#### COT 6405 Introduction to Theory of Algorithms

**Topic 4. Recurrences** 

#### Recurrences

• What is a recurrence?

 An equation that describes a function in terms of its value on smaller functions

• The time complexity of divide-and-conquer algorithms can be expressed as recurrences

#### **Recurrence Examples**

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases} \qquad s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

$$T(n) = \begin{cases} c & n=1 \\ T(n) = \begin{cases} c & n=1 \\ 2T\left(\frac{n}{2}\right) + c & n>1 \end{cases} \qquad T(n) = \begin{cases} c & n=1 \\ aT\left(\frac{n}{b}\right) + cn & n>1 \end{cases}$$

#### Solving the recurrences

- Substitution method
- Recursion Tree
- Master method

## Substitution method

- The substitution method comprises two steps:
  - 1. Guess the form of the solution
  - 2. Use mathematical induction to show the correctness of the guess

Example:

$$T(n) = \begin{cases} 1 & \text{if } n = 1, \\ 2T(n/2) + n & \text{if } n > 1. \end{cases}$$

- 1. *Guess:*  $T(n) = n \lg n + n$ . [Here, we have a recurrence with an exact function, rather than asymptotic notation, and the solution is also exact rather than asymptotic. We'll have to check boundary conditions and the base case.]
- 2. Induction:

Basis: 
$$n = 1 \Rightarrow n \lg n + n = 1 = T(n)$$

**Inductive step:** Inductive hypothesis is that  $T(k) = k \lg k + k$  for all k < n. We'll use this inductive hypothesis for T(n/2).

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$
  
=  $2\left(\frac{n}{2}\lg\frac{n}{2} + \frac{n}{2}\right) + n$  (by inductive hypothesis)  
=  $n\lg\frac{n}{2} + n + n$   
=  $n(\lg n - \lg 2) + n + n$   
=  $n\lg n - n + n + n$   
=  $n\lg n + n$ .

# Substitution method (cont'd)

- We generally express the solution by asymptotic notations
- We don't worry about boundary cases, nor do we show base cases in the substitution proof.
  - because we are ultimately interested in an asymptotic solution to a recurrence, it will always be possible to choose base cases that work.

*Example:*  $T(n) = 2T(n/2) + \Theta(n)$ . If we want to show an upper bound of T(n) = 2T(n/2) + O(n), we write  $T(n) \le 2T(n/2) + cn$  for some positive constant *c*.

#### 1. Upper bound:

*Guess:*  $T(n) \leq dn \lg n$  for some positive constant *d*. We are given *c* in the recurrence, and we get to choose *d* as any positive constant. It's OK for *d* to depend on *c*.

Substitution:

$$T(n) \leq 2T(n/2) + cn$$
  

$$\leq 2\left(d\frac{n}{2}\lg\frac{n}{2}\right) + cn$$
  

$$= dn \lg\frac{n}{2} + cn$$
  

$$= dn \lg n - dn + cn$$
  

$$\leq dn \lg n \qquad \text{if } -dn + cn \leq 0,$$
  

$$d \geq c$$

Guess  $T(n) = \Theta$  (*nlgn*) Prove: T(n) = O(nlgn) and  $\Omega(nlgn)$   Lower bound: Write T (n) ≥ 2T (n/2) + cn for some positive constant c. Guess: T (n) ≥ dn lg n for some positive constant d. Substitution:

$$T(n) \geq 2T(n/2) + cn$$
  

$$\geq 2\left(d\frac{n}{2}\lg\frac{n}{2}\right) + cn$$

$$= dn\lg\frac{n}{2} + cn$$

$$= dn\lg n - dn + cn$$
  

$$\geq dn\lg n \qquad \text{if } -dn + cn \geq 0,$$

$$dn \leq c$$

Therefore,  $T(n) = \Omega(n \lg n)$ .

Therefore,  $T(n) = \Theta(n \lg n)$ . [For this particular recurrence, we can use d = c for both the upper-bound and lower-bound proofs. That won't always be the case.]

#### Substitution method

- For the substitution method:
  - Show the upper and lower bounds separately.
    - Might need to use different constants for each.
- Making a good guess
  - Unfortunately, there is no general way to guess the correct solutions to recurrences.
  - Takes experience and creativity.

Make sure you show the same *exact* form when doing a substitution proof. Consider the recurrence

$$T(n) = 8T(n/2) + \Theta(n^2).$$

For an upper bound:

$$T(n) \le 8T(n/2) + cn^{2}.$$
  
Guess:  $T(n) \le dn^{3}.$   

$$T(n) \le 8d(n/2)^{3} + cn^{2}$$
  

$$= 8d(n^{3}/8) + cn^{2}$$
  

$$\le dn^{3} + cn^{2}$$
  
doesn't work!

Guess  $T(n) = \Theta(n^3)$ Prove:  $T(n) = O(n^3)$  and  $\Omega(n^3)$ 

How to fix this?

