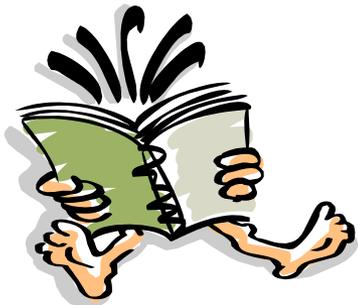


Asymptotic Notation & Master Method



Introduction to different cases of Analysis

- Worst case

- Provides an upper bound on running time
- An absolute **guarantee** that the algorithm would not run longer, no matter what the inputs are

- Best case

- Provides a lower bound on running time
- Input is the one for which the algorithm runs the fastest

$$\textit{Lower Bound} \leq \textit{Running Time} \leq \textit{Upper Bound}$$

- Average case

- Provides a **prediction** about the running time
- Assumes that the input is random

Asymptotic Analysis

- To compare two algorithms with running times $f(n)$ and $g(n)$, we need a **rough measure** that characterizes **how fast each function grows**.
- Hint: use *rate of growth*
- Compare functions in the limit, that is, **asymptotically!**
(i.e., for large values of n)

Rate of Growth

- Consider the example of buying *elephants* and *goldfish*:

Cost: cost_of_elephants + cost_of_goldfish

Cost ~ cost_of_elephants (**approximation**)

- The low order terms in a function are relatively insignificant for **large** n

$$n^4 + 100n^2 + 10n + 50 \sim n^4$$

i.e., we say that $n^4 + 100n^2 + 10n + 50$ and n^4 have the same **rate of growth**

Asymptotic Notation

- O notation: asymptotic “less than”:
 - $f(n)=O(g(n))$ implies: $f(n)$ “ \leq ” $g(n)$
- Ω notation: asymptotic “greater than”:
 - $f(n)=\Omega(g(n))$ implies: $f(n)$ “ \geq ” $g(n)$
- Θ notation: asymptotic “equality”:
 - $f(n)=\Theta(g(n))$ implies: $f(n)$ “ $=$ ” $g(n)$

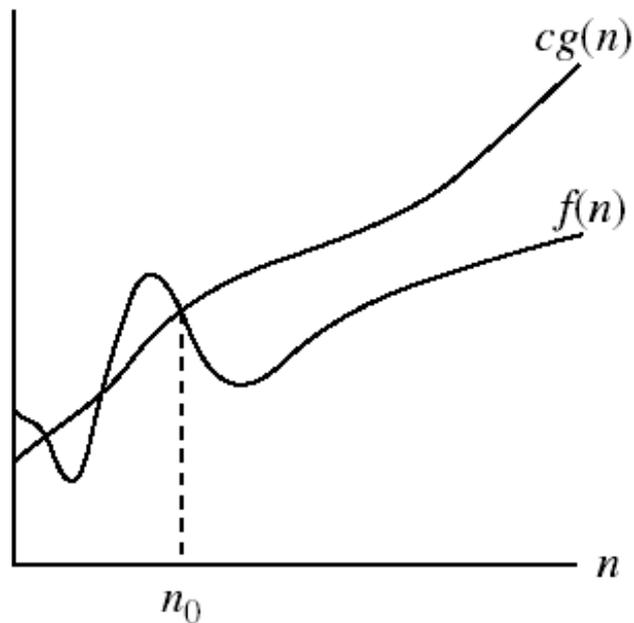
Big-O Notation

- We say $f_A(n)=30n+8$ is *order n* , or $O(n)$. It is, at most, roughly *proportional* to n .
- $f_B(n)=n^2+1$ is *order n^2* , or $O(n^2)$. It is, at most, roughly proportional to n^2 .
- In general, any $O(n^2)$ function is faster-growing than any $O(n)$ function.

Asymptotic notations

- *O-notation*

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



$g(n)$ is an *asymptotic upper bound* for $f(n)$.

Examples

- $2n^2 = O(n^3)$: $2n^2 \leq cn^3 \Rightarrow 2 \leq cn \Rightarrow c = 1$ and $n_0 = 2$

- $n^2 = O(n^2)$: $n^2 \leq cn^2 \Rightarrow c \geq 1 \Rightarrow c = 1$ and $n_0 = 1$

- $1000n^2 + 1000n = O(n^2)$:

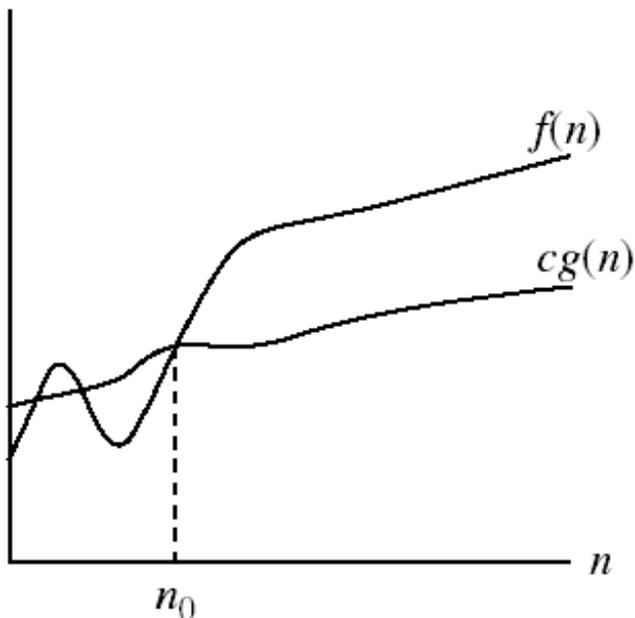
$$1000n^2 + 1000n \leq 1000n^2 + n^2 = 1001n^2 \Rightarrow c = 1001 \text{ and } n_0 = 1000$$

