

Minggu-6 – Notasi Asymptot



Algoritma & Pemrograman Saintifik

Gatot F. Hertono, Ph.D

Departemen Matematika

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Asymptotic Notation: Order of Growth

Background

Suppose, in worst case, a problem can be solved by using two different algorithms, with time complexity:

Algorithm A:

$$f(n) = 400n + 23$$

Algorithm B:

$$g(n) = 2n^2 - 1$$

Which one is better?

n	$3n^2$	$14n+17$
1	3	31
10	300	157
100	30,000	1,417
1000	3,000,000	14,017
10000	300,000,000	140,017

$$3n^2 > 14n+17$$

$\forall$ "large enough" n

Solution: Ignore the constants
& low-order terms.

→ we use Asymptotic Notation (Ω , Θ dan O)

Order of Growth

n	$\log_2 n$	n	$n \log_2 n$	n^2	n^3	2^n	$n!$
10	3.3	10^1	$3.3 \cdot 10^1$	10^2	10^3	10^3	$3.6 \cdot 10^6$
10^2	6.6	10^2	$6.6 \cdot 10^2$	10^4	10^6	$1.3 \cdot 10^{30}$	$9.3 \cdot 10^{157}$
10^3	10	10^3	$1.0 \cdot 10^4$	10^6	10^9		
10^4	13	10^4	$1.3 \cdot 10^5$	10^8	10^{12}		
10^5	17	10^5	$1.7 \cdot 10^6$	10^{10}	10^{15}		
10^6	20	10^6	$2.0 \cdot 10^7$	10^{12}	10^{18}		

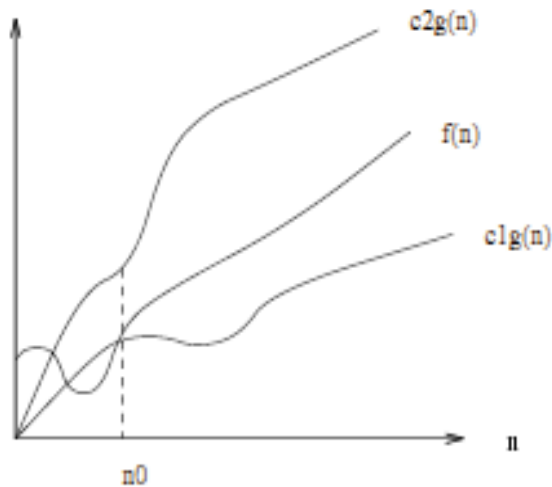
Values (some approximate) of several functions important for analysis of algorithms

Ω , Θ and O notations

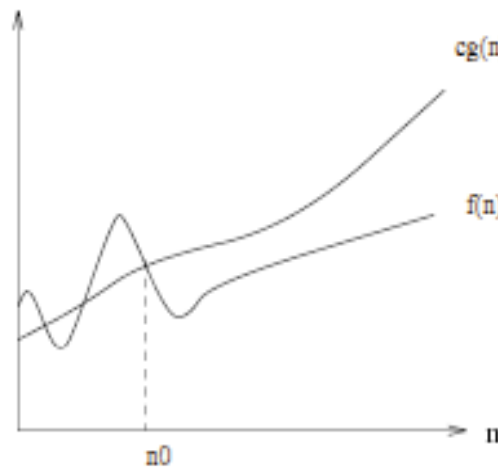
What does it mean?



O , Ω , and Θ

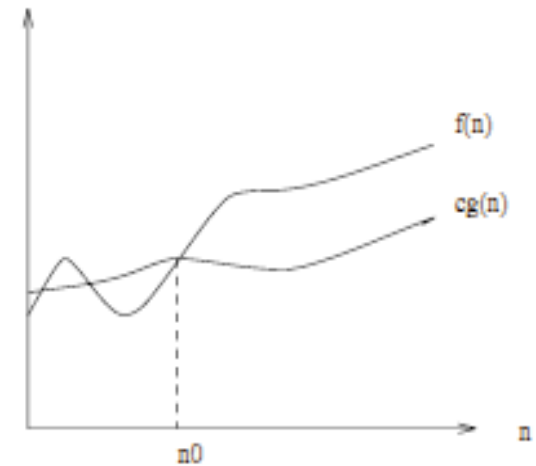


(a)



