

Lecture 3.

Growth of Functions - Asymptotic Analysis

Introduction to Algorithms
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Overview of Asymptotic Notation

- A way to describe behavior of functions in the limit
 - ▶ We are studying asymptotic efficiency
- Describe growth of functions
- Focus on what is important by abstracting away low-order terms and constant factors
- How we indicate running times of algorithms
- A way to compare “sizes” of functions

▶	O	\approx	\leq
	Ω	\approx	\geq
	Θ	\approx	$=$
	o	\approx	$<$
	ω	\approx	$>$

Asymptotic Notation

■ *Asymptotic efficiency of algorithms*

- ▶ Concerned with how the running time increases with the size of the input in the limit
 - ★ i.e., as the size of the input increases without bound.
- ▶ An asymptotically more efficient algorithm is the best choice

- Asymptotic running time of an algorithm is defined in terms of functions whose domains are the set of natural numbers
 $N = \{0, 1, 2, \dots\}$

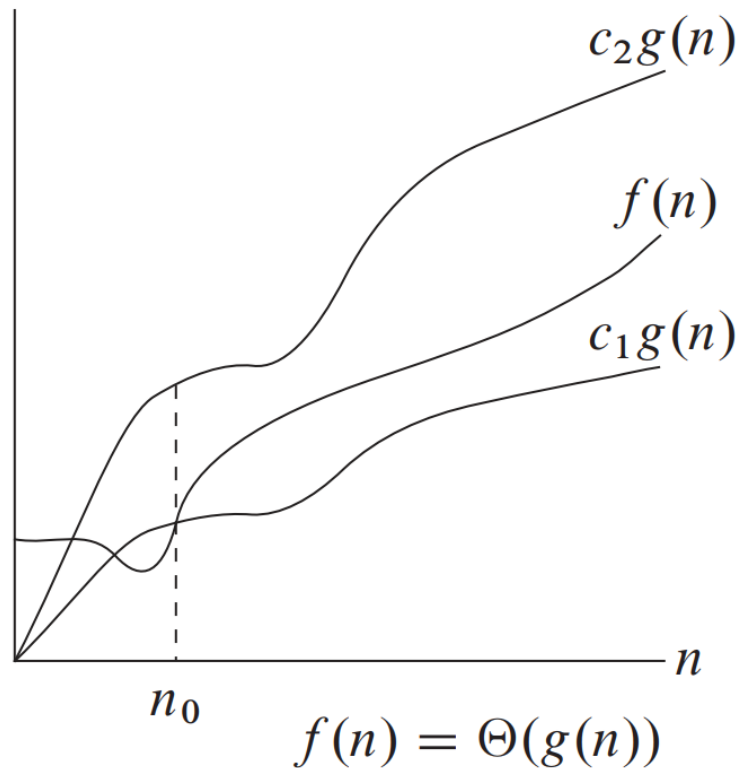
Θ - Notation (1/5)

■ Θ - notation (theta)

- ▶ $\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that}$
 $0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0\}$
- ▶ $\Theta(g(n))$ is the set of functions
- ▶ $f(n) = \Theta(g(n))$ really means that $f(n)$ belongs to $\Theta(g(n))$
- ▶ $g(n)$ is called an asymptotically tight bound for $f(n)$

Θ - Notation (2/5)

■ Θ - notation



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

Θ - Notation (3/5)

■ Θ - notation

► If $f(n)$ is a polynomial of degree d , then $f(n) = \Theta(n^d)$

► Example:

★ $n^2 - 2n = \Theta(n^2)$

★ $200n^2 - 100n = \Theta(n^2)$

$$c_1 n^2 \leq n^2 - 2n \leq c_2 n^2$$

$$c_1 \leq 1 - \frac{2}{n} \leq c_2$$

$$c_1 \leq \frac{1}{3} \quad c_2 \geq 1$$

$$n \geq 3 \quad n \geq 1$$

$$n_0 = 3$$

$$c_1 n^2 \leq 200n^2 - 100n \leq c_2 n^2$$

$$c_1 \leq 200 - \frac{100}{n} \leq c_2$$

$$c_1 \leq 100 \quad c_2 \geq 200$$

$$n \geq 1 \quad n \geq 1$$

$$n_0 = 1$$

★ Some choice for c_1 , c_2 , and n_0 exists, the functions are $\Theta(n^2)$

★ $\Theta(n^0)$?

