

# Announcements

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- Please sit in the front half
- Reading Assignments:  
Chapter 4 (Textbook: CLRS)
- Please turn in the survey form

# Asymptotic Notation



- $\Theta, O, \Omega, o, \omega$
- Used to describe the running times of algorithms
- Instead of exact running time, say  $\Theta(n^2)$
- Defined for functions whose domain is the set of natural numbers
- Determine **sets** of functions, in practice used to compare two functions

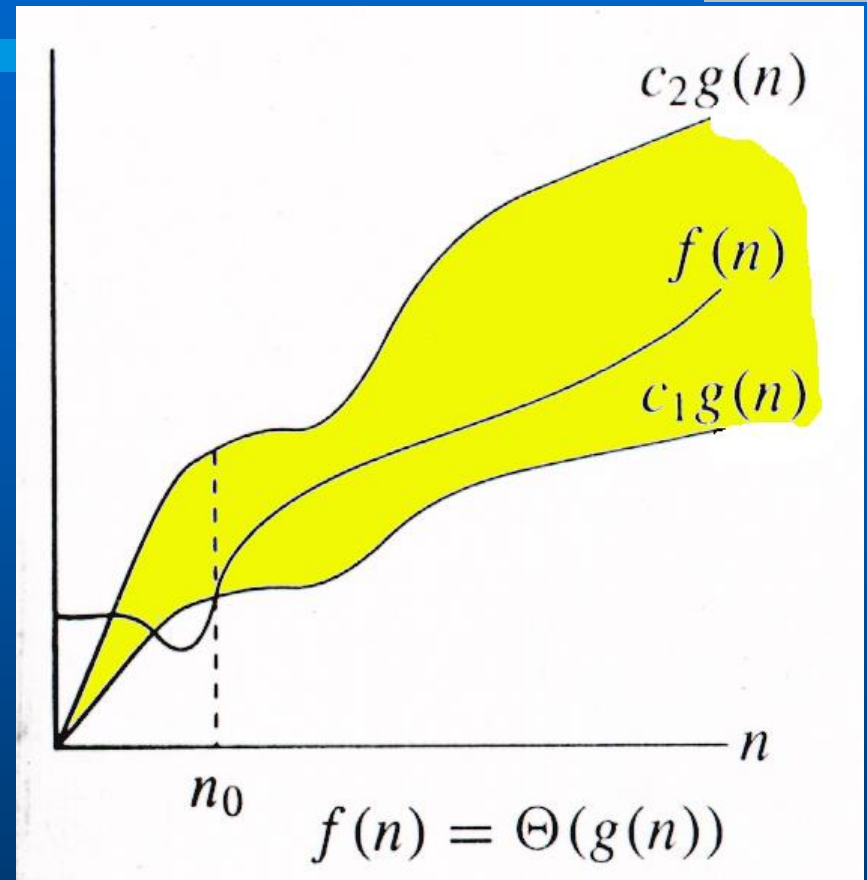
# $\Theta$ -notation



For a given function  $g(n)$ , we denote by  $\Theta(g(n))$  the set of functions

$\Theta(g(n)) = \{f(n): \text{there exist positive constants } c_1, c_2 \text{ and } n_0 \text{ such that}$

$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n),$   
for all  $n \geq n_0\}$



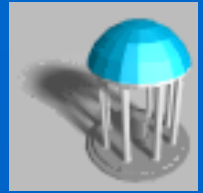
We say  $g(n)$  is an *asymptotically tight bound* for  $f(n)$

# Example



- $10n^2 - 3n = \Theta(n^2)$
- What constants for  $n_0$ ,  $c_1$ , and  $c_2$  will work?
- Make  $c_1$  a little smaller than leading coefficient, and  $c_2$  a little bigger
- *To compare orders of growth, look at leading term*

