

COMP3600/6466 – Algorithms

Asymptotic Analysis

[CLRS sec. 3.1]

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Topics

- What is asymptotic analysis?
 - The five asymptotic notations
 - Example on Insertion Sort
 - Additional notes on asymptotic notations
 - Using asymptotic notations to analyse algorithms
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What is Asymptotic Analysis?

- In general: Asymptotic analysis means understanding the behavior of a function in the limit
 - In this class: Asymptotic analysis means understanding how the running time (or memory consumption) of an algorithm increases with the size of the input *in the limit*—that is, when the input size increases without bound
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Asymptotic Analysis of Algorithms

- Recall: An algorithm transforms input to output
- Intuitively, asymptotic analysis of an algorithm means finding a function that maps input size to the required computational resources (time/memory) for the algorithm to transform the input to the desired output
- In this analysis, we are interested in the trend of the growth in computational resources rather than the exact function
 - They are easier to work with & sufficient
- Therefore, the functions we are interested in this analysis are essentially bounds of the actual requirements
 - Upper bound, lower bound, upper & lower bound

Why bother?

- In general, algorithms and computer programs are **not** for a single input, but for solving problems with varying input
 - Asymptotic analysis helps in:
 - Deciding which algorithms are better for solving a specific problem
 - Predicting time & memory requirements as input size grow. This could relate to the computers we need to purchase (the highest spec is not always the best, esp. when starting a business w. limited funding)
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Topics

✓ What is asymptotic analysis?

- The five asymptotic notations
 - Example on Insertion Sort
 - Additional notes on asymptotic notations
 - Using asymptotic notations to analyse algorithms
-

Five asymptotic notations

- Big-Oh (upper bound): $f(n) = O(g(n))$
- Little-Oh (strict upper bound): $f(n) = o(g(n))$
- Big-Omega (lower bound): $f(n) = \Omega(g(n))$
- Little-Omega (strict lower bound): $f(n) = \omega(g(n))$
- Theta (upper & lower bound): $f(n) = \Theta(g(n))$
- What does the notation mean? Let's take big-oh example
 - $O(g(n))$ is a set of functions that satisfy certain characteristics related to $g(n)$
 - $f(n) = O(g(n))$ means $f(n) \in O(g(n))$, and call $g(n)$ to be an asymptotic bound of $f(n)$. In the case of big-oh, asymptotic upper bound

Asymptotic notations: Big-Oh $O(g(n))$

Def.:

$$O(g(n)) = \{f(n) \mid \exists \text{ positive constants } c \text{ and } n_0 \\ \text{s.t. } 0 \leq f(n) \leq cg(n) \text{ for } \forall n \geq n_0\}$$