Θ - Notation (4/5)

■ $n^2 - 2n = \Theta(n^2)$

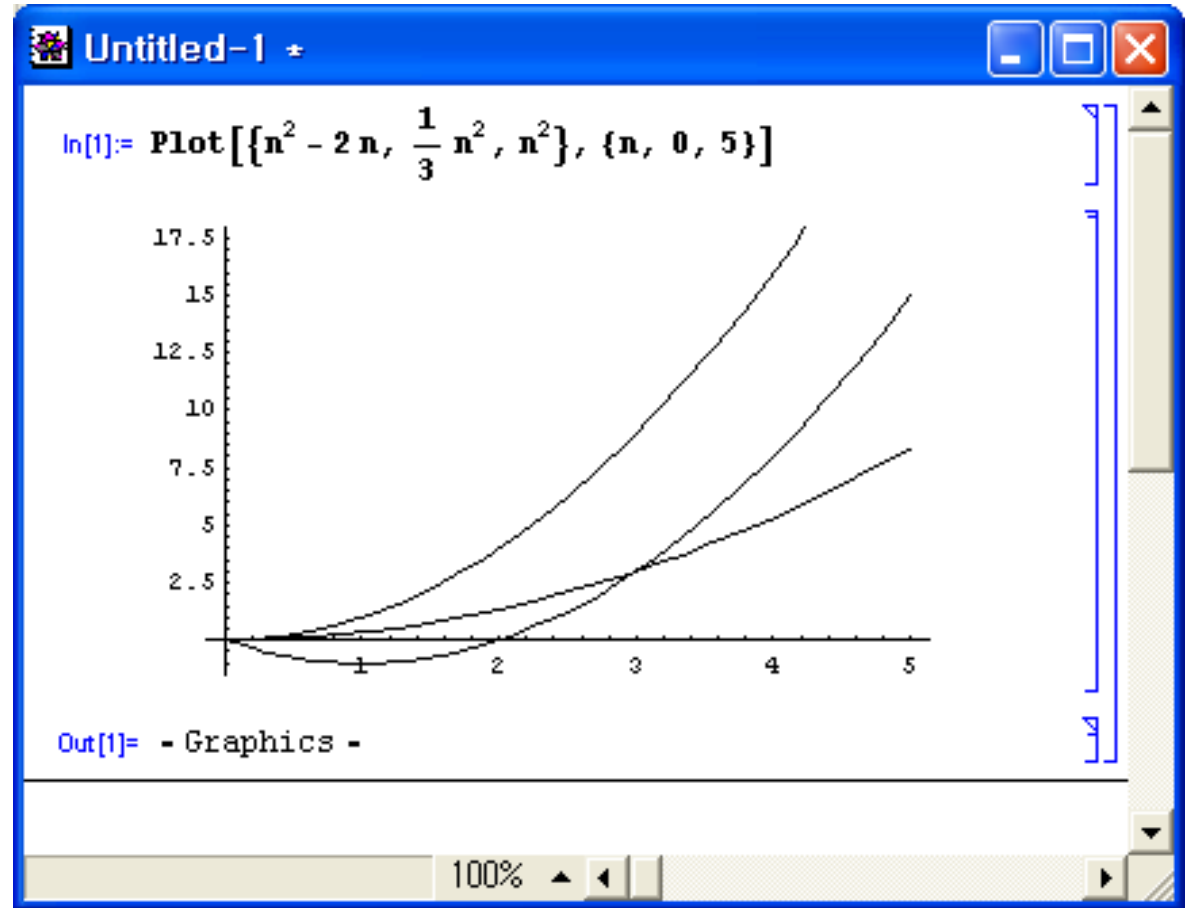
$$c_1 n^2 \leq n^2 - 2n \leq c_2 n^2$$

$$c_1 \leq 1 - \frac{2}{n} \leq c_2$$

$$c_1 \leq \frac{1}{3} \quad c_2 \geq 1$$

