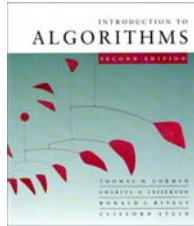




CS 3343 -- Spring 2009



Master Theorem

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Slides courtesy of Charles Leiserson with small changes by Carola Wenk



The divide-and-conquer design paradigm

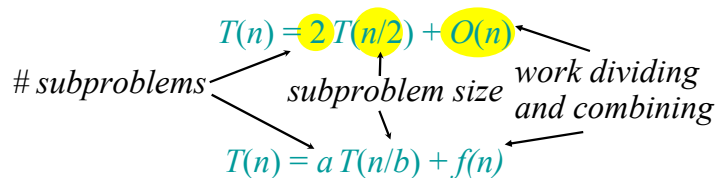
1. **Divide** the problem (instance) into subproblems.
 a subproblems, **each** of size n/b
2. **Conquer** the subproblems by solving them recursively.
3. **Combine** subproblem solutions.

Runtime for divide and combine is $f(n)$



Example: merge sort

1. **Divide:** Trivial.
2. **Conquer:** Recursively sort $a=2$ subarrays of size $n/2=n/b$
3. **Combine:** Linear-time merge, runtime $f(n) \in O(n)$



The master method

The master method applies to recurrences of the form

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

where $a \geq 1$, $b > 1$, and f is asymptotically positive.



Master theorem (summary)

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

CASE 1: $f(n) = O(n^{\log_b a - \epsilon})$

$$\Rightarrow T(n) = \Theta(n^{\log_b a}) .$$

CASE 2: $f(n) = \Theta(n^{\log_b a} \log^k n)$

$$\Rightarrow T(n) = \Theta(n^{\log_b a} \log^{k+1} n) .$$

CASE 3: $f(n) = \Omega(n^{\log_b a + \epsilon})$ and $af(n/b) \leq cf(n)$
for some constant $c < 1$.

$$\Rightarrow T(n) = \Theta(f(n)) .$$



Three common cases

Compare $f(n)$ with $n^{\log_b a}$:

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

- $f(n)$ grows polynomially slower than $n^{\log_b a}$ (by an n^ϵ factor).

Solution: $T(n) = \Theta(n^{\log_b a})$.

2. $f(n) = \Theta(n^{\log_b a} \log^k n)$ for some constant $k \geq 0$.

- $f(n)$ and $n^{\log_b a}$ grow at similar rates.

Solution: $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$.



Three common cases (cont.)

Compare $f(n)$ with $n^{\log_b a}$:

3. $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$.

- $f(n)$ grows polynomially faster than $n^{\log_b a}$ (by an n^ϵ factor),

and $f(n)$ satisfies the **regularity condition** that $af(n/b) \leq cf(n)$ for some constant $c < 1$.

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



Examples

Ex. $T(n) = 4T(n/2) + \text{sqrt}(n)$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = \text{sqrt}(n).$$

CASE 1: $f(n) = O(n^{2-\epsilon})$ for $\epsilon = 1.5$.

$$\therefore T(n) = \Theta(n^2).$$

Ex. $T(n) = 4T(n/2) + n^2$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2.$$

CASE 2: $f(n) = \Theta(n^2 \log^0 n)$, that is, $k = 0$.

$$\therefore T(n) = \Theta(n^2 \log n).$$



Examples

Ex. $T(n) = 4T(n/2) + n^3$
 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^3$.
CASE 3: $f(n) = \Omega(n^{2+\epsilon})$ for $\epsilon = 1$
and $4(n/2)^3 \leq cn^3$ (reg. cond.) for $c = 1/2$.
 $\therefore T(n) = \Theta(n^3)$.

Ex. $T(n) = 4T(n/2) + n^2/\log n$
 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2/\log n$.
 Master method does not apply. In particular,
 for every constant $\epsilon > 0$, we have $\log n \in o(n^\epsilon)$.



Example: merge sort

1. **Divide:** Trivial.
2. **Conquer:** Recursively sort 2 subarrays.
3. **Combine:** Linear-time merge.

$T(n) = 2T(n/2) + O(n)$

subproblems \swarrow subproblem size \nwarrow work dividing and combining

$n^{\log_b a} = n^{\log_2 2} = n^1 = n \Rightarrow$ **CASE 2** ($k = 0$)
 $\Rightarrow T(n) = \Theta(n \log n)$.



Recurrence for binary search

$T(n) = 1T(n/2) + \Theta(1)$

subproblems \swarrow subproblem size \nwarrow work dividing and combining

$n^{\log_b a} = n^{\log_2 1} = n^0 = 1 \Rightarrow$ **CASE 2** ($k = 0$)
 $\Rightarrow T(n) = \Theta(\log n)$.