

# Chapter 3: Growth of Functions

# About this lecture

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- Introduce Asymptotic Notation
  - $\Theta()$ ,  $O()$ ,  $\Omega()$ ,  $o()$ ,  $\omega()$

# Dominating Term

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Recall that for input size  $n$ ,

- Insertion Sort 's running time is:

$$An^2 + Bn + C, \quad (A, B, C \text{ are constants})$$

- Merge Sort 's running time is:

$$Dn \log n + En + F, \quad (D, E, F \text{ are constants})$$

- To compare their running times for large  $n$ , we can just focus on the dominating term (the term that grows fastest when  $n$  increases)
  - $An^2$  vs  $Dn \log n$

# Dominating Term

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- If we look more closely, the **leading constants** in the dominating term does not affect much in this comparison **(if  $n$  is sufficiently large)**
  - We may as well compare  $n^2$  vs  $n \log n$   
(instead of  $An^2$  vs  $Dn \log n$  )
- As a result, we conclude that Merge Sort is better than Insertion Sort when  $n$  is sufficiently large

# Asymptotic Efficiency

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- The previous comparison studies the **asymptotic** efficiency of two algorithms
- If algorithm P is **asymptotically** faster than algorithm Q, P is often a better choice
- To aid (and simplify) our study in the asymptotic efficiency, we now introduce some useful **asymptotic notations**

# Big-O notation

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✓ Definition: Given a function  $g(n)$ , we denote  $O(g(n))$  to be the set of functions:

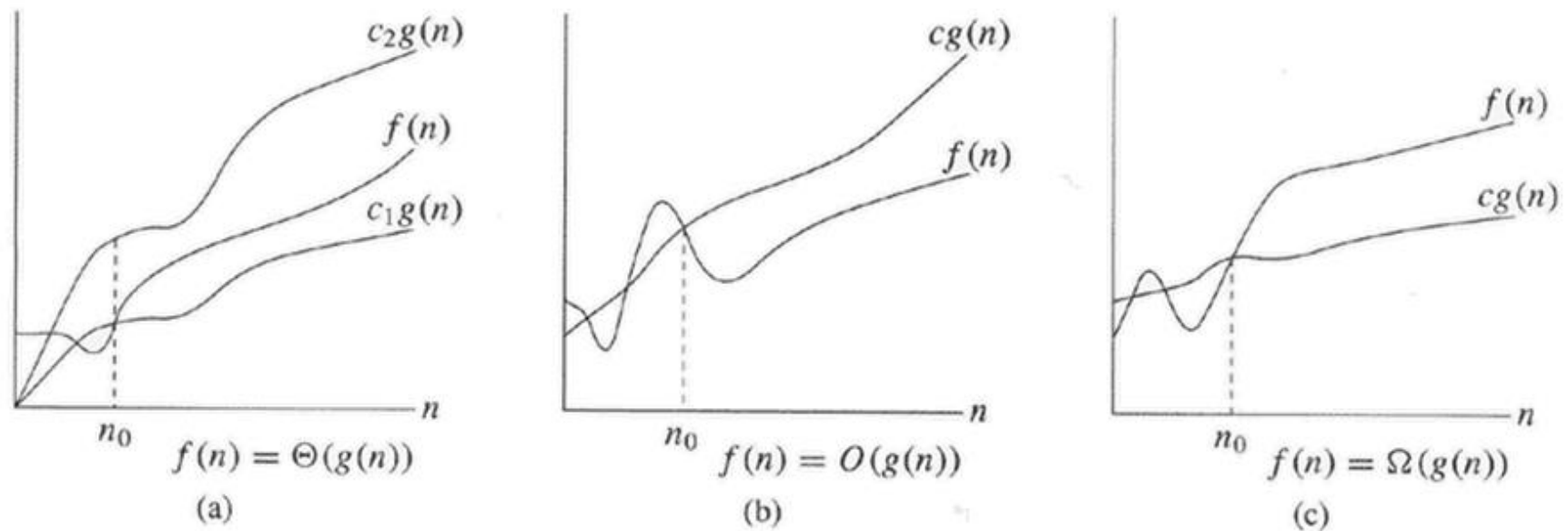
$\{f(n) \mid \text{there exists positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq c g(n) \text{ for all } n \geq n_0 \}$

