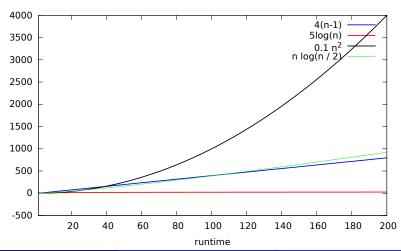
# Defining O-notation (recap) COMS20010 2020, Video lecture 1-3

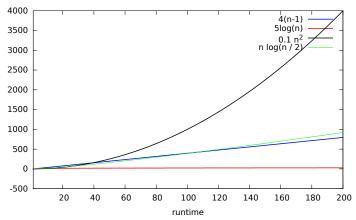
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## Why O-notation?

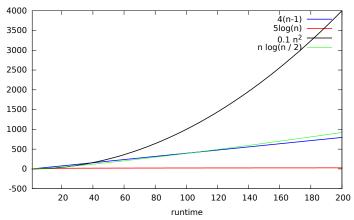
**Intuition:** As input sizes get large, asymptotic growth rate matters more than constant factors. Also, constant factors are implementation-dependent. So we focus on growth rate.



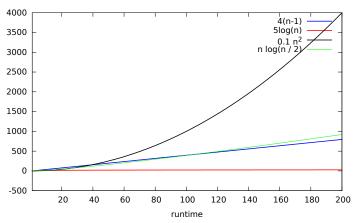
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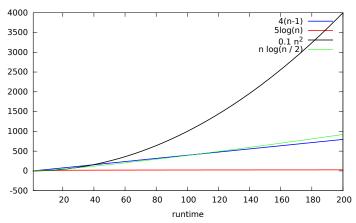
• f(n) "grows no faster than" g(n), ignoring constant factors.



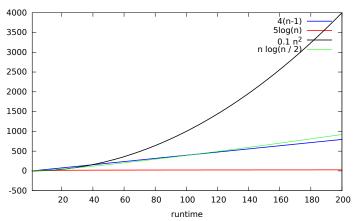
- f(n) "grows no faster than" g(n), ignoring constant factors.
- There exists C > 0 such that f(n) "grows no faster than"  $C \cdot g(n)$ .



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- There exists C > 0 such that  $f(n) \le C \cdot g(n)$  whenever n is sufficiently large.



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- There exist C,  $n_0 > 0$  such that  $f(n) \le C \cdot g(n)$  whenever  $n \ge n_0$ .



There exist  $C, n_0 > 0$  such that  $f(n) \leq C \cdot g(n)$  whenever  $n \geq n_0$ .

This rigorous definition is "just" a more precise version of our intuition.

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#### Other O-notation

 $f(n) \in O(g(n))$  is good notation for "f grows no faster than g, ignoring constants". But what if we want to say "g grows no slower than f"?

Notation	Intuitive meaning	Analogue
$f(n) \in O(g(n))$	f grows at most as fast as $g$	<u> </u>
$f(n) \in \Omega(g(n))$	f grows at least as fast as $g$	$\geq$
$f(n) \in \Theta(g(n))$	f at the same rate as $g$	=
$f(n) \in o(g(n))$	f grows strictly less fast than $g$	<
$f(n) \in \omega(g(n))$	f grows strictly faster than $g$	>

Notation	Formal definition
$f(n) \in O(g(n))$	$\exists C, n_0 \colon \forall n \geq n_0 \colon f(n) \leq C \cdot g(n)$
$f(n) \in \Omega(g(n))$	$\exists c, n_0 \colon \forall n \geq n_0 \colon f(n) \geq c \cdot g(n)$
$f(n) \in \Theta(g(n))$	$\exists c, C, n_0 : \forall n \geq n_0 : c \cdot g(n) \leq f(n) \leq C \cdot g(n)$
$f(n) \in o(g(n))$	$\forall C : \exists n_0 : \forall n \geq n_0 : f(n) \leq C \cdot g(n)$
$f(n) \in \omega(g(n))$	$\forall c \colon \exists n_0 \colon \forall n \geq n_0 \colon f(n) \geq c \cdot g(n)$

#### **Examples**

**Example 1:** Prove  $n^2 - 5n + 12 \in \Theta(n^2)$  directly from the definition.

Remember the definition: proving  $n^2 - 5n + 12 \in \Theta(n^2)$  means proving there exist c, C and  $n_0$  such that  $cn^2 \le n^2 - 5n + 12 \le Cn^2$  for all  $n \ge n_0$ .

