

Asymptotic order

BCSB-2019

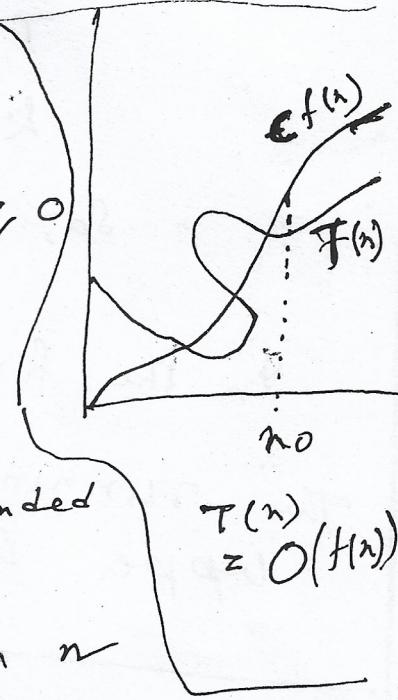
$O \rightarrow \text{Big Oh}$

$\Omega \rightarrow \text{Omega}$

$\Theta \rightarrow \text{Theta}$

Big-Oh: O

Def: $T(n) = O(f(n))$ if there exists Constants $C > 0$ & $n_0 \geq 0$
 so that for all $n \geq n_0$
 we have $T(n) \leq C f(n)$



- $T(n)$ is asymptotically upper bounded by f .
- note that C cannot depend on n

Ex: $T(n) = p n^r + q n + r$ (p, q, r are positive constants)

we can claim that

$$T(n) = O(n^r).$$

proof:

for all $n \geq 1$

$$\text{we have } qn < qn^r$$

$$r \leq rn^r$$

So,

$$\begin{aligned} T(n) &\leq p n^r + q n^r + r n^r \\ &\leq (p + q + r) n^r \end{aligned}$$

$$T(n) = O(n^r), \text{ where } C = p + q + r \\ n_0 = 1$$

It is also correct that

$$T(n) = \Theta(n^3)$$

because

$$T(n) \leq (p+q+r)n^2$$

$$\text{as we have } n^2 \leq n^3$$

$$\text{So, } T(n) \leq (p+q+r)n^3$$

In the first case, $T(n) = O(n^2)$

This running time is the tight upper bound (tight, possible running time).

Omega : Ω

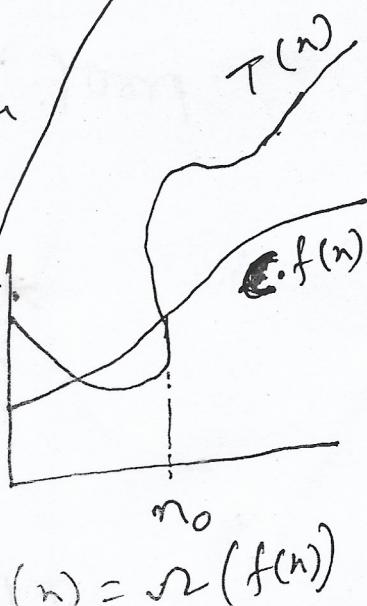
Def: $T(n) = \Omega(f(n))$ if there exist constants $c > 0$ and $n_0 > 0$ so that

for all $n \geq n_0$

$$\text{we have } T(n) \geq c \cdot f(n)$$

- $T(n)$ is asymptotically lower bounded by f .

- c is independent of n
(Bounding from below)



$$\text{Ex. } T(n) = pn^r + qn + r$$

$$> pn^r$$

[we are reducing the size of $T(n)$ until it looks like constant times n^r]

$$T(n) = \underline{\mathcal{O}}(n^r)$$

with

$$c = p > 0$$

$$n_0 = 0$$

In contrast,

to establish the upper bound, we need to involve "inflate" the terms in $T(n)$

Theta : Θ
(Asymptotically tight bound)

