# **Algorithms**

Recursion II

Emanuele Rodolà rodola@di.uniroma1.it



# "Solving" recursion

For merge sort, we have encountered the expression:

$$T(n) = \begin{cases} \Theta(1) & n = 1\\ 2T(\frac{n}{2}) + \Theta(n) & n > 1 \end{cases}$$

How to obtain asymptotic bounds on the solution?

# "Solving" recursion

For merge sort, we have encountered the expression:

$$T(n) = \begin{cases} \Theta(1) & n = 1\\ 2T(\frac{n}{2}) + \Theta(n) & n > 1 \end{cases}$$

How to obtain asymptotic bounds on the solution?

- Substitution method
- Recursion-tree method
- Master method

#### General idea:

- Guess an expression for the solution.
- Prove your guess by induction.

Good method when it is easy to guess.

#### General idea:

- Guess an expression for the solution.
- 2 Prove your guess by induction.

Good method when it is easy to guess.

#### Example:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

#### General idea:

- Guess an expression for the solution.
- 2 Prove your guess by induction.

Good method when it is easy to guess.

#### Example:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Similar to something we have seen, so we guess  $T(n) = O(n \lg n)$ .

#### General idea:

- Guess an expression for the solution.
- 2 Prove your guess by induction.

Good method when it is easy to guess.

#### Example:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Similar to something we have seen, so we guess  $T(n) = O(n \lg n)$ .

Since *O* measures upper bounds, we want to prove:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

for some c > 0.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le c n \lg n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

$$T(n) \le 2(c\lfloor n/2\rfloor \lg\lfloor n/2\rfloor) + n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

$$T(n) \le 2(c\lfloor n/2\rfloor \lg\lfloor n/2\rfloor) + n$$
  
$$\le cn \lg(n/2) + n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

$$T(n) \le 2(c\lfloor n/2\rfloor \lg\lfloor n/2\rfloor) + n$$
  
$$\le cn \lg(n/2) + n$$
  
$$= cn \lg n - cn \lg 2 + n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

$$T(n) \le 2(c\lfloor n/2\rfloor \lg\lfloor n/2\rfloor) + n$$

$$\le cn \lg(n/2) + n$$

$$= cn \lg n - cn \lg 2 + n$$

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

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

Assume the guess holds for  $T(\lfloor n/2 \rfloor)$ :

$$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor$$

$$\begin{split} T(n) &\leq 2(c \lfloor n/2 \rfloor \lg \lfloor n/2 \rfloor) + n \\ &\leq cn \lg (n/2) + n \\ &= cn \lg n - cn \lg 2 + n \\ &= cn \lg n - cn + n \\ &\leq cn \lg n \qquad \text{for } c \geq 1. \end{split}$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

We are not done yet! We only showed:

The guess holds for  $\lfloor n/2 \rfloor \Rightarrow$  The guess holds for n

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

We are not done yet! We only showed:

The guess holds for  $\lfloor n/2 \rfloor \Rightarrow$  The guess holds for n

We should also show:

The guess holds for  $\lfloor n/4 \rfloor \Rightarrow$  The guess holds for  $\lfloor n/2 \rfloor$ 

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

We are not done yet! We only showed:

The guess holds for  $\lfloor n/2 \rfloor \Rightarrow$  The guess holds for n

We should also show:

The guess holds for  $\lfloor n/8 \rfloor \Rightarrow$  The guess holds for  $\lfloor n/4 \rfloor$ 

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

We are not done yet! We only showed:

The guess holds for  $\lfloor n/2 \rfloor \Rightarrow$  The guess holds for n

We should also show:

The guess holds for  $\lfloor n/16 \rfloor \Rightarrow$  The guess holds for  $\lfloor n/8 \rfloor$  ...and so on.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Guess:

$$T(n) \le cn \lg n$$

We are not done yet! We only showed:

The guess holds for  $\lfloor n/2 \rfloor \Rightarrow$  The guess holds for n

We should also show:

The guess holds for  $\lfloor n/16 \rfloor \Rightarrow$  The guess holds for  $\lfloor n/8 \rfloor$  ...and so on.