$$n \geq 3 \quad n \geq 1$$

$$n_0 = 3$$



Θ - Notation (5/5)

■ $200n^2 - 100n = \Theta(n^2)$

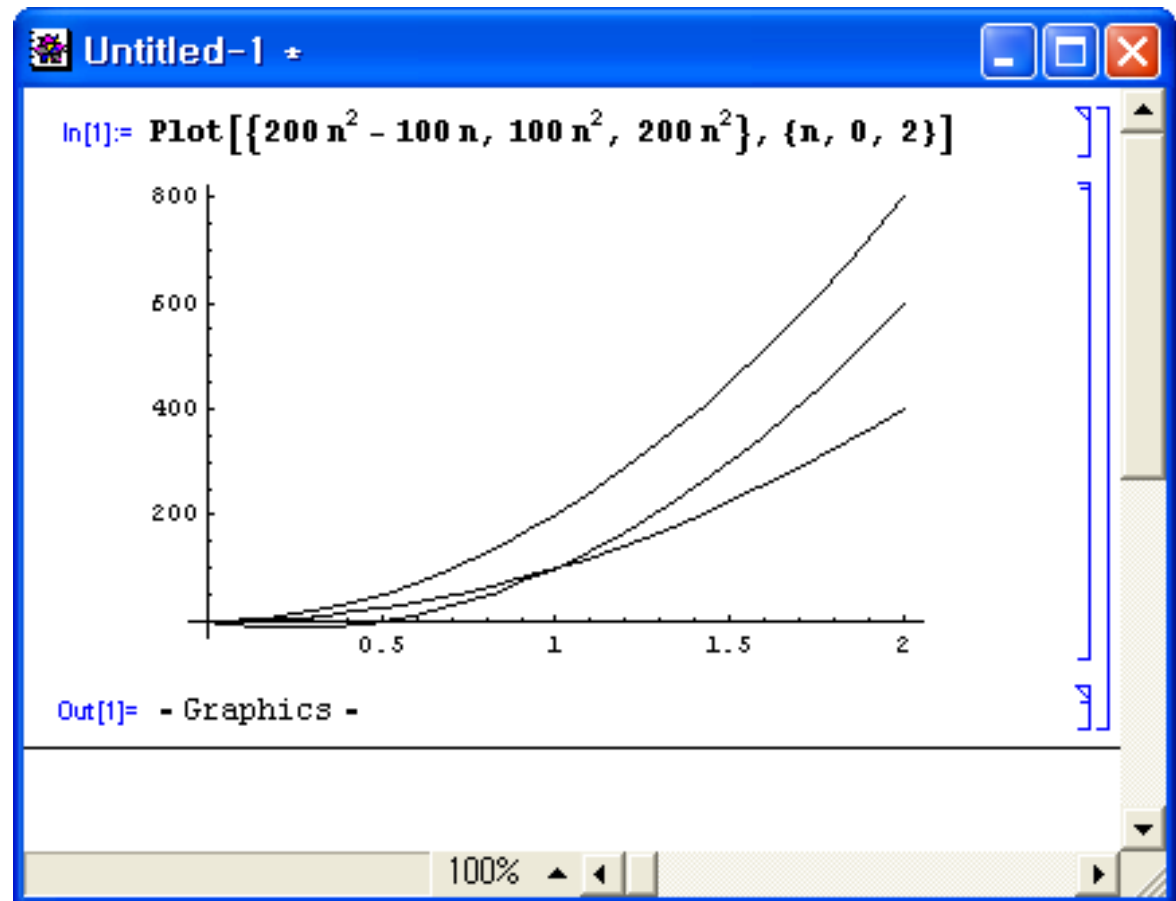
$$c_1 n^2 \leq 200n^2 - 100n \leq c_2 n^2$$