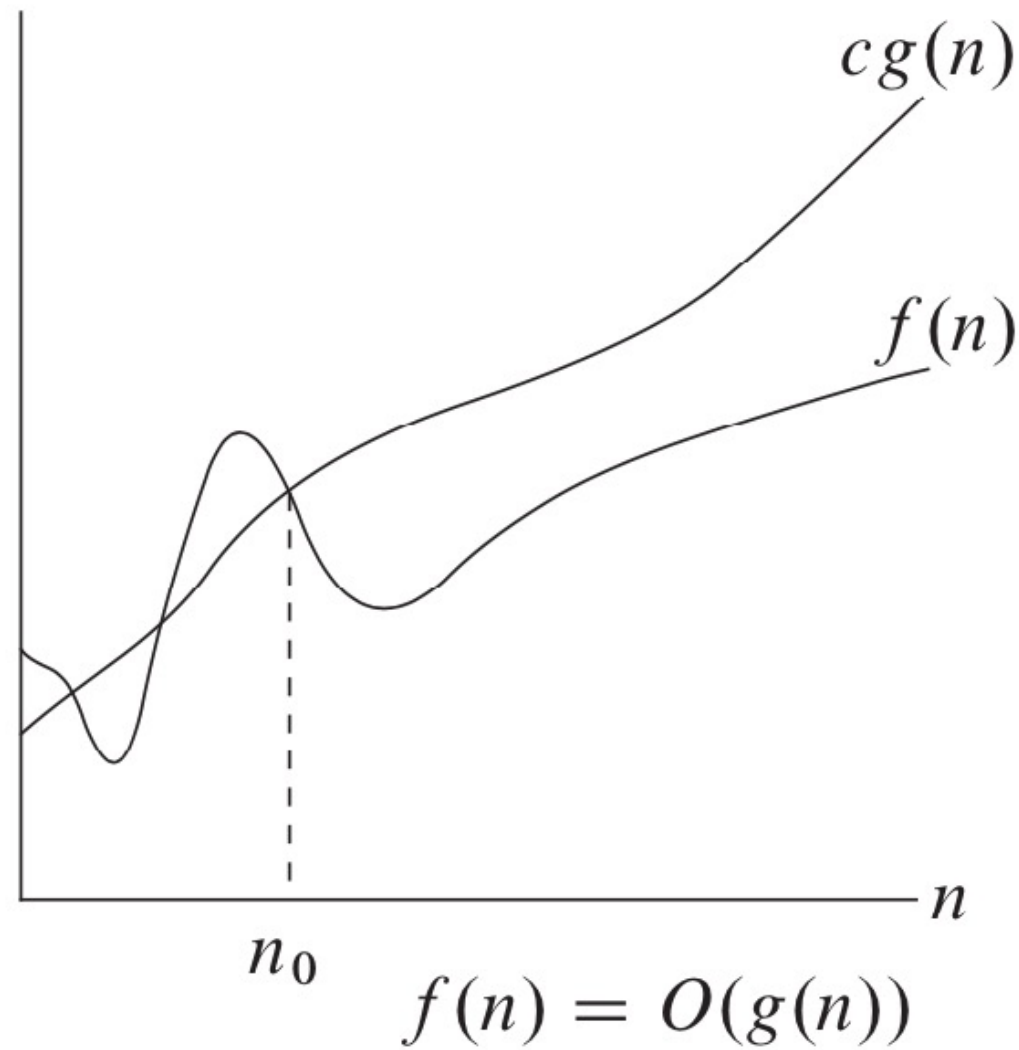
Intuitively:

$O(g(n))$ is a set of functions that is upper bounded by a constant multiplication of $g(n)$ for “large” n —that is, whenever $n \geq n_0$.

We write $f(n) = O(g(n))$ to mean $f(n) \in O(g(n))$, and call $g(n)$ to be an asymptotic upper bound of $f(n)$

This bound can be tight or not tight. The function $g(n)$ is an asymptotically tight upper bound of $f(n)$ whenever for $\forall n \geq n_0$, $f(n)$ is equal to $g(n)$ to within a constant factor

Visually,



Taken from: [CLRS] Figure 3.1

An Example: Back to Insertion Sort

- Recall Insertion Sort and its time complexity

InsertionSort(A)

1. for $j = 2$ to $A.length$
2. $Key = A[j]$
3. $i = j-1$
4. While $i > 0$ and $A[i] > key$
5. $A[i+1] = A[i]$
6. $i = i-1$
7. $A[i+1] = key$

Time complexity of Insertion Sort: $T(n) = Cn^2 + C'n - C''$
 $f(n)$

Is this information useful?

- The running-time complexity of an algorithm is at least $O(n^2)$

Membership test for Big-Oh $O(g(n))$

- How to figure out if $f(n) \in O(g(n))$?
 - Find the constants c and n_0
 - Use limit
 - $f(n) \in O(g(n))$ means $f(n)$ has smaller or equal growth rate, compared to $g(n)$
 - $f(n)$ has smaller growth rate than $g(n)$ means $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$
 - $f(n)$ has the same growth rate as $g(n)$ means $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c'$
for a positive constant c'
 - Examples:
 $2n ? O(n)$; $2n ? O(n^2)$; $2n^2 ? O(n)$; $2n^2 ? O(n^2)$
-

Just in case...

