

Algorithms and Data Structures Coursework

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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

(Can $f(n)$ be negative??) $T(n) = 64T(\frac{n}{8}) - n^2 \log(n)$ 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 ∞ because ϵ 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 ϵ 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 $n - k$ partition and k -sorted elements 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 will be also reverse-sorted which is again the worst-case for insertion sort. This overall gives the worst-case.