

CS5800: Algorithms SEC 05 Summer Full 2024 (Seattle)
Homework 1

Payal Chavan

05/15/2024

Question 1: In each of the following situations, indicate whether $f = \mathcal{O}(g)$, or $f = \Omega(g)$, or both (in which case $f = \Theta(g)$).

Part (a): $f(n) = n - 100$ $g(n) = n - 200$

Solution: Since $n-100$ and $n-200$ have the same power of n , we can say that $f = \Theta(g)$.

Part (b): $f(n) = n^{1/2}$ $g(n) = n^{2/3}$

Solution: If we compare the powers of n i.e., $1/2$ and $2/3$, we have $0.5 < 0.6$. Hence, we can say that $f = \mathcal{O}(g)$.

Part (c): $f(n) = 100n + \log n$ $g(n) = n + (\log n)^2$

Solution: As both the functions have $\log n$ terms, both are $\mathcal{O}(n)$. Hence, we have $f = \Theta(g)$.

Part (d): $f(n) = n \log n$ $g(n) = 10n \log 10n$

Solution: After solving $g(n) = 10n \log 10n$, we get $10n(\log 10 + \log n) = 10n \log 10 + 10n \log n$. Both of them have $f = \mathcal{O}(n \log n)$. Hence, $f = \Theta(g)$.

Part (e): $f(n) = \log 2n$ $g(n) = \log 3n$

Solution: When we solve $\log 2n = \log 2 + \log n$ and $\log 3n = \log 3 + \log n$, we get $\mathcal{O}(\log n)$. Because, $\log 2$ and $\log 3$ are constant terms. Therefore, $f = \Theta(g)$.

Part (f): $f(n) = 10 \log n$ $g(n) = \log(n^2)$

Solution: We have, $g(n) = \log(n^2)$ i.e., $2 \log n$. As both of them have $\mathcal{O}(\log n)$, we get $f = \Theta(g)$.

Part (g): $f(n) = n^{1.01}$ $g(n) = n \log^2 n$

Solution: If we divide, both sides by n , we get $n^{0.01}$ and $\log^2 n$. After comparing these values, ultimately, power function would always overtake the log function. Hence, we get $f = \Omega(g)$.

Part (h): $f(n) = n^2 / \log n$ $g(n) = n (\log n)^2$

Solution: We can neglect $\log n$ part in both the functions, as logs are not that significant. So, then we are left with n^2 and n . Hence, we get $f = \Omega(g)$.

Part (i): $f(n) = n^{0.1}$ $g(n) = (\log n)^{10}$

Solution: As power function would always be dominant over the log. Hence, we get $f = \Omega(g)$.

Part (j): $f(n) = (\log n)^{\log n}$ $g(n) = n / \log n$

Solution: Let $n > 1$, then for example $(\log 8)^{\log 8} > 8 / \log 8$ i.e., $27 > 2.6$
So any number when divided would result in a smaller value as compared to the exponential. Hence, we get $f = \Omega(g)$.

Part (k): $f(n) = \sqrt{n}$ $g(n) = (\log n)^3$

Solution: Since any polynomial dominates any logarithms, we get $f = \Omega(g)$.

Part (l): $f(n) = n^{1/2}$ $g(n) = 5^{\log_2(n)}$

Solution: The function $g(n) = 5^{\log_2(n)} = n^{\log_2(5)} = n^{2.32}$.
After comparing, we have $n^{0.5} < n^{2.32}$.
Hence, we get $f = \mathcal{O}(g)$.

Part (m): $f(n) = n2^n$ $g(n) = 3^n$

Solution: Let us assume, $n = k$ (some real integer). Now, we can neglect the multiplicative constant and so $f(n) = 2^n$.
We will just compare 2^n with 3^n .
By the mathematical induction, we get the result $3^n > 2^n$ for all n and $(n + 1)$ values. Therefore, $f = \mathcal{O}(g)$.

Part (n): $f(n) = 2^n$ $g(n) = 2^{n+1}$

Solution: Since, both differ just by the multiplicative constant 2, we get $f = \Theta(g)$.

