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CS590 Homework 1 Analyzing
Algorithms Reinforcement
Exercises

Due Date: January 30, 2022

Problem 1.6.7:

The following list of functions below were ordered by the big-Oh notation from increasing to decreasing. Section 1.2 Table 1.2.4 was used to order the following functions as well as reasoning and when in doubt between 2 functions, plug in large multiple numbers to check which functions is higher

Ordering the following list of functions by the big-Oh notation from increasing to decreasing:

- 1) 2^{2^n}
- 2) 4^n
- 3) 2^n
- 4) n^3
- 5) $(n^2) * (\log(n))$
- 6) $4^{\log(n)}$
- 7) $4n^{3/2}$
- 8) $2 * n * (\log(n))^2$
- 9) $6 * n * \log(n)$

- 10) $n \log_4(n)$
- 11) $5n$
- 12) $2^{\log(n)}$
- 13) $3 * n^{(0.5)}$
- 14) \sqrt{n}
- 15) $n^{(0.01)}$
- 16) $(\log(n))^{(2)}$
- 17) $\sqrt{\log(n)}$
- 18) $\log(\log(n))$
- 19) $2^{(100)}$
- 20) $1/n$

Grouping Together functions that are big-Theta of one another:

- 1) $6 * n * \log(n)$ and $n * \log_4(n)$ are big-Theta of one another (Reasoning below)
- 2) $2^{(\log(n))}$ and $5n$ are big-Theta of one another (Reasoning below)
- 3) $3 * (n^{0.5})$ and \sqrt{n} are big-Theta of one another (Reasoning below)

Reasoning for $3 * (n^{0.5})$ and \sqrt{n} are big-Theta of one another:

$$n^{0.5} = \sqrt{n}$$

This means the following:

$$3 * (n^{0.5}) = 3 * \sqrt{n}$$

$3 * \sqrt{n}$ and \sqrt{n} are big-Theta of one another because the only difference is that one of them is multiplied by a constant, which is 3 in this case.

Reasoning for $2^{\log(n)}$ and $5n$ are big-Theta of one another:

Let's consider $n = 2^x$, then we have the following:

$$2^{\log(n)} = 2^{\log(2^x)} = 2^x \text{ and } 5n = 5 * (2^x)$$

Both functions are big-Theta of one another because the only difference is that one is multiplied by a constant, which is 5 in this case.

Reasoning for $6n\log(n)$ and $n\log_4(n)$ are big-Theta of one another:

Let's consider as the previous example that

$$n = 2^x$$

This means the following:

$$\begin{aligned} 6 * n * \log(n) &= 6 * (2^x) * \log(2^x) \\ &= 6 * x * (2^x) \end{aligned}$$

$$n * \log_4(n) = (2^x) * (\log_4 2^x)$$

Now, let's divide x by 2 and multiply by 2 like nothing is done just so that we can square the 2 to get a 4:

$$\begin{aligned}
 n * \log_4(n) &= (2^x) \\
 &\quad * (\log_4 \left(2^{2 * \frac{x}{2}} \right)) = (2^x) * (\log_4 4^{\frac{x}{2}}) \\
 n * \log_4(n) &= (2^x) * \left(\frac{x}{2} \right) = \left(\frac{1}{2} \right) * (x) * (2^x)
 \end{aligned}$$

So, $6n \log(n)$ is as follows:

$$\begin{aligned}
 6 * n * \log(n) &= 6 * (2^x) * \log(2^x) \\
 &= 6 * x * (2^x)
 \end{aligned}$$

And $n \log_4(n)$ is as follows:

$$n \log_4(n) = \left(\frac{1}{2} \right) * (x) * (2^x)$$

So, from this, we can deduce that both functions are big-Theta of one another because the only difference is that both functions are multiplied by a different constant.

Problem 1.6.9:

Looking at Figure 1.3.2 in Section 1.3, I will start analyzing the primitive operations to determine the runtime of the Algorithm given in terms of n .

Figure 1.3.2: Algorithm

Algorithm arrayFind(x, A):

Input: An element x and an n -element array, A .

Output: The index i such that $x = A[i]$ or -1 if no element of A is equal to x .

$i \leftarrow 0$

while $i < n$ **do**

if $x = A[i]$ **then**

return i

else

$i \leftarrow i + 1$

return -1

[Feedback?](#)

One of the first primitive operation that come to mind is assigning $i = 0$. This would be counted as 1 primitive operation.

Another primitive operation that come to mind is the comparison between i and n . This primitive operation will run $n+1$ times because i starts at 0. n is the length of the array. The instructions inside of the while loop will run n times. However, the comparison runs 1 time more, which means $n+1$, because it would also get tested for n . So, the primitive operation for the comparison between i and n is $n + 1$.

Another primitive operation that comes to mind is checking whether x is equal to $A[i]$. This primitive operation will run n times in the worst case scenario. In the worse-case scenario, a value of -1 will be returned and no x will be matched with any row in the $n \times n$ array. So, the worst-case scenario is when we check whether x is equal to $A[i]$ for all the rows in the $n \times n$ array. So, the primitive operation of checking whether x is equal to $A[i]$ is n .

return i will never make it in the worst-case scenario. In fact, to get the worst-case scenario, the algorithm should skip over return i.

Another primitive operation that comes to mind is the 2 primitive operations produced by $i = i + 1$. First, we add i to 1, and this is one primitive operation and then we assign i with the value of $i + 1$. These 2 primitive operations will always be running (n times). So, the primitive operation produced by $i = i + 1$ is $(2) * (n)$.

Another primitive operation that comes to mind is return -1. This return -1 occurs in the worst-case scenario. It occurs after all values in a single row are checked and x does not match any value in the arrays row.

After Jotting all of them above, let's calculate the worst-case running time of arrayFind in terms of n,

$$(1) + (n + 1) + (3 * n) + 1 = 4n + 3 = O(n)$$

Now, it is mentioned in the Exercise that find2D iterates over the rows of A and calls the Algorithm arrayFind on each one until x is found or it has searched all rows of A.

This means that find2D will repeat the same thing that arrayFind did for each row. Since we have n rows, then the worst-case runtime of find2D in terms of n is $O(n^2)$.

Therefore, in conclusion the worst-case runtime of find2D in terms of n is $O(n^2)$.

Is this a linear Algorithm? Why or why not?

Since the worst-case runtime of find2D in terms of n is $O(n^2)$, the Algorithm is not considered linear. An Algorithm is said to have linear time if its worst-case runtime is $O(n)$.

Since the worst-case runtime for find2D is $O(n^2)$, then it is not a linear time Algorithm.

Problem 1.6.22:

As stated in Section 1.2 after Example 1.2.9, $f(n)$ is $o(g(n))$ if and only if

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$$

provided this limit exists

So, let's apply this into practice to show that

$$n \text{ is } o(n * \log(n))$$

By looking at the limit formula above and at the problem question that we want to solve, we can easily determine the following:

$$f(n) = n \text{ and } g(n) = n * \log(n)$$

So, let's solve for the limit and check if the limit exist and we get 0:

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{(n)}{(n * \log(n))} = \lim_{n \rightarrow \infty} \frac{1}{\log(n)}$$

$\log(n)$ increases when n increases and reaches very high values when n is very high. This means that $\log(n)$ at plus infinity is plus infinity. $1/\text{infinity}$ is 0:

$$\lim_{n \rightarrow \infty} \frac{1}{\log(n)} = 0$$

Therefore,

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$$

This shows that n is $o(n \log(n))$