

# THEORY OF ALGORITHMS SOLUTIONS TO THE PROBLEMS

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3.1-1 Let  $f(n)$  and  $g(n)$  be asymptotically nonnegative functions. Using the basic definition of  $\Theta$ -notation, prove that  $\max(f(n), g(n)) = \Theta(f(n) + g(n))$ .

*Proof.* We only need to show that there exist constants  $c_1, c_2$  such that for sufficiently large  $n$ , the relation

$$c_1(f(n) + g(n)) \leq \max(f(n), g(n)) \leq c_2(f(n) + g(n))$$

holds. It's easy to show that it's always hold for  $c_1 = 1/2$  and  $c_2 = 1$ .  $\square$

3.1-2 Show that for any real constants  $a$  and  $b$ , where  $b > 0$ ,

$$(n + a)^b = \Theta(n^b)$$

*Proof.* We only need to show that there exist constants  $c_1, c_2$  such that for sufficiently large  $n$ , the relation

$$c_1 n^b \leq (n + a)^b \leq c_2 n^b$$

holds, which is

$$\sqrt[b]{c_1} n - a \leq n \leq \sqrt[b]{c_2} n - a$$

It's easy to show that it holds for sufficiently large  $n$  and  $c_1 = (1/2)^b, c_2 = 2^b$ .  $\square$

3.1-4 Is  $2^{n+1} = O(2^n)$ ? Is  $2^{2n} = O(2^n)$ ?

*Solution.* Yes. No.  $\square$

3.1-6 Prove that the running time of an algorithm is  $\Theta(g(n))$  if and only if its worst-case running time is  $O(g(n))$  and its best-case running time is  $\Omega(g(n))$ .

*Proof.* Straightforward from the definitions.  $\square$

3.2-2 Prove the equation

$$a^{\log_b c} = c^{\log_b a}$$

*Proof.* It's sufficient to prove  $\log_b c \cdot \log_b a = \log_b a \cdot \log_b c$   $\square$

## 3.1 Asymptotic behavior of polynomials

Let

$$p(n) = \sum_{i=0}^d a_i n^i$$

where  $a_d > 0$ , be a degree- $d$  polynomial in  $n$ , and let  $k$  be a constant. Use the definition of asymptotic notations to prove the following properties.

- a. If  $k \geq d$ , then  $p(n) = O(n^k)$
- b. If  $k \leq d$ , then  $p(n) = \Omega(n^k)$

- c. If  $k = d$ , then  $p(n) = \Theta(n^k)$
- d. If  $k > d$ , then  $p(n) = o(n^k)$
- e. If  $k < d$ , then  $p(n) = \omega(n^k)$

*Proof.* First, we prove a more general lemma

**Lemma 0.1.** Suppose both  $P(n) = p_n x^n + \dots + p_0$  and  $Q(n) = q_n x^n + \dots + q_0$  are polynomials in  $n$  (where  $p_n > 0$  and  $q_n > 0$ ), let

$$l = \lim_{n \rightarrow \infty} \frac{P(n)}{Q(n)}$$

it is easy to show that for sufficiently large  $n$  and some constant  $c, c_1 > 0$ , we have

- (1) If  $l = 0$ , then  $0 < P(n) < cQ(n)$ .
- (2) If  $l = \infty$ , then  $0 < cQ(n) < P(n)$ .
- (3) If  $l = b$  for some  $b \in \mathbb{R}^+$ , then  $c_1 Q(n) < P(n) < cQ(n)$ .

*Proof.* Straightforward. □

It is easy to prove the proposition using the above lemma. □

### 3.2 Relative asymptotic growths

Indicate, for each pair of expressions  $(A, B)$  in the table below, whether  $A$  is  $O, o, \Omega, \omega$ , or  $\Theta$  of  $B$ . Assume that  $k \geq 1, \epsilon > 0$ , and  $c > 1$  are constants. Your answer should be in the form of the table with "yes" or "no" written in each box.

*Solution.*

	$A$	$B$	$O$	$o$	$\Omega$	$\omega$	$\Theta$
a	$\lg^k n$	$n^\epsilon$	no	no	yes	yes	no
b	$n^k$	$c^n$	no	no	yes	yes	no
c	$\sqrt{n}$	$n^{\sin n}$	no	no	no	no	no
d	$2^n$	$2^{n/2}$	yes	yes	no	no	no
e	$n^{\lg c}$	$c^{\lg n}$	yes	no	yes	no	yes
f	$\lg(n!)$	$\lg(n^n)$	yes	no	yes	no	yes

□