Part (o): $f(n) = n!$ $g(n) = 2^n$

Solution: $n!$ grows very fast than compared to a power of n . It's because the greater number is multiplied with the product each time.
 $(n + 1)! = 1.2 \dots n.(n + 1)$
Whereas, in the case of exponential function
 $c^{n+1} = c.c.c \dots$
Hence, we get $f = \Omega(g)$.

Part (p): $f(n) = (\log n)^{\log n}$ $g(n) = 2^{(\log_2 n)^2}$

Solution: Solving, $f(n) = (\log n)^{\log n} = n^{\log(\log n)}$ and
 $g(n) = 2^{(\log_2 n)^2} = 2^{(\log_2 n) * (\log_2 n)}$
 $g(n) = 2^{(\log_2 n)(\log_2 n)}$
By using Logarithmic rule, $b^{\log_b(k)} = k$, we get,
 $g(n) = n^{(\log_2 n)}$
Now comparing $f(n)$ and $g(n)$,
 $f(n) < g(n)$, because $\log(\log n)$ is much slower than that of $\log_2(n)$.
Therefore, we get, $f = \mathcal{O}(g)$.

Part (q): $f(n) = \sum_{i=1}^n i^k$ $g(n) = n^{k+1}$

Solution: Simplifying, $f(n) = \sum_{i=1}^n i^k = 1^k + 2^k + 3^k + \dots + n^k$
 $g(n) = n^{k+1} = n^k \cdot n$
Let us assume, $k = 1$ (some real integer) and $n = 10$, we get,
 $f(n) = 1 + 2 + 3 + \dots + 10 = 55$
 $g(n) = 10^2 = 10 * 10 = 100$
Therefore, as $\sum_{i=1}^n i^k < n^{k+1}$, we get $f = \mathcal{O}(g)$.

Question 2: Show that, if c is a positive real number, then $g(n) = 1 + c + c^2 + \dots + c^n$ is:

- (a) $\Theta(1)$ if $c < 1$
- (b) $\Theta(n)$ if $c = 1$
- (c) $\Theta(c^n)$ if $c > 1$

Solution:

a) if $c < 1$

Proof. given: $g(n) = 1 + c + c^2 + \dots + c^n$

Let us consider, $c = 0.5$

Now, $g(n) = 1 + (0.5) + (0.5)^2 + \dots + (0.5)^n$

The geometric series goes on decreasing. So, we could skip the decimal terms as they are negligible. The maximum number will be 1. With that, $g(n)$ becomes independent of n .

Hence, we get, $g(n) = \Theta(1)$ □

b) if $c = 1$

Proof. given: $g(n) = 1 + c + c^2 + \dots + c^n$

Let us consider, $c = 1$

Now, $g(n) = 1 + (1) + (1)^2 + \dots + (1)^n$

The geometric series is unchanging and will result in 'n' terms.

Hence, we get, $g(n) = \Theta(n)$ □

c) if $c > 1$

Proof. given: $g(n) = 1 + c + c^2 + \dots + c^n$

Let us consider, $c = 2$

Now, $g(n) = 1 + (2) + (2)^2 + \dots + (2)^n$

The geometric series is increasing and the largest term would be $(2)^n$.

Hence, we get, $g(n) = \Theta(c^n)$ □

Question 3: Is there a faster way to compute the nth Fibonacci number than by fib2? One idea involves matrices.

We start by writing the equations $F_1 = F_1$ and $F_2 = F_0 + F_1$ in matrix notation:

$$\begin{pmatrix} F_1 \\ F_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} F_0 \\ F_1 \end{pmatrix}$$

Similarly,

$$\begin{pmatrix} F_2 \\ F_3 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} F_1 \\ F_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}^2 \cdot \begin{pmatrix} F_0 \\ F_1 \end{pmatrix}$$

and in general

$$\begin{pmatrix} F_n \\ F_{n+1} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}^n \cdot \begin{pmatrix} F_0 \\ F_1 \end{pmatrix}$$

So, in order to compute F_n , it suffices to raise this 2×2 matrix, call it X , to the n th power.

Part (a): Show that two 2x2 matrices can be multiplied using 4 additions and 8 multiplications.