- How to compute the limit?

- $\lim_{n \rightarrow \infty} \frac{n+1}{n^2+1} = \frac{\infty}{\infty} = \textit{undefined}!!!$

- Recall L'Hopital's rule: $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{f'(n)}{g'(n)}$

Examples

- Is $8^n = O(2^n)$?
 - Is $\log(2n) = O(\ln(n))$?
-

Asymptotic notations: Little-Oh $o(g(n))$

Def.:

$$o(g(n)) = \{f(n) \mid \text{for any positive constant } c > 0, \\ \exists \text{ a constant } n_0 > 0 \text{ s.t. } 0 \leq f(n) < cg(n) \text{ for } \forall n \geq n_0\}$$

Intuitively:

$o(g(n))$ is a set of functions that is strictly upper bounded by a constant multiplication of $g(n)$ for “large” n —that is, whenever $n \geq n_0$ — in a non tight manner

We write $f(n) = o(g(n))$ to mean $f(n) \in o(g(n))$, and call $g(n)$ to be a non asymptotically tight upper bound of $f(n)$

Membership test for Little-Oh $o(g(n))$

- How to figure out if $f(n) \in o(g(n))$?
 - Find the constants c and n_0
 - Use limit
 - $f(n) \in o(g(n))$ means $f(n)$ has smaller growth rate compared to $g(n)$, i. e., $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$
- Example:
 $2n ? o(n) ; 2n ? o(n^2) ; 2n^2 ? o(n) ; 2n^2 ? o(n^2)$

For comparison

$$2n = O(n) ; 2n = O(n^2) ; 2n^2 \neq O(n) ; 2n^2 = O(n^2)$$

Asymptotic notations: Big-Omega $\Omega(g(n))$

Def.:

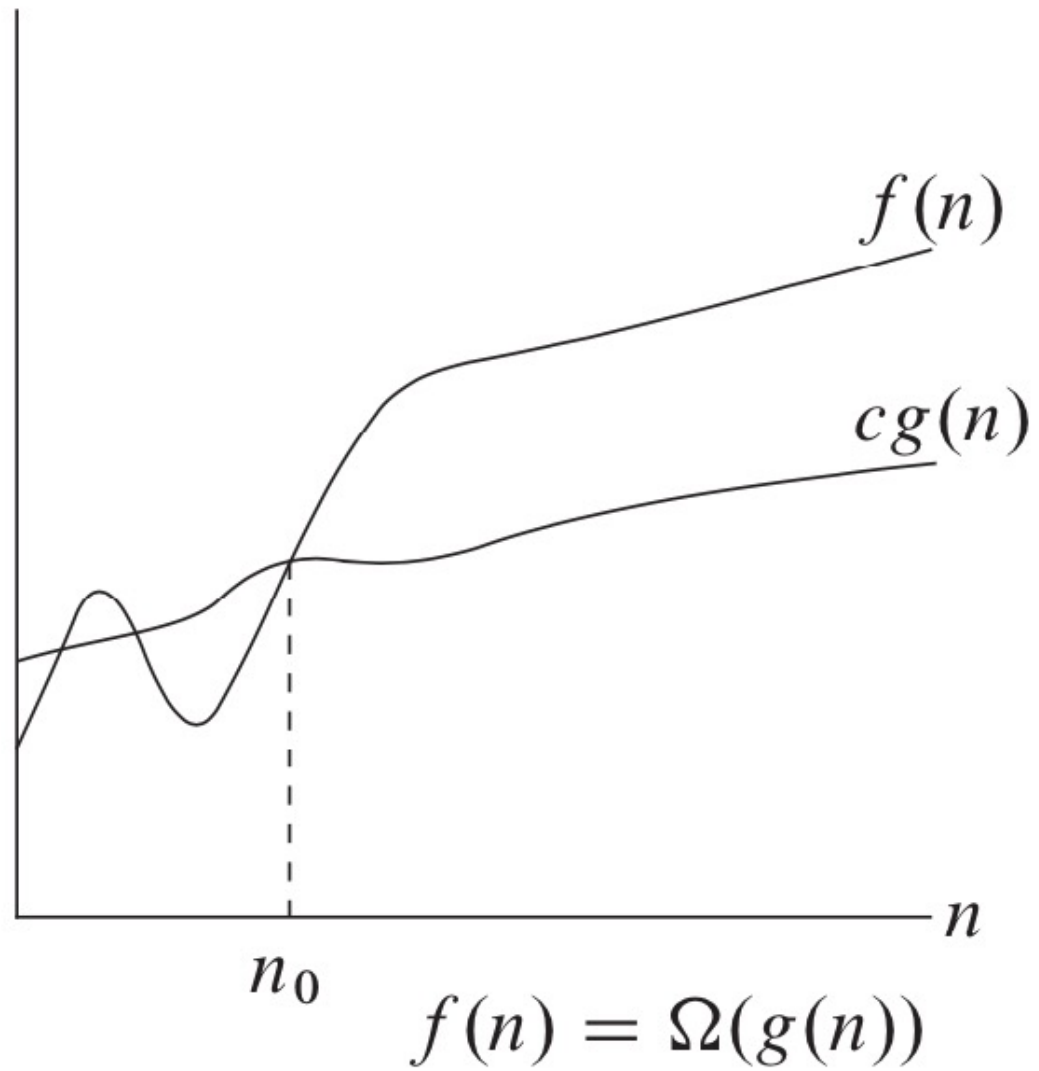
$$\Omega(g(n)) = \{f(n) \mid \exists \text{ positive constants } c \text{ and } n_0 \\ \text{s.t. } 0 \leq cg(n) \leq f(n) \text{ for } \forall n \geq n_0\}$$

Intuitively:

$\Omega(g(n))$ is a set of functions that is lower bounded by a constant multiplication of $g(n)$ for “large” n —that is, whenever $n \geq n_0$.

We write $f(n) = \Omega(g(n))$ to mean $f(n) \in \Omega(g(n))$, and call $g(n)$ to be an asymptotic lower bound of $f(n)$

Visually,



Membership test for Big-Omega $\Omega(g(n))$

- How to figure out if $f(n) \in \Omega(g(n))$?
 - Find the constants c and n_0
 - Use limit
 - $f(n) \in \Omega(g(n))$ means $f(n)$ has larger or equal growth rate, compared to $g(n)$
 - $f(n)$ has larger growth rate than $g(n)$ means $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$
 - $f(n)$ has the same growth rate as $g(n)$ means $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c'$ for a positive constant c'
- Example:
 $2n ? \Omega(n) ; 2n ? \Omega(n^2) ; 2n^2 ? \Omega(n) ; 2n^2 ? \Omega(n^2)$

For comparison

$$2n = O(n) ; 2n = O(n^2) ; 2n^2 \neq O(n) ; 2n^2 = O(n^2)$$

Asymptotic notations: Little-Omega $\omega(g(n))$

Def.:

$$\omega(g(n)) = \{f(n) \mid \text{for any positive constant } c > 0, \\ \exists \text{ a constant } n_0 > 0 \text{ s.t. } 0 \leq cg(n) < f(n) \text{ for } \forall n \geq n_0\}$$

Intuitively:

$\omega(g(n))$ is a set of functions that is lower bounded by a constant multiplication of $g(n)$ for “large” n —that is, whenever $n \geq n_0$ —in a non tight manner

We write $f(n) = \omega(g(n))$ to mean $f(n) \in \omega(g(n))$, and call $g(n)$ to be a non asymptotically tight lower bound of $f(n)$

