

# Algorithms and Data Structures Coursework

Finlay Boyle

November 2019

## Question 4

### Part a

If  $f(x) + g(x)$  is  $o(f(x)g(x))$  then  $\lim_{x \rightarrow \infty} \frac{f(x)+g(x)}{f(x)g(x)} = 0$

$$\begin{aligned}\lim_{x \rightarrow \infty} \frac{f(x) + g(x)}{f(x)g(x)} &= \lim_{x \rightarrow \infty} \left( \frac{f(x)}{f(x)g(x)} + \frac{g(x)}{f(x)g(x)} \right) \\ &= \lim_{x \rightarrow \infty} \left( \frac{1}{g(x)} + \frac{1}{f(x)} \right)\end{aligned}$$

This is not true for all asymptotically positive functions. A counter-example to this is  $f(x) = \frac{x}{x+2}$  and  $g(x) = x^4$ .  $f(x)$  is asymptotically positive since  $x^4 > 0 \forall x \neq 0$  and  $g(x)$  is asymptotically positive since  $\frac{x}{x+2} > 0 \forall x > 0$ .

$$\lim_{x \rightarrow \infty} \frac{1}{f(x)} = \lim_{x \rightarrow \infty} \frac{1}{x^4} = 0$$

and

$$\begin{aligned}\lim_{x \rightarrow \infty} \frac{1}{g(x)} &= \lim_{x \rightarrow \infty} \frac{1}{\frac{x}{x+2}} \\ &= \lim_{x \rightarrow \infty} \frac{x+2}{x} \\ &= \lim_{x \rightarrow \infty} \left( 1 + \frac{2}{x} \right) \\ &= 1\end{aligned}$$

hence

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

So these functions are a counter-example so  $x^4 + \frac{x}{x+2}$  is not  $o(x^4 * \frac{x}{x+2})$  hence it is **false**.

## Part b

If  $2^x x^2$  is  $o(2.1^x)$  then  $\lim_{x \rightarrow \infty} \frac{2^x x^2}{2.1^x} = 0$

$$\begin{aligned}\lim_{x \rightarrow \infty} \frac{2^x x^2}{2.1^x} &= \lim_{x \rightarrow \infty} x^2 \left( \frac{2}{2.1} \right)^x \\ &= \lim_{x \rightarrow \infty} x^2 \left( \frac{2.1}{2} \right)^{-x} \\ &= \lim_{x \rightarrow \infty} \frac{x^2}{\left( \frac{2.1}{2} \right)^x}\end{aligned}$$

Then since the exponential is monotonic increasing as its base is greater than 1 and exponentials beat powers as  $x \rightarrow \infty$  then  $\lim_{x \rightarrow \infty} \frac{2^x x^2}{2.1^x} = 0$  hence  $2^x x^2$  is  $o(2.1^x)$  is **true**.

## Part c

$$f(x) = x^2 \log(x), g(x) = x^2$$

Assume  $\exists C \in (0, \infty)$  and  $k \in (0, \infty)$  such that  $x^2 \log(x) \leq Cx^2 \forall x \geq k$  then  $\log(x) \leq C$ . However,  $\log(x)$  is monotonic increasing so  $\log(x) > C$  for sufficiently large  $x$  hence there does not exist a witness pair  $C, k$  which is a contradiction so  $x^2 \log(x)$  is not  $O(x^2)$ . The statement is **false**.

## Part d

$$f(x) = x^2 \log(x), g(x) = x^3 \text{ then}$$

$$\begin{aligned}x^2 \log(x) &\leq Cx^3 \\ \log(x) &\leq Cx\end{aligned}$$

This is **true** for  $x \geq 1$  as  $\log(1) = 0$  and  $x = 1$ ,  $x$  grows faster than  $\log(x)$  therefore giving the witness pair  $C = 1, k = 1$  so  $x^2 \log(x)$  is  $O(x^3)$

