-bits after the *i*th increment. Then, paralleling the computation at the bottom of page 461 in CLRS3,

$$\Phi(D_i) - \Phi(D_{i-1}) \le 2[(b_{i-1} - t_i) - b_{i+1}] = 2(1 - t_i).$$

and the amortized cost of an increment operation is

$$\hat{c}_{i} = c_{i} + \Phi(D_{i}) - \Phi(D_{i-1}) 
= (t_{i} + 1 + T(i)) + \Phi(D_{i}) - \Phi(D_{i-1}) 
\leq t_{i} + 1 + t_{i} + 2(1 - t_{i}) 
= 3$$

Thus the amortized time remains O(1) even with significant time wasting in line 9.