

CMPSCI 611 - Advanced Algorithms

Homework 1

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Question 1

Part A

The order of growth of functions is as follows

$$\begin{aligned}
 g_7 &= (n^{1/\log_2 n})^3 = (n^{\log_2 n / \log_n n})^3 = (2^{\log_n n / \log_n n})^3 = 2^3 = 8, \\
 g_{10} &= (\log_2 n)^2, \\
 g_1 &= 2^{\sqrt{\log_2 n}}, \\
 g_3 &= n(\log_2 n)^3, \\
 g_9 &= n^{1+1/\log_2 \log_2 n} = n \times n^{1+1/\log_2 \log_2 n}, \\
 g_4 &= n^{4/3} = n^{1+1/3} = n \times n^{1/3}, \\
 g_{11} &= (\log_2 n)^{\log_2 n}, \\
 g_5 &= n^{\log_2 n}, \\
 g_6 &= 2^n, \\
 g_{12} &= \pi^n = 3.14^n, \\
 g_8 &= n! \approx n^n, \\
 g_2 &= 2^{2^n}
 \end{aligned}$$

Explanations for the ordering

$$\begin{aligned}
 g_{10} &= O(g_1) : \\
 \lim_{n \rightarrow \infty} \frac{\log^2 n}{2^{\sqrt{\log_2 n}}} &= \lim_{n \rightarrow \infty} \frac{2 \log n}{2^{\sqrt{\log_2 n}}} = 2 \lim_{n \rightarrow \infty} \frac{\log \log n}{\log 2^{\sqrt{\log_2 n}}} = 2 \lim_{n \rightarrow \infty} \frac{\log \log n}{\sqrt{\log_2 n}} = \\
 2 \lim_{n \rightarrow \infty} \frac{\log \log \log n}{\log \log^{1/2} n} &= 2 \lim_{n \rightarrow \infty} \frac{\log \log \log n}{1/2 \log \log n} = 4 \lim_{n \rightarrow \infty} \frac{\log \log \log n}{1 \log \log n} = 0
 \end{aligned}$$

$\log \log \log n$ grows slower than $\log \log n$, hence the ratio test gives 0

$$\begin{aligned}
 g_1 &= O(g_3) : \\
 \lim_{n \rightarrow \infty} \frac{g_1}{g_3} &= \lim_{n \rightarrow \infty} \frac{2^{\sqrt{\log_2 n}}}{n \log^3 n} = \lim_{n \rightarrow \infty} \frac{\sqrt{\log_2 n}}{\log_2(n \log^3 n)} = \lim_{n \rightarrow \infty} \frac{\sqrt{\log_2 n}}{\log_2 n + \log_2 \log^3 n} = \\
 \lim_{n \rightarrow \infty} \frac{\sqrt{\log_2 n}}{\log_2 n + 3 \log_2 \log n} &= 0
 \end{aligned}$$

$\log n$ grows faster than $\sqrt{\log n}$, hence the ratio test gives 0, and $2^{\sqrt{\log n}} \in o(n \log^3 n)$

$$\begin{aligned}
g_3 &= O(g_9) : \\
n^{\frac{1}{\log_2 \log_2 n}} &\stackrel{?}{=} (\log_2 n)^3 \\
\frac{1}{\log_2 \log_2 n} &\stackrel{?}{=} \log_n (\log_2 n)^3 = \frac{\log_2 (\log_2 n)^2}{\log_2 n} = \frac{3 \log_2 (\log_2 n)}{\log_2 n} \\
\frac{1}{\log_2 \log_2 n} &\stackrel{?}{=} \frac{3 \log_2 \log_2 n}{\log_2 n} \\
\log_2 n &\stackrel{?}{=} 3 (\log_2 \log_2 n)^2 \\
\sqrt{\log_2 n} &\stackrel{?}{=} \sqrt{3} \sqrt{(\log_2 \log_2 n)^2} \\
\sqrt{\log_2 n} &\stackrel{?}{=} \sqrt{3} \log_2 \log_2 n; \{ \log_2 n = x \} \\
\lim_{x \rightarrow \infty} \frac{\sqrt{3} \log_2 x}{\sqrt{x}} &= 0
\end{aligned}$$

$$\begin{aligned}
g_9 &= O(g_4) : \\
\lim_{x \rightarrow \infty} \frac{1}{\log_2 \log_2 n} &= 0 \\
g_4 \text{ will be bigger than } g_9 \quad \forall n > n_0. & \left(\frac{1}{\log_2 \log_2 n_0} = 1/3 \right)
\end{aligned}$$

$$\begin{aligned}
g_4 &= O(g_{11}) : \\
\lim_{n \rightarrow \infty} \frac{n^{4/3}}{\log n^{\log n}} &= \lim_{n \rightarrow \infty} \frac{\log n^{4/3}}{\log \log n^{\log n}} = \lim_{n \rightarrow \infty} \frac{4/3 \log n}{\log^2 n} = \lim_{n \rightarrow \infty} \frac{4/3}{\log n} = 0
\end{aligned}$$