## Part e

$$f(x) = 7x^5, g(x) = 12x^4 + 5x^3 + 8$$

Assume  $\exists C \in (0, \infty)$  and  $k \in (0, \infty)$  such that  $7x^5 \leq C(12x^4 + 5x^3 + 8) \forall x \geq k$  then:

$$\begin{aligned}7x^5 &\leq C(12x^4 + 5x^3 + 8) \\ 7x &\leq 12C + \frac{5C}{x} + \frac{8C}{x^4}\end{aligned}$$

As  $x \rightarrow \infty$ ,  $\frac{1}{x} \rightarrow 0$  and  $\frac{1}{x^4} \rightarrow 0$  so:

$$\lim_{x \rightarrow \infty} \frac{5C}{x} = 0, \lim_{x \rightarrow \infty} \frac{8C}{x^4} = 0$$

So for sufficiently large  $x$ ,  $\frac{5C}{x}$  and  $\frac{8C}{x^4}$  become insignificant so the RHS is a decreasing function whereas the LHS is a monotonic increasing function hence  $7x^5 > C(12x^4 + 5x^3 + 8)$  for sufficiently large  $x$  so there cannot exist a witness pair  $C, k$  which is a contradiction therefore  $7x^5$  is not  $O(12x^4 + 5x^3 + 8)$  hence the statement is **false**.

## Question 5

### Part a

$T(n) = 64T(\frac{n}{8}) - n^2 \log(n)$  so  $a = 64$ ,  $b = 8$ ,  $f(n) = -n^2 \log(n)$

Since  $a$  and  $b$  are constants the Master Theorem will apply if the recurrence satisfies one of the cases:

$$n^{\log_b a} = n^{\log_8 64} = n^2$$
$$f(n) = -n^2 \log(n) = \Theta(n^2 \log^1(n))$$

It is clear that  $f(n) = \Theta(n^2 \log^1(n))$  because it only differs by a constant of  $-1$  so the recurrence satisfies  $\Theta(n^2 \log^1(n))$  hence the Master Theorem can be applied since  $k \geq 0$ :

$$T(n) = \Theta(n^{\log_b a} \log^k(n)) = \Theta(n^2 \log^2(n))$$

Therefore,  $T(n) = \Theta(n^2 \log^2(n))$

### Part b

$T(n) = 4T(\frac{n}{2}) + \frac{n}{\log(n)}$  so  $a = 4$ ,  $b = 2$  and  $f(n) = \frac{n}{\log(n)} = n \log^{-1}(n)$

The Master Theorem may apply here because  $a$  and  $b$  are both constants.

$$n^{\log_b a} = n^{\log_2 4} = n^2$$

This appears to be case 1 of the Master Theorem because for case 2,  $\log^k(n)$  requires  $k \geq 0$  whereas  $k = -1$  here.

To use case 1,  $f(n) = O(n^{2-\epsilon})$ :

$$\frac{n}{\log(n)} \leq C n^{2-\epsilon}$$
$$\frac{1}{\log(n)} \leq C n^{1-\epsilon}$$

Since  $n^{1-\epsilon}$  is monotonic increasing and tends towards  $\infty$  because  $\epsilon$  is a small positive constant and  $\log(n)$  is also monotonic increasing as  $n \rightarrow \infty$  so  $\frac{1}{\log(n)} \rightarrow 0$ . So the inequality holds for  $n \geq 2$ ,  $C = 1$  (assuming  $\log(n) = \log_2(n)$ ). Hence,

$$f(n) = O(n^{2-\epsilon})$$

The Master Theorem therefore applies so:

$$T(n) = \Theta(n^{\log_b(a)}) = \Theta(n^2)$$

### Part c

$T(n) = 2^n T(\frac{n}{2}) + n^n$  so  $a = 2^n$ ,  $b = 2$  and  $f(n) = n^n$

The Master Theorem cannot be applied here because  $a$  is not a constant which is required for the Master Theorem. Hence,  $T(n)$  cannot be solved by the Master Theorem.

## Part d

$T(n) = 3T(\frac{n}{4}) + n\log(n)$  so  $a = 3$ ,  $b = 4$  and  $f(n) = n\log(n)$

As  $a$  and  $b$  are constants, the Master Theorem may apply here if it satisfies one of the cases.

$$n^{\log_b(a)} = n^{\log_4(3)} \approx n^{0.792}$$

Since  $O(n^{\log_4(3)}) < O(n) < O(n\log(n))$ , this appears to case 3 of the Master Theorem.

