

Analysis of Algorithms - Midterm (Solutions)

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1. Consider the recurrence relation (6 points):

$$\begin{aligned}T(1) &= 1 \\T(n) &= 2 \cdot T(n-1) + 1, \quad n > 1\end{aligned}$$

Show that $T(n) = 2^n - 1$

Proof: Using induction:

Base case $T(1)$:

$$\begin{aligned}T(1) &= 1 \\T(1) &= 2^1 - 1 \\&= 2 - 1 \\&= 1\end{aligned}$$

Thus, the base case is true.

Let us assume that $T(k)$ is true, i.e.,

$$T(k) = 2^k - 1$$

We need to show that $T(k+1)$ is true.

$$\begin{aligned}T(k+1) &= 2 \cdot T(k+1-1) + 1 \\&= 2 \cdot T(k) + 1 \\&= 2 \cdot (2^k - 1) + 1 \text{ (using the inductive hypothesis)} \\&= 2^{k+1} - 2 + 1 \\&= 2^{k+1} - 1 \\T(k+1) &= 2^{k+1} - 1\end{aligned}$$

Thus, $P(k+1)$ is true and we have shown that $P(k) \rightarrow P(k+1)$; applying the principle of mathematical induction, we conclude that the conjecture is true. \square

2. Show that if $f(n) = O(g(n))$ and $e(n) = O(h(n))$, then $f(n) \cdot e(n) = O(g(n) \cdot h(n))$. (4 points)

Proof: By definition of 'O', $f(n) = O(g(n))$ implies that:

$$f(n) \leq c \cdot g(n)$$

Also, by definition of 'O', $e(n) = O(h(n))$ implies that:

$$e(n) \leq c' \cdot h(n)$$

Observe that:

$$\begin{aligned} f(n) \cdot e(n) &\leq c \cdot g(n) \cdot c' \cdot h(n) \\ &\leq c'' \cdot g(n) \cdot h(n) \end{aligned}$$

Then, by definition of 'O', $f(n) \cdot e(n) = O(g(n) \cdot h(n))$. \square

3. Let \mathbf{T} be a proper binary tree of height h , having n nodes. Show that $h \geq \log_2(n + 1) - 1$. (6 points)

Proof: Note that we want to find a lower bound on the height h of a proper binary tree containing n nodes. The height will be minimized when all n nodes are packed as tightly as possible, i.e. when the proper binary tree is also a full binary tree. In a full binary tree, of height h , the total number of nodes is: $2^0 + 2^1 + 2^2 + \dots + 2^h = 2^{h+1} - 1$, i.e., $h = \log_2(n + 1) - 1$. If the tree \mathbf{T} is not full, the height h will only increase. We can thus conclude that $h \geq \log_2(n + 1) - 1$, for any proper binary tree \mathbf{T} having n nodes. \square

4. Consider the binary tree \mathbf{T} in Figure (1). Write down the order of the nodes, when you traverse the tree in inorder, preorder and postorder. (6 points)

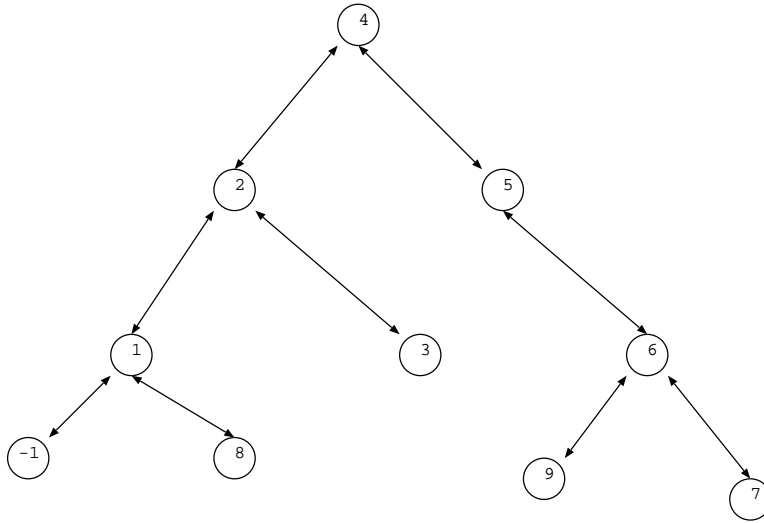


Figure 1: Binary Tree \mathbf{T}

Observe that in an inorder traversal, the left children of a node are visited before it is visited and the right children of a node are visited after it is visited. Applying this recursively, we conclude that the nodes in \mathbf{T} would be visited in the following order: $-1, 1, 8, 2, 3, 4, 5, 9, 6, 7$.

