Three time complexity functions

- The worse case time complexity of the algorithm is the function defined by the maximum number of operations performed, taken across all instances of size *n*.
- The **best case** time complexity of the algorithm is the function defined by the minimum number of operations performed, taken across all instances of size *n*.
- The **average-case** time complexity of the algorithm is the function defined by an average number of operations performed, taken across all instances of size *n*.

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Which complexity is most important?

- 1. The worst-case time complexity is an upper bound on the running time for any input.
- 2. For some algorithms, the worst case occurs fairly often. Binary search is an example.
- 3. Usually for a good algorithm, then average case is often roughly as bad as the worse case. Insertion sort is an example.

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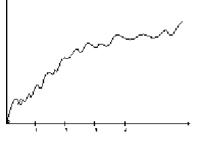
Exact analysis is hard and not necessary (in most cases)

The exact complexity of insertion sort is:

$$T(n) = \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right)n^2 + (c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8)n - (c_2 + c_4 + c_5 + c_8)$$

we prefer to write it as $an^2 + bn + c$

Another example of exact function $n^3 + 100n + 2 + 10 \lg n + 0.1 * 2^n$ may look like this:



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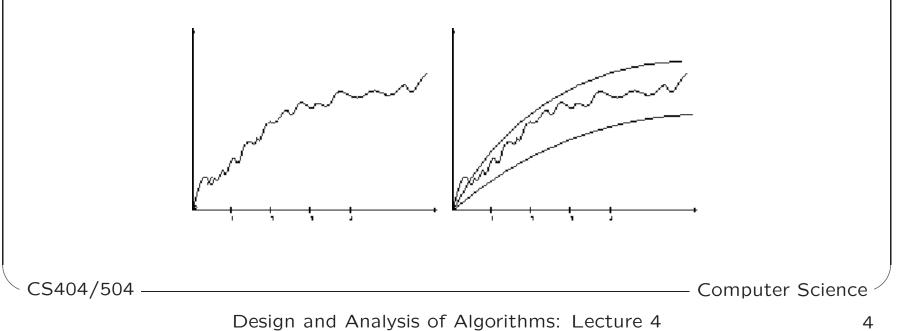
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What should we do? Simplify the analysis

- Ignore the constants:
 - We can erase the difference between $T_1(n) = 2n$ and $T_2(n) = 8n$ by using different machines.
- Only look at "dominant" terms:

$$-n^3 + 100n + 2$$
 is "dominated" by n^3 as $n \to \infty$



Asymptotic Notation: order of growth

f(n) = O(g(n)) means C × g(n) is an asymptotic upper bound on f(n).

f(n) = Ω(g(n)) means C × g(n) is an asymptotic lower bound on f(n).

f(n) = ⊖(g(n)) means C₁ × g(n) is an asymptotic upper bound on f(n) and C₂ × g(n) is an asymptotic lower bound on f(n).

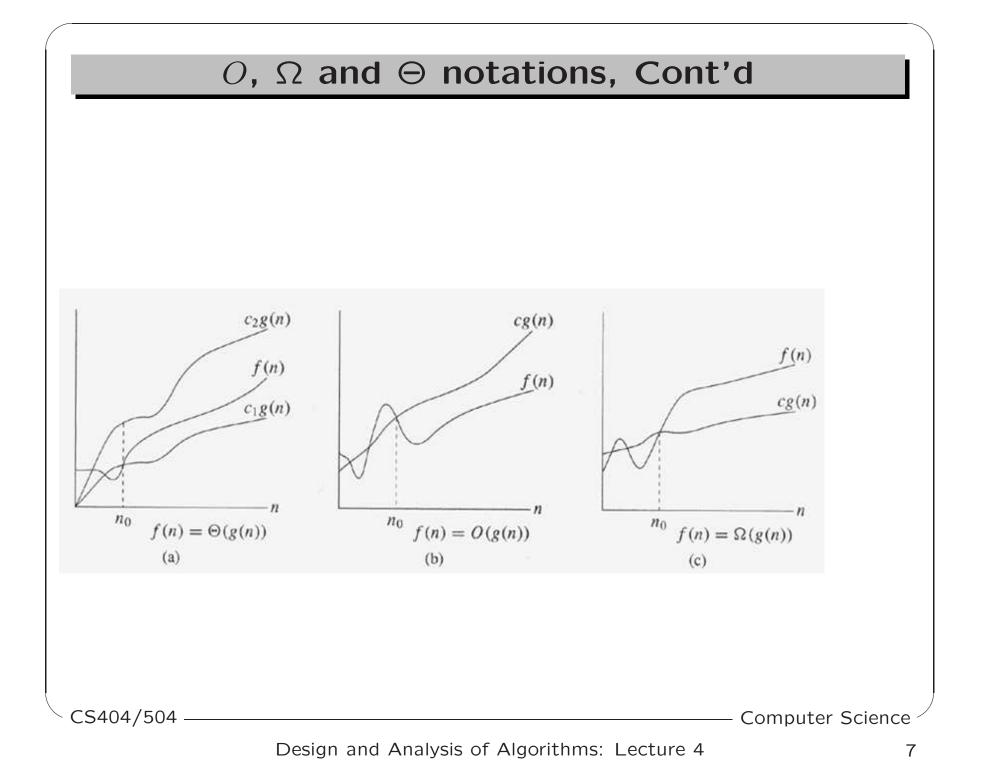
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O , Ω and Θ notations