To use case 3,  $f(n) = \Omega(n^{\log_4(3)})$

$$\begin{aligned} Cn\log(n) &\geq n^{\log_4 3} \\ C\log(n) &\geq n^{\log_4 3 - 1} \\ C\log(n) &\geq \frac{1}{n^{1 - \log_4 3}} \end{aligned}$$

This inequality holds for  $C = 1$  and  $n \geq 2$  (assuming  $\log(n) = \log_2(n)$ ) since  $C\log(n)$  and  $n^{1 - \log_4 3}$  are both monotonic increasing so  $\frac{1}{n^{1 - \log_4 3}} \rightarrow 0$  as  $n \rightarrow \infty$ . Hence,

$$f(n) = \Omega(n^{\log_4(3)})$$

Then for the second condition,  $af(\frac{n}{b}) \leq cf(n)$  for some  $c < 1$  and  $n \geq k$ :

$$\begin{aligned} 3f\left(\frac{n}{4}\right) &\leq cf(n) \\ \frac{3n}{4}\log\left(\frac{n}{4}\right) &\leq cn\log(n) \\ \frac{3}{4}(\log(n) - \log(4)) &\leq c\log(n) \\ \frac{3}{4}\log(n) - \frac{3}{4}\log(4) &\leq c\log(n) \end{aligned}$$

So for sufficiently large  $n$ , if  $\frac{3}{4} \leq c < 1$  then this holds true as  $\frac{3}{4}\log(4)$  becomes insignificant as it is a constant. Hence,  $\exists c < 1$  such that  $af(\frac{n}{b}) \leq cf(n)$  is true so this is case 3 of the Master Theorem:

$$T(n) = \Theta(f(n)) = \Theta(n\log(n))$$

## Part e

$T(n) = 3T(\frac{n}{3}) + \sqrt{n}$  so  $a = 3$ ,  $b = 3$  and  $f(n) = \sqrt{n}$

As  $a$  and  $b$  are constants, the Master Theorem may apply here if it satisfies one of the cases.

$$n^{\log_b(a)} = n^{\log_3(3)} = n$$

This is clearly case 1 so  $f(n) = O(n^{\log_b(a) - \epsilon})$ :

$$\begin{aligned} n^{0.5} &\leq Cn^{1 - \epsilon} \\ 1 &\leq Cn^{0.5 - \epsilon} \end{aligned}$$

This is true for  $C = 1$ ,  $n \geq 1$  and  $\epsilon < 0.5$  (but  $\epsilon$  is a small positive constant anyway). Hence,  $f(n) = O(n^{\log_b(a) - \epsilon})$  thus

$$T(n) = \Theta(n^{\log_b(a)}) = \Theta(n)$$

## Question 6

### Part b

The worse-case input for this algorithm for  $k$  pivots is:

- The first  $2k$  elements of the array are reverse-sorted.
- The next  $k$  elements are all greater than the first  $2k$  elements. These elements are also reverse sorted.
- Repeat this for as many  $k$  element blocks each one must contain elements greater than all previous elements in the array. The  $k$  elements must be reverse-sorted.

For example, for a 15-element array with  $k = 5$  a worst-case input would be [10, 9, 8, 7, 6, 5, 4, 3, 2, 1, 15, 14, 13, 12, 11, 20, 19, 18, 17, 16].

This is the worst-case input because when the algorithm selects the pivots it will select them from the right end. It will pick  $k$  elements which are reverse-sorted and then it will sort them using insertion sort. The worst-case for insertion sort is a reverse-sorted array.

Then it will partition the array between these pivots however none of the elements in the array will be put in the partitions producing a partition of size  $n - k$  and  $k$  partitions (each with one element in) on the right side of the array. It will repeat this until there are  $2k$  elements left. Each time it only reduces the problem by  $k$ -elements.

Once there are  $2k$  elements left it will sort them using insertion sort. These elements are also reverse-sorted which is again the worst-case for insertion sort. This overall gives the worst-case.