

COMP 250

Lecture 36

sets of $O()$ functions
rules for big O

big Omega Ω

April 6, 2022

Recall Formal Definition of Big O

Let $t(n)$ and $g(n)$ be two functions, where $n \geq 0$.

We say $t(n)$ is $O(g(n))$ if there exist two positive constants n_0 and c such that, for all $n \geq n_0$,

$$t(n) \leq c g(n).$$

We use functions $g(n)$ below.

Note: The following inequalities hold for n sufficiently large:

$$\underbrace{1 < \log_2 n < n}_{n \geq 3} < \underbrace{n < n \log_2 n < n^2 < n^3 < \dots}_{n \geq 3} < \underbrace{2^n < n!}_{n \geq 4}$$

Thus, we can write big O relationships between them,
e.g. n is $O(n \log_2 n)$

Sets of $O()$ functions

If $t(n)$ is $O(g(n))$, we often write

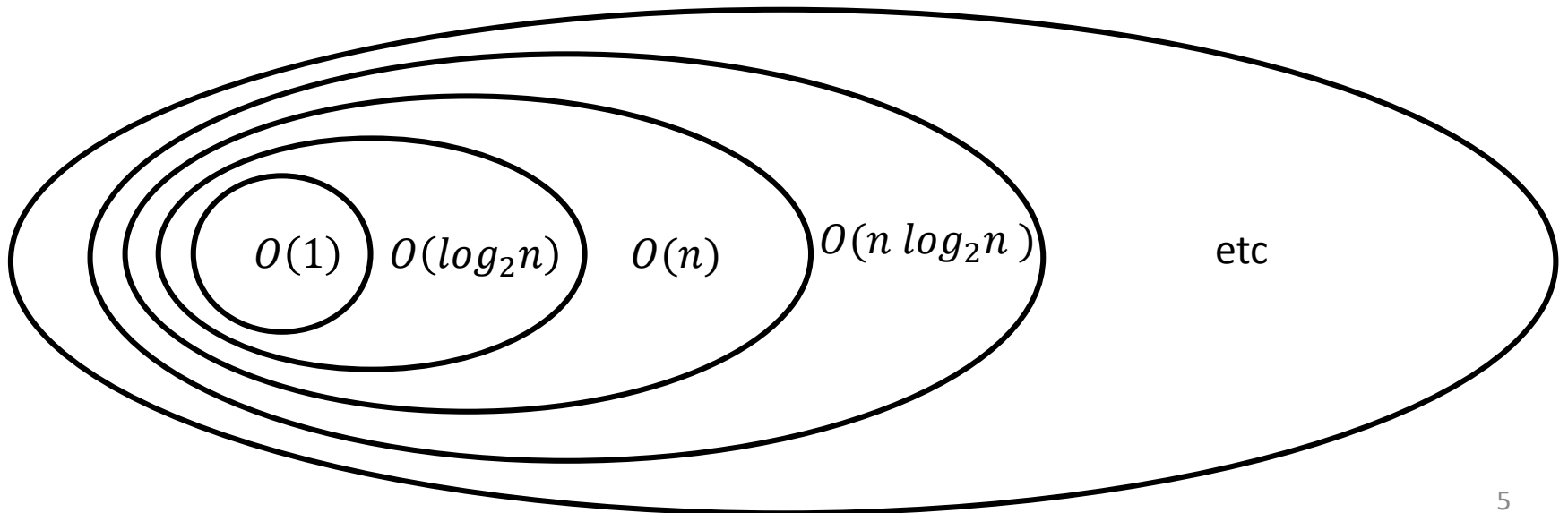
$$t(n) \in O(g(n)).$$

We say:

“ $t(n)$ is a member of the set of functions that are $O(g(n))$.”

