Chapter 3

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Exercise 3.1-1

Since we are requiring both f and g to be aymptotically non-negative, suppose that we are past some n_1 where both are non-negative (take the max of the two bounds on the n corresponding to both f and g). Let $c_1 = .5$ and $c_2 = 1$.

$$0 \le .5(f(n) + g(n)) \le .5(\max(f(n), g(n)) + \max(f(n), g(n)))$$
$$= \max(f(n), g(n)) \le \max(f(n), g(n)) + \min(f(n), g(n)) = (f(n) + g(n))$$

Exercise 3.1-2

Let $c=2^b$ and $n_0\geq 2a$. Then for all $n\geq n_0$ we have $(n+a)^b\leq (2n)^b=cn^b$ so $(n+a)^b=O(n^b)$. Now let $n_0\geq \frac{-a}{1-1/2^{1/b}}$ and $c=\frac{1}{2}$. Then $n\geq n_0\geq \frac{-a}{1-1/2^{1/b}}$ if and only if $n-\frac{n}{2^{1/b}}\geq -a$ if and only if $n+a\geq (1/2)^{a/b}n$ if and only if $(n+a)^b\geq cn^b$. Therefore $(n+a)^b=\Omega(n^b)$. By Theorem 3.1, $(n+a)^b=\Theta(n^b)$.

Exercise 3.1-3

There are a ton of different funtions that have growth rate less than or equal to n^2 . In particular, functions that are constant or shrink to zero arbitrarily fast. Saying that you grow more quickly than a function that shrinks to zero quickly means nothing.

Exercise 3.1-4

 $2^{n+1} \ge 2 \cdot 2^n$ for all $n \ge 0$, so $2^{n+1} = O(2^n)$. However, 2^{2n} is not $O(2^n)$. If it were, there would exist n_0 and c such that $n \ge n_0$ implies $2^n \cdot 2^n = 2^{2n} \le c2^n$, so $2^n \le c$ for $n \ge n_0$ which is clearly impossible since c is a constant.

Exercise 3.1-5

Suppose $f(n) \in \Theta(g(n))$, then $\exists c_1, c_2, n_0, \forall n \geq n_0, 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n)$, if we just look at these inequalities saparately, we have $c_1 g(n) \leq f(n)$ $(f(n) \in \Omega(g(n)))$ and $f(n) \leq c_2 g(n)$ $(f(n) \in O(g(n)))$.

Suppose that we had $\exists n_1, c_1, \forall n \geq n_1, c_1 g(n) \leq f(n)$ and $\exists n_2, c_2, \forall n \geq n_2, f(n) \leq c_2 g(n)$. Putting these together, and letting $n_0 = \max(n_1, n_2)$, we have $\forall n \geq n_0, c_1 g(n) \leq f(n) \leq c_2 g(n)$.

Exercise 3.1-6

Suppose the running time is $\Theta(g(n))$. By Theorem 3.1, the running time is O(g(n)), which implies that for any input of size $n \geq n_0$ the running time is bounded above by $c_1g(n)$ for some c_1 . This includes the running time on the worst-case input. Theorem 3.1 also imlpies the running time is $\Omega(g(n))$, which implies that for any input of size $n \geq n_0$ the running time is bounded below by $c_2g(n)$ for some c_2 . This includes the running time of the best-case input.

On the other hand, the running time of any input is bounded above by the worst-case running time and bounded below by the best-case running time. If the worst-case and best-case running times are O(g(n)) and $\Omega(g(n))$ respectively, then the running time of any input of size n must be O(g(n)) and $\Omega(g(n))$. Theorem 3.1 implies that the running time is $\Theta(g(n))$.

Exercise 3.1-7

Suppose we had some $f(n) \in o(g(n)) \cap \omega(g(n))$. Then, we have

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

a contradiction.

Exercise 3.1-8

 $\Omega(g(n,m)) = \{f(n,m) : \text{there exist positive constants } c, n_0, \text{ and } m_0 \text{ such that } f(n,m) \ge cg(n,m)$ for all $n \ge n_0$ or $m \ge m_0\}$

 $\Theta(g(n,m)) = \{f(n,m) : \text{there exist positive constants } c_1, c_2, n_0, \text{ and } m_0 \text{ such that } c_1g(n,m) \le f(n,m) \le c_2g(n,m) \text{ for all } n \ge n_0 \text{ or } m \ge m_0\}$

Exercise 3.2-1

Let $n_1 < n_2$ be arbitrary. From f and g being monatonic increasing, we know $f(n_1) < f(n_2)$ and $g(n_1) < g(n_2)$. So

$$f(n_1) + g(n_1) < f(n_2) + g(n_1) < f(n_2) + g(n_2)$$

Since $g(n_1) < g(n_2)$, we have $f(g(n_1)) < f(g(n_2))$. Lastly, if both are nonegative, then,

$$f(n_1)g(n_1) = f(n_2)g(n_1) + (f(n_2) - f(n_1))g(n_1)$$

= $f(n_2)g(n_2) + f(n_2)(g(n_2) - g(n_1)) + (f(n_2) - f(n_1))g(n_1)$

Since $f(n_1) \ge 0$, $f(n_2) > 0$, so, the second term in this expression is greater than zero. The third term is nonnegative, so, the whole thing is $f(n_2)g(n_2)$.