**Remedy:** Subtract off a lower-order term.  
Guess: 
$$T(n) \le dn^3 - d'n^2$$
.  
 $T(n) \le 8(d(n/2)^3 - d'(n/2)^2) + cn^2$   
 $= 8d(n^3/8) - 8d'(n^2/4) + cn^2$   
 $= dn^3 - 2d'n^2 + cn^2$   
 $= dn^3 - d'n^2 - d'n^2 + cn^2$   
 $\le dn^3 - d'n^2$  if  $-d'n^2 + cn^2 \le 0$ ,  
 $d' \ge c$ 

# **Avoiding Pitfalls**

- It is easy to err in the use of asymptotic notation
- Solve  $T(n) = 2T(n/2) + \Theta(n)$
- Guess: T(n) = O(n) and  $T(n) \le dn$  for some positive constant number d
- Induction:  $T(n) \le 2T(n/2) + cn$   $\le 2(d(n/2)) + cn$  $\le dn + cn = (d+c)n = O(n)$

#### Why wrong?

# Changing variables

- Sometimes, a little algebraic manipulations can make an unknown recurrence similar to one you have seen before.
- Solve the recurrence  $T(n) = 2T(\sqrt{n}) + lgn$ 
  - Renaming m = lgn yields  $T(2^m) = 2T(2^{m/2}) + m$
  - We can now rename  $S(m) = T(2^m)$  to produce the new recurrence S(m) = 2S(m/2) + m
  - $-S(m) = \Theta(mlgm)$
  - $-T(n)=T(2^m)=S(m)=\Theta(mlgm)=\Theta(lgnlglgn)$

#### Recursion tree method

- How to solve the recurrence of merge sort?
- By using substitution method, we can have -T(n) = 2T(n/2) + n = 2(2T(n/4) + n/2) + n = 4T(n/4) + 2n

#### Recursion tree method (cont'd)

- An alternative approach: draw a tree to diagram all the recursive calls that take place
   T(n) = 2T(n/2) + n
- For the original problem, we have a cost of n, plus the two subproblems, each costing n/2

#### Constructing the tree



For each of the size-n/2 subproblems, we have a cost of n/2, plus two subproblems, each costing n/4

#### Constructing the tree (cont'd)



#### Constructing the tree (cont'd)



#### Computing the cost

• We add up the costs over all levels to determine the cost for the entire tree



#### Example

• Solve  $T(n) = 3T(n/4) + cn^2$ 



# Example(cont'd)

- The subproblem size for a node at depth i is  $n/4^i$
- The subproblem size hits T(1), when  $n/4^i = 1$ , or  $i = \log_4 n$
- Thus, tree has  $1 + \log_4 n$  levels (*i* = 0,1,... $\log_4 n$ )



# Example(cont'd)

- Each node at level *i* has a cost of  $c(n/4^i)^2$
- Each level has 3<sup>*i*</sup> nodes
- Thus, the total cost of level *i* is  $3^i c(n/4^i)^2 = cn^2(3/16)^i$



# Example(cont'd)

- The bottom level has 3<sup>log<sub>4</sub> n</sup> = n<sup>log<sub>4</sub> 3</sup>nodes, each costing T(1)
- Assume T(1) is a constant. The total cost of the bottom level will be

T(1) 
$$n^{\log_4 3} = \Theta(n^{\log_4 3})$$

#### Total cost

- The total cost of level i is  $cn^2(3/16)^i$
- The total cost of the bottom level  $\Theta(n^{\log_4 3})$
- We add up the costs over all levels to determine the total cost for the entire tree:

$$T(n) = cn^{2} + \frac{3}{16}cn^{2} + (\frac{3}{16})^{2}cn^{2} + \dots + (\frac{3}{16})^{\log_{4} n - 1}cn^{2} + \Theta(n^{\log_{4} 3})$$

$$= \sum_{i=0}^{\log_4 n-1} (\frac{3}{16})^i cn^2 + \Theta(n^{\log_4 3})$$

$$=\frac{\left(\frac{3}{16}\right)^{\log_4 n-1}-1}{\frac{3}{16}-1}cn^2+\Theta(n^{\log_4 3})$$

#### How to simplify the answer

$$T(n) = \sum_{i=0}^{\log_4 n-1} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3})$$
  

$$\leq \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3})$$
  

$$= \frac{1}{1 - \frac{3}{16}} cn^2 + \Theta(n^{\log_4 3}) = \frac{16}{13} cn^2 + \Theta(n^{\log_4 3})$$
  

$$= O(n^2)$$

# How to simplify the answer (cont'd)

• On the other hand,

$$T(n) = 3T(n/4) + cn^2 \ge cn^2$$
  
Thus,  $T(n) = \Omega(n^2)$  and we conclude that  
 $T(n) = \Theta(n^2)$ 

How to use substitution method to verify?

#### Exercise

• Solve T(n) = aT(n/b) + f(n)



- The subproblem size for a node at depth i is  $n/b^i$
- The subproblem size hits T(1), when  $n/b^i = 1$ , or  $i = \log_b n$
- Thus, tree has  $1 + \log_b n$  levels ( $i = 0, 1, \dots, \log_b n$ )

- Each node at level *i* has a cost of  $f(n/b^i)$
- Each level has a<sup>i</sup> nodes
   Level 0: 1, level 1: a, level 2: a<sup>2</sup>, level 3: a<sup>3</sup>....
- Thus, the total cost of level *i* is  $a^i f(n/b^i)$

- The bottom level has a<sup>log<sub>b</sub> n</sup> = n<sup>log<sub>b</sub> a</sup> nodes, each costing T(1)
- Assume T(1) is a constant. The total cost of the bottom level will be
   T(1)n<sup>log<sub>b</sub> a</sup> = Θ(n<sup>log<sub>b</sub> a</sup>)

• We add up the costs over all levels to determine the total cost for the entire tree:

 $\mathsf{T}(n) = \mathsf{f}(n) + a\mathsf{f}(n/b) + a^2\mathsf{f}(n/b^2) + \dots + a^{\log_b n - 1}\mathsf{f}(n/b^{\log_b n - 1}) + \Theta(n^{\log_b a})$ 

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