

*This document contains slides from the lecture, formatted to be suitable for printing or individual reading, and with some supplemental explanations added. It is intended as a supplement to, rather than a replacement for, the lectures themselves — you should not expect the notes to be self-contained or complete on their own.*

## 1 Asymptotic notations

CLRS 3

In the analysis of algorithms, we are usually interested in how the performance of our algorithm changes as the problem size increases.

The primary tools for measuring the growth rate of a function that describes the run time of an algorithm are the **asymptotic notations**.

This provides a way of studying the algorithms themselves, independent of any specific hardware, operating system, compiler, programmer, etc.

### 2 Big-O

$O(g(n)) = \{f(n) \mid \text{there exist positive constants } c \text{ and } n_0, \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0. \}$

### 3 Big-Ω

$\Omega(g(n)) = \{f(n) \mid \text{there exist positive constants } c \text{ and } n_0, \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0. \}$

### 4 Big-Θ

$\Theta(g(n)) = \{f(n) \mid \text{there exist positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } 0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \text{ for all } n \geq n_0. \}$

More succinctly:  $\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$

## 5 Anonymous functions

These notations officially refer to sets of functions, but it's often useful to use them in larger arithmetic expressions.

$T(n) = O(n^2)$  means:

$$T(n) \in O(n^2)$$

$T(n) = 2n^2 + O(n)$  means:

There exists  $f(n) \in O(n)$  with  $T(n) = 2n^2 + f(n)$ .

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## 6 Little-o

$o(g(n)) = \{f(n) \mid \text{for any positive constant } c,$   
there exists a constant  $n_0$  such that  $0 \leq f(n) \leq cg(n)$   
for all  $n \geq n_0. \}$

This indicates a loose bound:  $f(n) = o(g(n))$  means  $f(n)$  grows strictly slower than  $g(n)$ .

## 7 Little- $\omega$

$\omega(g(n)) = \{f(n) \mid \text{for any positive constant } c,$   
there exists a constant  $n_0$  such that  $0 \leq cg(n) \leq f(n)$   
for all  $n \geq n_0. \}$

This indicates a loose bound:  $f(n) = \omega(g(n))$  means  $f(n)$  grows strictly faster than  $g(n)$ .

## 8 Little- $\theta$ ?

## 9 Limits

For functions that are eventually positive, we can compare asymptotic growth rates using limits.

Let  $L = \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ , if that limit exists.

Then  $f(n)$  is in ...

$L$	$\omega(g(n))?$	$\Omega(g(n))?$	$\Theta(g(n))?$	$O(g(n))?$	$o(g(n))?$
0				<b>X</b>	<b>X</b>
$(0, \infty)$		<b>X</b>	<b>X</b>	<b>X</b>	
$\infty$	<b>X</b>	<b>X</b>			

## 10 Informal summary of intuition

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