ESO207: Theoretical Assignment 1 - Part 2

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Problem 5

a) Let $f(n) = \min(n^2, 10^{12})$, g(n) = 1 and $M = 10^{12} + 1$. Then,

$$f(n) \le Mg(n), \forall n > 1, n \in \mathbb{N}$$

 $\implies f(n) = \mathcal{O}(g(n)) = \mathcal{O}(1)$

Hence Proved

b) Let $f(n) = n^2 + n \log n$. Now,

$$\log(n) < n, \forall n \in \mathbb{N}$$

$$\implies n \log n < n^2, \forall n \in \mathbb{N}$$

$$\therefore f(n) = n^2 + n \log n < 2n^2 \implies f(n) = \mathcal{O}(n^2)$$

Hence Proved

c) Proof by contraction.

Let us assume that $f(n) = n^3 + 3n^2 + 8 = \mathcal{O}(g(n)) = \mathcal{O}(n^2)$, where $g(n) = n^2$ then,

$$\exists M, n_0 \in \mathbb{N} \text{ such that } f(n) \leq Mn^2, \forall n \geq n_0, n \in \mathbb{N}$$

Consider k = Mn.

 $f(Mn) = M^3n^3 + 3M^2n^2 + 8 \ge M^2n^2 = g(Mn), \forall n > 1, n \in \mathbb{N}$ which is a clear contraction to our assumed statement.

Hence,
$$f(n) = n^3 + 3n^2 + 8 \neq \mathcal{O}(n^2)$$

d) Proof by contraction.

Let us assume that $f(n) = 4^n = \mathcal{O}(2^n)$ Then,

$$\exists M, n_0 \in \mathbb{N} \text{ such that } f(n) \leq M2^n, \forall n \geq n_0, n \in \mathbb{N}$$

$$2^{2n} \le M2^n$$

$$2^n \le M$$

Let $n_1 > n_0 + \log M$ then,

$$2^{n_1-n_0} > M$$

 $forn_1 > n_0$ contraction.

∴ our assumption is wrong.

Hence, $4^n \neq \mathcal{O}(2^n)$

e) $n! = n \cdot (n-1) \cdot (n-1) \cdot \dots \cdot 2 \cdot 1 \le n \cdot n \cdot \dots \cdot n \cdot n = n^n, \forall n \in \mathbb{N}$

$$\therefore \log(n!) \le \log(n^n) = n \log n, \forall n \in \mathbb{N} : \log(n!) = \mathcal{O}(n \log n)$$

Hence Proved

Problem 6

Description:

The algorithm is based off the divide and conquer paradigm. To understand my approach better consider the case where the array B is flipped (B'[k] = B[n-1-k]). Then the algorithm reduces to finding the index at which A[k] = B'[k]. This can be done with a divide and conquer algorithm which will take $T = \mathcal{O}(\log n)$ time. As the arrays are sorted we use the condition (1) A[k] > B[n-1-k]. If (1) is true, set R = k-1 else set L = K+1. The while loop is terminated when L = R and returns a -1 (failure) or when A[k] = B[n-1-k] and returns the value of k.

Algorithm:

```
function find_symmetric(A, B, n){
2
           L <- 0;
           R <- n-1;
3
           found <- false;
4
            i <- -1;
            while (!found){
6
                m < - (L+R)/2;
7
                if (A[mid] = B[n-1-mid]){
                                                         \\Found
8
                    i <- mid;
9
10
                    found <- true;
                }
                else {
12
13
                    if (L == R){
                                                         \\Not found
                        i <- -1;
14
                         found <- true;
16
                       (A[mid]>B[n-1-mid]){
                                                         \\Condition
                    if
18
                         R <- mid-1;
19
                    }
                    else {
20
                         L <- mid+1;
21
22
23
           }
24
           return i:
25
```

Listing 1: Find symmetric

Proof of correctness:

The two arrays A and B are sorted (assume increasing without loss of generality)

Assertion P(i): If there is a common element, it exists in the set $\{A[k]: L \leq k \leq R\}$ Base case (i = 0): L = 0, R = n - 1. If there exists a common element it belongs to A and B.