Exercise 3.2-2

$$a^{\log_b(c)} = a^{\frac{\log_a(c)}{\log_a(b)}} = c^{\frac{1}{\log_a(b)}} = c^{\log_b(a)}$$

Exercise 3.2-3

As the hint suggests, we will apply stirling's approximation

$$\begin{split} \lg(n!) &= \lg\left(\sqrt{(2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)\right) \\ &= \frac{1}{2}\lg(2\pi n) + n\lg(n) - n\lg(e) + \lg\left(\Theta\left(\frac{n+1}{n}\right)\right) \end{split}$$

Note that this last term is $O(\lg(n))$ if we just add the two expression we get when we break up the lg instead of subtract them. So, the whole expression is dominated by $n \lg(n)$. So, we have that $\lg(n!) = \Theta(n \lg(n))$.

$$\lim_{n\to\infty}\frac{2^n}{n!}=\lim_{n\to\infty}\frac{1}{\sqrt{2\pi n}(1+\Theta(\frac{1}{n}))}\left(\frac{2e}{n}\right)^n\leq\lim_{n\to\infty}\left(\frac{2e}{n}\right)^n$$

If we restrict to n > 4e, then this is

$$\leq \lim_{n\to\infty}\frac{1}{2^n}=0$$

$$\lim_{n \to \infty} \frac{n^n}{n!} = \lim_{n \to \infty} \frac{1}{\sqrt{2\pi n} (1 + \Theta(\frac{1}{n}))} e^n = \lim_{n \to \infty} O(n^{-.5}) e^n \ge \lim_{n \to \infty} \frac{e^n}{c_1 \sqrt{n}}$$
$$\ge \lim_{n \to \infty} \frac{e^n}{c_1 n} = \lim_{n \to \infty} \frac{e^n}{c_1} = \infty$$

Exercise 3.2-4

The function $\lceil \log n \rceil!$ is not polynomially bounded. If it were, there would exist constants c, a, and n_0 such that for all $n \ge n_0$ the inequality $\lceil \log n \rceil! \le cn^a$ would hold. In particular, it would hold when $n = 2^k$ for $k \in \mathbb{N}$. Then

this becomes $k! \leq c(2^a)^k$, a contradiction since the factorial function is not exponentially bounded.

We'll show that $\lceil \log \log n \rceil! \le n$. Without loss of generality assume $n = 2^{2^k}$. Then this becomes equivalent to showing $k! \leq 2^{2^k}$, or $1 \cdot 2 \cdots (k-1) \cdot k \leq 2^{2^k}$ $4 \cdot 16 \cdot 2^8 \cdots 2^{2^k}$, which is clearly true for $k \geq 1$. Therefore it is polynomially bounded.

Exercise 3.2-5

Note that $\lg^*(2^n) = 1 + \lg^*(n)$, so,

$$\lim_{n \to \infty} \frac{\lg(\lg^*(n))}{\lg^*(\lg(n))} = \lim_{n \to \infty} \frac{\lg(\lg^*(2^n))}{\lg^*(\lg(2^n))}$$

$$= \lim_{n \to \infty} \frac{\lg(1 + \lg^*(n))}{\lg^*(n)}$$

$$= \lim_{n \to \infty} \frac{\lg(1 + n)}{n}$$

$$= \lim_{n \to \infty} \frac{1}{1 + n}$$

$$= 0$$

So, we have that $\lg^*(\lg(n))$ grows more quickly

Exercise 3.2-6

$$\phi^2 = \left(\frac{1+\sqrt{5}}{2}\right)^2 = \frac{6+2\sqrt{5}}{4} = 1 + \frac{1+\sqrt{5}}{2} = 1 + \phi$$
$$\hat{\phi}^2 = \left(\frac{1-\sqrt{5}}{2}\right)^2 = \frac{6-2\sqrt{5}}{4} = 1 + \frac{1-\sqrt{5}}{2} = 1 + \hat{\phi}$$