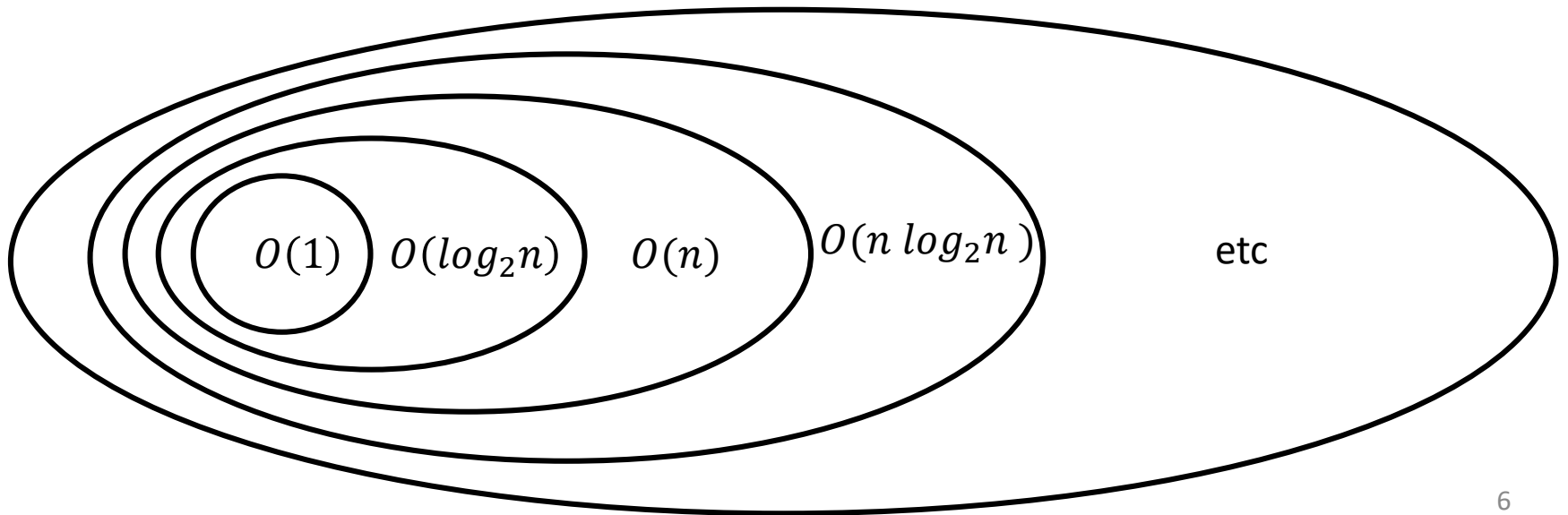
Thus we have the following *strict* subset relationships:

$$O(1) \subset O(\log_2 n) \subset O(n) \subset O(n \log_2 n) \subset O(n^2) \\ \dots \subset O(n^3) \subset \dots \subset O(2^n) \subset O(n!)$$



Tight Bounds

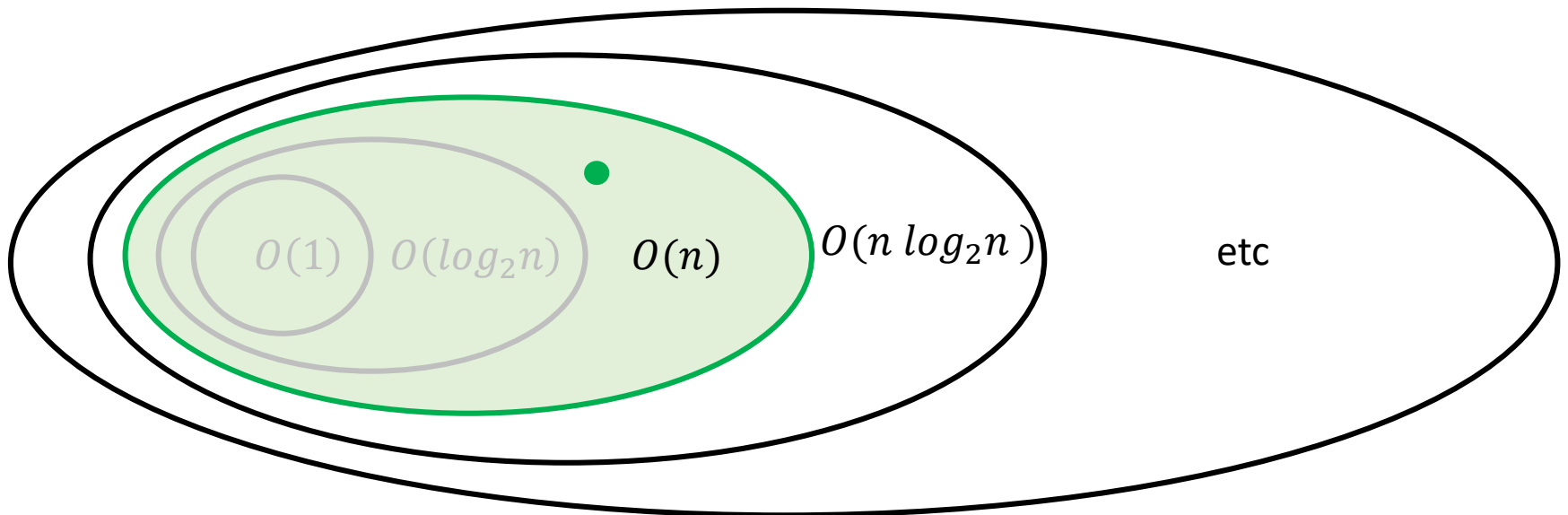
When we say “ $t(n)$ is $O(g(n))$ ”, typically we mean the *smallest set* that $t(n)$ belongs to, i.e. *tight bounds*.