➤ Rough Meaning:  $O(g(n))$  includes all functions that are upper bounded by  $g(n)$

# Big-O notation (example)

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- $4n \in O(5n)$  [ proof:  $c = 1, n \geq 1$  ]
- $4n \in O(n)$  [ proof:  $c = 4, n \geq 1$  ]
- $4n + 3 \in O(n)$  [ proof:  $c = 5, n \geq 3$  ]
- $n \in O(0.001n^2)$  [ proof:  $c = 10, n \geq 100$  ]
- $\log_e n \in O(\lg n)$  [ proof:  $c = 1, n \geq 1$  ]
- $\lg n \in O(\log_e n)$  [ proof:  $c = \lg e, n \geq 1$  ]
- **Remark:** Usually, we will slightly abuse the notation, and write  $f(n) = O(g(n))$  to mean  $f(n) \in O(g(n))$



**Figure 3.1** Graphic examples of the  $\Theta$ ,  $O$ , and  $\Omega$  notations. In each part, the value of  $n_0$  shown is the minimum possible value; any greater value would also work. (a)  $\Theta$ -notation bounds a function to within constant factors. We write  $f(n) = \Theta(g(n))$  if there exist positive constants  $n_0$ ,  $c_1$ , and  $c_2$  such that to the right of  $n_0$ , the value of  $f(n)$  always lies between  $c_1g(n)$  and  $c_2g(n)$  inclusive. (b)  $O$ -notation gives an upper bound for a function to within a constant factor. We write  $f(n) = O(g(n))$  if there are positive constants  $n_0$  and  $c$  such that to the right of  $n_0$ , the value of  $f(n)$  always lies on or below  $cg(n)$ . (c)  $\Omega$ -notation gives a lower bound for a function to within a constant factor. We write  $f(n) = \Omega(g(n))$  if there are positive constants  $n_0$  and  $c$  such that to the right of  $n_0$ , the value of  $f(n)$  always lies on or above  $cg(n)$ .



# Big-Omega notation

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- ✓ Definition: Given a function  $g(n)$ , we denote  $\Omega(g(n))$  to be the set of functions  $\{f(n) \mid \text{there exists positive constants } c \text{ and } n_0 \text{ such that } 0 \leq c g(n) \leq f(n) \text{ for all } n \geq n_0 \}$
- Rough Meaning:  $\Omega(g(n))$  includes all functions that are lower bounded by  $g(n)$

# Big- $\Omega$ notation (example)

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- $5n \in \Omega(4n)$  [ proof:  $c = 1, n \geq 1$  ]
- $n \in \Omega(4n)$  [ proof:  $c = 1/4, n \geq 1$  ]
- $4n + 3 \in \Omega(n)$  [ proof:  $c = 1, n \geq 1$  ]
- $0.001n^2 \in \Omega(n)$  [ proof:  $c = 1/10, n \geq 100$  ]
- $\log_e n \in \Omega(\lg n)$  [ proof:  $c = 1/\lg e, n \geq 1$  ]
- $\lg n \in \Omega(\log_e n)$  [ proof:  $c = 1, n \geq 1$  ]

# Big-O and Big-Omega

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- ✓ Similar to Big-O, we will slightly abuse the notation, and write  $f(n) = \Omega(g(n))$  to mean  $f(n) \in \Omega(g(n))$
- ✓ Relationship between Big-O and Big- $\Omega$  :  
$$f(n) = \Omega(g(n)) \Leftrightarrow g(n) = O(f(n))$$

# $\Theta$ notation (Big-O \ Big- $\Omega$ )

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✓ Definition: Given a function  $g(n)$ , we denote  $\Theta(g(n))$  to be the set of functions

$\{f(n) \mid \text{there exists positive constants } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}$