Define B'[k] = B[n-1-k]. As B is sorted in increasing order, B' is sorted in decreasing order.

Induction step: Assume P(i) to be true. Then consider the middle element m=(L+R)/2. If A[m] < B'[m] then A[k] < B'[k] for all k < m as A is increasing and B' is decreasing. Then let R=m-1 and divide again. If A[m] > B'[m] then A[k] > B'[k] for all k > m as A is increasing and B' is decreasing. Then let L=m+1 and divide again. If A[m] = B'[m] we have found our point. If L=R and $A[m] \neq B'[m]$ then there exists no such point. Return -1.

Problem 7

Description:

Initially lines A and B are parallel but oriented in some direction. Find slope θ and apply coordinate transformation $(T = \mathcal{O}(n))$ to orient new axes such that lines are parallel to new x axis. Then for each line y' = constant. Sort the points in time $T = \mathcal{O}(n \log n)$ according to their x coordinate. Traverse along the two lines together and keep comparing to get points a_{min} and b_{min} $(T = \mathcal{O}(n))$ apply inverse coordinate transformation to bring back into original form.

Algorithm:

```
function return_closest(A, B){
           theta <- Calculate slope from 2 points of A such that A is parallel to new x axis.
           R <- Rotation matrix from theta
3
           C <- Ra for a in A
                                                                   \T = O(n)
4
           D <- Rb for b in B
                                                                   \T = O(n)
6
           A_s <- Sort C according to x values (y is constant)
                                                                    \T = O(nlogn)
           B_s <- Sort D according to x values (y is constant)
                                                                   \T = O(nlogn)
           i <- 0;
           j <- 0;
9
           min_distance = MAX_int;
           a=0;
11
           b=0;
           while (i!=n-1 \&\& j!=n-1){
                                                                   \\Hasn't reached edges
13
               if(distance(A_s[i],B_s[j])<min_distance){</pre>
                                                                   \T = 0(2n) = 0(n)
14
                   min_distance = distance(A_s[i],B_s[j]);
15
                   a = i;
16
                   b = j
17
18
               if(A_s[i] < B_s[j] && i!=n-1){
19
                   i++;
20
21
               elseif(j!=n-1){
22
                   j++;
23
24
25
           R' <- Inverse rotation matrix.
                                                 \\Inverse rotation matrix
                                                 \\Rotate found points back into desired coordinates
           Return RA_s[a], RB_s[b];
26
```

Listing 2: Closest points

Proof of correctness:

We first apply a rotation to bring the points A and B to be parallel to the new axis. This allows for easier manipulation as one of the coordinates is made constant. The slope can be found in $\mathcal{O}(1)$ time. The matrix R can be calculated in $\mathcal{O}(1)$ time. Multiplication of the matrix R with a point P takes $\mathcal{O}(1)$ time. Hence RA (R applied to all elements in A) takes time $\mathcal{O}(n)$

We consider C = RA and D = RB. We then proceed to sort C and D according to their x values. This process takes $\mathcal{O}(n \log n)$ time by using the merge sort algorithm. A' = sort(C) and B' = sort(D).

Reduced problem: Find the pair of elements in two arrays A' and B' such that $(a'_x - b'_x)^2 + (a'_y - b'_y)^2$ is minimized. Equivalent to minimizing $|a'_x - b'_x|$, $a'_x \in A'$, $b'_x \in B'$

For i, j = 0 start comparing x component of elements of A' and B', as stated above and store the min distance and corresponding elements in $min_distance$ and a, b, while iterating. If A'[i] < B'[j] then i + +, else j + +. The loop will terminate in $T = \mathcal{O}(2n) = \mathcal{O}(n)$ as in each iteration i or j increases and we visit each element in A' and B' only once.

Let R' be the inverse rotation. Return R'a, R'b.