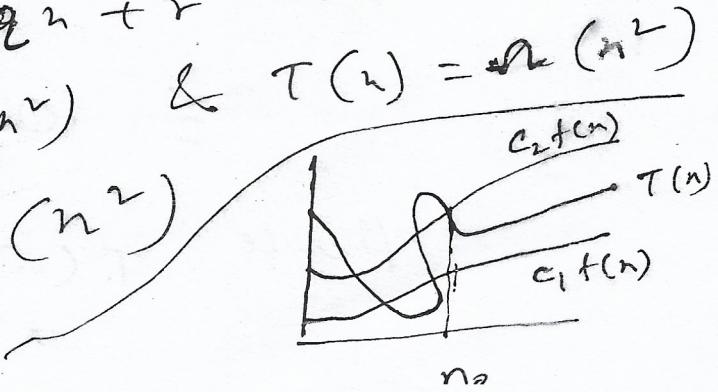
Def: $T(n) = \Theta(f(n))$ if $T(n)$ is both $O(f(n))$ and also $\Omega(f(n))$

Note: $T(n)$ grows exactly like $f(n)$ to within a constant factor

$$\text{Ex: } T(n) = pn^r + qn + r$$

$$T(n) = O(n^r)$$

$$T(n) = \Theta(n^r)$$



Limit Method : Big-O

$T(n) = O(f(n))$ if

$$\lim_{n \rightarrow \infty} \left(\frac{T(n)}{f(n)} \right) = c < \infty$$

Including the case
in which limit is 0

Intuition: $T(n)$ becomes insignificant

relative to $f(n)$ as n

approaches infinity, or

that is, the ratio converges

$\underset{n \rightarrow \infty}{\lim}$ to ~~a~~ positive
constant ~~c~~

as n goes to infinity.

Ex

Say $T(n) = 4 \cdot n^3 + 10 \cdot n^2 + 5 \cdot n + 1$

Consider $f(n) = n^4$

$$\lim_{n \rightarrow \infty} \left(\frac{4 \cdot n^3 + 10 \cdot n^2 + 5 \cdot n + 1}{n^4} \right)$$

$$= \underset{n \rightarrow \infty}{\lim} \left(\frac{4}{n} + \frac{10}{n^2} + \frac{5}{n^3} + \frac{1}{n^4} \right)$$

$$= 0$$

Hence $T(n) = O(n^4)$.

~~limit method~~
 ~~$\sim \Omega(\text{omega})$~~
~~(lower)~~

Limit method:

$\sim \Omega(\text{omega})$
(Lower bound that is not asymptotically tight)

$T(n) = \sim \Omega(f(n))$ if

$$\lim_{n \rightarrow \infty} \left(\frac{T(n)}{f(n)} \right) = c > 0$$

Including the case in which
the limit is ∞

Intuition:

$T(n)$ becomes arbitrarily large relative to $\Theta(f(n))$
or the ratio converges to a positive constant
as n goes to infinity.

Ex.

Say

$$T(n) = 4n^3 + 10n^2 + 5n + 1$$

$$g(n) = n^2$$

$$\lim_{n \rightarrow \infty} \left(\frac{4n^3 + 10n^2 + 5n + 1}{n^2} \right) = \infty$$

Hence $T(n) = \Omega(n^2)$.

Limit method: Θ

Theta (Θ).

(asymptotically tight bound)

$$T(n) = \Theta(f(n)) \text{ if}$$

$$\lim_{n \rightarrow \infty} \left(\frac{T(n)}{f(n)} \right) = c$$

where c is some constant s.t.
 $0 < c < \infty$.

Intuition: The ratio of $T(n)$ and $f(n)$ goes to ~~infinity~~
 converges to a positive constant as n goes to infinity.

