# Solving Recurrences: Master Theorem

# Solving Recurrences

- Iteration Method
- Master Method
- Recursion Tree Method

## Solving Recurrences: Iteration Method

- Given: a *divide and conquer* algorithm
  - An algorithm that divides the problem of size n into a subproblems, each of size n/b, where  $a \ge 1$ , b > 1
  - The a subproblems are solved recursively, each in time T(n/b)
  - Let the cost of each stage (i.e., the work to divide the problem + combine solved subproblems) be described by cn
- Then, the recurrence is

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

## Solving Recurrences: Iteration Method

- The "iteration method"
  - Expand the recurrence
  - Work some algebra to express as a summation
  - Evaluate the summation
- We show by using

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

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• 
$$T(n) =$$
 $aT(n/b) + cn$ 
 $a(aT(n/b/b) + cn/b) + cn$ 
 $a^2T(n/b^2) + cna/b + cn$ 
 $a^2T(n/b^2) + cn(a/b + 1)$ 
 $a^2(aT(n/b^2/b) + cn/b^2) + cn(a/b + 1)$ 
 $a^3T(n/b^3) + cn(a^2/b^2) + cn(a/b + 1)$ 
 $a^3T(n/b^3) + cn(a^2/b^2 + a/b + 1)$ 
...
 $a^kT(n/b^k) + cn(a^{k-1}/b^{k-1} + a^{k-2}/b^{k-2} + ... + a^2/b^2 + a/b + 1)$ 

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

So we have

$$T(n) = a^{k}T(n/b^{k}) + cn(a^{k-1}/b^{k-1} + ... + a^{2}/b^{2} + a/b + 1)$$

• For  $n/b^k = 1$ 

$$n = b^k \rightarrow k = \log_b n$$

■ 
$$T(n) = a^k T(1) + cn(a^{k-1}/b^{k-1} + ... + a^2/b^2 + a/b + 1)$$
  
 $= a^k c + cn(a^{k-1}/b^{k-1} + ... + a^2/b^2 + a/b + 1)$   
 $= ca^k b^k/b^k + cn(a^{k-1}/b^{k-1} + ... + a^2/b^2 + a/b + 1)$   
 $= cn a^k/b^k + cn(a^{k-1}/b^{k-1} + ... + a^2/b^2 + a/b + 1)$   
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$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with  $k = \log_b n$ 
  - $T(n) = cn(a^k/b^k + ... + a^2/b^2 + a/b + 1)$
- What if a = b?
  - T(n) = cn(k + 1)  $= cn(\log_b n + 1)$   $= \Theta(n \log n)$

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- So with  $k = \log_b n$ 
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- What if a < b?

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- So with  $k = \log_b n$ 
  - $T(n) = cn(a^k/b^k + ... + a^2/b^2 + a/b + 1)$
- What if a < b?
  - Recall that  $\Sigma(x^k + x^{k-1} + ... + x + 1) = (x^{k+1} 1)/(x-1)$

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- What if a < b?
  - Recall that  $\Sigma(x^k + x^{k-1} + ... + x + 1) = (x^{k+1} 1)/(x-1)$
  - So:

$$\frac{a^{k}}{b^{k}} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \frac{1 - (a/b)^{k+1}}{1 - (a/b)} < \frac{1}{1 - a/b}$$

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■ 
$$T(n) = cn \cdot \Theta(1) = \Theta(n)$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

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$$= cn \cdot \Theta(a^{\log_b n} / b^{\log_b n}) = cn \cdot \Theta(a^{\log_b n} / n)$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with  $k = \log_b n$ 
  - $T(n) = cn(a^k/b^k + ... + a^2/b^2 + a/b + 1)$
- What if a > b?

$$\frac{a^{k}}{b^{k}} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^{k})$$

= 
$$\operatorname{cn} \cdot \Theta(a^{\log_b n} / b^{\log_b n}) = \operatorname{cn} \cdot \Theta(a^{\log_b n} / n)$$
  
recall logarithm fact:  $a^{\log_b n} = n^{\log_b n}$ 

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with  $k = \log_b n$ 
  - $T(n) = cn(a^k/b^k + ... + a^2/b^2 + a/b + 1)$
- What if a > b?

$$\frac{a^{k}}{b^{k}} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^{k})$$

$$= \operatorname{cn} \cdot \Theta(\operatorname{a^{\log_b n}} / \operatorname{b^{\log_b n}}) = \operatorname{cn} \cdot \Theta(\operatorname{a^{\log_b n}} / \operatorname{n})$$

$$\operatorname{recall\ logarithm\ fact:\ } \operatorname{a^{\log_b n}} = \operatorname{n^{\log_b n}}$$

$$= \operatorname{cn} \cdot \Theta(\operatorname{n^{\log_b a}} / \operatorname{n}) = \Theta(\operatorname{cn} \cdot \operatorname{n^{\log_b a}} / \operatorname{n})$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

- So with  $k = \log_b n$ 
  - $T(n) = cn(a^k/b^k + ... + a^2/b^2 + a/b + 1)$
- What if a > b?

$$\frac{a^{k}}{b^{k}} + \frac{a^{k-1}}{b^{k-1}} + \dots + \frac{a}{b} + 1 = \frac{(a/b)^{k+1} - 1}{(a/b) - 1} = \Theta((a/b)^{k})$$

$$= \operatorname{cn} \cdot \Theta(\operatorname{a^{\log_b n}}/\operatorname{b^{\log_b n}}) = \operatorname{cn} \cdot \Theta(\operatorname{a^{\log_b n}}/\operatorname{n})$$

$$\operatorname{recall\ logarithm\ fact:\ } \operatorname{a^{\log_b n}} = \operatorname{n^{\log_b a}}$$

$$= \operatorname{cn} \cdot \Theta(\operatorname{n^{\log_b a}}/\operatorname{n}) = \Theta(\operatorname{cn} \cdot \operatorname{n^{\log_b a}}/\operatorname{n})$$

$$= \Theta(\operatorname{n^{\log_b a}})$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

