


CSCE 411
Design and Analysis of Algorithms



Andreas Klappenecker

Goal of this Lecture

- Recall the basic asymptotic notations such as Big Oh, Big Omega, Big Theta, and little oh.
- Recall some basic properties of these notations
- Give some motivation why these notions are defined in the way they are.
- Give some examples of their usage.

Summary

Let $g: \mathbf{N} \rightarrow \mathbf{C}$ be a real or complex valued function on the natural numbers.

$$O(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \exists u > 0 \exists n_0 \in \mathbf{N}$$

$$|f(n)| \leq u|g(n)| \text{ for all } n \geq n_0 \}$$

$$\Omega(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \exists d > 0 \exists n_0 \in \mathbf{N}$$

$$d|g(n)| \leq |f(n)| \text{ for all } n \geq n_0 \}$$

$$\Theta(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \exists u, d > 0 \exists n_0 \in \mathbf{N}$$

$$d|g(n)| \leq |f(n)| \leq u|g(n)| \text{ for all } n \geq n_0 \}$$

Time Complexity

- Estimating the time-complexity of algorithms, we simply want count the number of operations. We want to be
 - independent of the compiler used,
 - ignorant about details about the number of instructions generated per high-level instruction,
 - independent of optimization settings,
 - and architectural details.

This means that performance should only be compared up to multiplication by a constant.

- We want to ignore details such as initial filling the pipeline. Therefore, we need to ignore the irregular

Big Oh



Big Oh Notation

Let $f, g: \mathbb{N} \rightarrow \mathbb{R}$ be function from the natural numbers to the set of real numbers.

We write $f \in O(g)$ if and only if there exists some real number n_0 and a positive real constant u such that

$$|f(n)| \leq u|g(n)|$$

for all n in S satisfying $n \geq n_0$

Big Oh

Let $g: \mathbb{N} \rightarrow \mathbb{C}$ be a function.

Then $O(g)$ is the set of functions

$O(g) = \{ f: \mathbb{N} \rightarrow \mathbb{C} \mid \text{there exists a constant } u \text{ and a natural number } n_0 \text{ such that}$

$|f(n)| \leq u|g(n)| \text{ for all } n \geq n_0 \}$

Notation

We have

$$O(n^2) \subseteq O(n^3)$$

but it is usually written as

$$O(n^2) = O(n^3)$$

This does **not** mean that the sets are equal!!!! The equality sign should be read as 'is a subset of'.

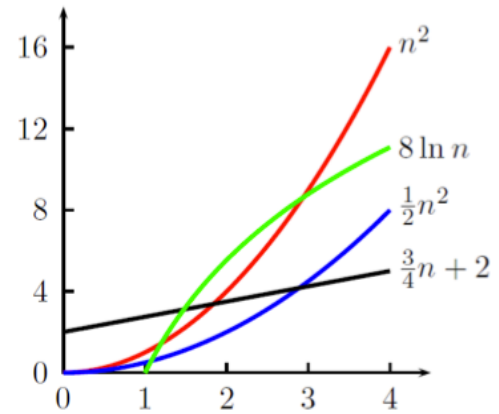
Notation

We write $n^2 = O(n^3)$,
[read as: n^2 is contained in $O(n^3)$]

But we **never** write

$$O(n^3) = n^2$$

Example $O(n^2)$



Big Oh Notation

The Big Oh notation was introduced by the number theorist Paul Bachman in 1894. It perfectly matches our requirements on measuring time complexity.

Example:

$$4n^3+3n^2+6 \text{ in } O(n^3)$$

The biggest advantage of the notation is that complicated expressions can be dramatically simplified.

Quiz



Does $O(1)$ contain only the constant functions?

Tool 1: Limits

Limit

Let (x_n) be a sequence of real numbers.

We say that μ is the **limit** of this sequence of numbers and write

$$\mu = \lim_{n \rightarrow \infty} x_n$$

if and only if for each $\varepsilon > 0$ there exists a natural number n_0 such that $|x_n - \mu| < \varepsilon$ for all $n \geq n_0$

μ? μ!



Limit - Again!

Let (x_n) be a sequence of real numbers.