$$c_1 \leq 200 - \frac{100}{n} \leq c_2$$

$$c_1 \leq 100 \quad c_2 \geq 200$$

$$n \geq 1 \quad n \geq 1$$

$$n_0 = 1$$



O - Notation (1 / 2)

■ *O* - notation (big-oh)

- ▶ $f(n) = O(g(n))$: $g(n)$ is an asymptotic upper bound for $f(n)$
- ▶ $O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}$

- $f(n) = \Theta(g(n))$ implies $f(n) = O(g(n))$,
but $f(n) = O(g(n))$ does NOT imply $f(n) = \Theta(g(n))$

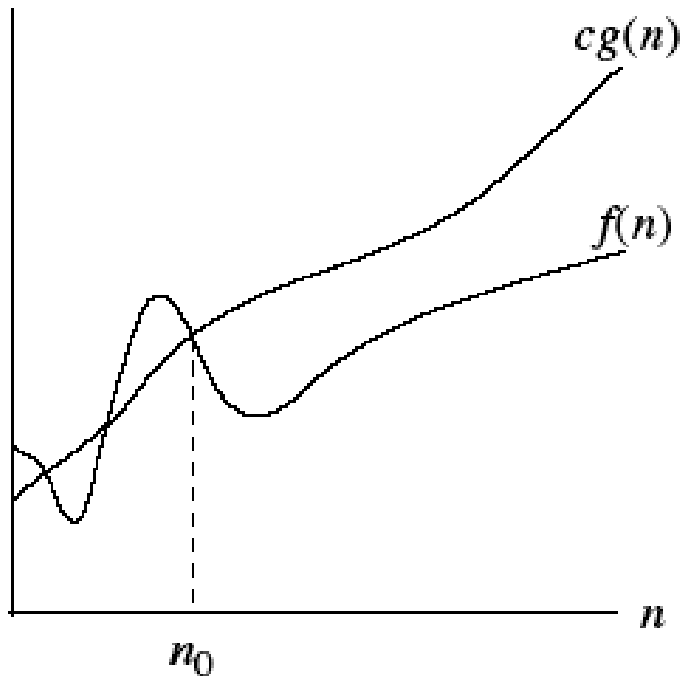
■ *Example:*

- ▶ $n^2 - 2n = O(n^2)$
- ▶ $200n^2 - 100n = O(n^2) = O(n^3) = \dots$
- ▶ $n = O(n^2)$

- *O* is good for describing the worst-case running time of an algorithm

O - Notation (2/2)

■ *O* - notation



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

Ω - Notation (1/2)

■ Ω - notation (omega)

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

- $f(n) = \Theta(g(n))$ implies $f(n) = \Omega(g(n))$,
but $f(n) = \Omega(g(n))$ does NOT imply $f(n) = \Theta(g(n))$