When ~~Θ~~ $T(n) = \Theta(f(n))$

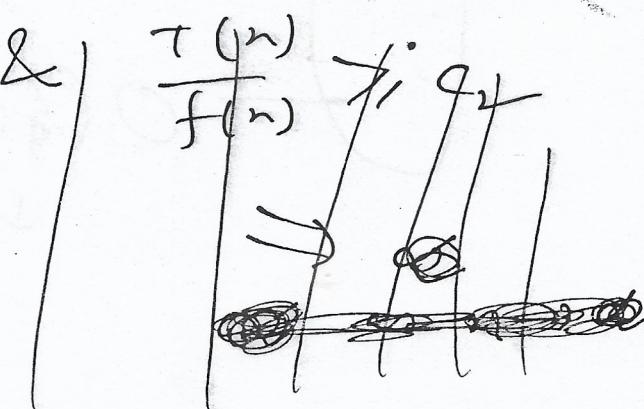
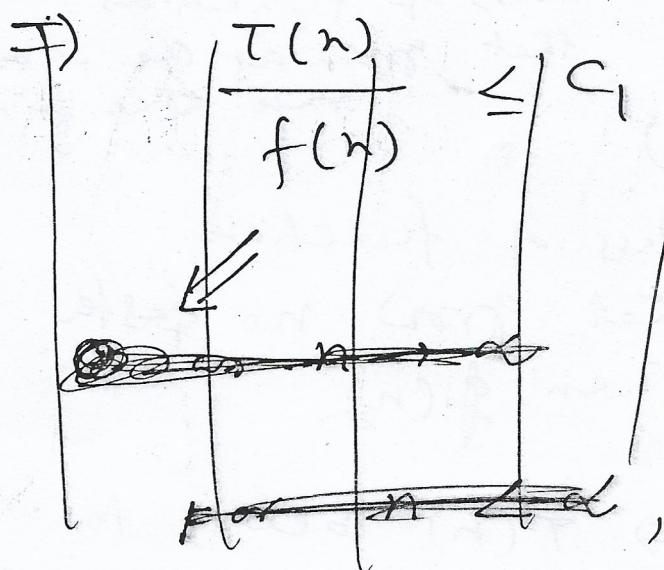
$T(n) = O(f(n))$ and

$T(n) = \Omega(f(n))$

That is, for some $c_1 > 0$ & $c_2 > 0$
 and $n > n_0$

$T(n) \leq c_1 f(n)$ &

$T(n) \geq c_2 f(n) \Rightarrow \underline{c_2 f(n) \leq T(n) \leq c_1 f(n)}$
 $\Rightarrow c_2 f(n) \leq T(n) \leq c_1 f(n)$



$$c_2 f(n) \leq T(n) \leq c_1 f(n)$$

$$c_2 \leq \frac{T(n)}{f(n)} \leq c_1.$$

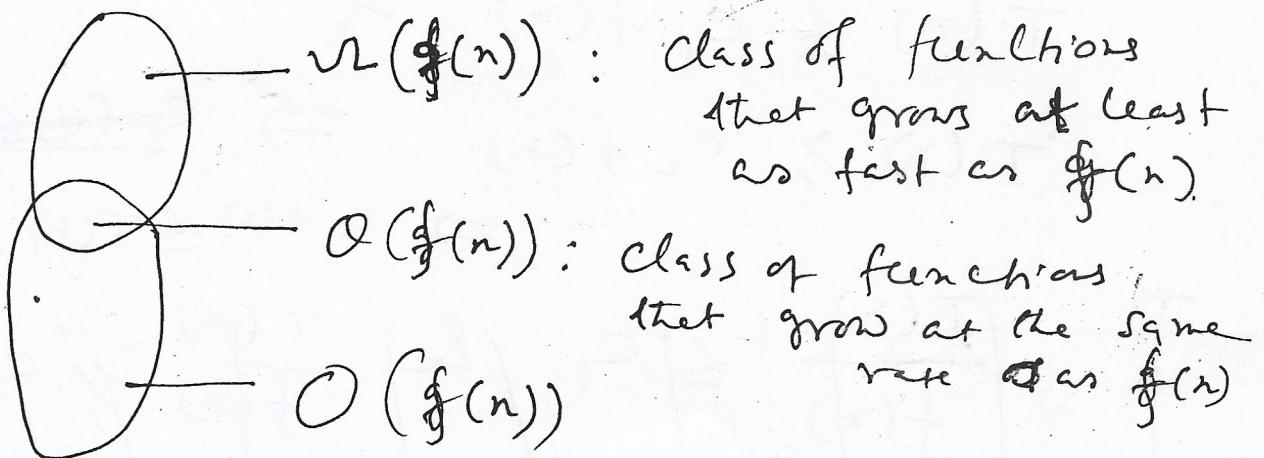
When $n \rightarrow \infty$

$$\frac{T(n)}{f(n)} = c \quad \text{where} \\ c_1 \leq c \leq c_2$$

$T(n)$ grows exactly like $f(n)$

to within a constant factor.

~~$\Theta(f(n))$~~



↓ class of functions
that grow no faster
than $f(n)$.