Membership test for Little-Omega $\omega(g(n))$

- How to figure out if $f(n) \in \omega(g(n))$?
 - Find the constants c and n_0
 - Use limit
 - $f(n) \in \omega(g(n))$ means $f(n)$ has larger growth rate compared to $g(n)$, i. e., $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$
- Example:
 $2n ? \omega(n) ; 2n ? \omega(n^2) ; 2n^2 ? \omega(n) ; 2n^2 ? \omega(n^2)$

For comparison

$$2n = O(n) ; 2n = O(n^2) ; 2n^2 \neq O(n) ; 2n^2 = O(n^2)$$

Asymptotic notations: Big-Theta $\Theta(g(n))$

Def.:

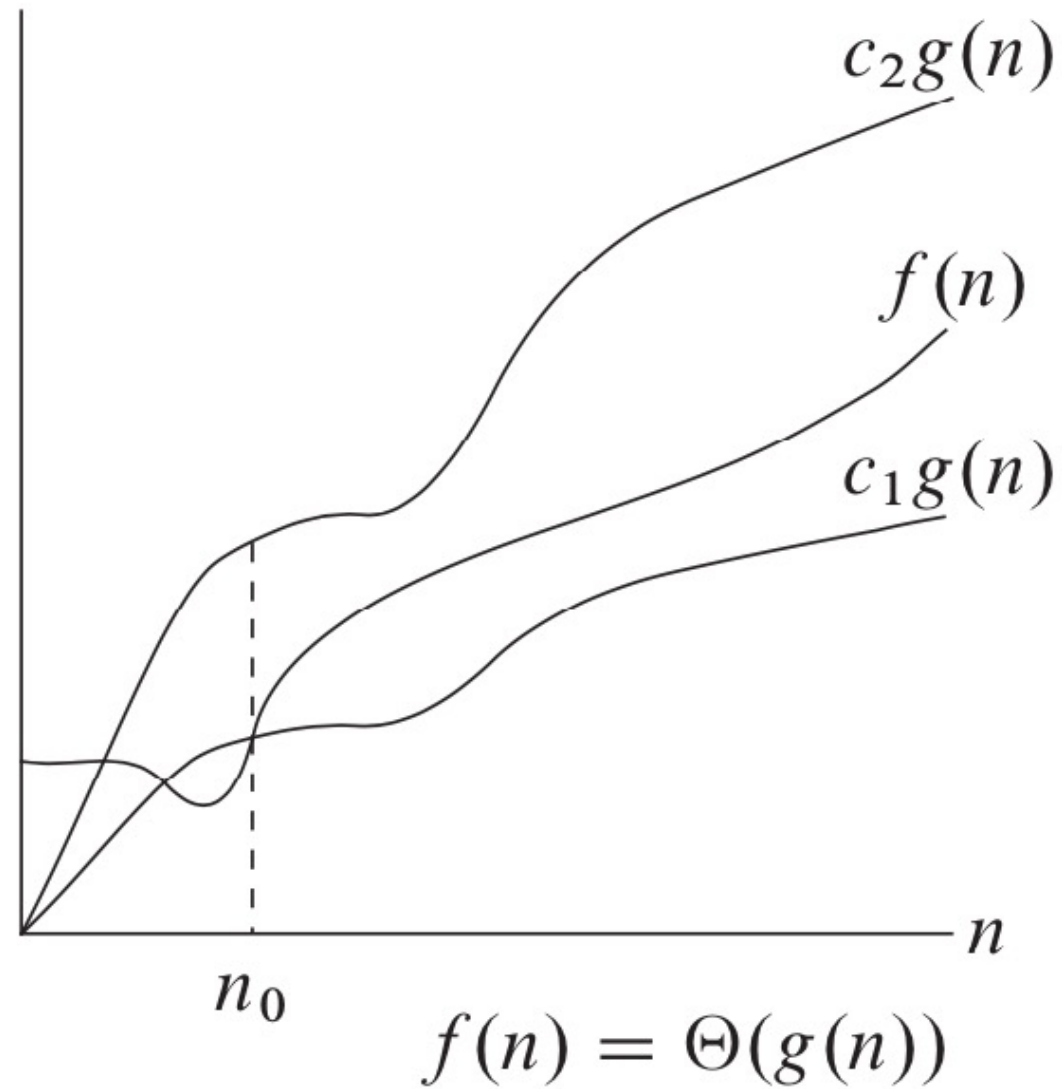
$$\Theta(g(n)) = \{f(n) \mid \exists \text{ positive constants } c_1, c_2, \text{ and } n_0 \\ \text{s.t. } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for } \forall n \geq n_0\}$$