Tight Bounds

When we say “ $t(n)$ is $O(g(n))$ ”, typically we mean the *smallest set* that $t(n)$ belongs to, i.e. *tight bounds*.

For example, if $t(n) = 5n + 7$, then the tight bound is $O(n)$ rather than $O(n \log_2 n)$ or something even larger.



If we have some function $t(n)$ that is defined by a complicated expression, we would like to say “ $t(n)$ is $O(g(n))$ ” where $g(n)$ is a simple function.

e.g. $t(n) = 5n \log_2(n + 3) + 17n + 4$ is $O(n \log_2 n)$.

What are the general rules to justify *using a simple function* ?

Scaling Rule

Suppose $f(n)$ is $O(g(n))$ and let $a > 0$.

Then $a f(n)$ is also $O(g(n))$.

So, multiplying a function by a scale factor doesn't change the big O set(s) that it belongs to.

If you understand the definition of big O, then this rule is obvious. Let's prove it anyhow.

Scaling Rule

By definition, if $f(n)$ is $O(g(n))$ then there exist two positive constants n_0 and c such that, for all $n \geq n_0$,

$$f(n) \leq c g(n).$$

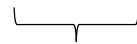
Thus... ?

Scaling Rule

By definition, if $f(n)$ is $O(g(n))$ then there exist two positive constants n_0 and c such that, for all $n \geq n_0$,

$$f(n) \leq c g(n)$$

or equivalently, $a f(n) \leq a c g(n)$ where $a > 0$.



This constant $a c$ satisfies the definition that $a f(n)$ is $O(g(n))$.

Sum Rule

Motivation: When terms are added, we only need to consider the term with the largest big O bound.

For example,

$$\begin{array}{ccc} & \mathbf{3 + 5n \text{ is } O(n)} & \\ & \nearrow \quad \nwarrow \quad \nearrow & \\ O(1) & & O(n) \end{array}$$

Sum Rule

Suppose $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$.

Then $f_1(n) + f_2(n)$ is $O(\max(g_1(n), g_2(n)))$.

Proof: There are constants n_1, c_1 and n_2, c_2 such that

$$f_1(n) \leq c_1 g_1(n) \text{ for all } n \geq n_1$$

$$f_2(n) \leq c_2 g_2(n) \text{ for all } n \geq n_2.$$

Thus, $f_1(n) + f_2(n) \leq (c_1 + c_2) \max(g_1(n), g_2(n))$

for all $n \geq \max(n_1, n_2)$

Product Rule

We want to be able to say, for example,

$$t(n) = (3 + 5n) \log_2(n + 7) \quad \text{is} \quad O(n \log_2 n) .$$

\nearrow \nwarrow
 $O(n)$ $O(\log_2 n)$

i.e. if two functions are multiplied together, then the $O()$ of their product is the product of their $O()$'s.

Product Rule

Suppose $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$.

Then $f_1(n) * f_2(n)$ is $O(g_1(n) * g_2(n))$.

Proof: Let n_1, c_1 and n_2, c_2 be constants such that

$$f_1(n) \leq c_1 g_1(n), \text{ for all } n \geq n_1$$

$$f_2(n) \leq c_2 g_2(n), \text{ for all } n \geq n_2.$$

So, $f_1(n) * f_2(n) \leq c_1 c_2 g_1(n) g_2(n)$
for all $n \geq \max(n_1, n_2)$

It is because of these rules that we can say, for example:

$$t(n) = 5 n \log_2 (n + 3) + 17n + 4 \quad \text{is} \quad O(n \log_2 n).$$

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“small omega” ω

“big omega” Ω

Big Omega (Ω): asymptotic lower bound

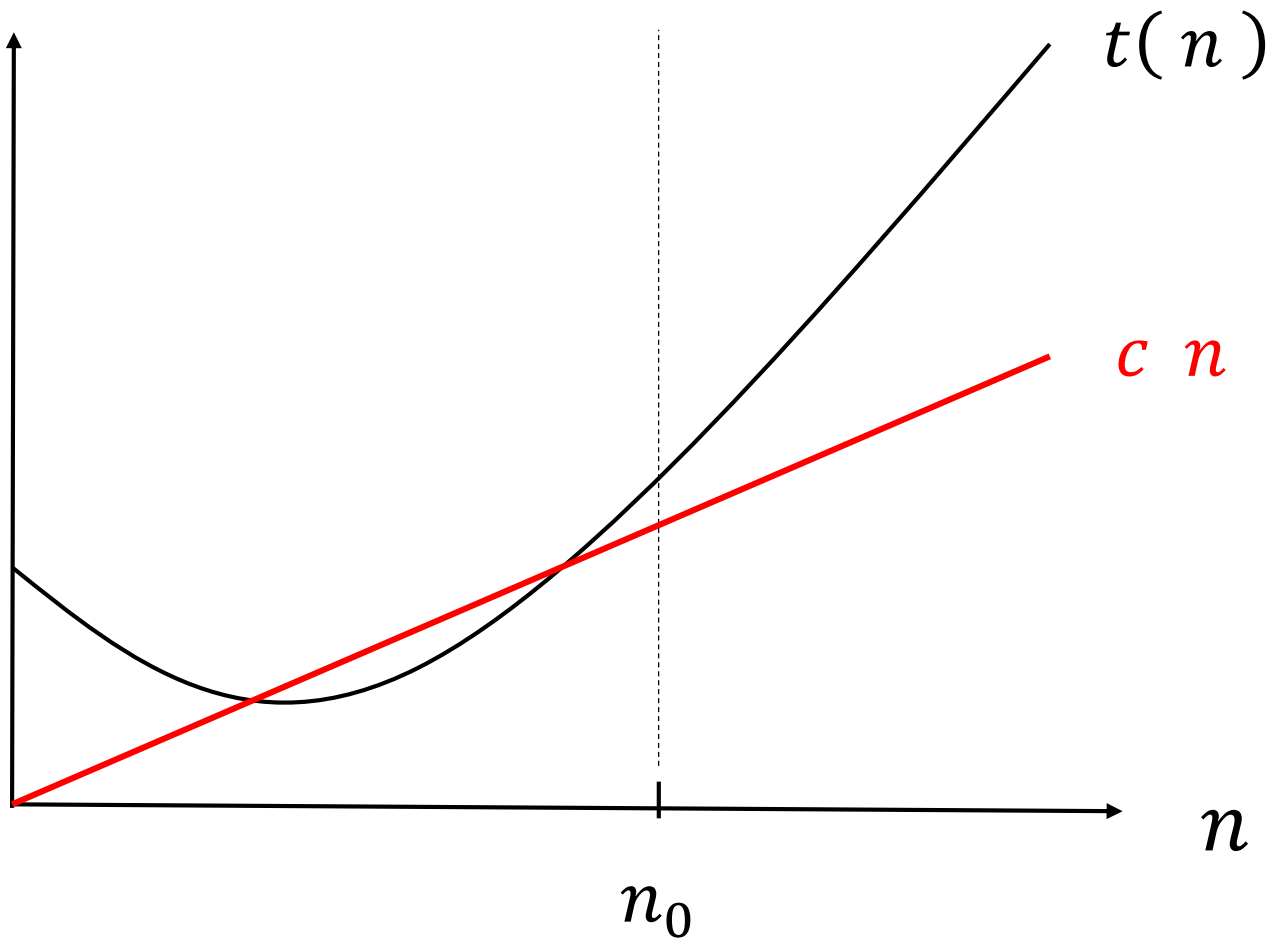
Sometimes we want to say that an algorithm takes *at least* a certain time to run, as a function of the input size n .

Example 1:

Let $t(n)$ be the time it takes for algorithm X to *find the maximum value* in an array of n numbers.

Then $t(n)$ is $\Omega(n)$. (This should be intuitively obvious.)

e.g. $t(n)$ is $\Omega(n)$



Big Omega (Ω): asymptotic lower bound

Example 2: (Comparison based sorting)