We say that μ is the **limit** of this sequence of numbers and write

$$\mu = \lim_{n \rightarrow \infty} x_n$$

if and only if for each $\varepsilon > 0$ there exists a natural number n_0 such that $|x_n - \mu| < \varepsilon$

How do we prove that $g = O(f)$?

Lemma 1. *Let f and g be functions from the positive integers to the complex numbers such that $g(n) \neq 0$ for all $n \geq n_0$ for some positive integer n_0 . If the limit $\lim_{n \rightarrow \infty} |f(n)/g(n)|$ exists and is finite then $f(n) = O(g(n))$.*

Proof. If $\lim_{n \rightarrow \infty} |f(n)/g(n)| = C$, then for each $\epsilon > 0$ there exists a positive integer $n_0(\epsilon)$ such that $C - \epsilon \leq |f(n)/g(n)| \leq C + \epsilon$ for all $n \geq n_0$; this shows that $|f(n)| \leq (C + \epsilon)|g(n)|$ for all integers $n \geq n_0(\epsilon)$. It follows that $f(n) = O(g(n))$. \square

Big versus Little Oh

$$O(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \exists u > 0 \exists n_0 \in \mathbf{N} \\ |f(n)| \leq u|g(n)| \text{ for all } n \geq n_0 \}$$
$$o(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \lim_{n \rightarrow \infty} |f(n)|/|g(n)| = 0 \}$$

Quiz



It follows that $o(f)$ is a subset of $O(f)$.

Why?

Quiz



What does $f = o(1)$ mean?

Hint:

$$o(g) = \{ f: \mathbf{N} \rightarrow \mathbf{C} \mid \lim_{n \rightarrow \infty} |f(n)|/|g(n)| = 0 \}$$

Quiz



Some computer scientists consider little oh notations too sloppy.

For example, $1/n+1/n^2$ is $o(1)$
but they might prefer $1/n+1/n^2 = O(1/n)$.

Why is that?

Tool 2: Limit Superior



Limits? There are no Limits!

The limit of a sequence might not exist.

For example, if $f(n) = 1 + (-1)^n$ then

$$\lim_{n \rightarrow \infty} f(n)$$

does not exist.

Least Upper Bound (Supremum)

The **supremum** b of a set of real numbers S is defined as the smallest real number b such that $b \geq x$ for all s in S .

We write $s = \sup S$.

- $\sup \{1, 2, 3\} = 3$,
- $\sup \{x : x^2 < 2\} = \sqrt{2}$,
- $\sup \{(-1)^n - 1/n : n \geq 0\} = 1$.

The Limit Superior

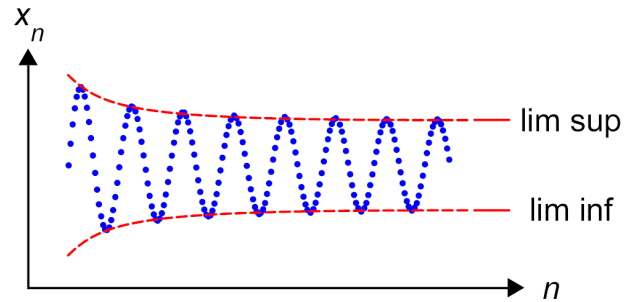
The limit superior of a sequence (x_n) of real numbers is defined as

$$\limsup_{n \rightarrow \infty} x_n = \lim_{n \rightarrow \infty} \left(\sup \{ x_m : m \geq n \} \right)$$

[Note that the limit superior always exists in the extended real line (which includes $\pm\infty$), as $\sup \{ x_m : m \geq n \}$ is a monotonically decreasing function of n and is bounded below by any

The Limit Superior

The limit superior of a sequence of real numbers is equal to the greatest accumulation point of the sequence.



Necessary and Sufficient Condition

Lemma 2. Let f and g be functions from the positive integers to the complex numbers such that $g(n) \neq 0$ for all $n \geq n_0$ for some positive integer n_0 . We have $\limsup_{n \rightarrow \infty} |f(n)/g(n)| < \infty$ if and only if $f(n) = O(g(n))$.