Problem 8

Description:

Preprocessing:

Sort the array A into A' in $T = \mathcal{O}(n \log n)$. Traverse along the array and count the different number of each element. Store these unique values and counts in two arrays key and count. This will take T = O(n) time as the A' is sorted, and space S = O(n). The array key will already be sorted because A' was sorted.

For problem 1. we calculate the cumulative sum *cumsum* of *counts*.

For problem 2. calculate the range maxima (like range minima) data structure of the array count as we discussed in class and store it as table. Instead of storing the maximum values in this table explicitly, store index i corresponding to element in counts (can be retrieved in $T = \mathcal{O}(1)$ Therefore overall preprocessing time complexity is $T = \mathcal{O}(n \log n)$ and space complexity is $S = \mathcal{O}(n + n \log n) = \mathcal{O}(n \log n)$

Example of Keys and counts:

A	1.5	1.5	-2	0	2	0	0	3.2	0	3	2.4	-1	1	1	1.7	1.5	1.2	-3	-2.1	-5
A'	-5.0	-3.0	-2.1	-2.0	-1.0	0.0	0.0	0.0	0.0	1.0	1.0	1.2	1.5	1.5	1.5	1.7	2.0	2.4	3.0	3.2

keys	-5.0	-3.0	-2.1	-2.0	-1.0	0.0	1.0	1.2	1.5	1.7	2.0	2.4	3.0	3.2
counts	1	1	1	1	1	4	2	1	3	1	1	1	1	1

Processing:

- (1) Given [a,b] We can find a',b' in keys using binary search in time $T = \mathcal{O}(logn)$, (a',b') are indexes of the elements in the keys). We then evaluate the total number of elements as cumsum[b'] cumsum[a'-1]. Handle for edge case if a' = 0 return cumsum[b']
- (2) Given [a, b] we can find a', b' in keys by using binary search in time $T = \mathcal{O}(logn)$ (a', b') are indexes of the elements in the keys). We then use the range maxima approach to get the max_index in between a' and b'. Return $keys[max_index]$

Algorithm:

```
function preprocess(A, n){
           A' <- sort(A)
                                                                   \mbox{\ensuremath{\mbox{\sc T}}} = O(\mbox{\sc nlogn})
2
           current_key = A'[0]
4
5
           keys <- array same size as A
6
           count <- array same size a A initialized as zeros.
           k < - 0:
                                                                   \\for counting keys
8
           for (int i = 0; i <n; i++){
                                                                   \gray T = 0
      (n)
               current_key;
               if (current_key!=keys[k]){
                   k++:
                   keys[k] = current_key;
13
14
               count[k]++;
16
           }
           keys <- keys, but now trimmed to size k;
                                                                   \Trim kevs T = O(n)
17
           counts <- counts, but now trimmed to size k;</pre>
                                                                   \Trim counts T = O(n)
19
           cumsum <- array of size(K) initialized to zero.</pre>
20
           cumsum[0] <- counts[0];</pre>
21
           for (int i = 1; i < k; i++){
                                                                   \\Generate cumulative sum of
22
      counts
               cumsum[k] = cumsum[k-1] + counts[0];
                                                                   \ T = O(n)
23
24
           }
25
           table <- generate range maxima table as in class, but with index of keys instead of
26
      value. \T = O(nlogn), S = O(nlogn)
27
           return keys, counts, cumsum, table;
28
29
30
       function total_elements(a, b, keys, cumsum) {
31
           a' <- find index a' in keys such that keys[i] >= a for i >= a'; \ Sinary search
32
           b' <- find index b' in keys such that keys[i] <= b for i <= b'; \ Sinary search
33
           if (a' == 0)
34
35
               return cumsum[b'];
           else
36
```

```
return cumsum[b'] - cumsum[a'-1]

function max_occour(keys, counts, keys', counts', table){
    a' <- find index a' in keys such that keys[i] >= a for i >= a'; \\Binary search
    b' <- find index b' in keys such that keys[i] <= b for i <= b'; \\Binary search

range_max_idx <- fetch range maximum from table using table and counts. \\T = O(1)

return keys[range_max_idx];
}</pre>
```

Listing 3: Problem 8