$f(n) = O(g(n))$

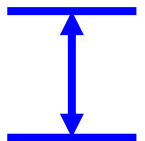
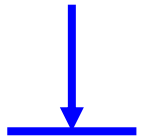
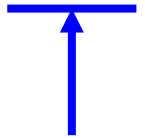
(b)



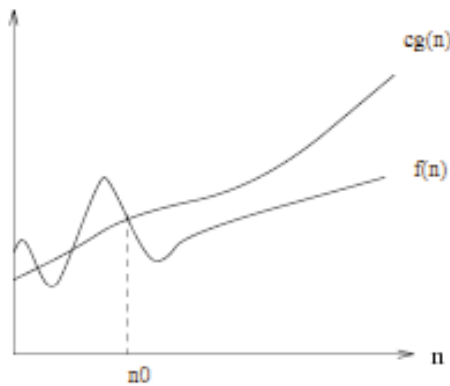
(c)

Names of Bounding Functions

- $f(n) = O(g(n))$ means $C \times g(n)$ is an upper bound on $f(n)$;
- $f(n) = \Omega(g(n))$ means $C \times g(n)$ is a lower bound on $f(n)$;
- $f(n) = \Theta(g(n))$ means $C_1 \times g(n)$ is an upper bound on $f(n)$ and $C_2 \times g(n)$ is a lower bound on $f(n)$.



O (big Oh) Notation



$f(n) = O(g(n))$: $g(n)$ is an *asymptotically upper bound* for $f(n)$. *i.e. f does not grow faster than g .*

Formal definition:

$$O(g(n)) = \{f(n) : \exists c > 0 \text{ and } n_0 > 0 \\ \text{such that } 0 \leq f(n) \leq cg(n) \\ \text{for all } n \geq n_0\}.$$

Remark: “ $f(n) = O(g(n))$ ” means that

$$f(n) \in O(g(n)).$$

Example: $f(n) = n^3 + 20n^2 + 100n$, then $f(n) = O(n^3)$.

Proof:

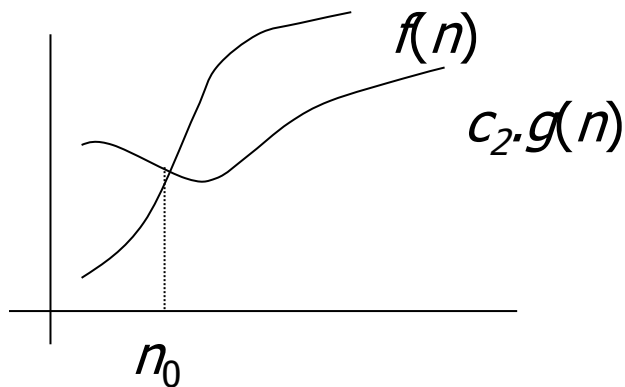
$$\forall n \geq 0, \quad n^3 + 20n^2 + 100n \leq n^3 + 20n^3 + 100n^3 = 121n^3.$$

Choose $c = 121$ and $n_0 = 0$, then it completes the definition.

Examples:

- (1) $n^2 + 2n = O(n^2)$.
- (2) $200n^2 - 100n = O(n^2)$.
- (3) $n \log_2 n = O(n^2)$.
- (4) $n^2 \log_2 n \neq O(n^2)$.
- (5) $\forall a, b > 1, \log_a n = O(\log_b n)$.

Ω (big Omega) Notation



$f(n) = \Omega(g(n))$: $g(n)$ is an asymptotically lower bound for $f(n)$.

Formal definition:

$$\Omega(g(n)) = \{f(n) : \exists c > 0 \text{ and } n_0 > 0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}.$$

Examples:

(1) $200n^2 - 100n = \Omega(n^2) = \Omega(n) = \Omega(1)$.

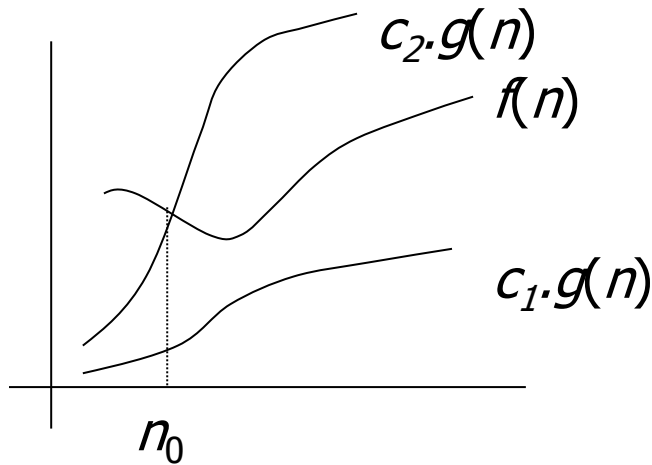
(2) $n^2 = \Omega(n)$.