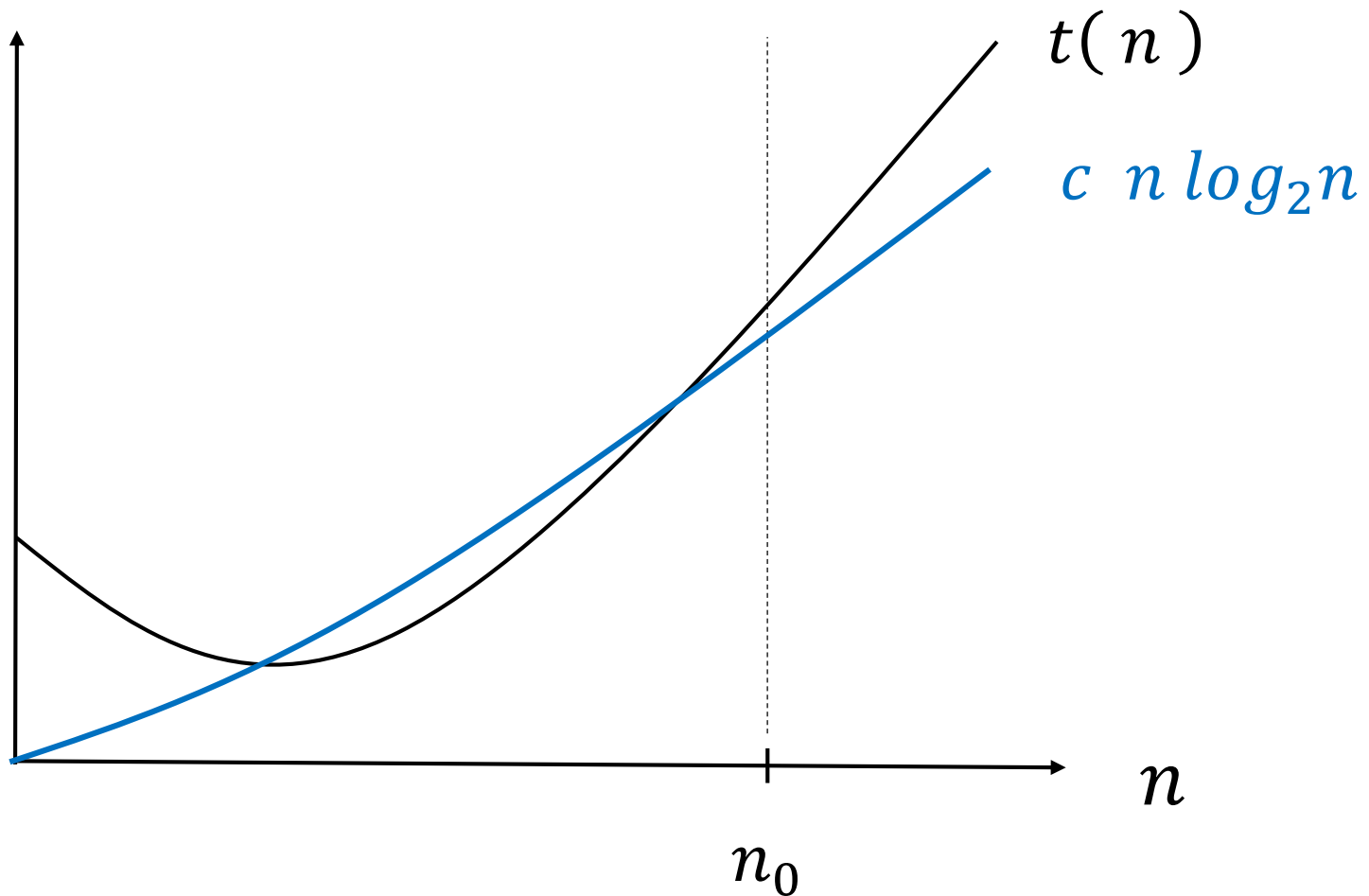
Let $t(n)$ be the number of element comparisons used by some algorithm (X) to sort an array of n numbers.

*One can prove** that $t(n)$ is $\Omega(n \log_2 n)$

That is, no faster comparison-based sorting algorithm is possible than the ones we have seen (e.g. X = merge/heap/quicksort).

*[**Updated after lecture:** Strictly speaking, this is a statement about *on average case*. You will cover this in COMP 251.]