We must show that the base case is satisfied.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

Let us check:

$$T(n) \le cn \lg n$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

Let us check:

$$T(1) \le c1 \lg 1$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

Let us check:

$$1 \le 0$$
 fail

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

If the base case was different, say T(2)=3, we would get:

$$T(2) \le c2 \lg 2$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

If the base case was different, say T(2) = 3, we would get:

$$3 \le c2$$

And we could easily find a c>0 satisfying the inequality.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

If the base case was different, say T(2) = 3, we would get:

$$3 \le c2$$

And we could easily find a c>0 satisfying the inequality.

Can we then "replace" the base case with something else?

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

Recall that the guess  $T(n) = O(n \lg n)$  means:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

We must show that the base case is satisfied.

Recall that the guess  $T(n) = O(n \lg n)$  means:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

Which allows us to use a different base case for the inductive proof.

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(n_0) = \cdots$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(n_0) = \cdots$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

$$n_0 = 1$$
 fail

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$T(n) \le cn \lg n$$
 for  $n \ge n_0$ 

$$n_0 = 1 \quad \text{fail}$$

$$n_0 = 2$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$T(2) \le c2 \lg 2$$
 for  $n \ge 2$ 

$$n_0 = 1 \quad \text{fail}$$

$$n_0 = 2$$

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$4 \leq c2 \qquad \text{for } n \geq 2$$

$$n_0 = 1$$
 fail  $n_0 = 2$  success

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$4 \le c2$$
 for  $n \ge 2$ 

So, let us find a good value for  $n_0$ .

$$n_0 = 1$$
 fail  $n_0 = 2$  success

However, we can not compute T(3) because we changed the base case.

## Substitution method

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$  and  $T(3) = 5$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$T(3) \le c3 \lg 3$$
 for  $n \ge 3$ 

So, let us find a good value for  $n_0$ .

$$n_0 = 1$$
 fail  $n_0 = 2$  success  $n_0 = 3$ 

## Substitution method

To solve:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$   
and  $T(2) = 4$  and  $T(3) = 5$ 

We must show that the base case is satisfied.

Recall that the guess  $O(n \lg n)$  means:

$$5 \le c3 \lg 3 \qquad \text{ for } n \ge 3$$

So, let us find a good value for  $n_0$ .

$$n_0 = 1$$
 fail  
 $n_0 = 2$  success  
 $n_0 = 3$  success

#### Exercise

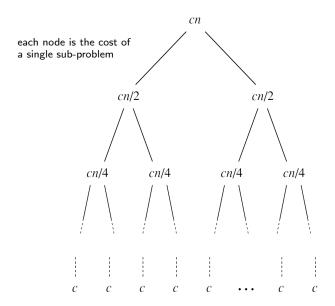
For the recursion:

$$T(n) = 2T(\lfloor n/2 \rfloor) + n$$
 and  $T(1) = 1$ 

Prove the "loose" worst-case complexity:

$$T(n) = O(n^2)$$

Use the substitution method for your proof.



Can be used to generate good guesses for the substitution method.

Can be used to generate good guesses for the substitution method.

#### Example:

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

We are interested in an upper bound for the cost.

Can be used to generate good guesses for the substitution method.

#### Example:

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

We are interested in an upper bound for the cost.

Therefore we consider the recursion:

$$T(n) = 3T(n/4) + cn^2$$

Can be used to generate good guesses for the substitution method.

#### Example:

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

We are interested in an upper bound for the cost.

Therefore we consider the recursion:

$$T(n) = 3T(n/4) + cn^2$$

We can also assume that  $n = 4^m$  for some m.

For merge sort, we also assumed that  $n=2^m$  for some m.