(3) $n^2 \neq \Omega(n^2 \log n)$.

(3) Does $f(n) = O(g(n))$ imply $g(n) = \Omega(f(n))$?

(4) Does $f(n) = \Omega(g(n))$ imply $g(n) = O(f(n))$?

Θ (Big Theta) Notation



$f(n) = \Theta(g(n))$: $g(n)$ is an *asymptotically tight bound* for $f(n)$.

Formal definition:

$$\Theta(g(n)) = \{f(n) : \exists n_0 > 0, c_1 > 0 \text{ and } c_2 > 0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}.$$

Examples:

(1) $5n^2 - 2n + 5 = \Theta(n^2)$

(2) $5n^2 - 2n + 5 = \Theta(n^2 + \log n)$

Examples of Θ

$$3n^2 + 7n + 8 = \Theta(n^2) ?$$

True

$$\exists c_1, c_2, \exists n_0, \forall n \geq n_0, c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\begin{array}{ccc} \updownarrow & \updownarrow & \updownarrow \\ 3 & 4 & 8 \end{array}$$

$$n \geq 8$$

$$3 \cdot n^2 \leq 3n^2 + 7n + 8 \leq 4 \cdot n^2$$

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

$$\exists c_1, c_2, \exists n_0, \forall n \geq n_0, c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\updownarrow$$

$$0$$



$$0 \cdot n^3 \leq n^2 \leq c_2 \cdot n^3$$

False, since C_1, C_2 must be > 0

Examples

$$3n^2 - 100n + 6 = O(n^2) \text{ because } 3n^2 > 3n^2 - 100n + 6$$

$$3n^2 - 100n + 6 = O(n^3) \text{ because } .01n^3 > 3n^2 - 100n + 6$$

$$3n^2 - 100n + 6 \neq O(n) \text{ because } c \cdot n < 3n^2 \text{ when } n > c$$

$$3n^2 - 100n + 6 = \Omega(n^2) \text{ because } 2.99n^2 < 3n^2 - 100n + 6$$

$$3n^2 - 100n + 6 \neq \Omega(n^3) \text{ because } 3n^2 - 100n + 6 < n^3$$

$$3n^2 - 100n + 6 = \Omega(n) \text{ because } 10^{10}n < 3n^2 - 100 + 6$$

$$3n^2 - 100n + 6 = \Theta(n^2) \text{ because } O \text{ and } \Omega$$

$$3n^2 - 100n + 6 \neq \Theta(n^3) \text{ because } O \text{ only}$$

$$3n^2 - 100n + 6 \neq \Theta(n) \text{ because } \Omega \text{ only}$$

Think of the equality as meaning *in the set of functions*.



Properties

Note that if

$$f(n) = \Theta(g(n)),$$

then

$$f(n) = \Omega(g(n)) \quad \text{and} \quad f(n) = O(g(n)).$$

In the other direction, if

$$f(n) = \Omega(g(n)) \quad \text{and} \quad f(n) = O(g(n)),$$

then

$$f(n) = \Theta(g(n)).$$

Transitivity.

- If $f = O(g)$ and $g = O(h)$ then $f = O(h)$.
- If $f = \Omega(g)$ and $g = \Omega(h)$ then $f = \Omega(h)$.
- If $f = \Theta(g)$ and $g = \Theta(h)$ then $f = \Theta(h)$.

Additivity.

- If $f = O(h)$ and $g = O(h)$ then $f + g = O(h)$.
- If $f = \Omega(h)$ and $g = \Omega(h)$ then $f + g = \Omega(h)$.
- If $f = \Theta(h)$ and $g = O(h)$ then $f + g = \Theta(h)$.

Assignment: Based on the definition of Ω , Θ and O , prove that

$$f(n) = \Theta(g(n)) \Leftrightarrow f(n) = O(g(n)) \text{ and } f(n) = \Omega(g(n))$$

What does asymptotic property imply for an algorithm?

