CS515 - Algorithms & Data Structures Practice Assignment 2

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A fixed point of an array A[1..n] is an index i such that A[i] = i. Given a sorted array of distinct integers A[1..n] as input, give a divide-and-conquer algorithm to determine if A has a fixed point that runs in time O(logn).

Description Let FP(i, j) be the function that checks if there exists a fixed point in A[i, j]. Imagine A[k] as a discrete function of index k with $i \geq k \geq j$, then the fixed point of A is the intersection between A[k] and the identity function I[k]. Because A is a distinct increasing array of integer, the rate of change of A at each index k must be greater than or equal to I[k], which is equal 1 for all k. Therefore, A[k] can only intersect I[k] if $A[i] \leq i$ and $A[j] \geq j$.

Recurrence Let $m = \frac{i+j}{2}$, the recursive formulation can be described as follows

$$FP(i,j) = \begin{cases} False & i > j \\ True & A[m] = m \\ FP(i,m) & A[m] > m & otherwise \\ FP(m+1,j) & A[m] < m \end{cases}$$

Pseudocode

```
Algorithm 1 FP(i, j)
```

```
if i > j then
    return False
end if
m \leftarrow (i+j)/2
if A[m] = m then
    return True
else if A[m] > m then
    return FP(i,m)
else
    return FP(m+1,j)
end if
```

Proof of Correctness

Base Case: first, if A only has one element, if there exists a fixed point, then it has to be that element. Second, the left index i greater than the right index j indicates an empty array, which implies that no fixed point exists.

Inductive Hypothesis: FP(i, m) and FP(m + 1, j) correctly determine if [i, m] and A[m + 1, j] contain a fixed point, respectively.

Inductive Step: It is trivial that if at least one of FP(i, m) and FP(m+1, j) is true, then FP(i, j) is true because they use the same indices and values of A.

Runing Time Analysis

On each recursive call, the algorithm splits the problem into two roughly equal halves and only solves one of them. It takes constant time to check if the midpoint is the fixed point. Therefore, the total running time of this algorithm is $T(n) = T(\frac{n}{2}) + O(1) = O(\log n)$.

For a sequence of n numbers $a_1, ..., a_n$, a significant inversion is a pair (a_i, a_j) such that i < j and $a_i > 2a_j$. Assuming each of the numbers a_i is distinct, give an O(nlogn) time algorithm to count the number of significant inversions in a sequence. (Hint: modify merge sort.)

Description Let SI(A) denotes the function that counts the number of significant inversions in A, and let BS(i,j,t) denote the function that can find the minimum number that is greater than or equal to target t. Assume that A is sorted, for each a_i , we can use BS to search for $2a_i + 1$, and then count the numbers of the following numbers in A. The final result is the sum of all counts.

Recurrence

Let $m = \frac{i+j}{2}$. The recurrence can be described as follows

$$BS(i,j,t) = \begin{cases} i+1 & i>j \\ \begin{cases} i & A[i] \geq t \\ i+1 & otherwise \end{cases} & i=j \\ \\ \begin{cases} m & A[m]=t \\ \begin{cases} min(BS(i,m-1,t),m) & t < A[m] \\ BS(m+1,j,t) & t > A[m] \end{cases} & otherwise \end{cases}$$

Pseudocode

$\overline{\textbf{Algorithm 2}} \ \overline{\textbf{S}I(A[1,n])}$

```
SortA
count \leftarrow 0
for \ i : [1, n-1] \ do
k = BS(i+1, n, 2A[i]+1)
if \ k \neq -1 \ then
count \leftarrow count + max(0, n-k+1)
end \ if
end \ for
return \ count
```

Proof of Correctness

First, we assume that BS(i, j, t) correctly determines the minimum number that is greater than or equal target t and we prove that SI(A[1, n]) can count the number of significant inversions. For each a_i , BS is use to find the minimum number that is greater or equal to $2a_i + 1$ in subarray A[i + 1, n]. Because A is sorted, we do not need to search in the subarray before a_i . If BS cannot find the minimum number that is greater or equal to $2a_i + 1$ and return -1, then there are no significant inversions that contain a_i . In case that BS can find such a number a_k , then all of the numbers following and including a_k in A can make a significant inversion with a_i . Applying this for every number in A and summing up the counts will produce the final result to the problem.

Algorithm 3 BS(i, j, t)

```
if i>j then return -1
else if i=j then
if A[i]>t then return i
elsereturn i+1
end if
else
m\leftarrow \frac{i+j}{2}
if t=A[m] then
return m
else if t< A[m] then
return BS(i,m-1,t)
else
return BS(m+1,j,t)
end if
```

Second, we prove that BS(i, j, t) correctly determines the minimum number that is greater than or equal target t in a sorted array A. Similar to binary search, the algorithm also searchs for target t in A[i, j]. If it can find t, then t is also the number of interest. It it cannot find t, then it returns the minimum number that is greater than t.

