Theory Assignment-1: ADA Winter-2024

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1 Assumptions

- 1. k ranges from 1 to 3n.
- 2. Array indices are from 0 to n-1.
- 3. Passing in arrays or subarrays of an array into a function is a constant time operation, because in practice we can implement this by using start and end pointers instead. Arrays are passed directly in the pseudocode to avoid overly messy code.
- 4. For an array A, |A| is the size of array A. Similar to the previous point, accessing the size of an array is also a constant-time operation.

2 Preprocessing

None required.

3 Algorithm Description

We start with the arrays A, B and C as the main problem.

The algorithm is recursive. Consider the current subarrays to be S_0 , S_1 and S_3 , and the current element to be found as the k^{th} smallest.

Define $mid(S_i) = \lfloor \frac{n_i}{2} \rfloor$ where n_i is the size of the array S_i , where i = 0, 1, 2.

We have two cases, $\sum_{i=0}^{2} mid(S_i) < k$ and $\sum_{i=0}^{2} mid(S_i) \ge k$.

1. Let's consider the first case, $\sum_{i=0}^{2} mid(S_i) < k$.

Without loss of generality, let $S_0[mid(S_0)]$ be the **minimum** of all $S_i[mid(S_i)]$ for i = 0, 1, 2.

Then, all elements of $S_0[0, \dots, mid(S_0)]$ can be written as the j^{th} smallest in $S_0 \cup S_1 \cup S_2$ where j < k (refer to Proof of Correctness).

So, our required element may not be in $S_0[0, \dots, mid(S_0)]$ because they can be at most the $(k-1)^{\text{th}}$ smallest, but it is definitely elsewhere in the remaining subarray and 2 arrays.

So, we recursively find the $(k-mid(S_0))^{\text{th}}$ smallest element in $S_0[mid(S_0)+1,\cdots,n_0-1]\cup S_1\cup S_2$.

2. Now for the latter case, $\sum_{i=0}^{2} mid(S_i) \geq k$.

Without loss of generality, let $S_0[mid(S_0)]$ be the **maximum** of all $S_i[mid(S_i)]$ for i = 0, 1, 2.

Then, all elements of $S_0[mid(S_0)+1,\dots,n_0-1]$ can be written as the j^{th} smallest in $S_0 \cup S_1 \cup S_2$ where j > k (refer to Proof of Correctness).

So, our required element may not be in $S_0[mid(S_0)+1,\dots,n_0-1]$ because they can be at least the $(k+1)^{\text{th}}$ smallest, but it is definitely elsewhere in the remaining subarray and 2 arrays.

So, we recursively find the k^{th} smallest element in $S_0[0, \dots, mid(S_0)] \cup S_1 \cup S_2$.

The base case for the recursion is simple. If the arrays S_0 , S_1 and S_2 all have a single element, then the elements smaller than them will be k-1 in total, so we return the minimum of these 3 elements.

Furthermore, for completeness, if at any point of the algorithm an array cannot be divided further, we only recurse further on arrays which can.

4 Recurrence Relation

Let $p = n_1 n_2 n_3$ be the current product of the sizes of A, B and C; n_1 , n_2 and n_3 respectively. Since the size of one of the arrays in the next subproblem will be approximately halved, size of the next subproblem $p' = \frac{p}{2}$. The only serial (non-recursive) operation is comparing two elements in $A \cup B \cup C$, which can be done in O(1) as defined, so the serial time taken is less than equal to c_0 for some positive real constant c_0 .

Therefore, we have:

$$T(p) = T(p') + c_0$$
$$T(p) = T(\frac{p}{2}) + c_0$$

5 Complexity Analysis

The above recurrence relation can be expressed as

$$T(p) = aT(\frac{p}{b}) + k_0 p^c$$

where a = 1, b = 2, c = 0.

By Master's theorem, since $0 = \log_2 1$ and thus $c = \log_b a$,

$$T(p) = O(\log p)$$

Since $p = n_1 n_2 n_3$, $\log p = \log n_1 + \log n_2 + \log n_3 = 3 \log n$.

Therefore, the overall problem with 3 arrays of size n can be solved in $O(3 \log n) = O(\log n)$ time.

6 Pseudocode

Algorithm 1 KthSmallest

```
1: function KTHSMALLEST(A, B, C, n, k)
       return KthIndex(A, B, C, k-1)
 3: end function
 4:
 5: function KthIndex(A, B, C, k)
 6:
       if A is empty and B is empty then return C[k]
 7:
 8:
 9:
10:
       if A is empty and C is empty then return B[k]
11:
12:
       if C is empty and B is empty then return A[k]
13:
14:
15:
16:
       for I \in \{A, B, C\} do
           if I is empty then
17:
              m_I \leftarrow \infty
18:
              M_I \leftarrow -\infty
19:
           else
20:
               m_I \leftarrow I[Mid(I)]
21:
               M_I \leftarrow I[Mid(I)]
22:
           end if
23:
       end for
24:
25:
       M \leftarrow \max(M_A, M_B, M_C)
26:
       m \leftarrow \min(m_A, m_B, m_C)
27:
28:
       n_{tot} \leftarrow Mid(A) + Mid(B) + Mid(C)
29:
30:
       if n_{tot} < k then
31:
32:
           if m_A = m then
33:
               return KthIndex(A[Mid(A)+1, \cdots, |A|-1], B, C, k - Mid(A) - 1)
34:
           else if m_B = m then
35:
               return KthIndex(A, B[Mid(B)+1, \cdots, |B|-1], C, k - Mid(B) - 1)
36:
           else
37:
               return KthIndex(A, B, C[Mid(C)+1, \cdots, |C|-1], k - Mid(C) - 1)
38:
           end if
39:
40:
       else
41:
42:
           if M_A = M then
43:
               return KthIndex(A[0, \cdots, Mid(A)], B, C, k)
44:
           else if M_B = M then
45:
               return KthIndex(A, B[0, \dots, Mid(B)], C, k)
46:
47:
           else
               return KthIndex(A, B, C[0, \dots, Mid(C)], k)
48:
           end if
49:
50:
51:
       end if
52:
53: end function
```

The Mid() function is defined as below:

```
1: function MID(Arr)
2: s \leftarrow |Arr|
3: return \lfloor \frac{s}{2} \rfloor
4: end function
```

7 Proof of Correctness

The basic explanation and proof of how the algorithm works is provided in the algorithm description section itself.

Here are some of the more minute details:

- For the proof of j < k, consider the opposite. If $j \ge k$, the middle elements are at least the $(k+1)^{\text{th}}$ smallest, and therefore our assumption is false.
- For the proof of j > k in case 2, consider the opposite. If $j \le k$, the middle elements are at most the k^{th} smallest, and therefore our assumption is false.