# *O*-notation

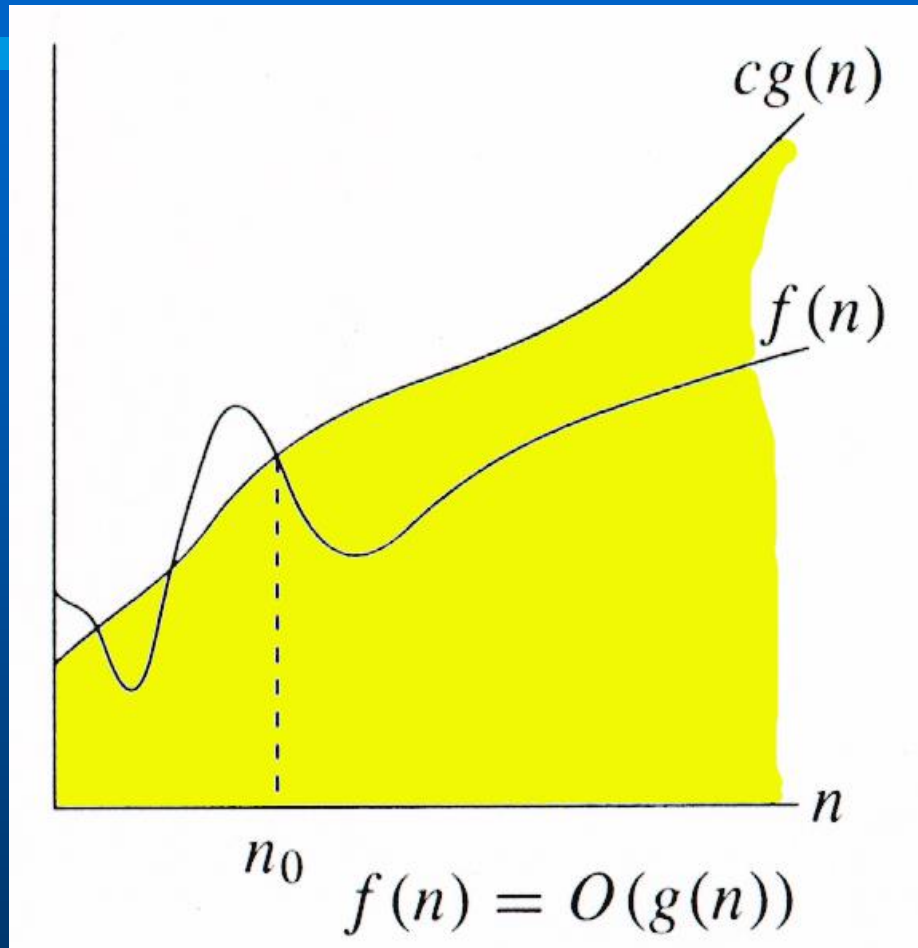


For a given function  $g(n)$ ,  
we denote by  $O(g(n))$   
the set of functions

$O(g(n)) = \{f(n): \text{there exist}$   
**positive constants  $c$  and  $n_0$**   
**such that**