Exercise 3.2-7

First, we show that $1+\phi=\frac{6+2\sqrt{5}}{4}=\phi^2$. So, for every $i,\ \phi^{i-1}+\phi^{i-2}=\phi^2$

$$\phi^{i-2}(\phi+1) = \phi^{i}. \text{ Similarly for } \hat{\phi}.$$
For $i=0, \frac{\phi^{0}-\hat{\phi}^{0}}{\sqrt{5}} = 0$. For $i=1, \frac{\frac{1+\sqrt{5}}{2}-\frac{1-\sqrt{5}}{2}}{\sqrt{5}} = \frac{\sqrt{5}}{\sqrt{5}} = 1$. Then, by induction,
$$F_{i} = F_{i-1} + F_{i-2} = \frac{\phi^{i-1}+\phi^{i-2}-(\hat{\phi}^{i-1}+\hat{\phi}^{i-2})}{\sqrt{5}} = \frac{\phi^{i}-\hat{\phi}^{i}}{\sqrt{5}}.$$

Exercise 3.2-8

Let c_1 and c_2 be such that $c_1 n \le k \ln k \le c_2 n$. Then we have $\ln c_1 + \ln n =$ $\ln(c_1 n) \leq \ln(k \ln k) = \ln k + \ln(\ln k)$ so $\ln n = O(\ln k)$. Let c_3 be such that $\ln n \le c_3 \ln k$. Then

$$\frac{n}{\ln n} \ge \frac{n}{c_3 \ln k} \ge \frac{k}{c_2 c_3}$$

so that $\frac{n}{\ln n} = \Omega(k)$. Similarly, we have $\ln k + \ln(\ln k) = \ln(k \ln k) \le \ln(c_2 n) = \ln(c_2) + \ln(n)$ so $\ln(n) = \Omega(\ln k)$. Let c_4 be such that $\ln n \ge c_4 \ln k$. Then

$$\frac{n}{\ln n} \le \frac{n}{c_4 \ln k} \le \frac{k}{c_1 c_4}$$

so that $\frac{n}{\ln n} = O(k)$. By Theorem 3.1 this implies $\frac{n}{\ln n} = \Theta(k)$. By symmetry, $k = \Theta\left(\frac{n}{\ln n}\right)$.

Problem 3-1

- a. If we pick any c > 0, then, the end behavior of $cn^k p(n)$ is going to infinity, in particular, there is an n_0 so that for every $n \ge n_0$, it is positive, so, we can add p(n) to both sides to get $p(n) < cn^k$.
- b. If we pick any c > 0, then, the end behavior of $p(n) cn^k$ is going to infinity, in particular, there is an n_0 so that for every $n \ge n_0$, it is positive, so, we can add cn^k to both sides to get $p(n) > cn^k$.
- c. We have by the previous parts that $p(n) = O(n^k)$ and $p(n) = \Omega(n^k)$. So, by Theorem 3.1, we have that $p(n) = \Theta(n^k)$.

d.

$$\lim_{n\to\infty}\frac{p(n)}{n^k}=\lim_{n\to\infty}\frac{n^d(a_d+o(1))}{n^k}<\lim_{n\to\infty}\frac{2a_dn^d}{n^k}=2a_d\lim_{n\to\infty}n^{d-k}=0$$

e.

$$\lim_{n \to \infty} \frac{n^k}{p(n)} = \lim_{n \to \infty} \frac{n^k}{n^d O(1)} < \lim_{n \to \infty} \frac{n^k}{n^d} = \lim_{n \to \infty} n^{k-d} = 0$$

Problem 3-2

A	B	O	0	Ω	ω	Θ
$\lg^k n$	n^{ϵ}	yes	yes	no	no	no
n^k	c^n	yes	yes	no	no	no
\sqrt{n}	$n^{\sin n}$	no	no	no	no	no
2^n	$2^{n/2}$	no	no	yes	yes	no
$n^{\log c}$	$c^{\log n}$	yes	no	yes	no	yes
$\log(n!)$	$\log(n^n)$	yes	no	yes	no	yes

The terms are in decreasing growth rate by row. Functions in the same row are Θ of each other.

b.

$$f(n) = \begin{cases} g_1(n)! & n \mod 2 = 0\\ 0 & n \mod 2 = 1 \end{cases}$$

- a. False. Counterexample: $n = O(n^2)$ but $n^2 \neq O(n)$.
- b. False. Counterexample: $n + n^2 \neq \Theta(n)$.
- c. True. Since f(n) = O(g(n)) there exist c and n_0 such that $n \ge n_0$ implies $f(n) \le cg(n)$ and $f(n) \ge 1$. This means that $\log(f(n)) \le \log(cg(n)) = \log(c) + \log(g(n))$. Note that the inequality is preserved after taking logs because $f(n) \ge 1$. Now we need to find d such that $f(n) \le d\log(g(n))$. It will suffice to make $\log(c) + \log(g(n)) \le d\log(g(n))$, which is achieved by taking $d = \log(c) + 1$, since $\log(g(n)) \ge 1$.