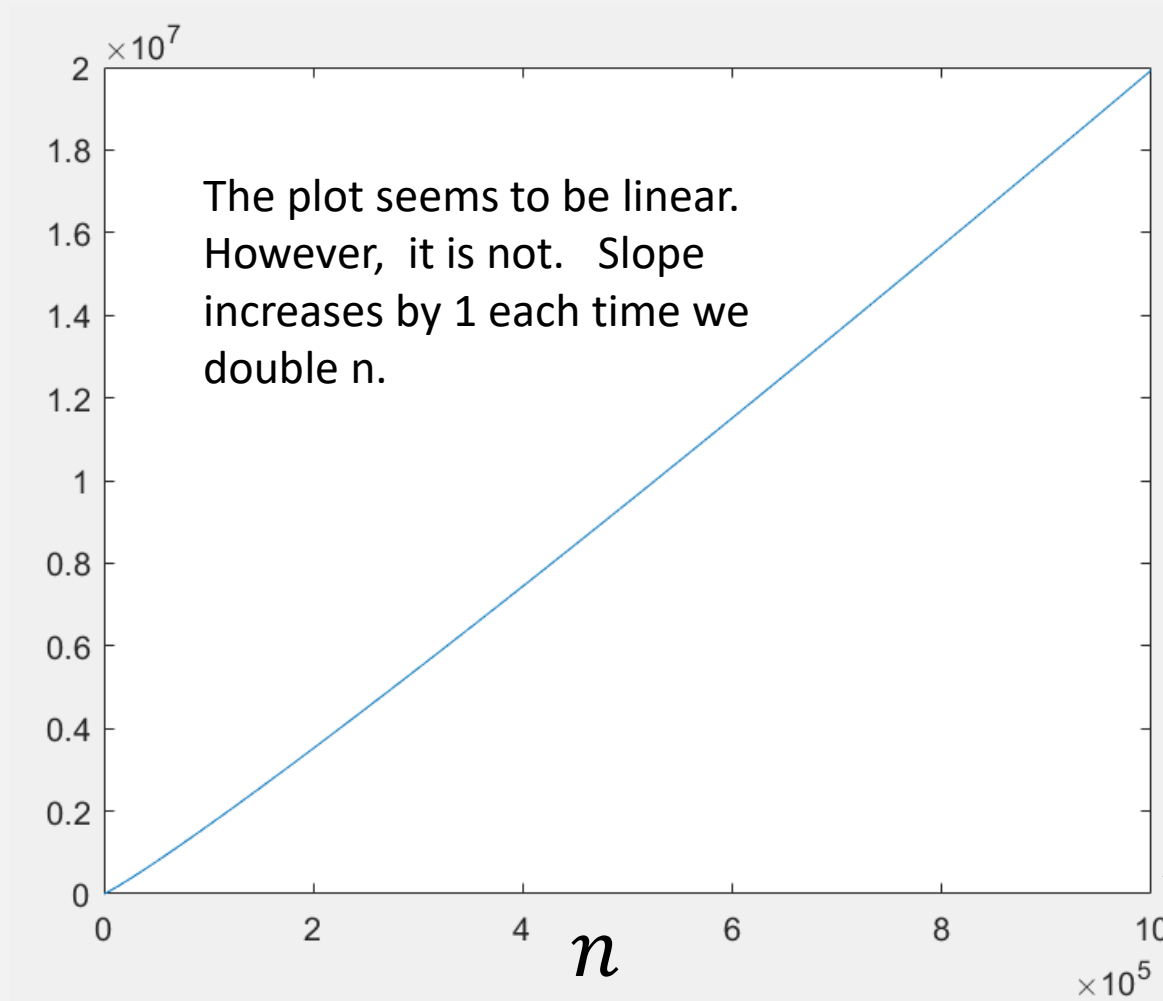
e.g. $t(n)$ is $\Omega(n \log_2 n)$



Plot of $n \log_2 n$ vs. n

$17 * 10^6$

$n \log_2 n$



$10 * 10^5$
 $= 10^6$
 $\approx 2^{17}$

Towards a Formal Definition of Big Ω

Let $t(n)$ and $g(n)$ be two functions, where $n \geq 0$.

We say $t(n)$ is ***asymptotically bounded below*** by $g(n)$ if there exist a constant n_0 such that, for all $n \geq n_0$,

$$t(n) \geq g(n).$$

Note that $g(n)$ here might not be a simple function.

Formal Definition of Big Omega (Ω)

Let $t(n)$ and $g(n)$ be two functions of $n \geq 0$.

We say $t(n)$ is $\Omega(g(n))$ if there exist two positive constants n_0 and c such that, for all $n \geq n_0$,

$$t(n) \geq c g(n).$$

As with big O, having a constant c lets $g(n)$ be a simple function.

Example : $t(n) = \frac{n(n-1)}{2}$. $t(n)$ is $\Omega(n^2)$.

Proof: How to choose c ?

$$\frac{n(n-1)}{2} \geq cn^2 ?$$

Example : $t(n) = \frac{n(n-1)}{2}$. $t(n)$ is $\Omega(n^2)$.

Proof: Try $c = \frac{1}{4}$.

$$\frac{n(n-1)}{2} \geq \frac{n^2}{4}$$

Heads up!

This inequality may be either true or false, depending on n .

$$\Leftrightarrow 2n(n-1) \geq n^2$$

$$\Leftrightarrow n^2 \geq 2n$$

$$\Leftrightarrow n \geq 2. \quad \text{So we can take } n_0 = 2.$$

“ \Leftrightarrow ” means “if and only if” i.e. same true/false value

Example : $t(n) = \frac{n(n-1)}{2}$. $t(n)$ is $\Omega(n^2)$.