Observe that in an preorder traversal, a node is visited before its children are visited and the left children of a node are visited before the right children are visited. Applying this recursively, we conclude that the nodes in **T** would be visited in the following order: 4, 2, 1, -1, 8, 3, 5, 6, 9, 7.

Observe that in an postorder traversal, a node is visited after its children are visited and the left children of a node are visited before its right children are visited. Applying this recursively, we conclude that the nodes in **T** would be visited in the following order: -1, 8, 1, 3, 2, 9, 7, 6, 5, 4.

5. Prove that Algorithm (0.1) correctly sorts an n -input sequence S provided as an n -element array **A** (in increasing order). You may assume that the n elements of the array are stored in the locations $A[1], A[2], \dots, A[n]$. What is the worst-case running time of the algorithm? (8 points)

Hint: You may either use the Loop Invariant Technique or induction (second principle!) on the number of elements in the array!

Function BUBBLE-SORT(**A**, n)

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1: for ( $i = 1$  to  $n - 1$ ) do
2:   for ( $j = i + 1$  to  $n$ ) do
3:     if ( $A[i] > A[j]$ ) then
4:        $temp = A[i]$ 
5:        $A[i] = A[j]$ 
6:        $A[j] = temp$ 
7:     end if
8:   end for
9: end for
```

Algorithm 0.1: Bubble Sort Algorithm

Proof: We shall discuss correctness of the BUBBLE-SORT() Algorithm using the Loop invariant technique (Please see Pg. 27 of [GT02]).

We use the following loop invariant:

$S(i)$: The first $i - 1$ elements are in their correct positions in **A**.

The key difference between our approach and the approach in [GT02], is that we start from $S(1)$ since our elements are stored in $A[1], A[2], \dots, A[n]$ as opposed to $A[0], A[1], \dots, A[n - 1]$.

$S(1)$ is trivially true, since $A[0]$ does not exist. Consider the working of the outer loop in iteration $i = k$. Prior to the start of this iteration, we have $A[1] \leq A[2] \leq \dots \leq A[k - 1]$, with $A[k - 1]$ being the $(k - 1)^{th}$ smallest element in **A**. As iteration $i = k$ proceeds, we scan through the array to determine the smallest element in $A[k]$ through $A[n]$ and put it in $A[k]$. Hence, if $S(1), \dots, S(k)$ are true, then $S(k + 1)$ is true, i.e. after the $i = k$ iteration (and before the $i = k + 1$ iteration), we have $A[1] \leq A[2] \leq \dots \leq A[k - 1] \leq A[k]$ and $A[k]$ is the k^{th} smallest element in **A**. It follows that $S(n)$ is true, i.e. at the end of the iteration $i = n - 1$, the first $n - 1$ elements are in their correct positions in **A**. This forces $A[n]$ to be in its correct place!

Thus, we have shown that the algorithm is correct by applying the principle of loop invariants.

A rough approximation to the running time can be obtained by observing that the i loop runs at most n times and so does the j loop. Further, within the nested **for** loops, at most 4 statements are executed. So the total running time cannot exceed $4 \cdot n^2$, i.e., $O(n^2)$. We give a more formal analysis below. Let $T(n)$ denote the worst-case running time of Algorithm (0.1). We then have

$$\begin{aligned}
 T(n) &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n 4 \\
 &= 4 \cdot \sum_{i=1}^{n-1} (n - i)
 \end{aligned}$$

$$\begin{aligned}
&= 4 \cdot \left(\sum_{i=1}^{n-1} n - \sum_{i=1}^{n-1} i \right) \\
&= 4 \cdot \left(n \cdot \sum_{i=1}^{n-1} 1 - \frac{n \cdot (n-1)}{2} \right) \\
&= 4 \cdot \left(n \cdot (n-1) - \frac{n \cdot (n-1)}{2} \right) \\
&= 4 \cdot \frac{n \cdot (n-1)}{2} \\
&= O(n^2)
\end{aligned}$$

*In passing, we note that there is no good input for this algorithm. The **if** statement within the double **for** loop is executed $\Omega(n^2)$ times. \square*

References

- [GT02] Michael T. Goodrich and Roberto Tamassia. *Algorithm Design: Foundations, Analysis and Internet Examples*. John Wiley & Sons, 2002.