

Analysis and Design of Algorithms

Homework 1

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1. Bounding Summations

Give tight bounds on the following summations. Assume $r \geq 0$ and $s \geq 0$ are constants.

(a) $\sum_{k=1}^n k^r$

We can use integrals to find an upper and lower bound to the summation:

$$\begin{aligned}\int_0^n k^r dk &\leq \sum_{k=1}^n k^r \leq \int_1^{n+1} k^r dk \\ \frac{k^{r+1}}{r+1} \Big|_0^n &\leq \sum_{k=1}^n k^r \leq \frac{k^{r+1}}{r+1} \Big|_1^{n+1} \\ \frac{n^{r+1}}{r+1} &\leq \sum_{k=1}^n k^r \leq \frac{(n+1)^{r+1}}{r+1} - \frac{1^{r+1}}{r+1}\end{aligned}$$

We take away the constants to get something like this:

$$\begin{aligned}n^{r+1} &\leq \sum_{k=1}^n k^r \leq (n+1)^{r+1} \\ \Omega(n^{r+1}) &\leq \sum_{k=1}^n k^r \leq O(n^{r+1})\end{aligned}$$

Because Ω is equal to O , we can then say that the tight bound of the summation is: $\Theta(n^{r+1})$

(b) $\sum_{k=1}^n l g^s k$

Expanding the summation we get the following terms:

$$lg^s 1 + lg^s 2 + lg^s 3 + \dots + lg^s n$$

As it is a sum of logarithms it can be put like this:

$$lg^s(1 * 2 * 3 * \dots * n) = lg^s n!$$

Let's prove this by mathematical induction:

$$\begin{aligned} \sum_{k=1}^{n+1} lg^s k &= \sum_{k=1}^n lg^s k + lg^s(n+1) \\ &= lg^s n! + lg^s(n+1) \\ &= lg^s(1 * 2 * \dots * n * (n+1)) \\ &= lg^s(n+1)! \end{aligned}$$

So we can say that the complexity of this summation is: $\Theta(lg^s(n!))$

(c) $\sum_{k=0}^{lgn} \lceil \frac{n}{2^k} \rceil$

This summation can be bounded using a geometric series, because the following is true:

$$\begin{aligned} \frac{a_{k+1}}{a_k} &\leq r, \quad 0 < r < 1 \text{ is constant} \\ \frac{\frac{n}{2^{k+1}}}{\frac{n}{2^k}} &= \frac{n * 2^k}{n * 2^{k+1}} = \frac{2^k}{2^{k+1}} = 2^{k-(k+1)} = 2^{-1} = \frac{1}{2} \end{aligned}$$

Then the following applies:

$$\sum_{k=0}^{lgn} \lceil \frac{n}{2^k} \rceil \leq \sum_{k=0}^{\infty} a_0 r^k, \text{ where } a_0 \text{ is the first term of the series}$$

$$a_0 = \frac{n}{2^0} = n$$

$$\begin{aligned} \sum_{k=0}^{lgn} \lceil \frac{n}{2^k} \rceil &\leq \sum_{k=0}^{\infty} n * r^k \\ &\leq n * \frac{1}{1 - \frac{1}{2}} = 2n \end{aligned}$$

Havin $2n$ as a bounding expression, we can define constants c_0 and c_1 such that the conditions for Θ are met:

$$c_0 n \leq \sum_{k=0}^{\lg n} \lceil \frac{n}{2^k} \rceil \leq c_1 n$$

Because each term of the series is the previous one divided over two, the series won't ever result in $2n$. So having $c_1 = 2$ would do, and $c_0 \leq 1$ would do, such that:

$$\sum_{k=0}^{\lg n} \lceil \frac{n}{2^k} \rceil = \Theta(n)$$

2. Given the Tree Sort Algorithm:

(a) Give the step count

TreeSort(A):

1. <i>Tree</i> = <i>BinaryTree</i>	1
2. for $i = 0$ to <i>A.length</i> :	n
3. <i>Tree.insert</i> (<i>A</i> [i])	$(n - 1)(\log(n))$ or $(n - 1)(n)$
4. <i>A</i> = <i>Tree.inOrder</i> ()	n

(b) Complexities

The complexities of Tree Sort depend on the form of its input, a randomized array will take less time than an already sorted array.

Best Case Analysis:

In the best case of Tree Sort the input is an array such that the resulting tree of lines 2 - 3 is a balanced tree, in which each insertion takes $O(\log(n))$. Such that we have the following summation:

$$1 + n + \sum_{k=0}^{n-1} \log(n) + n = 1 + n + (n - 1)\log(n) + n$$

This case takes $O(n\log(n))$.