Solution: Let us consider the matrices X and Y.

$$X = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \text{ and } Y = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

Now, multiplying X and Y, we get,

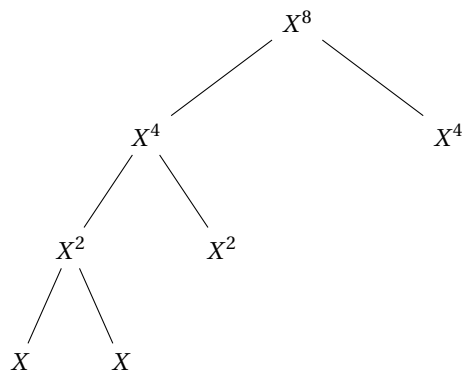
$$X \cdot Y = \begin{bmatrix} (1 * 5) + (2 * 7) & (1 * 6) + (2 * 8) \\ (3 * 5) + (4 * 7) & (3 * 6) + (4 * 8) \end{bmatrix}$$

Hence, from the above 2x2 matrix, we have got 4 additions and 8 multiplications.

Part (b): But how many matrix multiplications does it take to compute X^n ?

Show that $f = \mathcal{O}(\log n)$ matrix multiplications suffice for computing X^n . (Compute X^8).

Solution: Suppose, we want to compute X^8 .



From the above figure, we can write X^8 as,

$$X^8 = X^4 * X^4$$

$$\Rightarrow (X^2 * X^2) * X^4$$

$$\Rightarrow (X * X) * X^2 * X^4$$

As we can see, this gives us a total of 3 matrix multiplications.

Then, $X^8 = 3$ multiplications.

To prove, $f = \mathcal{O}(\log n)$,

Here $n = 8$

Taking log on both sides,

$$\log_2(n) \Rightarrow \log_2(8)$$

$$\Rightarrow \log_2(2)^3 = 3$$

Hence proved $f = \mathcal{O}(\log n)$ for X^n .

If we have to compute Fibonacci of, say 7.

We already know the value of matrix X^8 .

All we have to do is compute $X^2 = X * X$ and $X^4 = X^2 * X^2$ and $X^8 = X^4 * X^4$, which gives us a total of 3 matrix multiplication operations as before.

for Even $X^8 \Rightarrow (X * X) * X^2 * X^4$

for Odd $X^7 = X^{8-1}$

$$\begin{aligned} &\Rightarrow X^1 * X^2 * X^4 \\ &\Rightarrow X^1 * (X * X) * X^4 \end{aligned}$$

So, as proven before for $n = \text{even}$, similarly for $(n-1)$ is odd, we get same number of matrix multiplications which is $O(\log n)$.

To generalize for even and odd cases.

$2^n + 1$ to 2^{n+1} exponent ranges where we will have same number of matrix multiplications.

Let's assume $n = 3$, $2^3 + 1$ to $2^{3+1} \Rightarrow X^9$ to X^{16} we will have same number of matrix multiplications that is 4.

Reflection:

From exercise 0.1, I gained the understanding of "Asymptotic Analysis" and the difference between Big \mathcal{O} vs Big Theta Θ vs Big Omega Ω .

From exercise 0.2, I grasped the concept of geometric series.

And finally from exercise 0.4, I understood the working of Fibonacci series and how to compute matrices using the recursive method.

Acknowledgements:

I would like to express my sincere gratitude to the following individuals and resources for their invaluable contributions to this assignment:

- 1) Prof. Bruce Maxwell: Thank you, Professor Bruce Maxwell for your guidance in this assignment. Your expertise in algorithms greatly influenced my work.
- 2) Classmates and TA's: I appreciate the discussions of my classmates, and TA's who clarified my doubts.
- 3) DPV Algorithms Textbook: This textbook was a useful resource for me to understand the basics of algorithms.
- 4) <https://www.geeksforgeeks.org/difference-between-big-oh-big-omega-and-big-theta/>: This website provided clear explanations of various notations, enhancing my understanding.
- 5) <https://www.mathsisfun.com/algebra/matrix-multiplying.html>: This website helped me to learn matrix multiplication.
- 6) <https://www.chilimath.com/lessons/advanced-algebra/logarithm-rules/>: This website provided me the logarithmic rules to solve exercise 0.1 questions.