Intuitively:

$\Theta(g(n))$ is a set of functions that is asymptotically upper and lower bounded by constants multiplication of $g(n)$ for “large” n —that is, whenever $n \geq n_0$.

We write $f(n) = \Theta(g(n))$ to mean $f(n) \in \Theta(g(n))$, and call $g(n)$ to be an asymptotically tight bound of $f(n)$

Visually,



Membership test for Big-Theta $\Theta(g(n))$

- How to figure out if $f(n) \in \Theta(g(n))$?
 - Find the constants c_1, c_2 , and n_0
 - Use limit
 - $f(n) \in \Theta(g(n))$ means $f(n)$ has the same growth rate as $g(n)$, i. e., $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c'$ for a positive c'
- Example:
 $2n \in \Theta(n) ; 2n \in \Theta(n^2) ; 2n^2 \in \Theta(n) ; 2n^2 \in \Theta(n^2)$

For comparison

$$2n = O(n) ; 2n = O(n^2) ; 2n^2 \neq O(n) ; 2n^2 = O(n^2)$$

Asymptotic notations: Little-Theta?

- Actually, little-theta does not exist. And the previous notation is usually called “Theta” rather than “Big-Theta”
 - Why?

Summary of Asymptotic notations

Asymptotic bounds of $f(n)$	Upper bound	Lower bound
May be tight or not / inclusive	$O(g(n))$	$\Omega(g(n))$
Non-tight / strict	$o(g(n))$	$\omega(g(n))$
Tight	$\Theta(g(n))$	

- These notations are essentially sets of functions
 - Membership test:
 - Finding the appropriate constants
 - Limit definition
-

Summary of Membership Tests for Asymptotic notations

Asymptotic bounds of $f(n)$ & membership test based on limit	Upper bound	Lower bound
May be tight or not / inclusive	$O(g(n))$ $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0 \text{ or } c'$	$\Omega(g(n))$ $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty \text{ or } c'$
Non-tight / strict	$o(g(n))$ $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$	$\omega(g(n))$ $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$
Tight	$\Theta(g(n))$ $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c'$	

c' : Positive constant

Topics

- ✓ What is asymptotic analysis?
 - ✓ The five asymptotic notations
 - Example on Insertion Sort
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 - Using asymptotic notations to analyse algorithms
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Asymptotic Analysis of Algorithms: Back to Insertion Sort Example

- Recall Insertion Sort and its time complexity

InsertionSort(A)

1. for $j = 2$ to $A.length$
2. $Key = A[j]$
3. $i = j - 1$
4. While $i > 0$ and $A[i] > key$
5. $A[i + 1] = A[i]$
6. $i = i - 1$
7. $A[i + 1] = key$

Time complexity of Insertion Sort: $T(n) = Cn^2 + C'n - C''$

Tight asymptotic bound of the running time of Insertion Sort: ?

Topics

- ✓ What is asymptotic analysis?
- ✓ The five asymptotic notations
- ✓ Example on Insertion Sort

To be continued next Monday...

- Additional notes on asymptotic notations
 - Using asymptotic notations to analyse algorithms
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