Runing Time Analysis

We can use merge sort or quick sort to sort A in O(nlogn). BS takes O(logn) to search. Because it is called O(n) times, the entire algorithm will take O(nlogn).

You are given two sorted arrays of size m and n. Give an O(log m + log n) time algorithm for computing the k-th smallest element in the union of the two arrays.

Description

Let us call the two sorted arrays A and B with i and j as their indices, respectively. We can apply binary search on both arrays to find two points $1 \le i \le m$ and $1 \le j \le n$ such that i + j = k. Then the k-th smallest element in the union of the two arrays will be $max(A_i, B_j)$.

To begin with, i and j are the midpoints of the subarrays of interest of A and B. The algorithm starts off considering the entire two arrays and assigns their midpoints to i and j. Given i and j, we know that $G = max(A_i, B_j)$ is the $(i + j)^{th}$ element of the union because G is larger than all elements on the left of A_i and those of B_j . When i + j > k, we eliminate the right half of the subarray that contains $max(A_i, B_j)$. On the other hand, if i + j < k, we eliminate the left half of the subarray that contains $S = min(A_i, B_j)$ by half.

Pseudocode

```
Algorithm 4 FK(A[1:m], B[1:n]), k
```

```
al, ar, bl, br \leftarrow 1, m, 1, n
while al < ar \lor bl < br do
    i \leftarrow (al + ar)/2
    j \leftarrow (bl + br)/2
    if i + j == k then return max(A[i], B[j])
    else if i + j < k then
        L \leftarrow min(A[i], B[j])
        if L == A[i] then
             al \leftarrow i
        else
             bl \leftarrow j
        end if
    else
        G \leftarrow max(A[i], B[j])
        if G == A[i] then
             ar \leftarrow i
        else
             br \leftarrow j
        end if
    end if
```

Proof of Correctness

end while

To prove that this algorithm, we will prove that it always correctly eliminates numbers that are not the target. In the first case when i + j < k, assume that $A_i < B_j$, then the union $U = A[1:i] \cup B[1:j] = A[1:i] \cup B[1:i]$

..., $A_i, ..., B_j$ Therefore, $A[1:i-1] < A_i < target$ can be safely removed from consideration. In the second case when i+j>k, assume that $A_i < B_j$, then the union $U=A[1:i] \cup B[1:j] = ..., A_i, ..., B_j$ Therefore, $B[j+1:m] > B_j > target$ can be safely removed from consideration. The algorithm will terminate because it eliminates half of one of the subarrays at each iteration.

Runing Time Analysis Because the algorithm eliminates half of one of the two arrays at each iteration, it takes O(logm + logn) to eliminate all elements from both arrays.

You are given an $n \times n$ matrix A[1..n, 1..n] where all elements are distinct. We say that an element A[x] is a *local minimum* if it is less than its (at most) four neighbors, i.e. its up, down, left and right neighbors. Give an O(n) time algorithm to find a local minimum of A.

Description The idea is to narrow down the position of the local minimum by elimination. First, We divide the matrix into 4 parts by a cross A[1:n,n/2] and A[n/2,1:n]. Then we search for the smallest element s in the cross. If s is the center of the cross then it is also the local minimum. Otherwise, if both of its neighbors are greater, then s is also a local minimum. Other than those cases, we find the quadrant that contains s's smallest neighbor, and recursively apply the algorithm on it until a local minimum is found.

Pseudocode

```
Algorithm 5 LM(A[1:n,1:n]), k
```

```
xmin \leftarrow 1
xmax \leftarrow n
ymin \leftarrow 1
ymax \leftarrow n
while xmin < xmax \lor ymin < ymax do
   xmid \leftarrow (xmax - xmin)/2
   ymid \leftarrow (ymax - ymin)/2
   minCell \leftarrow MAXINT
   minRow \leftarrow xmid
   minCol \leftarrow ymid
   for A[i,j] \in A[m,1:n] \cup A[1:n,m] do
       if a < minCell then
           minCell \leftarrow a
           minRow \leftarrow i
           minCol \leftarrow i
       end if
   end for
   if minRow == xmid \land minCol == ymid then return A[minRow, minCol]
   end if
   minNeighbor \leftarrow MinNeighbor(minRow, minCol)
   Update xmin,xmax,ymin,ymax to those of the quadrant that contains minNeighbor
end while
```

Proof of Correctness

The quadrant that contains the minimum neighbor must also contain a local minimum because if we follow a downward path from s to its minimum neighbor and so on, we eventually reach a cell c where we cannot move to another cell because c < s, which is smaller than all cells on the cross. In other words, the cross is the boundary of the quadrant. If we recursively drawing crosses to divide the matrix, eventually the local minimum will lie on the cross.

Runing Time Analysis This algorithm takes $2n + 2\frac{n}{2} + 2\frac{n}{4} + ... = O(n)$ (geometric series) time to find the minimum number in the crosses. It takes constant time to check cases and find the smallest neighbor of s. Totally, the algorithm takes O(n).