$$0 \leq f(n) \leq cg(n)$$

**for all  $n \geq n_0$  }**



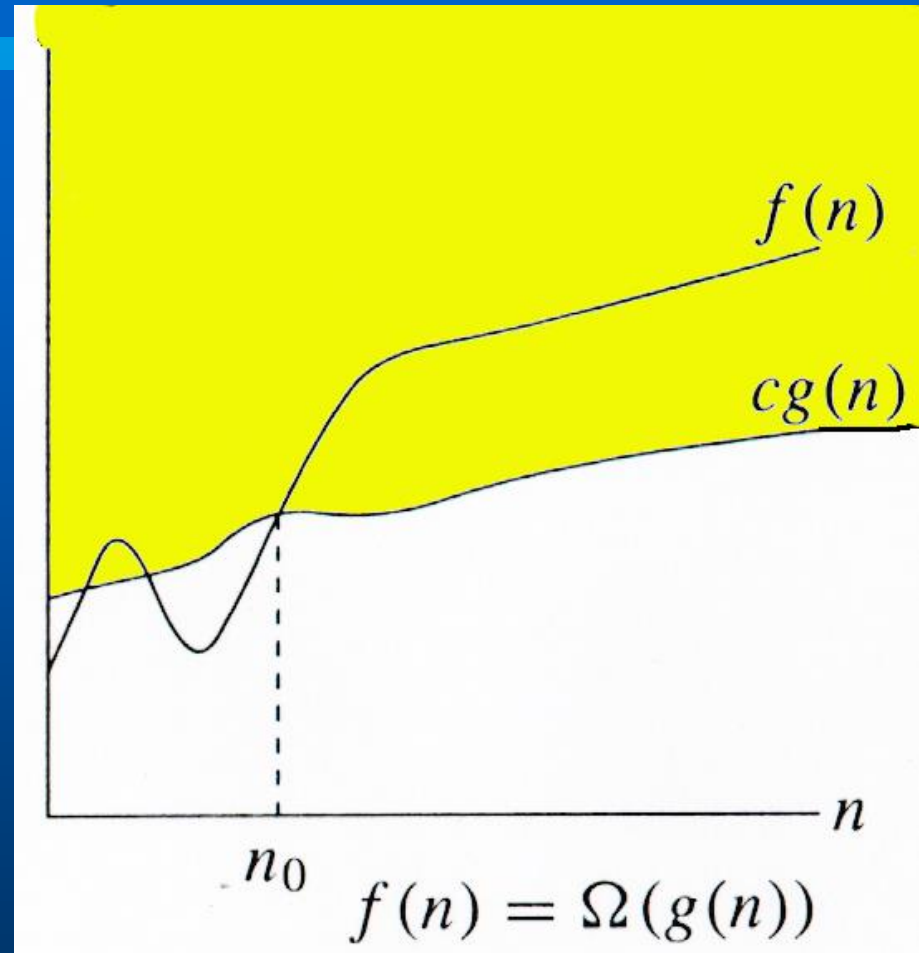
We say  $g(n)$  is an **asymptotic upper bound** for  $f(n)$

# $\Omega$ -notation



For a given function  $g(n)$ , we denote by  $\Omega(g(n))$  the set of functions

$\Omega(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}$



We say  $g(n)$  is an *asymptotic lower bound* for  $f(n)$

# Relations Between $\Theta$ , $\Omega$ , $O$



- For any two functions  $g(n)$  and  $f(n)$ ,  
 $f(n) = \Theta(g(n))$  if and only if  $f(n) = O(g(n))$   
and  $f(n) = \Omega(g(n))$ .
- I.e.,  $\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$

# Running Times



- “The running time is  $O(f(n))$ ”  
 $\Rightarrow$  Worst case is  $O(f(n))$
- “Running time is  $\Omega(f(n))$ ”  $\Rightarrow$  Best case is  $\Omega(f(n))$
- Can still say “Worst case running time is  $\Omega(f(n))$ ”
  - Means worst case running time is given by some unknown function  $g(n) \in \Omega(f(n))$

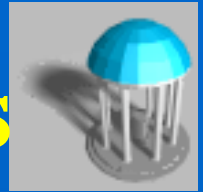


# Example



- **Insertion sort** takes  $\Theta(n^2)$  worst case time, so sorting (as a *problem*) is  $O(n^2)$
- Any sort algorithm must look at each item, so sorting is  $\Omega(n)$
- In fact, using (e.g.) merge sort, sorting is  $\Theta(n \lg n)$  in the worst case

# Asymptotic Notation in Equations



- Used to replace expressions containing lower-order terms

- For example,

$$\begin{aligned} 4n^3 + 3n^2 + 2n + 1 &= 4n^3 + 3n^2 + \Theta(n) \\ &= 4n^3 + \Theta(n^2) = \Theta(n^3) \end{aligned}$$

- In equations,  $\Theta(f(n))$  always stands for an **anonymous function**  $g(n) \in \Theta(f(n))$ 
  - In the example above,  $\Theta(n^2)$  stands for  $3n^2 + 2n + 1$

# *o*-notation



For a given function  $g(n)$ , we denote by  $o(g(n))$  the set of functions

$o(g(n)) = \{f(n): \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq f(n) < cg(n) \text{ for all } n \geq n_0\}$

$f(n)$  becomes insignificant relative to  $g(n)$  as  $n$  approaches infinity:  $\lim_{n \rightarrow \infty} [f(n) / g(n)] = 0$

We say  $g(n)$  is an **upper bound** for  $f(n)$  that is not asymptotically tight.