Can be used to generate good guesses for the substitution method.

#### Example:

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

We are interested in an upper bound for the cost.

Therefore we consider the recursion:

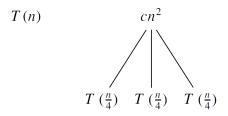
$$T(n) = 3T(n/4) + cn^2$$

We can also assume that  $n = 4^m$  for some m.

For merge sort, we also assumed that  $n=2^m$  for some m.

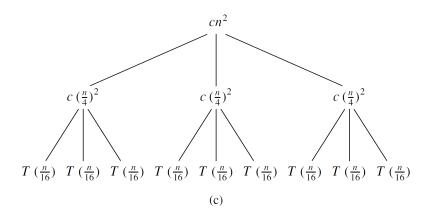
Further, we assume the base case is T(1).

$$T(n) = 3T(n/4) + cn^2$$

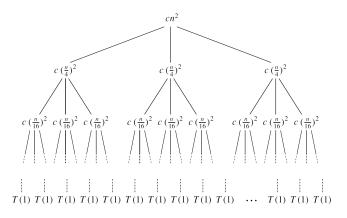


(a) (b)

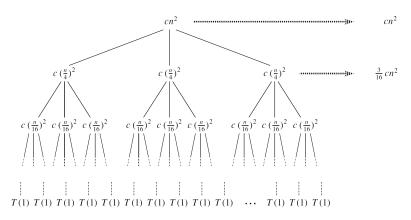
$$T(n) = 3T(n/4) + cn^2$$



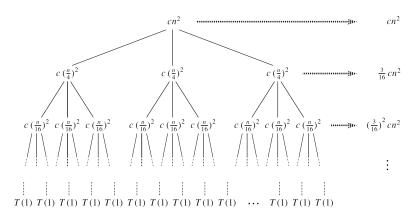
$$T(n) = 3T(n/4) + cn^2$$



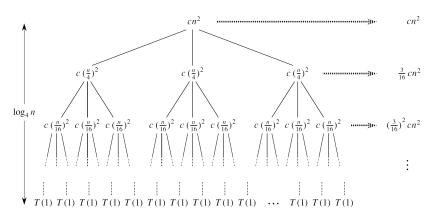
$$T(n) = 3T(n/4) + cn^2$$



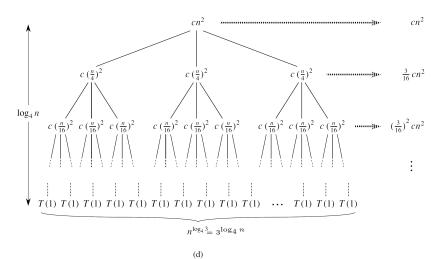
$$T(n) = 3T(n/4) + cn^2$$



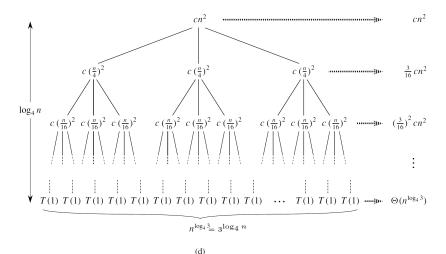
$$T(n) = 3T(n/4) + cn^2$$



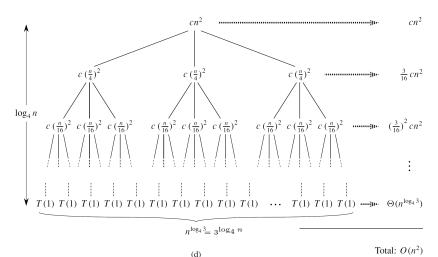
$$T(n) = 3T(n/4) + cn^2$$



$$T(n) = 3T(n/4) + cn^2$$



$$T(n) = 3T(n/4) + cn^2$$



$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

Use the substitution method with the guess  $T(n) = O(n^2)$ , that is:

$$T(n) \le dn^2$$
 for some  $d > 0$ 

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

Use the substitution method with the guess  $T(n) = O(n^2)$ , that is:

$$T(n) \le dn^2$$
 for some  $d > 0$ 

$$T(n) \le 3d\lfloor n/4\rfloor^2 + cn^2$$

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

Use the substitution method with the guess  $T(n)={\cal O}(n^2)$ , that is:

$$T(n) \le dn^2$$
 for some  $d > 0$ 

$$T(n) \le 3d\lfloor n/4\rfloor^2 + cn^2$$
  
$$\le 3d(n/4)^2 + cn^2$$

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

Use the substitution method with the guess  $T(n) = O(n^2)$ , that is:

$$T(n) \le dn^2$$
 for some  $d > 0$ 

$$T(n) \le 3d\lfloor n/4\rfloor^2 + cn^2$$
$$\le 3d(n/4)^2 + cn^2$$
$$= \frac{3}{16}dn^2 + cn^2$$

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

Use the substitution method with the guess  $T(n) = O(n^2)$ , that is:

$$T(n) \le dn^2$$
 for some  $d > 0$ 

$$\begin{split} T(n) & \leq 3d \lfloor n/4 \rfloor^2 + cn^2 \\ & \leq 3d(n/4)^2 + cn^2 \\ & = \frac{3}{16} dn^2 + cn^2 \\ & \leq dn^2 \qquad \text{for } d \geq \frac{16}{13} c \end{split}$$

Applies to recursion of the form:

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

with  $a \ge 1, b > 1$ .

Applies to recursion of the form:

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

with  $a \ge 1, b > 1$ .

We always assume f(n) to be asymptotically positive.

ullet We have a sub-problems.

Applies to recursion of the form:

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

with  $a \ge 1, b > 1$ .

- We have a sub-problems.
- Each has size n/b.

Applies to recursion of the form:

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

with  $a \ge 1, b > 1$ .

- We have a sub-problems.
- Each has size n/b.
- The cost of dividing + combining is f(n) (i.e. the root of the tree).

Applies to recursion of the form:

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

with  $a \ge 1, b > 1$ .

- ullet We have a sub-problems.
- Each has size n/b.
- The cost of dividing + combining is f(n) (i.e. the root of the tree).
- By n/b we mean either  $\lfloor n/b \rfloor$  or  $\lceil n/b \rceil$ .

$$\bullet \ \ \, \text{If } f(n)=O(n^{\log_b a-\epsilon}) \text{ for some } \epsilon>0 \text{, then } T(n)=\Theta(n^{\log_b a}).$$

- $\bullet \ \, \text{If} \, \, f(n) = O(n^{\log_b a \epsilon}) \, \, \text{for some} \, \, \epsilon > 0, \, \text{then} \, \, T(n) = \Theta(n^{\log_b a}).$
- 2 If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .

- $\bullet \ \, \text{If} \, \, f(n) = O(n^{\log_b a \epsilon}) \, \, \text{for some} \, \, \epsilon > 0 \text{, then} \, \, T(n) = \Theta(n^{\log_b a}).$
- 2 If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .
- $\textbf{ 3} \ \, \text{If} \, \, f(n) = \Omega(n^{\log_b a + \epsilon}) \, \, \text{for some} \, \, \epsilon > 0, \, \text{and if} \, \, af(n/b) \leq cf(n) \, \, \text{for some} \, \, c < 1 \, \, \text{and large} \, \, n, \, \, \text{then} \, \, T(n) = \Theta(f(n)).$

We use the master theorem, which can be applied in three cases:

- $\bullet \ \, \text{If} \, \, f(n) = O(n^{\log_b a \epsilon}) \, \, \text{for some} \, \, \epsilon > 0 \text{, then} \, \, T(n) = \Theta(n^{\log_b a}).$
- 2 If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .
- $\textbf{ If } f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ for some } \epsilon > 0 \text{, and if } af(n/b) \leq cf(n) \text{ for some } c < 1 \text{ and large } n \text{, then } T(n) = \Theta(f(n)).$

Intuitively: compare f(n) with  $n^{\log_b a}$  and see who's bigger.