- $n = O(n^2)$: $n \leq cn^2 \Rightarrow cn \geq 1 \Rightarrow c = 1$ and $n_0 = 1$

Asymptotic notations (cont.)

- Ω - notation

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



$\Omega(g(n))$ is the set of functions with larger or same order of growth as $g(n)$

$g(n)$ is an *asymptotic lower bound* for $f(n)$.

Examples

- $5n^2 = \Omega(n)$

$\exists c, n_0$ such that: $0 \leq cn \leq 5n^2 \Rightarrow cn \leq 5n^2 \Rightarrow c = 1$ and $n_0 = 1$

- $100n + 5 \neq \Omega(n^2)$

$\exists c, n_0$ such that: $0 \leq cn^2 \leq 100n + 5$

$100n + 5 \leq 100n + 5n \ (\forall n \geq 1) = 105n$

$cn^2 \leq 105n \Rightarrow n(cn - 105) \leq 0$

Since n is positive $\Rightarrow cn - 105 \leq 0 \Rightarrow n \leq 105/c$

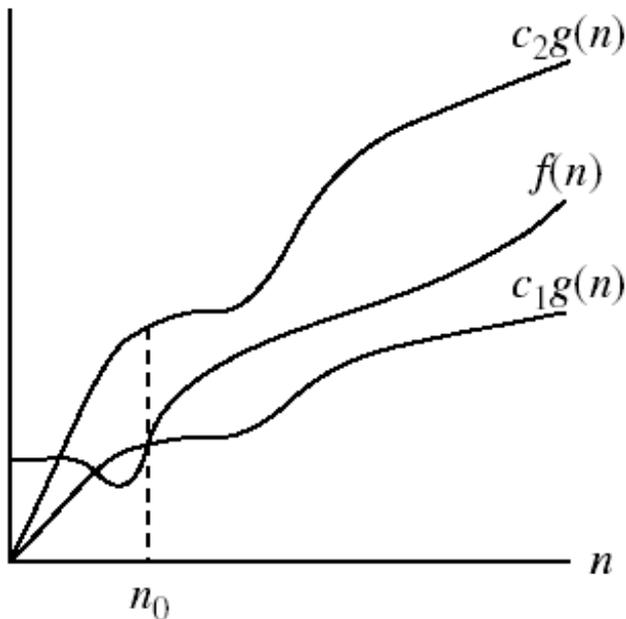
\Rightarrow contradiction: n cannot be smaller than a constant

- $n = \Omega(2n), n^3 = \Omega(n^2), n = \Omega(\log n)$

Asymptotic notations (cont.)

- Θ -notation

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



$\Theta(g(n))$ is the set of functions with the same order of growth as $g(n)$

$g(n)$ is an *asymptotically tight bound* for $f(n)$.

Examples

- $n^2/2 - n/2 = \Theta(n^2)$

• $\frac{1}{2} n^2 - \frac{1}{2} n \leq \frac{1}{2} n^2 \quad \forall n \geq 0 \quad \Rightarrow \quad c_2 = \frac{1}{2}$