- d. False. Counterexample: 2n = O(n) but $2^{2n} \neq 2^n$ as shown in exercise 3.1-4.
- e. False. Counterexample: Let $f(n) = \frac{1}{n}$. Suppose that c is such that $\frac{1}{n} \le c \frac{1}{n^2}$ for $n \ge n_0$. Choose k such that $kc \ge n_0$ and k > 1. Then this implies $\frac{1}{kc} \le \frac{c}{k^2c^2} = \frac{1}{k^2c}$, a contradiction.
- f. True. Since f(n) = O(g(n)) there exist c and n_0 such that $n \ge n_0$ implies $f(n) \le cg(n)$. Thus $g(n) \ge \frac{1}{c}f(n)$, so $g(n) = \Omega(f(n))$.
- g. False. Counterexample: Let $f(n) = 2^{2n}$. By exercise 3.1-4, $2^{2n} \neq O(2^n)$.
- h. True. Let g be any function such that g(n) = o(f(n)). Since g is asymptotically positive let n_0 be such that $n \geq n_0$ implies $g(n) \geq 0$. Then $f(n) + g(n) \geq f(n)$ so $f(n) + o(f(n)) = \Omega(f(n))$. Next, choose n_1 such that $n \geq n_1$ implies $g(n) \leq f(n)$. Then $f(n) + g(n) \leq f(n) + f(n) = 2f(n)$ so f(n) + o(f(n)) = O(f(n)). By Theorem 3.1, this implies f(n) + o(f(n)) = O(f(n)).

- a. Suppose that we do not have that f = O(g(n)). This means that $\forall c > 0, n_0, \exists n \geq n_0, f(n) > cg(n)$. Since this holds for every c, we can let it be arbitrary, say 1. Initially, we set $n_0 = 1$, then, the resulting n we will call a_1 . Then, in general, let $n_0 = a_i + 1$ and let a_{i+1} be the resulting value of n. Then, on the infinite set $\{a_1, a_2, \ldots\}$, we have f(n) > g(n), and so, $f = \overset{\infty}{\Omega}(g(n))$
 - This is not the case for the usual definition of Ω . Suppose we had $f(n) = n^2(n \mod 2)$ and g(n) = n. On all the even values, g(n) is larger, but on all the odd values, f(n) grows more quickly.
- b. The advantage is that you get the result of part a which is a nice property. A disadvantage is that the infinite set of points on which you are making claims of the behavior could be very sparse. Also, there is nothing said about the behavior when outside of this infinite set, it can do whatever it wants.
- c. A function f can only be in $\Theta(g(n))$ if f(n) has an infinite tail that is non-negative. In this case, the definition of O(g(n)) agrees with O'(g(n)). Similarly, for a function to be in $\Omega(g(n))$, we need that f(n) is non-negative for some infinite tail, on which O(g(n)) is identical to O'(g(n)). So, we have that in both directions, changing O to O' does not change anything.

d. Suppose $f(n) \in \overset{\sim}{\Theta}(g(n))$, then $\exists c_1, c_2, k_1, k_2, n_0, \forall n \geq n_0, 0 \leq \frac{c_1 g(n)}{\lg^{k_1}(n)} \leq f(n) \leq c_2 g(n) \lg^{k_2}(n)$, if we just look at these inequalities separately, we have $\frac{c_1 g(n)}{\lg^{k_1}(n)} \leq f(n) \ (f(n) \in \overset{\sim}{\Omega}(g(n)))$ and $f(n) \leq c_2 g(n) \lg^{k_2}(n) \ (f(n) \in \tilde{O}(g(n)))$. Now for the other direction. Suppose that we had $\exists n_1, c_1, k_1 \forall n \geq n_1, \frac{c_1 g(n)}{\lg^{k_1}(n)} \leq f(n)$ and $\exists n_2, c_2, k_2, \forall n \geq n_2, f(n) \leq c_2 g(n) \lg^{k_2}(n)$. Putting these together, and letting $n_0 = \max(n_1, n_2)$, we have $\forall n \geq n_0, \frac{c_1 g(n)}{\lg^{k_1}(n)} \leq f(n) \leq c_2 g(n) \lg^{k_2}(n)$.

f(n)	c	$f_c^*(n)$
n-1	0	$\lceil n \rceil$
$\log n$	1	$\log^* n$
n/2	1	$\lceil \log(n) \rceil$
n/2	2	$\lceil \log(n) \rceil - 1$
\sqrt{n}	2	$\log \log n$
\sqrt{n}	1	undefined
$n^{1/3}$	2	$\log_3 \log_2(n)$
$n/\log n$	2	$\Omega\left(\frac{\log n}{\log(\log n)}\right)$