$T(n) \in O(f(n)) \Rightarrow T(n)$ belongs to
a class of functions that grow at the
same rate as $f(n)$.

$$T(n) = p n^r + q n + r$$

~~$f(n) = n^r$~~ show that

$$T(n) = \Theta(n^r)$$

$$\lim_{n \rightarrow \infty} \left(\frac{T(n)}{n^r} \right)$$

$$= \lim_{n \rightarrow \infty} \left(\frac{p n^r + q n + r}{n^r} \right)$$

$$= \lim_{n \rightarrow \infty} \left(p + \frac{q}{n} + \frac{r}{n^r} \right)$$

$$= p$$

$$\text{Hence } T(n) = \Theta(n^r)$$

proved.

Little oh

$$T(n) = \Theta(f(n))$$

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

Little omega (ω)

$$T(n) = \omega(f(n))$$

$$\text{if } \lim_{n \rightarrow \infty} \frac{T(n)}{f(n)} = \infty$$

Show That

$$\text{for } f_1(n) \in \Theta(g_1(n))$$

$$\text{and } f_2(n) \in \Theta(g_2(n))$$

$$f_1(n) + f_2(n) \in \Theta(\max(g_1(n), g_2(n)))$$

Proof

This
implies
that we
can ignore
lower
order
terms
that is,
 $p(n) = a_n n^k$,
 $p(n) = o(n^m)$

$$f_1(n) \leq c_1 g_1(n) \quad \& \quad f_2(n) \leq c_2 g_2(n)$$

$$\Rightarrow f_1(n) + f_2(n) \leq (c_1 + c_2) (g_1(n) + g_2(n))$$

$$\Rightarrow f_1(n) + f_2(n) \leq 2(c_1 + c_2) \max(g_1(n), g_2(n))$$

$$\Rightarrow f_1(n) + f_2(n) = \Theta(\max(g_1(n), g_2(n)))$$

Asymptotic Proof Techniques

Definitional Proof - Example II

Proof.

- If $n \geq 1$ it is clear that $n \leq n^3$ and $n^2 \leq n^3$.
- Therefore, we have that

$$n^2 + n \leq n^3 + n^3 = 2n^3$$

- Thus, for $n_0 = 1$ and $c = 2$, by the definition of Big-O, we have that $f(n) \in O(g(n))$. \square

Asymptotic Proof Techniques

Definitional Proof - Example III

Example

Let $f(n) = n^3 + 4n^2$ and $g(n) = n^2$. Find a tight bound of the form $f(n) \in \Delta\Omega(g(n))$.

Here, our intuition should tell us that

$$f(n) \in \Omega(g(n))$$

Asymptotic Proof Techniques

Definitional Proof - Example III

Proof.

- If $n \geq 0$ then $n^3 \leq n^3 + 4n^2$
- As before, if $n \geq 1$, $n^2 \leq n^3$
- Thus, when $n \geq 1$, $n^2 \leq n^3 \leq n^3 + 3n^2$
- Thus by the definition of Big- Ω , for $n_0 = 1, c = 1$, we have that $f(n) \in \Omega(g(n))$. \square

Asymptotic Proof Techniques

Trick for polynomial of degree 2

If you have a polynomial of degree 2 such as $an^2 + bn + c$, you can prove it is $\Theta(n^2)$ using the following values:

- $c_1 = \frac{a}{4}$
- $c_2 = \frac{7a}{4}$
- $n_0 = 2 \cdot \max(\frac{|b|}{a}, \sqrt{\frac{|c|}{a}})$

Limit Method

Now try this one:

$$\begin{aligned} f(n) &= n^{50} + 12n^3 \log^4 n - 1243n^{12} + \\ &\quad 245n^6 \log n + 12 \log^3 n - \log n \\ g(n) &= 12n^{50} + 24 \log^{14} n^4 3 - \frac{\log n}{n^5} + 12 \end{aligned}$$

Using the formal definitions can be very tedious especially when one has very complex functions. It is much better to use the *Limit Method* which uses concepts from calculus.