Worst Case Analysis:

The worst case of the Tree Sort is when the input is an array already sorted or almost sorted, this results in a degenerate tree, where child

nodes hang either only on the left or right side of the Tree, an insertion in this Tree takes $O(n)$. Such that we have the following summation:

$$1 + n + \sum_{k=0}^{n-1} n + n = 1 + n + (n-1)n + n$$

This case takes $O(n^2)$.

Average Case:

The average case is when we have a randomized array, per se. For this case we can assume that we have a uniform distribution, meaning every element of the array has the same probability of being in the correct position to result in a balanced tree:

For every element i in the original array, we have i possible spaces in the tree for i to be placed in. We define an indicator function $I(A) \begin{cases} P(A) & P(\tilde{A}) \end{cases}$ in which $P(A)$ would be $\frac{1}{i}$ given that we are at element i .

To get the expected value of an insertion we must take into account insertions 1 through n , we can do so by adding up the expected value at each step, where $X_i = I(A_i)$: $E[x_n] = E[x_1] + E[x_2] + \dots + E[x_n] = E[\sum_{i=1}^n x_i] = 1 + \int_2^n \frac{1}{i} di = 1 + \log n - \log 2 = \log n$.

So the average case for inserting n elements would be: $O(n \log n)$.

(c) Prove the correctness of the algorithm

To prove the correctness of Tree Sort using loop invariants we must also prove the correctness of the method *Tree.insert(item)* inside Tree Sort in order to prove the loop invariant set for lines 2 - 3.

