

CSC263: Assignment 1

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Question 1

(a)

$T(n)$ is $O(n^2)$.

Justify: For every input, array A of size n that contains at least $n \geq 2$ arbitrary integers, the outer while loop (line 3 – 13) takes at most $(n - 1)$ iterations. And each such iteration can make the inner while loop (line 5 – 8) execute at most $(n - 2)$ times. Thus, the whole loop is executed at most $(n - 1)(n - 2)$ times, and since statements at all other lines execute constant time, so executing the procedure `doStuff` takes at most n^2 time. Then, $T(n)$ is $O(n^2)$.

(b)

$T(n)$ is $\Omega(n^2)$.

Justify: As a worst case, the outer while loop (line 3 – 13) should not end early. So there exists an array A of size $n \geq 2$, which contains n 1's, namely $A[1..n] = \{1, 1, \dots, 1, 1\}$. For this input array A , the total runtime of the loop (line 3 – 13) is $1 + 2 + 3 + \dots + (n - 2) = \frac{[1+(n-2)](n-2)}{2} = \frac{(n-1)(n-2)}{2}$, and since statements at all other lines execute constant time, so there are at least proportional to n^2 steps, thus executing the procedure `doStuff` takes at least n^2 time. Then, $T(n)$ is $\Omega(n^2)$.

Question 2

(a)

The data structure I used is Max Heap.

My algorithm will be 3 parts:

A. Build-up an array S of ordered x position points of given rectangles.

Each element is a tuple of (s, h, e, ptToRectangle, ableToOutput) which:

- s: the x positions of given rectangles (including both left and right edges) in an increasing order (if duplicate s value occur, the order depends on l/r by the rules mentioned below)
- h: the height of these points in relative rectangles, (this h will be updated in part B as the exact height of required outline points).
- e: represent this point is either left edge x position (denote as L) or right edge (denote as R)
- ptToRectangle: a pointer points to the relative given rectangle.
- ableToOutput: represents if this point is our final output point, while set ableToOutput as TRUE now (this TRUE will be updated in part B as if it exactly can be output.)

We can first make an array S contains all the x position of given rectangles as the point tuple of (s, h, e, ptToRectangle, ableToOutput), that is each rectangle T0 (x1, x2, h) making 2 point tuples (x1, h, L, T0, TRUE) and (x2, h, R, T0, TRUE), and then use Heap-Sort to make these tuples as increasing x orders (the sorting KEY value is the first element value in the tuple). Notice that if duplicated KEY values occur, we sort the node according to the edge e in order to make part B efficient to reduce duplication:

- a) left edge x position (e = L) is always prior to the right edge (e = R)
- b) if both left edge (e = L), greater height h valued point has priority.
- c) if both right edge (e = R), smaller height h valued point has priority.

Then we store these (s, h, e, ptToRectangle, ableToOutput) in the array S. And since the given array of rectangles is sized n, there are 2n x positions to be sorted, then the size of S should be 2n.

Now each point in S should be all our potential output points since the outline point's x position must be either one of the rectangle edge positions, then S is ready for being updated.

B. Build up Searching Max Heap H to get the required output x position's height.

Our Searching Max Heap H is actually tracking the overlapping sets of given rectangles for each potential output point in S, that is to say, each point in S will be included in some rectangles, and we only need to get the highest height among such overlapping rectangles for each point. Each node of H represents

a given rectangle and the KEY is the height of this rectangle. We update our Max Heap in the loop.

So for each point p in S :

- if p is the left edge of a rectangle ($p.e = L$), use the insert operation of Heap to insert this rectangle ($p.ptToRectangle$) as a node in H

- if p is the right edge of a rectangle ($p.e = R$), remove $p.ptToRectangle$ from our Heap H and swap nodes relatively to always keep the Max-Heapify rule

Then use the getMax operation of Heap to get the value h_0 from our current Heap H and set it as the p 's height h . If our Heap H is empty, set p 's height h as 0.