$$\begin{aligned}
g_{11} &= O(g_5) : \\
\log n^{\log n} &= \overbrace{\log n \times \log n \times \dots \times \log n}^{k=\log n \text{ times}} \\
n^{\log n} &= \overbrace{n \times n \times \dots \times n}^{k=\log n \text{ times}} \text{ Since } \log n \in O(n), \text{ clearly } \log n^{\log n} \in O(n^{\log n})
\end{aligned}$$

$$\begin{aligned}
g_5 &= O(g_6) : \\
\lim_{n \rightarrow \infty} \frac{n^{\log n}}{2^n} &= \lim_{n \rightarrow \infty} \frac{\log n^{\log n}}{n} = \lim_{n \rightarrow \infty} \frac{\log n \times \log n}{n} = \lim_{n \rightarrow \infty} \frac{\log^2 n}{n} = 0
\end{aligned}$$

If the top function would have been $(\sqrt{n})^2$, the ratio in the lim would have been equal to 1, but it is known that $\log n \in O(\sqrt{n})$, hence $\log n$ grows slower and the ratio is equal to 0

$$\begin{aligned}
g_{12} &= O(g_8) : \\
\pi^n &= \overbrace{\pi \times \pi \times \dots \times \pi}^{n \text{ times}} \\
n! &= 1 \times 2 \times 3 \times 4 \times \dots \times n
\end{aligned}$$

According to Stirling's approximation $n!$ can be viewed as $n! = \overbrace{n \times n \times \dots \times n}^{n \text{ times}}$. Each number larger than 4 will increase the product much more significantly than just another multiplication by π

$$\begin{aligned}
g_8 &= O(g_2) : \\
\text{According to Stirling's approximation } n! &= O(n^n). \text{ So, it is sufficient to show} \\
\text{that } n^n &\in O(2^{2^n}) \\
\lim_{n \rightarrow \infty} \frac{n^n}{2^{2^n}} &= \lim_{n \rightarrow \infty} \frac{\log n^n}{2^n} = \lim_{n \rightarrow \infty} \frac{\log_n n^n / \log_n 2}{2^n} = \lim_{n \rightarrow \infty} \frac{n}{2^n \cdot \log_n 2} = 0
\end{aligned}$$

Part B

1. $\log f(n)$ is $O(\log g(n))$ - **True**

$$O(\log(g(n))) = \log(cf(n)) = \log c + \log f(n)$$

$$\lim_{x \rightarrow \infty} \frac{\log c + \log f(n)}{\log f(n)} = \lim_{x \rightarrow \infty} \left(\frac{\log c}{\log f(n)} + \frac{\log f(n)}{\log f(n)} \right) = 0 + 1 = 1$$

2. $2^{f(n)}$ is $O(2^{g(n)})$ - **False**

Let us take $f(n) = x$ and $g(n) = 2x$ then:

$$\lim_{x \rightarrow \infty} \frac{2^x}{x} = \lim_{x \rightarrow \infty} 2^{\frac{x}{2}} = 2$$

$$\lim_{x \rightarrow \infty} \frac{2^{2x}}{2^x} = \lim_{x \rightarrow \infty} \frac{2^x * 2^x}{2^x} = \lim_{x \rightarrow \infty} 2^x = \infty$$

3. $(f(n))^2$ is $O((g(n))^2)$ - **True**

$$O((g(n))^2) = (cf(n))^2 = c^2(f(n))^2$$

$$\lim_{x \rightarrow \infty} \frac{c^2(f(n))^2}{(f(n))^2} = c^2$$

Question 2

Proposed Algorithm

Assumption: The array index counters start at 1

- set $\text{partitionSize} = n/4$
- set $\text{medianA} = \text{medianB} = n/2$
- while not $\text{partitionSize} == 0$
 - if $\text{arrayA}[\text{medianA}] > \text{arrayB}[\text{medianB}]$
 $\text{medianA} -= \text{partitionSize}$
 $\text{medianB} += \text{partitionSize}$
 else
 $\text{medianA} += \text{partitionSize}$
 $\text{medianB} -= \text{partitionSize}$
 - $\text{partitionSize} *= 0.5$
- if $\text{arrayA}[\text{medianA}] > \text{arrayB}[\text{medianB}]$
 $\text{medianB} ++$
 else
 $\text{medianA} ++$
- if $\text{arrayA}[\text{medianA}] > \text{arrayB}[\text{medianB}]$
 return $\text{arrayB}[\text{medianB}]$
 else
 return $\text{arrayA}[\text{medianA}]$

Justification: total running time is $O(\log n)$ as each time that the while loop is called, the examined size of the arrays decreases in half, and all other operations are basic comparison and arithmetic operations.