Proof. If $\limsup_{n \rightarrow \infty} |f(n)/g(n)| = C$, then for each $\epsilon > 0$ we have

$$|f(n)|/|g(n)| > C + \epsilon$$

for at most finitely many positive integers; so $|f(n)| \leq (C + \epsilon)|g(n)|$ holds for all integers $n \geq n_0(\epsilon)$ for some positive integer $n_0(\epsilon)$, and this proves that $f(n) = O(g(n))$.

Conversely, if $f(n) = O(g(n))$, then there exists a positive integer n_0 and a constant C such that $g(n) \neq 0$ and $|f(n)|/|g(n)| \leq C$ for all $n \geq n_0$. This implies that $\limsup_{n \rightarrow \infty} |f(n)/g(n)| \leq C$. \square

Big Omega



Big Omega Notation

Let $f, g: \mathbb{N} \rightarrow \mathbb{R}$ be functions from the set of natural numbers to the set of real numbers.

We write $g \in \Omega(f)$ if and only if there exists some real number n_0 and a positive real constant C such that

$$|g(n)| \geq C|f(n)|$$

for all n in \mathbb{N} satisfying $n \geq n_0$.

Big Omega

Theorem: $f \in \Omega(g)$ iff $\liminf_{n \rightarrow \infty} |f(n)/g(n)| > 0$.

Proof: If $\liminf |f(n)/g(n)| = C > 0$, then we have for each $\varepsilon > 0$ at most finitely many positive integers satisfying $|f(n)/g(n)| < C - \varepsilon$. Thus, there exists an n_0 such that

$$|f(n)| \geq (C - \varepsilon)|g(n)|$$

holds for all $n \geq n_0$, proving that $f \in \Omega(g)$.

The converse follows from the definitions.

Big Theta



Big Theta Notation

Let S be a subset of the real numbers (for instance, we can choose S to be the set of natural numbers).

If f and g are functions from S to the real numbers, then we write $g \in \Theta(f)$ if and only if

there exists some real number n_0 and positive real constants C and C' such that

$$C|f(n)| \leq |g(n)| \leq C'|f(n)|$$

for all n in S satisfying $n \geq n_0$.

Thus, $\Theta(f) = O(f) \cap \Omega(f)$

Examples



Sums

- $1+2+3+\dots+n = n(n+1)/2$
- $1^2 + 2^2 + 3^2 + \dots + n^2 = n(n+1)(2n+1)/6$

We might prefer some simpler formula, especially when looking at sum of cubes, etc.

The first sum is approximately equal to $n^2/2$, as $n/2$ is much smaller compared to $n^2/2$ for large n . The second sum is approximately equal to $n^3/3$ plus smaller terms.

Approximate Formulas

(complicated function of n)

= (simple function of n)

+ (bound for the size of the error in
terms of n)

Approximate Formulas

Instead of

$$1^2 + 2^2 + 3^2 + \dots + n^2 = n^3/3 + n^2/2 + n/6$$

we might write

$$1^2 + 2^2 + 3^2 + \dots + n^2 = n^3/3 + O(n^2)$$

Approximate Formulas

If we write $f(n) = g(n) + O(h(n))$, then this means that there exists a constant $u > 0$ and a natural number n_0 such that

$$|f(n) - g(n)| \leq u|h(n)|$$

for all $n \geq n_0$.