Proof (2): Try $c = \frac{1}{3}$

$$\frac{n(n-1)}{2} \geq \frac{n^2}{3}$$

\Leftrightarrow : \leftarrow you can fill this in

$\Leftrightarrow n \geq 3$

So take $n_0 = 3$, $c = \frac{1}{3}$.

Relationship of Big O and Big Omega (Ω)

Let $f(n)$ and $g(n)$ be two functions of $n \geq 0$.

The following are equivalent statements:

$f(n)$ is $O(g(n))$

$g(n)$ is $\Omega(f(n))$.

Why?

$$f(n) < c g(n) \quad \equiv \quad g(n) > \frac{1}{c} f(n)$$

Sets of $\Omega()$ functions

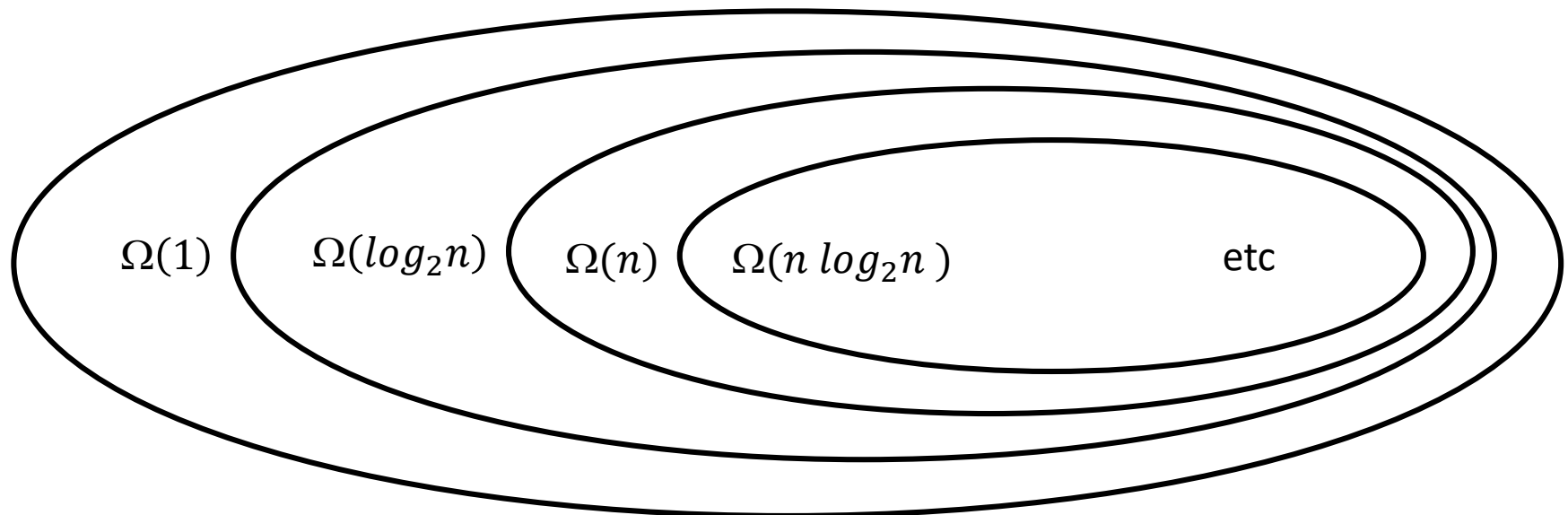
If $t(n)$ is $\Omega(g(n))$, we often write or say:

$$t(n) \in \Omega(g(n)).$$

$t(n)$ is a member of the set of functions that are $\Omega(g(n))$.

As with big O, we have strict subset relationships :