Limit Method Process

Say we have functions $f(n)$ and $g(n)$. We set up a limit quotient between f and g as follows:

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \begin{cases} 0 & \text{then } f(n) \in O(g(n)) \\ c > 0 & \text{then } f(n) \in \Theta(g(n)) \\ \infty & \text{then } f(n) \in \Omega(g(n)) \end{cases}$$

- Justifications for the above can be proven using calculus, but for our purposes the limit method will be sufficient for showing asymptotic inclusions.
- Always try to look for algebraic simplifications first.
- If f and g both diverge or converge on zero or infinity, then you need to apply l'Hôpital's Rule.

for all $k > 0$

Show that

$$(\ln n)^k = O(n^\alpha)$$

limit method

$$\lim_{n \rightarrow \infty} \frac{(\ln n)^k}{n^\alpha}$$

$$= \left(\lim_{n \rightarrow \infty} \frac{\ln n}{n^{\alpha/k}} \right)^k$$

$$= \left(\lim_{n \rightarrow \infty} \frac{1/n}{(\alpha/k) n^{\alpha/k-1}} \right)^k \quad \begin{matrix} (\text{using} \\ \text{L'Hopital} \\ \text{rule}) \end{matrix}$$

$$= \lim_{n \rightarrow \infty} \left(\frac{1}{(\alpha/k) n^{\alpha/k}} \right)^k = 0$$

Hence $(\ln n)^k = O(n^\alpha)$

$$\text{Let } T(n) = 2^n$$

$$f(n) = 3^n$$

Show that

$$T(n) = O(3^n).$$

Limit method.

$$\lim_{n \rightarrow \infty} \frac{T(n)}{f(n)} = \lim_{n \rightarrow \infty} \frac{2^n}{3^n}$$

Using L' Hospital

Rule

$$\lim_{n \rightarrow \infty} \frac{T(n)}{f(n)} = \lim_{n \rightarrow \infty} \frac{2^n}{3^n} = \frac{(\ln 2)^{2^n}}{(\ln 3)^{3^n}}.$$

(still diverse)

Using algebra

$$\lim_{n \rightarrow \infty} \frac{2^n}{3^n} = \lim_{n \rightarrow \infty} \left(\frac{2}{3}\right)^n$$

$$\therefore 0$$

[since $\frac{2}{3} < 1$.
Theorem: $\lim_{n \rightarrow \infty} (d^n) = \begin{cases} 0 & \text{if } d^1 \\ 1 & \text{if } d = 1 \\ \infty & \text{if } d > 1 \end{cases}$]

$$\text{Hence } T(n) = O(3^n).$$

Efficiency Classes

Constant	$\mathcal{O}(1)$
Logarithmic	$\mathcal{O}(\log(n))$
Linear	$\mathcal{O}(n)$
Polylogarithmic	$\mathcal{O}(\log^k(n))$
Quadratic	$\mathcal{O}(n^2)$
Cubic	$\mathcal{O}(n^3)$
Polynomial	$\mathcal{O}(n^k)$ for any $k > 0$
Exponential	$\mathcal{O}(2^n)$
Super-Exponential	$\mathcal{O}(2^{f(n)})$ for $f(n) = n^{(1+\epsilon)}$, $\epsilon > 0$ For example, $n!$

Table: Some Efficiency Classes

Summary

Asymptotics is easy, but remember:

- Always look for algebraic simplifications
- You must always give a rigorous proof
- Using the limit method is always the best
- Always show l'Hôpital's Rule if need be
- Give as simple (and tight) expressions as possible

Say you have computed $T(n)$
for an algorithm and found
that $T(n) = \mathcal{O}(f(n))$
What does it mean actually?

→ $T(n)$ belongs to a class of
functions that grow at
the same rate as $f(n)$.

Similarly
 $T(n) = \mathcal{O}(f(n))$

$T(n)$ belongs to a class of
functions that grow no faster
than $f(n)$

$$T(n) = \omega(f(n))$$

$T(n)$ belongs to a class of functions that grow at least as fast as $f(n)$.

If $T(n) = O(n^2)$

means

$T(n)$ belongs to a class of functions that grow no faster than n^2 .

If $T(n) = \Theta(n^2)$

means

$T(n)$ belongs to a class of functions that grow at the same rate as $\Theta(n^2)$

if $T(n) = \Omega(n^r)$

means

$T(n)$ belongs to a class
of functions that grow
at least as fast as $f(n)$

How to compare two algorithms?

For the same problem. { Algorithm 1 $\rightarrow T(n) = O(n^r)$
Algorithm 2 $\rightarrow T(n) = \Theta(n^r)$

Which one is better?

\rightarrow Algorithm 2 \rightarrow due to

its asymptotically
tight upper bound.

Actually big-O is more
informative than Big- Θ .

But it is harder to prove $T(n)$
is $\Theta(f(n))$ than $O(f(n))$.