# $\omega$ -notation



For a given function  $g(n)$ , we denote by  $\omega(g(n))$  the set of functions

$\omega(g(n)) = \{f(n): \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq cg(n) < f(n) \text{ for all } n \geq n_0\}$

$f(n)$  becomes arbitrarily large relative to  $g(n)$  as  $n$  approaches infinity:  $\lim_{n \rightarrow \infty} [f(n) / g(n)] = \infty$

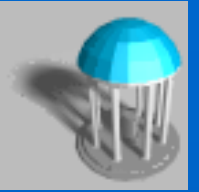
We say  $g(n)$  is a **lower bound** for  $f(n)$  that is not asymptotically tight.

# Limits



- $\lim_{n \rightarrow \infty} [f(n) / g(n)] = 0 \Rightarrow f(n) \in o(g(n))$
- $\lim_{n \rightarrow \infty} [f(n) / g(n)] < \infty \Rightarrow f(n) \in O(g(n))$
- $0 < \lim_{n \rightarrow \infty} [f(n) / g(n)] < \infty \Rightarrow f(n) \in \Theta(g(n))$
- $0 < \lim_{n \rightarrow \infty} [f(n) / g(n)] \Rightarrow f(n) \in \Omega(g(n))$
- $\lim_{n \rightarrow \infty} [f(n) / g(n)] = \infty \Rightarrow f(n) \in \omega(g(n))$
- $\lim_{n \rightarrow \infty} [f(n) / g(n)]$  undefined  $\Rightarrow$  can't say

# Comparison of Functions



$$f \leftrightarrow g \approx a \leftrightarrow b$$

$$f(n) = O(g(n)) \approx a \leq b$$

$$f(n) = \Omega(g(n)) \approx a \geq b$$

$$f(n) = \Theta(g(n)) \approx a = b$$

$$f(n) = o(g(n)) \approx a < b$$

$$f(n) = \omega(g(n)) \approx a > b$$

# Properties



- **Transitivity**

$$f(n) = \Theta(g(n)) \ \& \ g(n) = \Theta(h(n)) \Rightarrow f(n) = \Theta(h(n))$$

$$f(n) = O(g(n)) \ \& \ g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$$

$$f(n) = \Omega(g(n)) \ \& \ g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$$

$$f(n) = o(g(n)) \ \& \ g(n) = o(h(n)) \Rightarrow f(n) = o(h(n))$$

$$f(n) = \omega(g(n)) \ \& \ g(n) = \omega(h(n)) \Rightarrow f(n) = \omega(h(n))$$

- **Symmetry**

$$f(n) = \Theta(g(n)) \text{ if and only if } g(n) = \Theta(f(n))$$

- **Transpose Symmetry**

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

$$f(n) = o(g(n)) \text{ if and only if } g(n) = \omega(f(n))$$

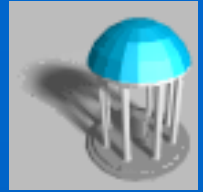
# Useful Facts



- For  $a \geq 0, b > 0$ ,  $\lim_{n \rightarrow \infty} (\lg^a n / n^b) = 0$ , so  $\lg^a n = o(n^b)$ , and  $n^b = \omega(\lg^a n)$ 
  - Prove using L'Hopital's rule repeatedly
- $\lg(n!) = \Theta(n \lg n)$



# Examples



A

- $5n^2 + 100n$

- $\log_3(n^2)$

- $n^{\lg 4}$

- $\lg^2 n$

B

$$3n^2 + 2$$

$$\log_2(n^3)$$

$$3^{\lg n}$$

$$n^{1/2}$$

# Examples



A

B

- $5n^2 + 100n$

$$3n^2 + 2$$

$$A \in \Theta(B)$$

$$A \in \Theta(n^2), n^2 \in \Theta(B) \Rightarrow A \in \Theta(B)$$

- $\log_3(n^2)$

$$\log_2(n^3)$$

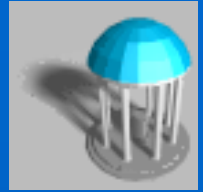
- $n^{\lg 4}$

$$3^{\lg n}$$

- $\lg^2 n$

$$n^{1/2}$$

# Examples



A

B

- $5n^2 + 100n$        $3n^2 + 2$        $A \in \Theta(B)$

$A \in \Theta(n^2), n^2 \in \Theta(B) \Rightarrow A \in \Theta(B)$

- $\log_3(n^2)$        $\log_2(n^3)$        $A \in \Theta(B)$

$\log_b a = \log_c a / \log_c b$ ;  $A = 2 \lg n / \lg 3$ ,  $B = 3 \lg n$ ,  $A/B = 2/(3 \lg 3)$