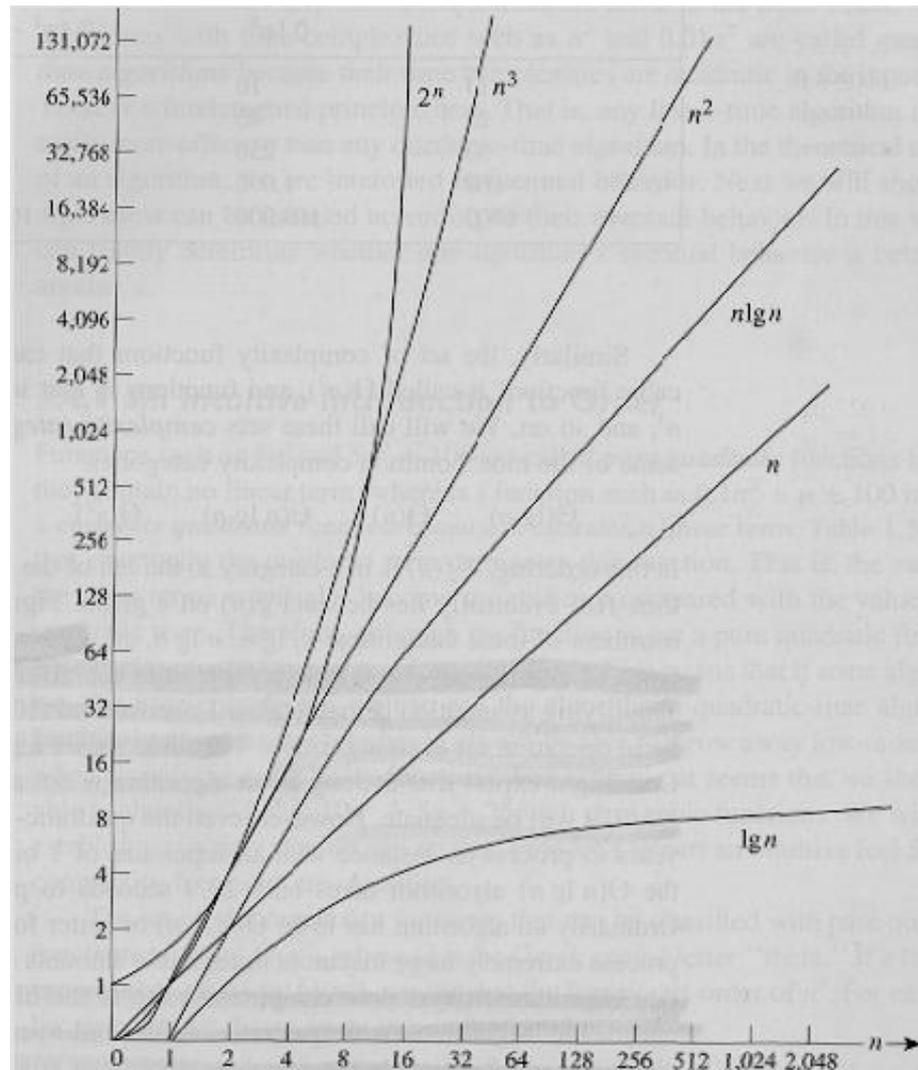
• $\frac{1}{2} n^2 - \frac{1}{2} n \geq \frac{1}{2} n^2 - \frac{1}{2} n * \frac{1}{2} n \quad (\forall n \geq 2) = \frac{1}{4} n^2$

$\Rightarrow \quad c_1 = \frac{1}{4}$

- $n \neq \Theta(n^2): c_1 n^2 \leq n \leq c_2 n^2$

\Rightarrow only holds for: $n \leq 1/c_1$

Common orders of magnitude



Master Method

- for $T(n) = aT(n/b) + f(n)$, n/b may be $\lceil n/b \rceil$ or $\lfloor n/b \rfloor$.
where $a \geq 1$, $b > 1$ are positive integers, $f(n)$ be a non-negative function.
 1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
 2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
 3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$.

Implications of Master Theorem

- Comparison between $f(n)$ and $n^{\log_b a}$ ($<, =, >$)
- Must be asymptotically smaller (or larger) by a polynomial, i.e., n^ϵ for some $\epsilon > 0$.
- In case 3, the “regularity” must be satisfied, i.e., $af(n/b) \leq cf(n)$ for some $c < 1$.
- There are gaps
 - between 1 and 2: $f(n)$ is smaller than $n^{\log_b a}$, but not polynomially smaller.
 - between 2 and 3: $f(n)$ is larger than $n^{\log_b a}$, but not polynomially larger.
 - in case 3, if the “regularity” fails to hold.

Application of Master Theorem

- $T(n) = 9T(n/3) + n;$
 - $a=9, b=3, f(n) = n$
 - $n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$
 - $f(n) = O(n^{\log_3 9 - \epsilon})$ for $\epsilon=1$
 - By case 1, $T(n) = \Theta(n^2)$.
- $T(n) = T(2n/3) + 1$
 - $a=1, b=3/2, f(n) = 1$
 - $n^{\log_b a} = n^{\log_{3/2} 1} = \Theta(n^0) = \Theta(1)$
 - By case 2, $T(n) = \Theta(\lg n)$.

Application of Master Theorem

- $T(n) = 3T(n/4) + n \lg n$;
 - $a=3, b=4, f(n) = n \lg n$
 - $n^{\log_b a} = n^{\log_4 3} = \Theta(n^{0.793})$
 - $f(n) = \Omega(n^{\log_4 3 + \epsilon})$ for $\epsilon \approx 0.2$
 - Moreover, for large n , the “regularity” holds for $c=3/4$.
 - $af(n/b) = 3(n/4) \lg(n/4) \leq (3/4)n \lg n = cf(n)$
 - By case 3, $T(n) = \Theta(f(n)) = \Theta(n \lg n)$.

Scope of research: To find the solution for Exception to Master Theorem

- $T(n) = 2T(n/2) + n \lg n$;
 - $a=2, b=2, f(n) = n \lg n$
 - $n^{\log_b a} = n^{\log_2 2} = \Theta(n)$
 - $f(n)$ is asymptotically larger than $n^{\log_b a}$, but not polynomially larger because
 - $f(n)/n^{\log_b a} = \lg n$, which is asymptotically less than n^ε for any $\varepsilon > 0$.
 - Therefore, this is a gap between 2 and 3.

Assignment

- What is master method?
- How to find Big-O notation for given problem