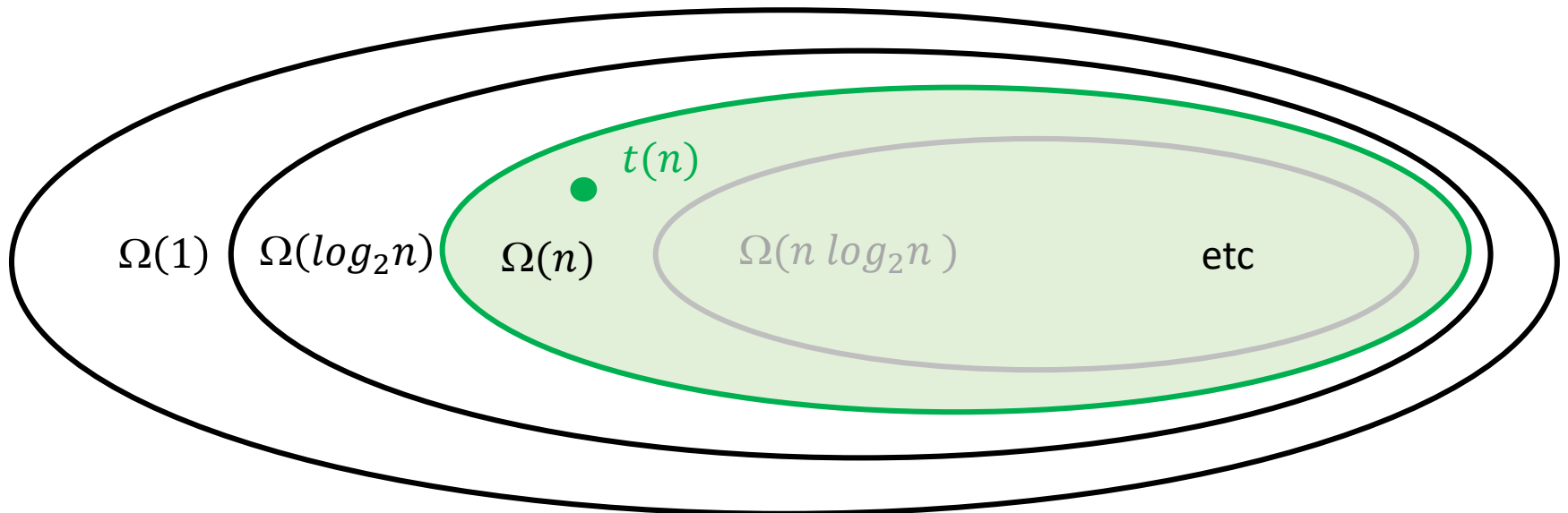
$$\begin{aligned} & \Omega(1) \supset \Omega(\log_2 n) \supset \Omega(n) \supset \Omega(n \log_2 n) \\ & \supset \Omega(n^2) \dots \supset \Omega(n^3) \supset \dots \supset \Omega(2^n) \supset \Omega(n!) \end{aligned}$$



Note the biggest set is now $\Omega(1)$. *e.g.* any positive non-decreasing function $t(n)$ will be bounded below by a constant.

For example, if $t(n)$ belongs to $\Omega(n)$, then $t(n)$ also belongs to $\Omega(\log_2 n)$ and to $\Omega(1)$.

We can again talk about **tight lower bounds**. For example, $\Omega(n)$ is a tight lower bound for $t(n)$ in the example below.
i.e. $\Omega(n)$ is the smallest $\Omega(\)$ set that contains $t(n)$.



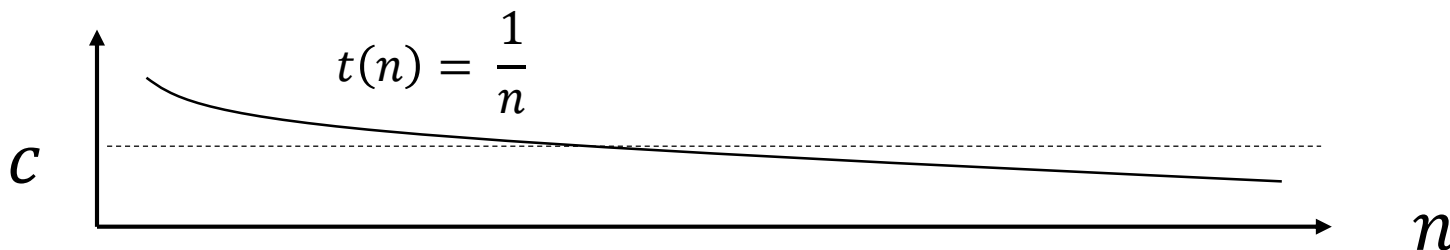
Exercises (see PDF)

Q 12: Let $t(n) = \frac{1}{n}$. Is $t(n) \in \Omega(1)$?

A: No. Apply the definition:

$t(n)$ is $\Omega(1)$ if there exist two constants $n_0 > 0$ and $c > 0$ such that, for all $n \geq n_0$, $t(n) \geq c$.

But this is impossible. See below.



Coming up...

Lectures

Fri : April 8

big Theta, best and worst cases

Mon April 11 (class cancelled)

I will try to have OH before final exam.

Assessments

Assignment 4 due today.

Final Exam Thurs. April 21 (2 weeks)