➤ Meaning: Those functions which can be both upper bounded and lower bounded by  $g(n)$

# Big-O, Big- $\Omega$ , and $\Theta$

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✓ Similarly, we write  $f(n) = \Theta(g(n))$  to mean  $f(n) \in \Theta(g(n))$

✓ Relationship between Big-O, Big- $\Omega$ , and  $\Theta$ :

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



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

# $\Theta$ notation (example)

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- $4n = \Theta(n)$  [  $c_1 = 1, c_2 = 4, n \geq 1$  ]
- $4n + 3 = \Theta(n)$  [  $c_1 = 1, c_2 = 5, n \geq 3$  ]
- $\log_e n = \Theta(\lg n)$  [  $c_1 = 1/\lg e, c_2 = 1, n \geq 1$  ]
- Running Time of Insertion Sort =  $\Theta(n^2)$ 
  - If not specified, running time refers to the **worst-case** running time
- Running Time of Merge Sort =  $\Theta(n \lg n)$

# To remember the notation

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$O$  is like  $\leq$  :  $f(n) = O(g(n))$  means  $f(n) \leq cg(n)$

$\Omega$  is like  $\geq$  :  $f(n) = \Omega(g(n))$  means  $f(n) \geq cg(n)$

$\Theta$  is like  $=$  :  $f(n) = \Theta(g(n)) \Leftrightarrow g(n) = \Theta(f(n))$

$o$  is like  $<$  :  $f(n) = o(g(n))$  means  $f(n) < cg(n)$

$\omega$  is like  $>$  :  $f(n) = \omega(g(n))$  means  $f(n) > cg(n)$

✓ Note: Not any two functions can be compared asymptotically (E.g.,  $\sin x$  vs.  $\cos x$ )

# Little-o notation

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✓ Definition: Given a function  $g(n)$ , we denote  $o(g(n))$  to be the set of functions

$\{f(n) \mid \text{for any positive } c, \text{ there exists positive constant } n_0 \text{ such that } 0 \leq f(n) < c g(n), \text{ for all } n \geq n_0 \}$

Note the similarities and differences with Big O



# Little-o (equivalent definition)

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✓ Definition: Given a function  $g(n)$ ,  $o(g(n))$  is the set of functions

$$\{f(n) \mid \lim_{n \rightarrow \infty} (f(n)/g(n)) = 0\}$$

Examples:

- $4n = o(n^2)$
- $n \lg n = o(n^{1.0000001})$
- $n \lg n = o(n \lg^2 n)$

# Little-omega notation

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✓ Definition: Given a function  $g(n)$ , we denote  $\omega(g(n))$  to be the set of functions

$\{f(n) \mid \text{for any positive } c, \text{ there exists positive constant } n_0 \text{ such that } 0 \leq c g(n) < f(n), \text{ for all } n \geq n_0\}$

Note the similarities and differences with the Big-Omega definition

# Little-omega (equivalent definition)

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✓ Definition: Given a function  $g(n)$ ,  $\omega(g(n))$  is the set of functions

$$\{f(n) \mid \lim_{n \rightarrow \infty} (g(n)/f(n)) = 0\}$$

✓ Relationship between Little-o and Little- $\omega$  :

$$f(n) = \omega(g(n)) \Leftrightarrow g(n) = o(f(n))$$

# Practice at home

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- Exercises: 3.1-1, 3.1-3, 3.1-4
- Problem: 3-1 (a, d), 3-2 (b, f), 3-4(c, f, g)

# Practice at home

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1. Given that

$$f(n) = a_m n^m + a_{m-1} n^{m-1} + \dots + a_1 n + a_0, \text{ where } a_m > 0.$$

Show that  $f(n) = \Theta(n^m)$ .

2. Write a program to solve the Hanoi Tower Problem

3. What is wrong with the following argument?

“Since  $n = O(n)$ , and  $2n = O(n)$ , ..., we have

$$\sum_{k=1}^n k \cdot n = \sum_{k=1}^n O(n) = O(n^2).”$$