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BinaryTree
Insert(item):
1. node = Tree.root
2. while node  $\neq$  null:
3.     if item < node:
4.         node = node.leftSubtree
5.     else:
6.         node = node.rightSubtree
7. node = item

```

We define the following invariant for the while loop of lines 2-6:

“Before iteration of the while loop of lines 2 - 6 *node.leftSubtree* contains elements less than *node* and *n.rightSubtree* contains elements greater than *node*”

Initialization: before the first iteration, *node* is the root of a Binary Tree, if no insertions have been made, the root is null and the properties of a Binary Tree are true.

Maintenance: before each iteration, *node* is placed as the root of another subtree, either right or left depending on the value of *item* and so *node* is now the root of a Binary Tree, so the invariant is validated.

Termination: at termination, *node* is null, so we have arrived at the place where *item* should be, as we ended up here through a Binary Search, *item* is now the root of a right or left subtree of it's parent node. The invariant is validated.

The loop invariant for the for loop of lines 2 - 3 is thus:

“Before each iteration of the for loop of lines 2 - 3 *Tree* is ordered in such way that each node is a root of a Binary Tree, where any node's left subtree has values smaller than the root, and the right subtree has values bigger than the root.”

Initialization: before the first iteration of the loop, *Tree* is an empty tree, no nodes and whose root is null, so it follows the properties of a Binary Tree.

Maintenance: before each iteration *i* each element of the sub-array $A[1..i - 1]$ is now a node in *Tree* through *Tree.insert(item)* such that it follows the properties of a Binary Tree. Each node is the root of a Binary Tree, where the right subtree has values bigger than the root and the left subtree has values smaller than the root.

Termination: at termination, $i = A.length$ and the elements of the array $A[1...A.length - 1]$ are nodes in *Tree* sorted in since their insertion using *Tree.insert(item)*, resulting in a Binary Tree where each node is the root of a Binary Tree, the right subtree having values greater than the root and the left subtree having values smaller than the root. Such that in line 4, *Tree.inOrder()* outputs the sorted version of array *A*.

3. Book Problems

(a) Exercise 2.3-3:

Use mathematical induction to show that when n is an exact power of 2, the solution of the recurrence $T(n)$ is $nlgn$

$$\begin{aligned} T(n) &= 2 \quad \text{if } n = 2, \\ T(n) &= 2T(\frac{n}{2}) + n \quad \text{if } n = 2^k, \text{ for } k > 1 \end{aligned}$$

$$\begin{aligned} T(n) &= \sum_{i=0}^{lg n} (\frac{n}{2})^i = \sum_{i=0}^{lg n} n = nlgn \\ T(2^k) &= \sum_{i=0}^{lg 2^k} 2^k = 2^k lg(2^k) = k2^k, \text{ where } k \text{ is } \log n \text{ when } n = 2^k \\ T(2^{k+1}) &= \sum_{i=0}^{lg 2^{k+1}} 2^{k+1} = (k+1)2^{k+1}, \text{ where } k+1 \text{ is } \log n \text{ when } n = 2^{k+1} \end{aligned}$$

Thus, $T(n)$ is exactly $nlgn$ when n is an exact power of 2.

(b) Exercise 3.1-7:

Prove that $o(g(n)) \cap \omega(g(n))$ is the empty set.

We can prove this by looking at the definition of each $o(g(n))$ and $\omega(g(n))$.

$o(g(n))$ is defined as the set $\{f(n) \text{ such that } 0 \leq f(n) < cg(n) \text{ for any } c > 0, n_0 > 0\}$

$\omega(g(n))$ is defined as the set $\{f(n) \text{ such that } 0 \leq cg(n) < f(n) \text{ for any } c > 0, n_0 > 0\}$

So we get the following equation:

$$\omega(g(n)) < f(g(n)) < o(g(n))$$

Then $\omega(g(n)) < o(g(n))$. Thus $\omega(g(n))$ is not a subset of $o(g(n))$ and $o(g(n)) \cap \omega(g(n)) = \emptyset$.

4. Give asymptotic upper and lower bounds for $T(n)$ in:

(a) $T(n) = 9T(\frac{n}{81}) + \log n$

Using the master theorem we can make this fit in one of the cases:

$$f(n) = O(n^{\log_b a - \varepsilon}), \text{ for some } \varepsilon > 0$$

$$f(n) = \log n, \quad b = 81, \quad a = 9$$

$$\log_b a = \log_{81} 9 = \frac{1}{2}$$

$$\text{So: } \log(n) = O(n^{\frac{1}{2} - \varepsilon}) \text{ for some } \varepsilon > 0$$

For cn^k to be greater than $\log(n)$ k must be greater than 0 :

$$\frac{1}{2} - \varepsilon > 0$$

$$\varepsilon < \frac{1}{2}, \text{ and } \varepsilon > 0 \text{ so:}$$

$$0 < \varepsilon < \frac{1}{2}$$

Because there is an ε that satisfies the condition for the case, then, following the master theorem $T(n) = \Theta(n^{\log_b a})$, so in this case $T(n) = \Theta(n^{\log_{81} 9}) = \Theta(n^{\frac{1}{2}}) = \Theta(\sqrt{n})$

(b) $T(n) = T(n-1) + n^c$, where $c \leq 1$ is a constant.

Expanding $T(n)$ we get the following:

$$T(n-3) + (n-2)^c + (n-1)^c + n^c$$

So we get the following summation:

$$\sum_{k=0}^{n-1} (n-k)^c$$

We get the following bounds using the minimum and maximum term of the summation:

$$a_{max} = n^c \quad a_{min} = 1^c$$

$$\sum_{k=0}^{n-1} 1^c \leq \sum_{k=0}^{n-1} (n-k)^c \leq \sum_{k=0}^{n-1} n^c$$

$$(n-1) * 1^c \leq \sum_{k=0}^{n-1} (n-k)^c \leq n^c * (n-1)$$

$$n - 1 \leq \sum_{k=0}^{n-1} (n - k)^c \leq n^c n - n^c$$

$$n - 1 \leq \sum_{k=0}^{n-1} (n - k)^c \leq n^{c+1} - n^c$$

And so we get:

$$\Omega(n) \leq \sum_{k=0}^{n-1} (n - k)^c \leq O(n^{c+1})$$

$$(c) \ T(n) = T(n^{\frac{1}{10}}) + 1$$

Using substitution we get the following:

$$m = \log_{10} n$$

$$10^m = 10^{\log_{10} n}$$

$$10^m = n^{\log_{10} 10}$$

$$10^m = n$$

$$(10^m)^{\frac{1}{10}} = 10^{\frac{m}{10}}$$

$$S(m) = T(10^m)$$

$$S(m) = S\left(\frac{m}{10}\right) + 1$$

Now we use the recursion tree structure to solve $S(m)$:

There'll be a point where a division would be equal to 1, we're going to get that value to get the last term of the series:

$$\frac{m}{10}, \frac{m}{10^2}, \frac{m}{10^3} \cdots \frac{m}{10^h}, \text{ where } h \text{ is the height of the tree.}$$

$$\frac{m}{10^h} = 1$$

$$m = 10^h$$

$$\log_{10} m = \log_{10} 10^h$$

$$\log_{10} m = h$$

So we can solve $S(m)$ as follows:

$$\sum_{k=0}^{\log_{10} m} 1 = \Theta(\log_{10} m)$$

Then we substitute to get back to terms of n :

$$\Omega(\log_{10}(\log_{10} n)) \leq \sum_{k=0}^{\log_{10} m} 1 \leq O(\log_{10}(\log_{10} n))$$

$$\sum_{k=0}^{\log_{10} m} 1 = \Theta(\log_{10}(\log_{10} n))$$