■ **Example:**

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

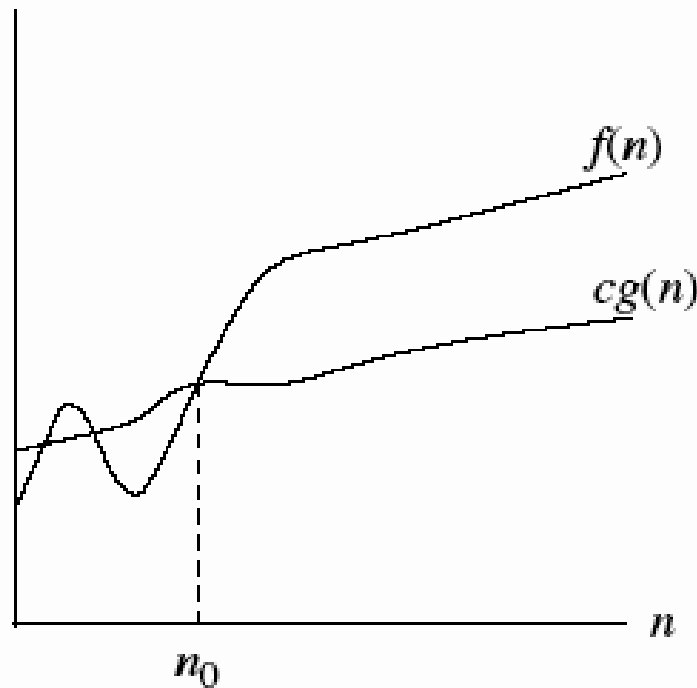
- Ω is good for describing the best case running time of an algorithm

■ **Theorem**

- ▶ $f(n) = \Theta(g(n))$ if and only if $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$

Ω - Notation (2/2)

■ Ω - notation (omega)



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

Asymptotic Notation in Equations

- On the right-hand-side
 - ▶ $n = O(n^2)$ means “ n belongs to $O(n^2)$ ”
- In general, asymptotic notation stands for some anonymous function
- **Example:**
 - ▶ $2n^2 + 3n + 1 = 2n^2 + \Theta(n)$ means
 - ★ there is some function $f(n) \in \Theta(n)$ that makes the equation true, i.e. $f(n) = 3n + 1$
 - ▶ $2n^2 + \Theta(n) = \Theta(n^2)$ means
 - ★ for any function $f(n) \in \Theta(n)$, there is some function $g(n) \in \Theta(n^2)$ such that $2n^2 + f(n) = g(n)$ for all n
 - ▶ $2n^2 + 3n + 1 = 2n^2 + \Theta(n) = \Theta(n^2)$ (a chain of relationships)

o - Notation

- ***o* - notation (little-oh)**

- ▶ ***O* - notation** may or may not be asymptotically tight
- ▶ $2n^2 = O(n^2)$ is asymptotically tight, but $2n = O(n^2)$ is not
- ▶ $f(n) = o(g(n))$: $g(n)$ is an upper bound of $f(n)$ that is not asymptotically tight
- ▶ $o(g(n)) = \{f(n): \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq f(n) < cg(n) \text{ for all } n \geq n_0\}$
- ▶ $2n = o(n^2)$, but $2n^2 \neq o(n^2)$

- ***O***: for some constant c , ***o***: for all constant c

- In the ***o*-notation**, the function $f(n)$ becomes insignificant relative to $g(n)$ as n approaches infinity

- ▶ i.e., $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$

ω - Notation

■ ω - notation (little-omega)

- ▶ $2n^2 = \Omega(n^2)$ is asymptotically tight, but $2n^3 = \Omega(n^2)$ is not
- ▶ $f(n) = \omega(g(n))$: $g(n)$ is a lower bound of $f(n)$ that is not asymptotically tight
- ▶ $\omega(g(n)) = \{f(n): \text{for any positive constant } c > 0, \text{ there exists a constant } n_0 > 0 \text{ such that } 0 \leq cg(n) < f(n) \text{ for all } n \geq n_0\}$
- ▶ $2n^2 = \omega(n)$, but $2n^2 \neq \omega(n^2)$

■ In the ω -notation, the function $f(n)$ becomes arbitrarily large relative to $g(n)$ as n approaches infinity

- ▶ i.e., $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$

Comparison of Functions (1/2)

- *Many of the relational properties of real numbers apply to asymptotic comparisons too*

- ▶ ***Reflexivity***

- ★ $f(n) = \Theta(f(n))$
- ★ *This is true for O , Ω*

- ▶ ***Symmetry***

- ★ $f(n) = \Theta(g(n))$ if and only if $g(n) = \Theta(f(n))$

- ▶ ***Transpose symmetry***

- ★ $f(n) = O(g(n))$ if and only if $g(n) = \Omega(f(n))$
- ★ $f(n) = o(g(n))$ if and only if $g(n) = \omega(f(n))$