Now we reduce the duplication: 1), we only need each x position once; 2) if two adjacent x positions has the same height. So we also need a pointer to keep track the scanning point after the previous loop, denote as $prePoint$.

- if current $p.s == prePoint.s$, set $prePoint.ableToOutput$ as FALSE (we only need the latest point of same x position points);

- if current $p.h == prePoint.h$, set $p.ableToOutput$ as FALSE (we only need the first point of same height points);

Update the $prePoint$ as p and end the loop.

Now our array S is updated and ready to output.

C. Output

Finally, we output all the x position s and height h of the point p in S as a tuple (s, h) if $p.ableToOutput$ is TRUE, and the array of all satisfied points is what we need.

(b)

I will explain the worst-case time complexity related to my algorithm.

In part A we use heap sort on $2n$ keys, so we just do $2n$ times insert operation of Heap (which is $O(\log n)$ per insert that we learned from class) , notice that the updated key comparison rules cos constant time per insert, and $2n$ times extract-min operations (which is $O(1)$ per extract that we learned from class), so the total cost of this part is $O(n \log n)$.

In part B, for the outer loop, we execute $2n$ times loops to update the array S . For each inner loop, we do either one insert operation of Heap or Delate operation of Heap while keeping the Binary Tree property and Max-Heapify property (which is both $O(\log n)$ per operation that we learned from class), then do getMax operation of Heap (which is $O(1)$ per operation that we learned from class), and the duplication tracking costs constant time. So the total cost of this part is $O(n \log n)$.

In part C we visit all $2n$ elements in array S , each visit cost constant time to output the left end point, so the total cost of this part is $O(n)$.

Consider all parts, the total cost of my algorithm is $O(n \log n)$.

Question 3

(a)

Let $\text{Binary}(n)$ denote the binary representation of n . According to CLRS, for the binomial tree B_k , there are 2^k nodes, which means each binomial tree has a size which is a power of 2 and the binomial trees required to represent n nodes are uniquely determined. Since we can have at most one of each binomial trees and at most 1 as any digit of $\text{Binary}(n)$, we can use $\text{Binary}(n)$ to represent the number of binomial trees in a binomial heap — include B_k if and only if the k th position of $\text{Binary}(n)$ is 1.

Therefore, a binomial heap with n elements has $\alpha(n)$ binomial trees, where $\alpha(n)$ is the number of 1's in the binary representation of n . Let T_i denote the trees of the binomial heap, where $1 \leq i \leq \alpha(n)$. Since the number of edges in a tree is the number of vertices minus one, T_i with n_i vertices has $n_i - 1$ edges.

$$\sum_{i=1}^{\alpha(n)} n_i - 1 = \sum_{i=1}^{\alpha(n)} n_i - \alpha(n) = n - \alpha(n)$$

Therefore, the total number of edges in the binomial heap is exactly $n - \alpha(n)$.

(b)

Since we do pairwise comparisons when we union two separate binomial trees, the number of pairwise comparisons between the keys of the binomial heap that is required to do k consecutive insertions is equal to the number of new edges created during these insertions.

By part (a), the binomial heap H has $n - \alpha(n)$ edges before insertion. After k consecutive insertions, H with $(n+k)$ vertices has $(n+k) - \alpha(n+k)$ edges. So the number of new edges created equals to $(n+k) - \alpha(n+k) - (n - \alpha(n)) = k + \alpha(n) - \alpha(n+k)$.

Since $\alpha(n)$ is the number of 1's in the binary representation of n , $n \leq \lfloor \log_2 n \rfloor + 1$. Therefore, k consecutive insertions require at most $k + \alpha(n)$ comparisons, which is $k + \lfloor \log_2 n \rfloor + 1$ comparisons. When $k > \log_2 n$, k dominates the equation $k + \lfloor \log_2 n \rfloor + 1$ so that k consecutive insertions require $O(k)$ pairwise comparisons.