Inf 2B: Asymptotic notation

Lecture 2 of ADS thread

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Linear Search in Pseudocode

Input: Integer array A, integer k being searched.
Output: The least index i such that $A[i] = k$; otherwise -1.

Algorithm linSearch(A, k)

1. **for** $i \leftarrow 0$ **to** $A.length - 1$ **do**
2. **if** $A[i] = k$ **then**
3. **return** i
4. **return** -1

Assume each line takes constant time to execute once.
Let c_i be the time for line i. Then

$$(c_1 + c_2)n + \min\{c_3, c_1 + c_4\} \leq T_{\text{linSearch}}(n) \leq (c_1 + c_2)n + \max\{c_3, c_1 + c_4\}.$$

The Big-O Notation

Definition

Let $f, g : \mathbb{N} \to \mathbb{R}$ be functions. We say that f *is* $O(g)$ if there is some $n_0 \in \mathbb{N}$ and some $c > 0$ from $\mathbb{R}$ such that for all $n \geq n_0$ we have

$$0 \leq f(n) \leq c\, g(n).$$

In other words:

$$O(g) = \left\{ f : \mathbb{N} \to \mathbb{R} \mid \text{there is an } n_0 \in \mathbb{N} \text{ and } c > 0 \text{ in } \mathbb{R} \text{ such that } 0 \leq f(n) \leq cg(n), \text{ for all } n \geq n_0. \right\}$$

Then "f is $O(g)$" means $f \in O(g)$.
Informally, we say "*for sufficiently large n*" instead of "there is some $n_0 \in \mathbb{N}$ such that for all $n \geq n_0 \ldots$".

Intention

"f is $O(g)$" tells us that the growth rate of f is *no worse* than that of g. Could be better.

- c allows us to adjust for constants: n^2 obviously has same growth rate as $3n^2$, $20n^2$, $100n^2 \ldots$
  - consider $n \to an$ then

$$\begin{aligned} n^2 &\to a^2 \cdot n^2 \\ 3n^2 &\to a^2 \cdot 3n^2 \\ 20n^2 &\to a^2 \cdot 20n^2 \\ &\vdots \end{aligned}$$

- Positivity condition ($0 \leq f(n) \leq \cdots$) required to ensure some useful properties, always satisfied by runtimes!
- n_0 allows a settling in period of atypical behaviour.
- O allows us to concentrate on the big picture rather than details (many being implementation dependent).

Notational Convention

Write

$$f = O(g),$$

instead of

$$f \in O(g).$$

- Makes it convenient to have chains reasoning with inequalities etc.
- Notation here is from *left to right*. $f = O(g)$ does *not* mean that $O(g) = f$!
- $f = f_1 + O(g) = f_2 + O(g)$ does *not* imply that $f_1 = f_2$.
- Seems strange but easy to get used to it and *very* useful.

Examples of O

1. $3n^3 = O(n^3)$.
   Need c and n_0 so that $3n^3 \leq cn^3$ for all $n \geq n_0$.
   Take $c = 3$, $n_0 = 0$.
2. $3n^3 + 8 = O(n^3)$.
   For a constant $c > 0$ we have

   $$3n^3 + 8 \leq cn^3 \iff 3 + \frac{8}{n^3} \leq c \quad \text{provided } n > 0.$$

   As n increases $8/n^3$ decreases. Thus

   $$3 + \frac{8}{n^3} \leq 11 \quad \text{for all } n > 0.$$

   So we take $c = 11$, $n_0 = 1$.
   We *can* also take $c = 4$, $n_0 = 2$ or $c = 3 + 8/27$, $n_0 = 3$ etc.

More Examples of O

3. $\lg(n) = O(n)$
   Intuitively: $\lg(n) < n$ for all $n \geq 1$.
   Need a proof. Well

   $$\lg(n) < n \iff n < 2^n, \quad \text{for all } n > 0.$$

   Use induction on n to prove rhs.
   - Base case $n = 1$ is clearly true.
   - For induction step assume claim holds for n. Then

     $$2^{n+1} = 2 \cdot 2^n > 2n, \quad \text{by induction hypothesis.}$$

     To complete the proof just need to show that $2n \geq n + 1$.
     Now

     $$2n \geq n + 1 \iff n \geq 1,$$

     and we have finished.
     So we take $c = 1$ and $n_0 = 1$.

More Examples of O

4. $8n^2 + 10n \lg(n) + 100n + 10000 = O(n^2)$.
   We have

   $$\begin{aligned} 8n^2 &+ 10n \lg(n) + 100n + 10000 \\ &< 8n^2 + 10n \cdot n + 100n + 10000, \quad \text{for all } n > 0 \\ &\leq 8n^2 + 10n^2 + 100n^2 + 10000n^2 \\ &= (8 + 10 + 100 + 10000)n^2 \\ &= 10118n^2. \end{aligned}$$

   Thus we can take $n_0 = 1$ and $c = 10118$.
   - Value for c seems rather large.
   - Any $c > 8$ will do, the closer c is to 8 the larger n_0 has to be.
   - For big-O notation, no point at all in expending more effort just to reduce some constant.
5. $2^{100} = O(1)$.
   Take $n_0 = 0$ and $c = 2^{100}$.