Definition:

- f(n) = O(g(n)) if \exists positive constants n_0 and c such that $0 \le f(n) \le cg(n)$ for all $n \ge n_0$.
- $f(n) = \Omega(g(n))$ if \exists positive constants n_0 and c such that $f(n) \ge cg(n) \ge 0$ for all $n \ge n_0$.
- $f(n) = \Theta(g(n))$ if \exists positive constants c_1 , c_2 , and n_0 such that $0 \le c_1 g(n) \le f(n) \le c_2 g(n)$ for all $n \ge n_0$.

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Examples

Example 1:
$$f(n) = 3n^2$$
, $g(n) = 10n^2 + 5n$
Claim: $f(n) = O(g(n))$
Proof:
 $c = 1$; $n_0 = 1$ $3n^2 \le 10n^2 + 5n$ for all $n \ge 1$

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Examples (Cont'd)

Example 2:
$$f(n) = 10n^2 + 5n, g(n) = 3n^2$$

Claim: f(n) = O(g(n))

Proof: c = 4; $10n^2 + 5n \le 12n^2$, for all $n \ge 10$

Example 3: $10n^2 + 4n + 2 = O(n^2)$

Proof: $10n^2 + 4n + 2 \le 11n^2$ for all $n \ge 5$.

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Example (Cont'd)

Example 4:
$$f(n) = a_d n^d + a_{d-1} n^{d-1} + \dots + a_1 n + a_0$$
,

Claim:
$$f(n) = O(n^d)$$

Proof:

$$\begin{split} f(n) &= a_d n^d + a_{d-1} n^{d-1} + \ldots + a_1 n + a_0 \\ &\leq |a_d| n^d + |a_{d-1}| n^{d-1} + \ldots + |a_1| n + |a_0| \\ &= \sum_{i=0}^d |a_i| n^i \\ &= n^d \sum_{i=0}^d |a_i| n^{i-d} \\ &\leq n^d \sum_{i=0}^d |a_i| \text{ for } n \geq 1, \text{ because } n^{i-d} \leq 1 \text{ when } n \geq 1. \end{split}$$

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Alternative Definitions

Definition:

$$f(n) = O(g(n)) \iff \lim_{n \to \infty} \frac{f(n)}{g(n)} = c;$$

where c is a constant $c \geq 0$.

Definition:

$$f(n) = \Omega(g(n)) \iff \lim_{n \to \infty} \frac{g(n)}{f(n)} = c;$$

where c is a constant $c \geq 0$.

Theorem:
$$f(n) = O(g(n)) \iff g(n) = \Omega(f(n)).$$

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Examples

Claim:
$$12n^2 + 5n = \Theta(100n^2)$$

Proof:

$$\lim_{n \to \infty} \frac{12n^2 + 5n}{100n^2} = \lim_{n \to \infty} \frac{12}{100} + \lim_{n \to \infty} \frac{5}{100n} = \frac{12}{100}.$$
(1)

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Examples (Cont'd)

Claim:
$$10n^2 + 5n = \Theta(n^3)$$
?

Proof:

$$\lim_{n \to \infty} \frac{10n^2 + 5n}{n^3} = \lim_{n \to \infty} \frac{10}{n} + \lim_{n \to \infty} \frac{5}{n^2} = 0 + 0 = 0$$

But:

$$\lim_{n \to \infty} \frac{n^3}{10n^2 + 5n} = \frac{1}{\lim_{n \to \infty} \frac{10n^2 + 5n}{n^3}} = \infty$$

Hence, $10n^2 + 5n$ is only $O(n^3)$.

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Examples (Cont'd)

Compare: $2^n, n^2$ and $\lg n$.

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Examples, Cont'd

$$2n^{2} - 10n + 5 = O(n^{2}), \text{ (Yes)}$$

$$2n^{2} - 10n + 5 = O(n^{3}), \text{ (Yes)}$$

$$2n^{2} - 10n + 5 = O(n), \text{ (No)}$$

$$2n^{2} - 10n + 5 = \Omega(n^{2}), \text{ (Yes)}$$

$$2n^{2} - 10n + 5 = \Omega(n^{3}), \text{ (No)}$$

$$2n^{2} - 10n + 5 = \Theta(n^{3}), \text{ (Yes)}$$

$$2n^{2} - 10n + 5 = \Theta(n^{3}), \text{ (No)}$$

$$2n^{2} - 10n + 5 = \Theta(n^{3}), \text{ (No)}$$

$$2n^{2} - 10n + 5 = \Theta(n), \text{ (No)}$$

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