Master Method

Adapted from the CLRS book slides

MASTER METHOD

Used for many divide-and-conquer *master recurrences* of the form T(n) = aT(n/b) + f(n), where $a \ge 1$, b > 1, and f(n) is an asymptotically nonnegative function defined over all sufficiently large positive numbers.

Master recurrences describe recursive algorithms that divide a problem of size n into a subproblems, each of size n/b. Each recursive subproblem takes time T(n/b) (unless it's a base case). Call f(n) the *driving function*.

Based on the *master theorem* (Theorem 4.1):

Let a, b > 0 be constants, f(n) be a driving function defined and nonnegative on all sufficiently large reals. Define recurrence T(n) on $n \in \mathbb{N}$ by

$$T(n) = aT(n/b) + f(n),$$

and where aT(n/b) actually means $a'T(\lfloor n/b \rfloor) + a''T(\lceil n/b \rceil)$ for some constants $a', a'' \ge 0$ satisfying a = a' + a''.

Case 1: $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$.

(f(n) is polynomially smaller than $n^{\log_b a}$.)

Then you can solve the recurrence by comparing $n^{\log_b a}$ vs. f(n):

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Solution: T(n) = \Theta(n^{\log_b a}). (Intuitively: cost is dominated by leaves.)

Case 2: f(n) = \Theta(n^{\log_b a} \lg^k n), where k \ge 0 is a constant. (f(n) is within a polylog factor of n^{\log_b a}, but not smaller.)

Solution: T(n) = \Theta(n^{\log_b a} \lg^{k+1} n). (Intuitively: cost is n^{\log_b a} \lg^k n at each level, and there are \Theta(\lg n) levels.)

Simple case: k = 0 \Rightarrow f(n) = \Theta(n^{\log_b a}) \Rightarrow T(n) = \Theta(n^{\log_b a} \lg n).
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Case 3: $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$ and f(n) satisfies the regularity condition $af(n/b) \le cf(n)$ for some constant c < 1 and all sufficiently large n.

(f(n)) is polynomially greater than $n^{\log_b a}$.)

Solution: $T(n) = \Theta(f(n))$.

(Intuitively: cost is dominated by root.)

What's with the Case 3 regularity condition?

- Generally not a problem.
- It always holds whenever $f(n) = n^k$ and $f(n) = \Omega(n^{\log_b a + \epsilon})$ for constant $\epsilon > 0$. So you don't need to check it when f(n) is a polynomial.

Call $n^{\log_b a}$ the *watershed function*. Master method compares the driving function f(n) with the watershed function $n^{\log_b a}$.

- If the watershed function grows *polynomially faster* than the driving function, then case 1 applies.
- If the driving function grows *polynomially faster* than the watershed function and the regularity condition holds, then case 3 applies.
- If the driving function is within a polylog factor of the watershed function but not smaller, then case 2 applies.
- There are gaps between cases 1 and 2 and between cases 2 and 3.

MASTER METHOD EXAMPLES

$$T(n) = 5T(n/2) + \Theta(n^2)$$

- 1 $n^{\log_2 5}$ vs. n^2 Since $\log_2 5 - \epsilon = 2$ for some constant $\epsilon > 0$, use case $1 \Rightarrow T(n) = \Theta(n^{\lg 5})$
- $T(n) = 27T(n/3) + \Theta(n^3 \lg n)$ $n^{\log_3 27} = n^3 \text{ vs. } n^3 \lg n$ Use case 2 with $k = 1 \Rightarrow T(n) = \Theta(n^3 \lg^2 n)$

$$T(n) = 5T(n/2) + \Theta(n^3)$$

 $n^{\log_2 5}$ vs. n^3

Now $\lg 5 + \epsilon = 3$ for some constant $\epsilon > 0$ Check regularity condition (don't really need to since f(n) is a polynomial): $af(n/b) = 5(n/2)^3 = 5n^3/8 \le cn^3$ for c = 5/8 < 1Use case $3 \Rightarrow T(n) = \Theta(n^3)$

$$T(n) = 27T(n/3) + \Theta(n^3/\lg n)$$

 $n^{\log_3 27} = n^3 \text{ vs. } n^3 / \lg n = n^3 \lg^{-1} n \neq \Theta(n^3 \lg^k n) \text{ for any } k \geq 0.$ Cannot use the master method.