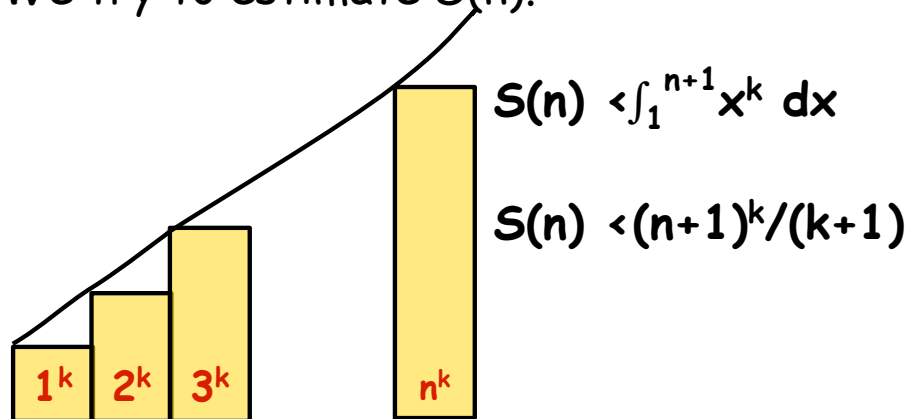
Bold Conjecture

$$1^k + 2^k + 3^k + \dots + n^k = \frac{n^{k+1}}{k+1} + O(n^k)$$

Proof

Write $S(n) = 1^k + 2^k + 3^k + \dots + n^k$

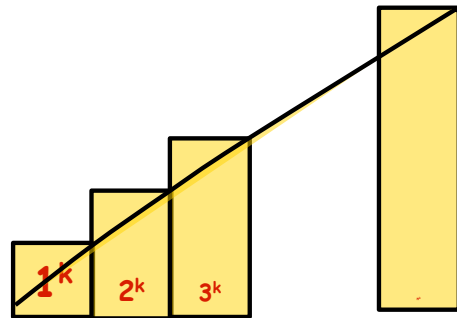
We try to estimate $S(n)$.



Proof

Write $S(n) = 1^k + 2^k + 3^k + \dots + n^k$

We try to estimate $S(n)$.



$$S(n) > \int_0^n x^k dx$$

$$S(n) > n^{k+1}/(k+1)$$

Proof

We have shown that

$$n^{k+1}/(k+1) < 1^k + 2^k + 3^k + \dots + n^k < (n+1)^{k+1}/(k+1).$$

Let's subtract $n^{k+1}/(k+1)$ to get

$$\begin{aligned} 0 &< 1^k + 2^k + 3^k + \dots + n^k - n^{k+1}/(k+1) \\ &< ((n+1)^{k+1} - n^{k+1}) / (k+1) \end{aligned}$$

Proof

$$((n+1)^{k+1} - n^{k+1}) / (k+1) = (k+1)^{-1} \sum_{i=0}^{k+1} C(k+1, i) n^i - n^{k+1}$$

$$\leq \sum_{i=0}^k C(k+1, i) n^k$$

// simplify and enlarge terms

$$\leq (\text{some mess involving } k) n^k$$

It follows that

$$S(n) - n^{k+1}/(k+1) < (\text{mess involving } k) n^k$$

End of Proof

Since the mess involving k does not depend on n , we have proven that

$$1^k + 2^k + 3^k + \dots + n^k = n^{k+1}/(k+1) + O(n^k)$$

holds!

Harmonic Number

The Harmonic number H_n is defined as

$$H_n = 1 + 1/2 + 1/3 + \dots + 1/n.$$

We have

$$H_n = \ln n + \gamma + O(1/n)$$

where γ is the Euler-Mascheroni constant

$$\gamma = \lim_{n \rightarrow \infty} \left[\left(\sum_{k=1}^n \frac{1}{k} \right) - \ln(n) \right] = \int_1^{\infty} \left(\frac{1}{[x]} - \frac{1}{x} \right) dx. = 0.577\dots$$

log n!

Recall that $1! = 1$ and $n! = (n-1)! n$.

Theorem: $\log n! = \Theta(n \log n)$

Proof:

$$\log n! = \log 1 + \log 2 + \dots + \log n$$

$$\leq \log n + \log n + \dots + \log n = n \log n$$

Hence, $\log n! = O(n \log n)$.

log n!

On the other hand,

$$\log n! = \log 1 + \log 2 + \dots + \log n$$

$$\geq \log (\lfloor (n+1)/2 \rfloor) + \dots + \log n$$

$$\geq (\lfloor (n+1)/2 \rfloor) \log (\lfloor (n+1)/2 \rfloor)$$

$$\geq n/2 \log(n/2)$$

$$= \Omega(n \log n)$$

For the last step, note that

$$\liminf_{n \rightarrow \infty} (n/2 \log(n/2)) / (n \log n) = \frac{1}{2}.$$

Reading Assignment

- Read Chapter 1-3 in [CLRS]
- Chapter 1 introduces the notion of an algorithm
- Chapter 2 analyzes some sorting algorithms
- Chapter 3 introduces Big Oh notation