We expect  $n^2-5n+12\approx n^2$  for large n, so we could e.g. set c=1/2 and C=2 and solve the quadratic. But let's be lazy! No need to optimise. We have

$$n^{2} - 5n + 12 \le n^{2} + 12 = n^{2} \left(1 + \frac{12}{n^{2}}\right),$$
  

$$n^{2} - 5n + 12 \ge n^{2} - 5n = n^{2} \left(1 - \frac{5}{n}\right).$$

Looking at it like this, it's much easier to see that

$$n^2 - 5n + 12 \le 13n^2$$
 for all  $n \ge 1$ ,  
 $n^2 - 5n + 12 \ge n^2/2$  for all  $n \ge 10$  (so  $\frac{5}{n} \le \frac{1}{2}$ ).

So we prove  $n^2 - 5n + 12 \in \Theta(n^2)$  by taking  $c = \frac{1}{2}$ , C = 13, and  $n_0 = 10$ .

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### Examples

**Example 2:** Prove  $n! \in \omega(2^n)$  directly from the definition.

Remember the definition: proving  $n! \in \omega(2^n)$  means proving that for all c > 0, there exists  $n_0$  such that for all  $n \ge n_0$ ,  $n! \ge c \cdot 2^n$ .

So we're given a constant c, and we need to show  $n! \ge c \cdot 2^n$  when n is sufficiently large. Remember we have

$$n! = \underbrace{n \cdot (n-1) \cdot \dots \cdot 1}_{n \text{ terms}}, \qquad 2^n = \underbrace{2 \cdot 2 \cdot \dots \cdot 2}_{n \text{ terms}}.$$

So we have a lot of wiggle room to bound things term-by-term.

Let's use the fact that  $n! \ge 4^{n-3} = 2^n \cdot 2^{n-6}$ .

Thus  $n! \ge c \cdot 2^n$  whenever  $2^{n-6} \ge c$ , i.e. whenever  $n \ge \log c + 6$ .

So we prove  $n! = \omega(2^n)$  by taking  $n_0 \ge \log c + 6$ .

#### Multi-variable O-notation

We will often need O-notation for functions of more than one variable.

For example, an algorithm running on an n-vertex m-edge graph will often have running time depending on both m and n.

What does it mean to say that e.g.  $f(m, n) \in O(mn)$  or  $f(m, n) \in \Theta(m^2 \log n)$ ?

The only difference is that instead of requiring n to be sufficiently large, we require **all** variables to be sufficiently large.

For example,  $f(m, n) \in O(g(m, n))$  when there exist C,  $m_0$  and  $n_0$  such that  $f(m, n) \le C \cdot g(m, n)$  whenever  $m \ge m_0$  and  $n \ge n_0$ .

All the useful properties of single-variable O-notation (see next video!) carry over to multi-variable O-notation, so e.g. if  $f(m,n) \in O(g(m,n))$  and  $f(m,n) \in O(g(m,n))$  then we still have  $f(m,n) \in \Theta(g(m,n))$ .

## An important clarification (added after recording)

O-notation can behave strangely with negative functions.

But we only care about O-notation for running times, which are positive!

So whenever you are asked to prove something general about O-notation in this course, you can assume the functions involved are non-negative.

But logarithms get used to bound running times all the time, and e.g.  $n \log(n/100)$  is negative for small n. Since it's positive for large n, we'd still like to be able to say e.g.  $n \log(n/100) \in \Theta(n \log n)$ .

So the formal requirement is that the functions involved are **eventually non-negative** — that is, before we can say  $f(n) \in O(g(n))$  or similar, we require that  $f(n), g(n) \ge 0$  for all sufficiently large n.

Any fact that holds about O-notation for non-negative functions will also hold for eventually non-negative functions, by taking  $n_0$  large enough that "eventually non-negative" becomes "non-negative".