

CSCI 5454: PS1

Robert Werthman

1.

Let's say these algorithms solve an array sorting problem.

- Let algorithm A be bubblesort with a worst-case runtime of n^2 .
- Let algorithm B be mergesort with a worst-case runtime of $n * \log(n)$.
- Let C be the newly designed sorting algorithm with a worst-case runtime of $h(n)$.

In this case, $O(\min(f(n), g(n)))$ will become $O(n * \log(n))$ because it is the smaller of the two runtimes.

If $h(n)$ is $\log(n)$ then $h(n)$ achieves the running time $O(\min(f(n), g(n)))$ because $\log(n)$ does not grow faster than $n * \log(n)$ and is therefore bounded above by it.

Yes, you can achieve a running time exactly $\min(f(n), g(n))$. Algorithm C would need to be designed in such a way that its running was equal to $\min(f(n), g(n))$.

2.

Proposition/Claim: For any real constants a and b , where $b > 0$, the asymptotic relation $(n + a)^b = \Theta(n^b)$ is true.

Theorem: The asymptotic relation $(n + a)^b = \Theta(n^b)$ is true iff:

- There exists positive constants c_1, c_2, n_0 such that $0 \leq c_1(n^b) \leq (n + a)^b \leq c_2(n^b)$ for all $n \geq n_0$.

In order to prove the proposition above we must find some constants c_1, c_2, n_0 to satisfy the above bulleted sentence.

Proof:

First we want to find the floor and ceiling of $n + a$ so we can create an inequality similar to the one in the theorem above.

1. If $|a| \leq n$ then we can say that $n + a \leq n + |a| \leq 2n$ (Ceiling of $n + a$).
2. If $|a| \leq \frac{1}{2}n$ then we can say that $n + a \geq n - |a| \geq \frac{1}{2}n$ (Floor of $n + a$).

Now if $2|a| \leq n$ then we can combine the floor and ceilings into an compound inequality that holds true :

$$0 \leq \frac{1}{2}n \leq n + a \leq 2n$$

The only thing missing from this new equation is a power of b . Raising the new equation to a power of b gives:

$$0 \leq \left(\frac{1}{2}n\right)^b \leq (n + a)^b \leq (2n)^b \Rightarrow 0 \leq \left(\frac{1}{2}\right)^b n^b \leq (n + a)^b \leq (2)^b n^b$$

Extracting the constants c_1, c_2, n_0 from this equation yields $c_1 = \left(\frac{1}{2}\right)^b$, $c_2 = 2^b$, and $n_0 = 2|a|$ since $n \geq 2|a|$. These represent one solution.

3.

$f(n) = \Omega g(n)$ means that for all values to the right of some n_0 the value of $f(n)$ is on or above $cg(n)$.

$n!$	e^n	$(\frac{3}{2})^n$	$(\lg n)!$	n^2	$n \lg n$	$\lg(n!)$	n	$(\sqrt{2})^{\lg n}$	$2^{\lg^* n}$	$n^{1/\lg n}$	1
------	-------	-------------------	------------	-------	-----------	-----------	-----	----------------------	---------------	---------------	---

Equivalence Classes

$$\lg(n!) = \Theta(n \lg n)$$

$$n^{1/\lg n} = \Theta(1)$$

4.

a.

$$T(n) = T(n-1) + n, \quad T(1) = 1$$

I will use a recurrence tree to solve this recurrence relation.



The height of the tree is n and the cost at the root starts at n and decreases by 1 each level in the tree.

This means that the total cost of the tree is n .

So $T(n) = O(\text{cost} * \text{depth}) = O(n^2)$.

b.

$$T(n) = 2T(n/2) + n^3, T(1) = 1$$

I will use the master method to solve this recurrence relation.

$$a = 2, b = 2, f(n) = n^3$$

$$\text{so } n^{\log_b a} = n^{\log_2 2} = n$$

This tells us that the first 2 rules of the master theorem do not apply.

$$1. f(n) \neq O(n^{1-\epsilon})$$

$$2. f(n) \neq \Theta(n)$$

This leaves the 3rd rule of the master theorem as the solution.

$$3. f(n) = n^3 = \Omega(n^{1+\epsilon}) \text{ if } \epsilon = 1. \text{ And } 2f(n/2) \leq cf(n) \Rightarrow 2(n/2)^3 \leq cn^3 \\ \text{if } c = \frac{1}{2} \text{ and } n \geq 1.$$

Therefore, $T(n) = \Theta(n^3)$.

5.