- ▶ ***Transitivity***

- ★ $f(n) = \Theta(g(n))$ and $g(n) = \Theta(h(n))$ imply $f(n) = \Theta(h(n))$
- ★ *This is true for O , Ω , o , and ω*

Comparison of Functions (2/2)

■ *Analogy to real numbers*

- ▶ $f(n) = O(g(n)) \approx a \leq b$
- ▶ $f(n) = \Omega(g(n)) \approx a \geq b$
- ▶ $f(n) = \Theta(g(n)) \approx a = b$
- ▶ $f(n) = o(g(n)) \approx a < b$
- ▶ $f(n) = \omega(g(n)) \approx a > b$

■ *Example: $f(n) = 3n^3 + 4$*

- ▶ $f(n) = \Theta(n^3)$
- ▶ $f(n) = O(n^3) = O(n^4) = \dots$
- ▶ $f(n) = \Omega(n^3) = \Omega(n^2) = \Omega(n) = \Omega(1)$
- ▶ $f(n) = o(n^4) = o(n^5) = \dots$
- ▶ $f(n) = \omega(n^2) = \omega(n) = \omega(1)$

Standard Notation and Common Functions (1/2)

■ **Monotonicity:** A function $f(n)$ is:

- ▶ monotonically increasing if $a \leq b$ implies $f(a) \leq f(b)$
- ▶ monotonically decreasing if $a \leq b$ implies $f(a) \geq f(b)$
- ▶ strictly increasing if $a < b$ implies $f(a) < f(b)$
- ▶ strictly decreasing if $a < b$ implies $f(a) > f(b)$

■ **Floors and ceilings**

- ▶ Floor: $\lfloor x \rfloor$ is the greatest integer $\leq x$
- ▶ Ceiling: $\lceil x \rceil$ is the least integer $\geq x$

▶ **Examples:** $\lfloor 3 \rfloor = 3$ $\lceil 3 \rceil = 3$

$$\lfloor 3.3 \rfloor = 3 \quad \lceil 3.3 \rceil = 4$$

$$\lfloor 3.9 \rfloor = 3 \quad \lceil 3.9 \rceil = 4$$

$$\lfloor n/2 \rfloor + \lfloor n/2 \rfloor = n, \quad \lceil \lceil n/a \rceil / b \rceil = \lceil n/ab \rceil, \quad \lfloor \lfloor n/a \rfloor / b \rfloor = \lfloor n/ab \rfloor$$

Standard Notation and Common Functions (2/2)

■ **Logarithms**

- ▶ $\log_b n$: logarithm of n base b
- ▶ $\lg n = \log_2 n$ (binary logarithm)
- ▶ $\ln n = \log_e n$ (natural logarithm, $e = 2.7182\dots$)

■ **Factorials**

- ▶
$$n! = \begin{cases} 1 & \text{if } n = 0 \\ n(n-1)! & \text{if } n > 0 \end{cases}$$
- ▶ $n! = 1 * 2 * 3 * \dots * n$
- ▶ $n! \leq n^n$, thus $\mathcal{O}(n^n)$

■ **Stirling's Approximation**

- ▶
$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)$$
- ▶ From Stirling's approximation, the followings hold:
 - ★ $n! = \mathcal{o}(n^n)$
 - ★ $n! = \omega(2^n)$
 - ★ $\lg(n!) = \Theta(n \lg n)$

Practice Problems

■ Prove or disprove the following:

a. $4 + 2n + 3n^2 \in O(n^2)$

b. $2 + 4 + 6 + 8 + \dots + 2n + 2n^2 \in O(n^2)$

c. $2^{2n} = O(2^n)$

Will be solved in Q&A session

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