Correctness: We start from looking at the element in the middle of each of the arrays. In the array that has the smaller median, all the numbers smaller than that number are definitely smaller than the median in the other array as well. In the other array, everything bigger than the median is also bigger than the median of the first array. Thus, by excluding these parts from the search area, we ensure that we ignore only that which trivially is not close to the median. This repeats until no other items can be discarded, and thus we find the n -th smallest element in the array

Question 3

The algorithm to solve this problem is very similar to the MergeSort algorithm, with one additional variable to count the number of inverted pairs, and one additional line in the algorithm to track the value of the variable.

The suggested algorithm is a modification to an array based algorithm of the MergeSort algorithm on Wikipedia.org

```
function CountInvertedPairs(intArray intArr, int start, int end)
// if list size is 1 consider it sorted and return it
if sizeof(intArr) <= 1
return intArr
// else array size is > 1, so split the array into two subarrays
// The counter of inverted pairs, as global
var integer invertedCount = 0
// The start and end of the left of the subarray
var int leftStart, leftEnd
// The start and end of the right of the subarray
var int rightStart, rightEnd
var integer middle = floor((end - start) / 2)
// recursively call CountInvertedPairs() to further split each
// sublist until subarray size is 1
left = CountInvertedPairs(intArr, start, middle)
right = CountInvertedPairs(intArr, middle + 1, end)
// merge the sublists returned from prior calls to
// CountInvertedPairs() and return the resulting merged sublist
return merge(left right invertedCount)
```

```
function merge(intArray left, intArray right, invertedCount)
var intArray result
// Indexes to each of the arrays - left, right and result
var int leftIndex = 0, rightIndex = 0, resIndex = 0
while sizeof(left) > 0 or sizeof(right) > 0
if sizeof(left) > 0 and sizeof(right) > 0
if left[leftIndex] <= right[rightIndex]
result[resIndex] = left[leftIndex]
leftIndex++
resIndex++
else
result[resIndex] = right[rightindex]
rightIndex++
resIndex++
invertedCount = rightIndex - resIndex
else if sizeof(left) > 0
```

```
result[resIndex] = left[leftIndex]
leftIndex++
resIndex++
else if sizeof(right) > 0
result[resIndex] = right[rightindex]
rightIndex++
resIndex++
invertedCount = rightIndex - resIndex
end while
return result
```

Question 4

Assumption:

- The set S is stored in an array
- All the numbers in the given set are less than x . If not, a simple modification to the MergeSort algorithm can exclude all irrelevant numbers.
- None of the numbers in the set appears more than one. If not, a simple modification to the MergeSort algorithm can exclude all irrelevant numbers.

The algorithm:

- Run arrayMergeSort(set S) - complexity of $O(n \log n)$
(The same algorithm as presented in the solution to Question 3, without the counter and operations for counting inverted pairs)
- Use a binary search to find *half* - all members of the array to the left of this number are smaller or equal to $x/2$, and the members to the right of the array are bigger or equal to $x/2$ - complexity of $O(\log(n))$
- If $\text{indexOf}(\text{half}, \text{completeArray}) < (\text{sizeof}(\text{completeArray}))/2$ then
 $\text{matchToArr}[\text{start}] = 0$
 $\text{matchToArr}[\text{end}] = \text{indexOf}(\text{half}, \text{completeArray})$
 $\text{matchFromArr}[\text{start}] = \text{indexOf}(\text{half}, \text{completeArray}) + 1$
 $\text{matchFromArr}[\text{end}] = \text{sizeof}(\text{completeArray})$
 Else
 $\text{matchToArr}[\text{start}] = \text{indexOf}(\text{half}, \text{completeArray}) + 1$
 $\text{matchToArr}[\text{end}] = \text{sizeof}(\text{completeArray})$
 $\text{matchFromArr}[\text{start}] = 0$
 $\text{matchFromArr}[\text{end}] = \text{indexOf}(\text{half}, \text{completeArray})$
- For Each Item in matchToArray - complexity of $O(n \log(n))$
 - Use binarySearch in the matchFromArray to find the value $x - \text{Item}$
 - If binarySearch returns NOT NULL
 Return TRUE and break
 - Return FALSE

The highest complexity is encountered when the sorting of the complete set is executed. The sum matching process will run $m \log(k)$ times, where at most $m = k = n/2$, and in most cases $m < n/2$, hence the total matching process will run faster than $O((n/2) \log(n/2))$

Question 5

This is a simple proof by induction. For $n = 1$: There is only one edge, hence it's possible to construct only one minimal spanning tree

For $n = k - 1$: Assume that $(k-1)$ edges were chosen by now, and they resulted only in one minimal spanning tree

For $n = k$: An additional edge needs to be chosen at this point. Kruskal's algorithm is a greedy algorithm and it always chooses the edge with the minimal weight available. It is known that there are no two edges with the same weight, hence when coming to choose, the group of edges with minimal weight is of size 1, and only that edge will be added to the minimal spanning tree. This will repeat for each k , $0 < k < n$ according to the problem statement.