"Laws" of Big-O

Theorem: Let $f_1, f_2, g_1, g_2 : \mathbb{N} \to \mathbb{R}$ be functions. Then:

1. For any constant $a > 0$ in $\mathbb{R}$: $f_1 = O(g_1) \implies af_1 \in O(g_1)$.
2. $f_1 = O(g_1)$ and $f_2 = O(g_2) \implies f_1 + f_2 = O(g_1 + g_2)$.
3. $f_1 = O(g_1)$ and $f_2 = O(g_2) \implies f_1 f_2 = O(g_1 g_2)$.
4. $f_1 = O(g_1)$ and $g_1 = O(g_2) \implies f_1 = O(g_2)$.
5. For any $d \in \mathbb{N}$: if f_1 is polynomial of degree d with positive leading coefficient then $f_1 = O(n^d)$.
6. For any *constants* $a > 0$ and $b > 1$ in $\mathbb{R}$: $n^a = O(b^n)$.
7. For any *constant* $a > 0$ in $\mathbb{R}$: $\lg(n^a) = O(\lg(n))$.
8. For any *constants* $a > 0$ and $b > 0$ in $\mathbb{R}$: $\lg^a(n) = O(n^b)$.

Example (using Laws for O)

$$871n^3 + 13n^2 \lg^5(n) + 18n + 566 = O(n^3).$$

$$\begin{aligned}
&871n^3 + 13n^2 \lg^5(n) + 18n + 566 \\
&\quad = 871n^3 + 13n^2 O(n) + 18n + 566 \qquad \text{by (8)} \\
&\quad = 871n^3 + O(n^3) + 18n + 566 \qquad \text{by (3)} \\
&\quad = 871n^3 + 18n + 566 + O(n^3) \\
&\quad = O(n^3) + O(n^3) \qquad \text{by (5)} \\
&\quad = O(n^3) \qquad \text{by (2) \& (1)}
\end{aligned}$$

Big-Ω and Big-Θ

Definition

Let $f, g : \mathbb{N} \to \mathbb{R}$ be functions.

1. We say that f *is* $\Omega(g)$ if there is an $n_0 \in \mathbb{N}$ and $c > 0$ in $\mathbb{R}$ such that for all $n \geq n_0$ we have
$$f(n) \geq c\, g(n) \geq 0.$$
2. We say that f *is* $\Theta(g)$, or f *has the same asymptotic growth rate as* g, if f is $O(g)$ and $\Omega(g)$.

- Have corresponding 'laws of Ω' (see notes).
- $f = \Omega(g) \iff g = O(f)$. [**Prove** this.]

Examples of $f = \Omega(g)$

1. Let $f(n) = 3n^3$ and $g(n) = n^3$.
(combining this with Ex 1. for O gives $3n^3 = \Theta(n^3)$)

   Let $n_0 = 0$ and $c = 1$. Then for all $n \geq n_0$,
$f(n) = 3n^3 \geq cg(n) = g(n)$.
2. Let $f(n) = \lg(n)$ and $g(n) = \lg(n^2)$.

   Well
$$\lg(n^2) = 2\log(n).$$
   So take $c = 1/2$ and $n_0 = 1$. Then for every $n \geq n_0$ we have,
$$f(n) = \lg(n) = \frac{1}{2} 2\lg(n) = \frac{1}{2}\lg(n^2) = \frac{1}{2} g(n).$$

Quick quiz: True or False ?

$$\sqrt{n^3} = O(n^2)?$$

True. $\sqrt{n^3} = n^{3/2} \leq n^2$.

$$2^{\lfloor \lg n \rfloor} = O(n)?$$

True. $2^{\lfloor \lg n \rfloor} \leq 2^{\lg n} = n$.

$$2^{\lfloor \lg n \rfloor} = \Omega(n)?$$

True: Let $n \geq 2$. Then $\lfloor \lg n \rfloor \geq (\lg n) - 1 \geq 0$. Hence $2^{\lfloor \lg n \rfloor} \geq 2^{(\lg n)-1} = n/2$. So take $n_0 = 2, c = 1/2$.

$$n \lg n = \Theta(n^2)?$$

False: We do have $n \lg n = O(n^2)$ but $n \lg n$ is **not** $\Omega(n^2)$.

Further Reading

- Lecture notes 2 (handed out).
- If you have Goodrich & Tamassia [GT]:
  All of the chapter on “Analysis Tools” (especially the “Seven functions” and “Analysis of Algorithms” sections).

  NB: the title of the book is as given in slides of lecture 1, not as in note 1.
- If you have [CLRS]:
  Read chapter 3 on “Growth of Functions”.
- `Wikipedia` has a page about asymptotic notation:
  `en.wikipedia.org/wiki/Asymptotic_notation`