# COMP3711: Design and Analysis of Algorithms

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Asymptotic notations		
Asymptotic upper bound		
Definition (big-Ob)		

f(n) = O(g(n)): There exists constant c > 0 and  $n_0$  such that  $f(n) \le c \cdot g(n)$  for  $n \ge n_0$ 

#### Asymptotic lower bound

Definition (big-Omega)

 $f(n) = \Omega(g(n))$ : There exists constant c > 0 and  $n_0$  such that  $f(n) \ge c \cdot g(n)$  for  $n \ge n_0$ .

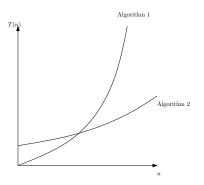
### Asymptotic tight bound

### Definition (big-Theta)

$$f(n) = \Theta(g(n))$$
:  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ 

## Comparing time complexity

Example:



Algorithm 2 is clearly superior

- T(n) for Algorithm 1 is  $O(n^3)$
- T(n) for Algorithm 2 is  $O(n^2)$
- Since  $n^3$  grows much more rapidly, we expect Algorithm 1 to take much more time than Algorithm 2 when *n* increases

For all real  $a \neq 0$ , m and n, we have the following identities:

$$a^{0} = 1$$

$$a^{1} = a$$

$$a^{-1} = 1/a$$

$$(a^{m})^{n} = (a^{n})^{m} = a^{mn}$$

$$a^{m}a^{n} = a^{m+n}$$

$$a^{1/n} = \sqrt[n]{a}$$

## Some Basic mathematic background on logarithms

For all real a > 0, b > 0, c > 0, and n:

$$a = b^{\log_b a}$$
$$\log_c(ab) = \log_c a + \log_c b$$
$$\log_b a^n = n \log_b a$$
$$\log_b a = \frac{\log_c a}{\log_c b}$$
$$\log_b(1/a) = -\log_b a$$
$$\log_b a = \frac{1}{\log_a b}$$
$$a^{\log_b n} = n^{\log_b a}$$

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For each of the following statement, answer whether the statement is true or false.

(a) 
$$1000n + n \log n = O(n \log n)$$
.  
(b)  $n^2 + n \log(n^3) = O(n \log(n^3))$ .  
(c)  $n^3 = \Omega(n)$ .  
(d)  $n^2 + n = \Omega(n^3)$ .  
(e)  $n^3 = O(n^{10})$ .  
(f)  $n^3 + 1000n^{2.9} = \Theta(n^3)$   
(g)  $n^3 - n^2 = \Theta(n)$ 

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Solution 1		

- (a) True.
- (b) False.
- (c) True.
- (d) False.
- (e) True.
- (f) True.
- (g) False.

### Question 2

For each pair of expressions (A, B) below, indicate whether A is O,  $\Omega$ , or  $\Theta$  of B. Note that zero, one, or more of these relations may hold for a given pair; list all correct ones. Justify your answers.

(a) 
$$A = n^3 + n \log n$$
;  $B = n^3 + n^2 \log n$   
(b)  $A = \log \sqrt{n}$ ;  $B = \sqrt{\log n}$ .  
(c)  $A = n \log_3 n$ ;  $B = n \log_4 n$ .  
(d)  $A = 2^n$ ;  $B = 2^{n/2}$ .  
(e)  $A = \log(2^n)$ ;  $B = \log(3^n)$ .

	A	Relation:	В
(a)	$n^3 + n \log n$	$\Omega, \Theta, O$	$n^3 + n^2 \log n$
(b)	$\log \sqrt{n}$	Ω	$\sqrt{\log n}$
(c)	n log <sub>3</sub> n	$\Omega, \Theta, O$	n log <sub>4</sub> n
(d)	2 <sup>n</sup>	Ω	2 <sup>n/2</sup>
(e)	$\log(2^n)$	$\Omega, \Theta, O$	$\log(3^n)$

## Solution 2: Step by step

Notes:

- (a) Both are  $\Theta(n^3)$ , the lower order terms can be ignored. Note that if  $A(n) = \Theta(B(n))$ , then automatically A(n) = O(B(n)) and  $A(n) = \Omega(B(n))$ .
- (b) After simplifying, A is  $(1/2) \log n$ , and B is  $\sqrt{\log n}$ . Substituting  $m = \log n$ , we can see ratio A/B grows as  $m/2\sqrt{m} = \sqrt{m}/2$  which tends to infinity as n (and hence m) tends to infinity, i.e.,  $A(n) = \Omega(B(n))$ .
- (c) Log base conversion only introduces a constant factor.
- (d) The ratio is  $2^n/2^{n/2} = (2)^{n/2}$  which goes to infinity in the limit.
- (e) After simplifying these are  $n \log 2$  and  $n \log 3$ , both of which are  $\Theta(n)$ .

Question 3

Suppose  $T_1(n) = O(f(n))$  and  $T_2(n) = O(f(n))$ . Which of the following are true? Justify your answers.

(a) 
$$T_1(n) + T_2(n) = O(f(n))$$
  
(b)  $\frac{T_1(n)}{T_2(n)} = O(1)$   
(c)  $T_1(n) = O(T_2(n))$ 

## Solution 3

(a) True. Since  $T_1(n) = O(f(n))$  and  $T_2(n) = O(f(n))$ , it follows from the definition that there exist constants  $c_1, c_2 > 0$  and positive integers  $n_1, n_2$  such that  $T_1(n) \le c_1 f(n)$  for  $n \ge n_1$ and  $T_2(n) \le c_2 f(n)$  for  $n \ge n_2$ . This implies that,  $T_1(n) + T_2(n) \le (c_1 + c_2)f(n)$  for  $n \ge \max(n_1, n_2)$ . Thus,  $T_1(n) + T_2(n) = O(f(n))$ .

(b) False. For a counterexample to the claim, let 
$$T_1(n) = n^2$$
,  $T_2(n) = n$ ,  $f(n) = n^2$ . Then  $T_1(n) = O(f(n))$  and  $T_2(n) = O(f(n))$  but  $\frac{T_1(n)}{T_2(n)} = n \neq O(1)$ .

(c) False. We can use the same counterexample as in part (b). Note that  $T_1(n) \neq O(T_2(n))$ 

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

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For any value of n,  $\max(f(n), g(n))$  is either equal to f(n) or equal to g(n). Therefore, for all n,

 $\max(f(n),g(n)) \leq f(n) + g(n).$ 

Using c = 1 and  $n_0 = 1$  in the big-oh definition, it follows that  $\max(f(n), g(n)) = O(f(n) + g(n)).$ 

Also, for all n,

$$\max(f(n),g(n)) \ge f(n)$$

and

$$\max(f(n),g(n)) \ge g(n).$$

Adding we have

$$2 \times \max(f(n),g(n)) \ge f(n) + g(n).$$

Therefore,

$$\max(f(n),g(n)) \geq \frac{1}{2}(f(n)+g(n)).$$

Using c = 1/2 and  $n_0 = 1$  in the Omega definition, it follows that

$$\max(f(n),g(n)) = \Omega(f(n) + g(n)).$$

Since 
$$\max(f(n), g(n)) = O(f(n) + g(n))$$
 and  
 $\max(f(n), g(n)) = \Omega(f(n) + g(n))$ , it implies that  
 $\max(f(n), g(n)) = \Theta(f(n) + g(n))$ .