

# Design and Analysis of Algorithms

## Tutorial 1



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# Asymptotic notations

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## Asymptotic upper bound

### Definition (big-Oh)

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

## Asymptotic lower bound

### Definition (big-Omega)

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

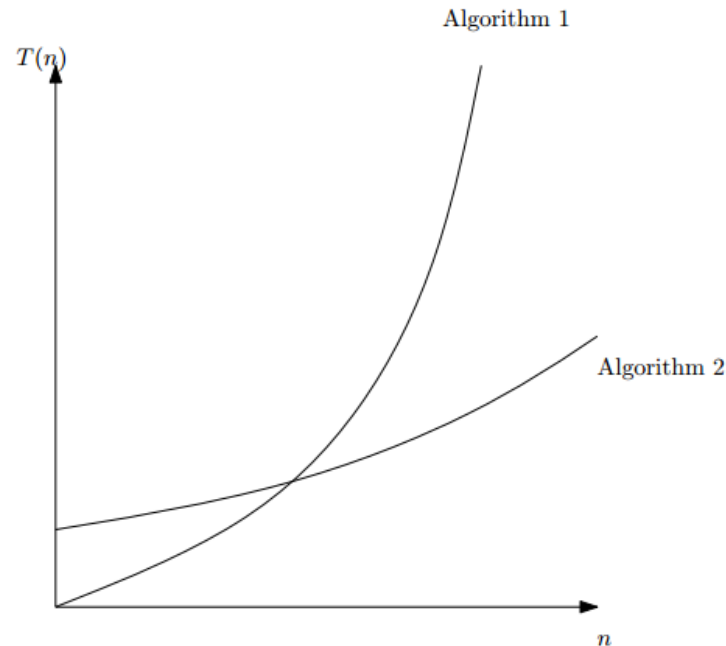
## Asymptotic tight bound

### 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

# Some Basic mathematic background on exponentials

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

$$a^0 = 1$$

$$a^1 = a$$

$$a^{-1} = \frac{1}{a}$$

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

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

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

# Some Basic mathematic background on logarithms

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

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

$$\log_c(ab) = \log_c a + \log_c b$$

$$\log_b a^n = n \log_b a$$

$$\log_b a = \frac{\log_c a}{\log_c b}$$

$$\log_b\left(\frac{1}{a}\right) = -\log_b a$$

$$\log_b a = \frac{1}{\log_a b}$$

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

# Question 1

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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)  $n^3 - n^2 = \Theta(n)$

# Solution 1

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

## Question 2

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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)  $A = 2^n; B = 2^{\frac{n}{2}}$

(e)  $A = \log(2^n); B = \log(3^n)$



## Solution 2

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	A	Relation:	B
(a)	$n^3 + n \log n$	$\Omega, \Theta, O$	$n^3 + n^2 \log n$
(b)	$\log \sqrt{n}$	$\Omega$	$\sqrt{\log n}$
(c)	$n \log_3 n$	$\Omega, \Theta, O$	$n \log_4 n$
(d)	$2^n$	$\Omega$	$2^{\frac{n}{2}}$
(e)	$\log(2^n)$	$\Omega, \Theta, O$	$\log(3^n)$

## Solution 2: Step by step

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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  
 $\frac{m}{2\sqrt{m}} = \frac{\sqrt{m}}{2}$  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  $\frac{2^n}{2^{\frac{n}{2}}} = (2)^{\frac{n}{2}}$  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

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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

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(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  $T_1(n) \leq c_1 f(n)$  for  $n \geq n_1$  and  $T_2(n) \leq c_2 f(n)$  for  $n \geq n_2$ . This implies that,  $T_1(n) + T_2(n) \leq (c_1 + c_2)f(n)$  for  $n \geq \max(n_1, n_2)$ . 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  $T_1(n) \neq O(T_2(n))$

## Question 4

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

## Solution 4

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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  $n$ ,

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

And

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

Adding we have

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

Therefore,

$$\max(f(n), g(n)) \geq 1/2(f(n) + g(n))$$

## Solution 4

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