[illegible]

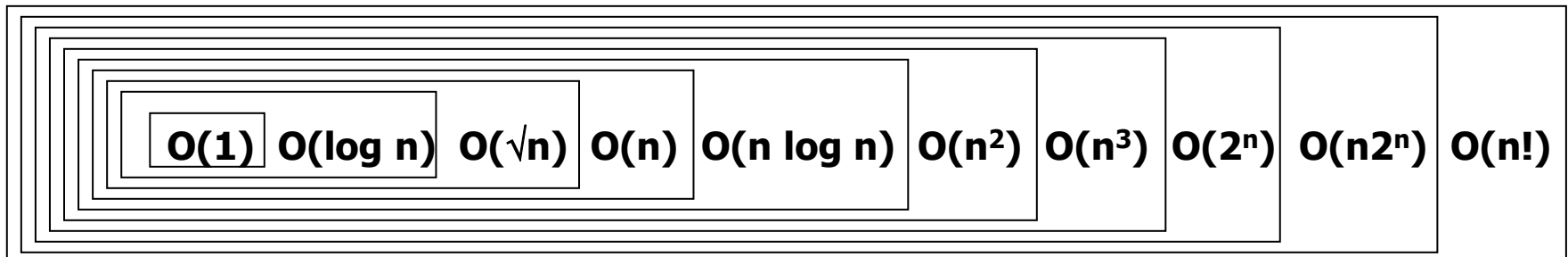
$$t_1(n) \in O(g_1(n))$$

$$t_2(n) \in O(g_2(n))$$

How is the overall efficiency of this algorithm?

Basic asymptotic efficiency classes

1	constant
$\log n$	logarithmic
n	linear
$n \log n$	n -log- n
n^2	quadratic
n^3	cubic
2^n	exponential
$n!$	factorial



Time efficiency of nonrecursive algorithms

General Plan for Analysis

- Decide on parameter n indicating input size
- Identify algorithm's basic operation
- Determine worst, average, and best cases for input of size n
- Set up a sum for the number of times the basic operation is executed
- Simplify the sum using standard formulas and rules

Example 1: Maximum element

ALGORITHM *MaxElement*($A[0..n - 1]$)

//Determines the value of the largest element in a given array

//Input: An array $A[0..n - 1]$ of real numbers

//Output: The value of the largest element in A

$maxval \leftarrow A[0]$

for $i \leftarrow 1$ **to** $n - 1$ **do**

if $A[i] > maxval$

$maxval \leftarrow A[i]$

return $maxval$

$$T(n) = O(?)$$

Example 2: Element uniqueness problem

ALGORITHM *UniqueElements*($A[0..n - 1]$)

//Determines whether all the elements in a given array are distinct

//Input: An array $A[0..n - 1]$

//Output: Returns “true” if all the elements in A are distinct

// and “false” otherwise

for $i \leftarrow 0$ **to** $n - 2$ **do**

for $j \leftarrow i + 1$ **to** $n - 1$ **do**

if $A[i] = A[j]$ **return false**

return true

$$T(n) = O(?)$$

Example 3: Matrix multiplication

```
ALGORITHM MatrixMultiplication( $A[0..n-1, 0..n-1]$ ,  $B[0..n-1, 0..n-1]$ )  
  //Multiplies two  $n$ -by- $n$  matrices by the definition-based algorithm  
  //Input: Two  $n$ -by- $n$  matrices  $A$  and  $B$   
  //Output: Matrix  $C = AB$   
  for  $i \leftarrow 0$  to  $n - 1$  do  
    for  $j \leftarrow 0$  to  $n - 1$  do  
       $C[i, j] \leftarrow 0.0$   
      for  $k \leftarrow 0$  to  $n - 1$  do  
         $C[i, j] \leftarrow C[i, j] + A[i, k] * B[k, j]$   
  return  $C$ 
```

$$T(n) = O(?)$$

Example 5: Counting binary digits

ALGORITHM *Binary*(n)

//Input: A positive decimal integer n

//Output: The number of binary digits in n 's binary representation

$count \leftarrow 1$

while $n > 1$ **do**

$count \leftarrow count + 1$

$n \leftarrow \lfloor n/2 \rfloor$

return $count$

It cannot be investigated the way the previous examples are.