• So...

$$T(n) = \begin{cases} \Theta(n) & a < b \\ \Theta(n \log_b n) & a = b \\ \Theta(n^{\log_b a}) & a > b \end{cases}$$

#### The Master Theorem

- Given: a divide and conquer algorithm
  - An algorithm that divides the problem of size n into a subproblems, each of size n/b, where  $a \ge 1$ , b > 1
  - The a subproblems are solved recursively, each in time T(n/b)
  - Let the cost of each stage (i.e., the work to divide the problem + combine solved subproblems) be described by the function f(n), where f is asymptotically positive
  - $\blacksquare$  T(n) is monotonically increasing function
- Then, the Master Theorem gives us a *cookbook* for the algorithm's running time.

## The Master Theorem: Pitfalls

- You cannot use the Master Theorem if
  - T(n) is not monotone, e.g.  $T(n) = \sin(x)$
  - f(n) is not a polynomial, e.g.,  $T(n) = 2T(n/2) + 2^n$
  - b cannot be expressed as a constant, e.g.

$$T(n) = T(\sqrt{n})$$

 Note that the Master Theorem does not solve all recurrence equations

## The Master Theorem

Let  $a \ge 1$  and b > 1 be constants, let f(n) be a function, and let T(n) be defined on the nonnegative integers by the recurrence T(n) = a T(n/b) + f(n).

Then T(n) has the following asymptotic bounds:

- o If  $f(n) = O(n^{\log_b a \varepsilon})$  for some constant  $\varepsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$
- o If  $f(n) = \Theta(n^{\log_b a})$  then  $T(n) = \Theta(n^{\log_b a} \log_n n)$
- o If  $f(n) = O(n^{\log_b a + \varepsilon})$  for some constant  $\varepsilon > 0$  then  $T(n) = \Theta(f(n))$

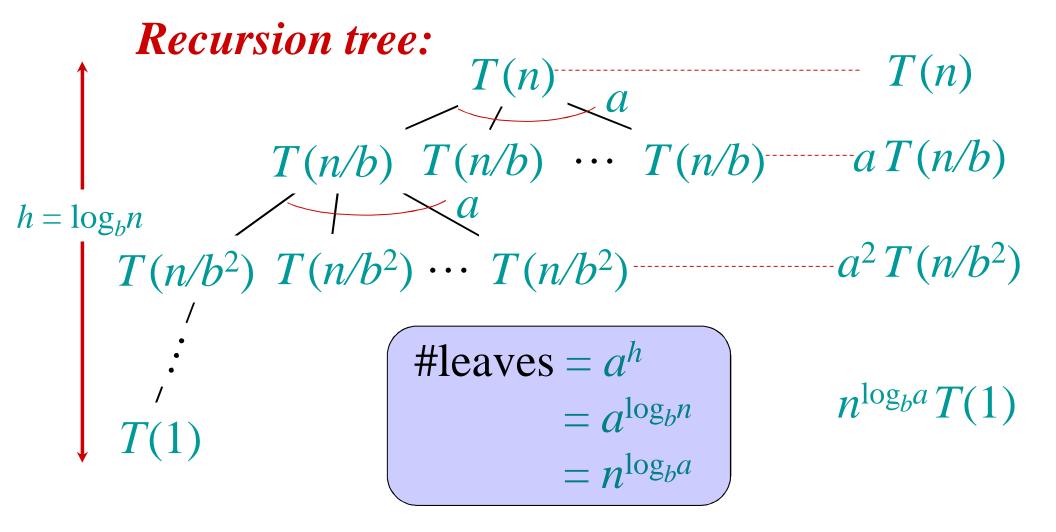
# Excuse me, what did it say ???

• Essentially, the Master theorem compares the function f(n) with the function  $g(n) = n^{\log_b(a)}$ .

Roughly, the theorem says:

- If  $f(n) \ll g(n)$  then  $T(n) = \Theta(g(n))$
- If  $f(n) \approx g(n)$  then  $T(n) = \Theta(g(n) \log_b n)$
- If f(n) >> g(n) then  $T(n) = \Theta(f(n))$
- Now go back and memorize the theorem !!!

#### Idea of Master Theorem



#### Idea of Master Theorem

Let us iteratively substitute the recurrence:

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

$$= a(aT(n/b^{2})) + f(n/b)) + f(n)$$

$$= a^{2}T(n/b^{2}) + af(n/b) + f(n)$$

$$= a^{3}T(n/b^{3}) + a^{2}f(n/b^{2}) + af(n/b) + f(n)$$

$$= ...$$

$$= a^{\log_{b} n}T(1) + \sum_{i=0}^{(\log_{b} n) - 1} a^{i}f(n/b^{i})$$

$$= n^{\log_{b} a}T(1) + \sum_{i=0}^{(\log_{b} n) - 1} a^{i}f(n/b^{i})$$

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#### Idea of Master Theorem

Thus, we obtained

$$T(n) = n^{\log_b(a)} T(1) + \sum_i a^i f(n/b^i)$$

The proof proceeds by distinguishing three cases:

■ The first term is dominant:

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

■ Each term of the summation is equally dominant:

$$f(n) = \Theta(n^{\log_b(a)})$$

The second term is dominant and can be bounded by a geometric series:

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

#### Master Theorem: Three Common Cases

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  $n^{\log_b a}$

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

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

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

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

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

## Master Theorem: Examples

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

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

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  $\epsilon = 1$   
 $\therefore T(n) = \Theta(n^3).$