

ASSIGNMENT – 2

Introduction to Algorithms (CS 430)

1. To determine the worst-case time complexity of an alternative version of SELECT algorithm where the elements are grouped in groups of size $2m + 1$ (with $m \geq 2$ being an integer constant);

Analysis:

1. Grouping and finding medians:
 - Divide n elements into $\frac{n}{2m+1}$ groups of $2m + 1$ elements each, with at most one group containing less than $2m + 1$ elements.
 - We will sort each group and will find median of each group. *Time complexity: $O(n)$*
2. Using recursive function for medians:
 - Let M be the array of medians of all these groups. The size of M is $\frac{n}{2m+1}$.
 - Recursively find the median of M . Time complexity: $T\left(\frac{n}{2m+1}\right)$

Recurrence Relation:

- The recurrence relation for time complexity $T(n)$ is: $T(n) \leq T\left(\frac{n}{2m+1}\right) + T\left(\frac{2m * n}{2m+1}\right) + O(n)$

Solving the Recurrence

- $T(n) \leq T\left(\frac{n}{2m+1}\right) + T\left(\frac{2m * n}{2m+1}\right) + O(n)$
- Where; $a = 2$, $b = 2m+1$, $f(n) = O(n)$
- $c = \log_{(2m+1)} 2$
- Since, $f(n) = O(n)$ and $c = \log_{(2m+1)} 2$ //the work done at each level of recursion tree is $O(n)$, height of tree is $O(\log n)$

Conclusion:

The worst-case time complexity of SELECT algorithm, grouping elements in groups of size $2m+1$ where $m \geq 2$ is: $T(n) = O(n)$.

2. $O(n \lg k)$ -time algorithm to merge k sorted arrays into one sorted array, where n is the total number of elements in all the input arrays.

MERGE K SORTED ARRAYS(A_1, A_2, \dots, A_k)

1. Initialize a min-heap H with the first elements of each arrays A_1, A_2, \dots, A_k
2. Initialize an empty output array B
3. while H is not empty
4. $x = \text{HEAP-EXTRACT-MIN}(H)$
5. Append x to the end of B
6. $i = \text{index of the array that } x \text{ was extracted from}$
7. if there are more elements in array A_i
8. Insert the next element from A_i into H
9. Return B

The key steps are:

- Initializing a min-heap H with the first elements of each of the k sorted input arrays.
Time taken: $O(k)$
 - Extracting the minimum elements repeatedly from the min-heap, appending it to the output array B , and inserting the next element from that array into min-heap.
Time taken for; Extracting minimum: $O(\log k)$, Inserting an element: $O(\log k)$.
 - This is done n time, once for each element in the input arrays.
- ❖ ***The total time complexity is $O(k + n \log k) = O(n \log k)$.***

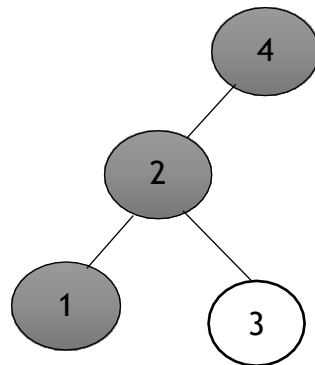
3. Pseudo-code for MAX-HEAP-DELETE (A, i) that deletes the element in $A[i]$ from a binary max-heap A .

1. if $i > \text{heap_size}(A)$ or $i < 1$
2. error “index out of bounds”
3. $A[i] = A[\text{heap_size}(A)]$
4. Decrease $\text{heap_size}(A)$ by 1
5. if $i > 1$ and $A[\text{PARENT}(i)] < A[i]$
6. $\text{HEAP-INCREASE-KEY}(A, i, A[i])$
7. else
8. $\text{MAX-HEAPIFY}(A, i)$

Time Complexity:

- HEAP-INCREASE-KEY: $O(\log n)$
- MAX-HEAPIFY: $O(\log n)$
- ❖ *Time complexity of MAX-HEAP-DELETE is $O(\log n)$.*

4. Professor Bunyan thinks he has discovered a remarkable property of binary search trees. Suppose that the search for key k in a binary search tree ends up in a leaf. Consider three sets: A , the keys to the left of the search path; B , the keys on the search path; and C , the keys to the right of the search path. Professor Bunyan claims that any three keys $a \in A$, $b \in B$, and $c \in C$ must satisfy $a \leq b \leq c$. Give the smallest possible (with fewest nodes) counterexample to the professor's conjecture.



- We are searching for key $k = 1$
- The path traversed from root to leaf is the shaded nodes.
- 'A' indicates the keys on the left of the search path. Set $A = \{ \}$; is null set/empty set.
- 'B' indicates the keys on the search path. 'B' = $\{4, 2, 1\}$
- 'C' indicates the keys on the right of the search path. 'C' = $\{3\}$
- **Since, $3 < 4$, Prof. Bunyan's claim is wrong. As, it does not satisfy the given condition.**