- $n^{\lg 4}$        $3^{\lg n}$

- $\lg^2 n$        $n^{1/2}$

# Examples



A

B

- $5n^2 + 100n$        $3n^2 + 2$        $A \in \Theta(B)$

$$A \in \Theta(n^2), n^2 \in \Theta(B) \Rightarrow A \in \Theta(B)$$

- $\log_3(n^2)$        $\log_2(n^3)$        $A \in \Theta(B)$

$$\log_b a = \log_c a / \log_c b; A = 2 \lg n / \lg 3, B = 3 \lg n, A/B = 2/(3 \lg 3)$$

- $n^{\lg 4}$        $3^{\lg n}$        $A \in \omega(B)$

$$a^{\log b} = b^{\log a}; B = 3^{\lg n} = n^{\lg 3}; A/B = n^{\lg(4/3)} \rightarrow \infty \text{ as } n \rightarrow \infty$$

- $\lg^2 n$        $n^{1/2}$

# Examples



A

B

- $5n^2 + 100n$        $3n^2 + 2$        $A \in \Theta(B)$

$$A \in \Theta(n^2), n^2 \in \Theta(B) \Rightarrow A \in \Theta(B)$$

- $\log_3(n^2)$        $\log_2(n^3)$        $A \in \Theta(B)$

$$\log_b a = \log_c a / \log_c b; A = 2 \lg n / \lg 3, B = 3 \lg n, A/B = 2/(3 \lg 3)$$

- $n^{\lg 4}$        $3^{\lg n}$        $A \in \omega(B)$

$$a^{\log b} = b^{\log a}; B = 3^{\lg n} = n^{\lg 3}; A/B = n^{\lg(4/3)} \rightarrow \infty \text{ as } n \rightarrow \infty$$

- $\lg^2 n$        $n^{1/2}$        $A \in o(B)$

$$\lim_{n \rightarrow \infty} (\lg^a n / n^b) = 0 \text{ (here } a = 2 \text{ and } b = 1/2) \Rightarrow A \in o(B)$$

# Review on Summation



- Why do we need summation formulas?

We need them for computing the running times of some algorithms. (CLRS/Chapter 3)

- Example: Maximum Subvector

Given an array  $a[1..n]$  of numeric values (can be positive, zero and negative) determine the subvector  $a[i..j]$  ( $1 \leq i \leq j \leq n$ ) whose sum of elements is maximum over all subvectors.

1	-2	2	2
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# Reviews on Summation



```
MaxSubvector(a, n)
  maxsum ← 0;
  for i ← 1 to n
    for j = i to n
      sum ← 0;
      for k ← i to j
        sum += a[k]
      maxsum ← max(sum, maxsum);
  return maxsum;
```

- $T(n) = \sum_{i=1}^n \sum_{j=i}^n \sum_{k=i}^j 1$

- **NOTE:** This is not a simplified solution. What *is* the final answer?

# Reviews on Summation



- Constant Series: For  $n \geq 0$ ,

$$\sum_{i=a}^b 1 = b - a + 1$$

- Linear Series: For  $n \geq 0$ ,

$$\sum_{i=1}^n i = 1 + 2 + \dots + n = n(n+1) / 2$$



# Reviews on Summation



- Quadratic Series: For  $n \geq 0$ ,

$$\sum_{i=1}^n i^2 = 1^2 + 2^2 + \dots + n^2 = (2n^3 + 3n^2 + n) / 6$$

- Linear-Geometric Series: For  $n \geq 0$ ,

$$\sum_{i=1}^n ic^i = c + 2c^2 + \dots + nc^n = [(n-1)c^{(n+1)} - nc^n + c] / (c-1)^2$$