

# Design & Analysis of Algorithms COMP 482 / ELEC 420



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## Math Background: Review & Beyond

1. Asymptotic notation
2. Math used in asymptotics
3. Recurrences
4. Probabilistic analysis

To do:  
[CLRS] 2, 3.1  
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## Asymptotes

Want to describe an algorithm's running time.  
(or space, ...)

Described by  $O(n^2)$ ,  $\Omega(n \log n)$ ,  $\Theta(n)$ , ...



What do these mean?  
Why do we use these notations?



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## Upper Bounds

Creating an algorithm proves we can solve the problem within a given bound.

But another algorithm might be faster.

E.g., sorting an array.  
Insertion sort  $\rightarrow O(n^2)$

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## Lower Bounds

Sometimes can prove that we cannot compute something without a sufficient amount of time.

That doesn't necessarily mean we know how to compute it in this lower bound.

E.g., sorting an array.  
# comparisons needed in worst case  $\rightarrow \Omega(n \log n)$

Will prove this soon...

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## Upper & Lower Bounds: Informal Summary

Upper bounds:

$$\leq O() \quad < o()$$

Lower bounds:

$$\geq \Omega() \quad > \omega()$$

Upper & lower ("tight") bounds:

$$= \Theta()$$

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### Definitions: O, Ω

$T(n) \in O(g(n))$

↔

∃ constants  $c, n_0 > 0$   
such that

∀  $n \geq n_0, T(n) \leq c \times g(n)$

$T(n) \in \Omega(g'(n))$

↔

∃ constants  $c', n'_0 > 0$   
such that

∀  $n \geq n'_0, T(n) \geq c' \times g'(n)$

$n_0 \quad n'_0$

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### Bounding allows abstractions of details

$2n+13 \in O(?)$	$O(n)$	Also, $O(n^2), O(5n), \dots$ Can always weaken the bound.
$2n+13 \in \Omega(?)$	$\Omega(n)$ , also $\Omega(\log n), \Omega(1), \dots$	
$2^n \in O(n) ? \Omega(n) ?$		Given a $c, 2^n \geq c \times n$ , for all but small $n$ . $\Omega(n)$ , not $O(n)$ .
$n^{\log n} \in O(n^5) ?$		No. Given a $c, \log n \geq c \times 5$ , for all large enough $n$ . Thus, $\Omega(n^5)$ .

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### Definitions: o, ω

Might know that our upper & lower bounds aren't tight.

$T(n) \in o(g(n))$

↔

∀ constants  $c > 0, \exists$  constant  $n_0 > 0$ ,  
such that

∀  $n \geq n_0, T(n) < c \times g(n)$

Also,  $T(n) \in o(g(n))$

↔

$\lim_{n \rightarrow \infty} \frac{T(n)}{g(n)} = 0$

Never equal, for any constant.

$T(n) \in \omega(g'(n))$

↔

∀ constants  $c' > 0, \exists$  constant  $n'_0 > 0$ ,  
such that

∀  $n \geq n'_0, T(n) > c' \times g'(n)$

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### Examples: o, ω

$2n+13 \in o(n) ? \omega(n) ?$	No to both!
	$2n+13 < c \times n$ fails for many $c$ .
	$2n+13 > c \times n$ fails for many $c$ .
$2n+13 \in o(?) \omega(?)$	$o(n \log n), o(n^2), \dots$ $\omega(\log n), \omega(1), \dots$
$1 / \log n \in o(1) ?$	Yes.
	$\lim_{n \rightarrow \infty} \frac{T(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{1/\log n}{1} = \lim_{n \rightarrow \infty} \frac{1}{\log n} = 0$

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### Definition: Θ

$T(n) \in \Theta(g(n))$

↔

$T(n) \in O(g(n))$  and  $T(n) \in \Omega(g(n))$

Ideally, find algorithms that are asymptotically as good as possible.

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### Standard notational abuse

$O(), \Omega(), \dots$  are sets of functions.

$T(n) = O(\dots)$	=	$T(n) \in O(\dots)$
$T(n) = f(n) + O(\dots)$	=	$T(n) = f(n) + g(n)$ $\exists g(n) \in O(\dots)$

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### Three types of measures

Most common.

Worst case  $T(n) = \max \{T(x) \mid x \text{ is an instance of size } n\}$

Best case  $T(n) = \min \{T(x) \mid x \text{ is an instance of size } n\}$

Average case  $T(n) = \sum_{|x|=n} \Pr\{x\} \times T(x)$

Determining the input probability distribution can be difficult.