

Analysis of Algorithms

Complexity and Sorting

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September 17, 2020

Outline

1 Algorithmic Complexity Analysis

- Introduction
- Asymptotic Bounds
- Correctness of Algorithms
- Exercises



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Algorithm efficiency

When measuring an algorithm efficiency we must consider:

- Speed.
- Memory usage.
- Scalability.



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- 2 Big Ω (Lower bound)
- 3 Big Θ (Expected bound)



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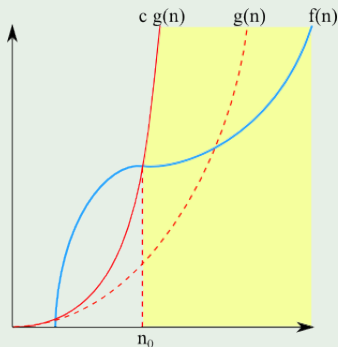
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Big O (Upper bound)

Intuition

- Let $f(n)$ and $g(n)$ be two real valued functions, lets build intuition on the meaning of $f(n) \in O(g(n))$.



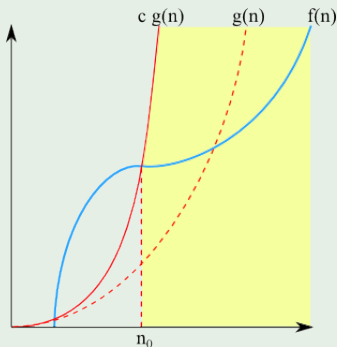
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- $f(n) \in O(g(n))$ if there exists $c, n_0 > 0$ such that $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

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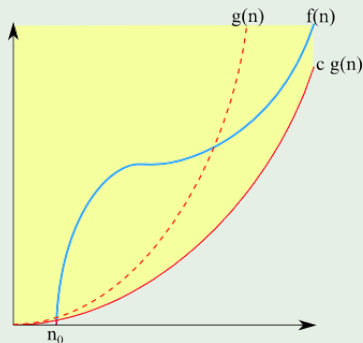
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Big Ω (Lower bound)

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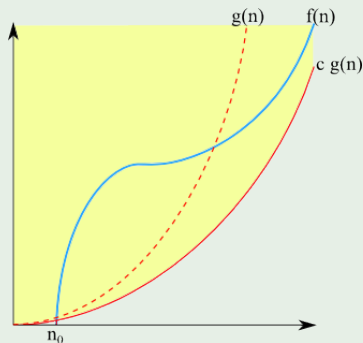
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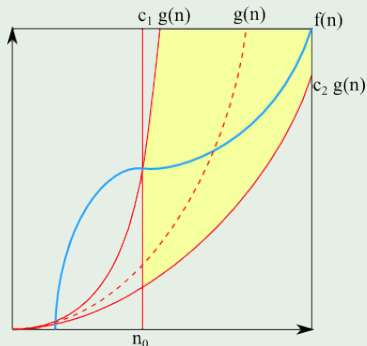
Definition

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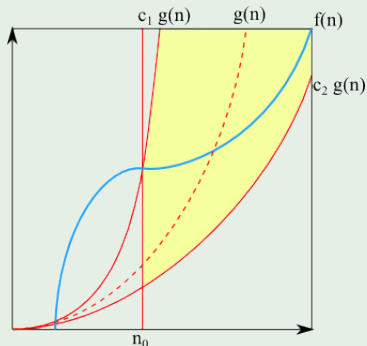
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Correctness of an algorithm

Loop invariant and loop conditional

- A loop invariant is a condition that is necessarily true immediately before and immediately after each iteration of a loop.
- A loop conditional is a statements that controls the termination of the loop.
- Both loop invariant and loop conditional must be different conditions.

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Proof

- To prove an algorithm is correct we must find a loop conditional that ensures the algorithm terminates.
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Well, now you know the basics. Time to work!

From Dasgupta's Algorithms book, exercise 0.1

Using the definition in each of the following situations indicate whether $f = O(g)$, or $f = \Omega(g)$, or both (in which case $f = \Theta(g)$).

1. $f(n) = n - 100$, $g(n) = n - 200$

2. $f(n) = n2^n$, $g(n) = 3^n$



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Exercise

Let's try this one!

Show that $\sum_{k=1}^n \frac{1}{k^2}$ is bounded by a constant. (help me here!).



Exercise

From Cormen's book exercise 2.3-4

We can express insertion sort as a recursive procedure as follows. In order to sort $A[1..n]$, we recursively sort $A[1..n-1]$ and then insert $A[n]$ into the sorted array. Write a recurrence for the running time of this recursive version of insertion sort.



Exercise

From Cormen's book exercise 3.1-7

Prove that $o(g(n)) \cap \omega(g(n))$ is the empty set.



Remember o

$$o(g(n)) = \{f(n) \mid \text{For any } c > 0 \text{ there exists } n_0 > 0 \\ \text{s.t. } 0 \leq f(n) < cg(n) \forall n \geq n_0\}$$

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Exercise

From Cormen's book exercise 3.2-8

Show that $k \ln k = \Theta(n)$ implies that $k = \Theta\left(\frac{n}{\ln n}\right)$.



Proof

By definition

- Exist c_1, c_2 and n_0 such that for a specific k

$$c_1 n \leq k \ln k \leq c_2 n, \quad n > n_0$$

Then we can choose:

$$k \leq n_0$$

Therefore

$$\log k \leq \log n \Rightarrow \frac{1}{\log k} > \frac{1}{\log n}$$



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Exercise

From Cormen's book exercise 4.3-1, 4.3-6

1 Show that the solution of $T(n) = T(\lceil \frac{n}{2} \rceil) + 1$ is $O(\lg n)$.

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Proof

By substitution method $T(n) \leq c \lg n$

$$\begin{aligned}T(n) &\leq 2c \lg \left(\left\lfloor \frac{n}{2} \right\rfloor \right) + 1 \\&\leq 2c (\lg n - \lg 2) + 1 \\&\leq 2c \lg n - 2c + 1\end{aligned}$$

Thus, we need to have

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Exercise

From Cormen's book exercise 2.1-3

Consider the searching problem:

Input: A sequence of n numbers $A = \langle a_1, a_2, \dots, a_n \rangle$ and a value v .

Output: An index i such that $v = A[i]$ or the special value NIL if v does not appear in A .

Write pseudocode for linear search, which scans through the sequence, looking for v . Using a loop invariant, prove that your algorithm is correct. Make sure that your loop invariant fulfills the three necessary properties.

