

CMPS 6610/4610 – Fall 2016

Master Theorem

Carola Wenk

Slides courtesy of Charles Leiserson
with changes and additions by Carola Wenk

The divide-and-conquer design paradigm

- 1. *Divide*** the problem (instance) into subproblems of sizes that are fractions of the original problem size.
- 2. *Conquer*** the subproblems by solving them recursively.
- 3. *Combine*** subproblem solutions.

⇒ Runtime recurrences

The master method

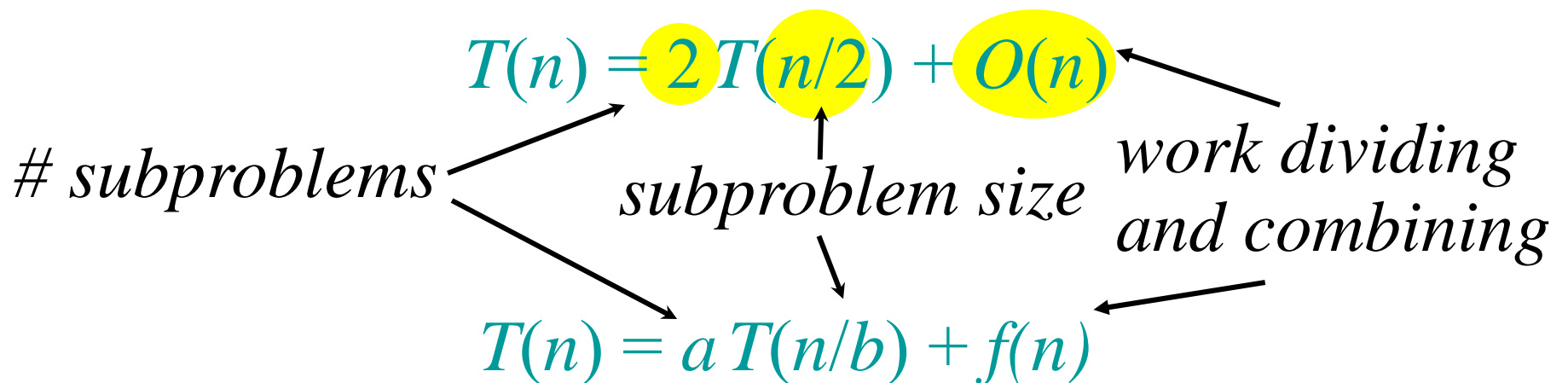
The master method applies to recurrences of the form

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

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

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 $f(n) \in O(n)$



Master Theorem

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

CASE 1:

$$f(n) = O(n^{\log_b a - \varepsilon}) \quad \Rightarrow \quad T(n) = \Theta(n^{\log_b a})$$

for some $\varepsilon > 0$

CASE 2:

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

for some $k \geq 0$

CASE 3:

$$(i) f(n) = \Omega(n^{\log_b a + \varepsilon})$$

for some $\varepsilon > 0$

$$\text{and (ii) } a f(n/b) \leq c f(n)$$

for some $c < 1$

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

How to apply the theorem

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

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

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

How to apply the theorem

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

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

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

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

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

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 \nearrow 2 \nearrow $T(n/2)$ \nearrow $n/2$ \nearrow $O(n)$ \longleftarrow work dividing and combining

subproblem size

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

Example: binary search

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

subproblems *subproblem size* *work dividing and combining*

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

Master theorem: Examples

Ex. $T(n) = 4T(n/2) + \sqrt{n}$
 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = \sqrt{n}$.
CASE 1: $f(n) = O(n^{2-\varepsilon})$ for $\varepsilon = 1.5$.
 $\therefore T(n) = \Theta(n^2)$.

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, $k = 0$.
 $\therefore T(n) = \Theta(n^2 \log n)$.

Master theorem: 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: $f(n) = \Omega(n^{2 + \varepsilon})$ for $\varepsilon = 1$

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 $\varepsilon > 0$, we have $\log n \in o(n^\varepsilon).$

Conclusion

- Divide and conquer is just one of several powerful techniques for algorithm design.
- Divide-and-conquer algorithms can be analyzed using recurrences and the master method .
- Can lead to more efficient algorithms