We use the master theorem, which can be applied in three cases:

- $\bullet \ \, \text{If} \,\, f(n) = O(n^{\log_b a \epsilon}) \,\, \text{for some} \,\, \epsilon > 0, \,\, \text{then} \,\, T(n) = \Theta(n^{\log_b a}).$
- $\textbf{2} \ \text{ If } f(n) = \Theta(n^{\log_b a}) \text{, then } T(n) = \Theta(n^{\log_b a} \lg n).$
- $\textbf{ If } f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ for some } \epsilon > 0 \text{, and if } af(n/b) \leq cf(n) \text{ for some } c < 1 \text{ and large } n \text{, then } T(n) = \Theta(f(n)).$

Intuitively: compare f(n) with  $n^{\log_b a}$  and see who's bigger.

We can read the theorem as follows:

① The total cost is dominated by the base cases. f(n) must be polynomially smaller than  $n^{\log_b a}$  (i.e. by a factor  $n^{\epsilon}$ ).

We use the master theorem, which can be applied in three cases:

- $\bullet \ \, \text{If} \, \, f(n) = O(n^{\log_b a \epsilon}) \, \, \text{for some} \, \, \epsilon > 0, \, \text{then} \, \, T(n) = \Theta(n^{\log_b a}).$
- 2 If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .
- $\textbf{ If } f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ for some } \epsilon > 0 \text{, and if } af(n/b) \leq cf(n) \text{ for some } c < 1 \text{ and large } n \text{, then } T(n) = \Theta(f(n)).$

Intuitively: compare f(n) with  $n^{\log_b a}$  and see who's bigger.

We can read the theorem as follows:

- ① The total cost is dominated by the base cases. f(n) must be polynomially smaller than  $n^{\log_b a}$  (i.e. by a factor  $n^{\epsilon}$ ).
- 2 The total cost is distributed across the levels of the recursion tree.

We use the master theorem, which can be applied in three cases:

- $\bullet \ \, \text{If} \, \, f(n) = O(n^{\log_b a \epsilon}) \, \, \text{for some} \, \, \epsilon > 0, \, \text{then} \, \, T(n) = \Theta(n^{\log_b a}).$
- 2 If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .
- ① If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some  $\epsilon > 0$ , and if  $af(n/b) \le cf(n)$  for some c < 1 and large n, then  $T(n) = \Theta(f(n))$ .

Intuitively: compare f(n) with  $n^{\log_b a}$  and see who's bigger.

We can read the theorem as follows:

- ① The total cost is dominated by the base cases. f(n) must be polynomially smaller than  $n^{\log_b a}$  (i.e. by a factor  $n^{\epsilon}$ ).
- 2 The total cost is distributed across the levels of the recursion tree.
- **3** The total cost is dominated by the root. f(n) must be polynomially larger than  $n^{\log_b a}$  and be regular.

$$T(n) = 9T(n/3) + n$$

We get  $n^{\log_b a} = \Theta(n^2)$  and  $f(n) = O(n^{\log_3 9 - \epsilon})$  with  $\epsilon = 1$ . Thus, we are in case (1) and the solution is  $T(n) = \Theta(n^2)$ .

$$T(n) = 9T(n/3) + n$$

We get  $n^{\log_b a} = \Theta(n^2)$  and  $f(n) = O(n^{\log_3 9 - \epsilon})$  with  $\epsilon = 1$ . Thus, we are in case (1) and the solution is  $T(n) = \Theta(n^2)$ .

$$T(n) = T(2n/3) + 1$$

Here  $n^{\log_b a} = 1$  and case (2) applies, hence  $T(n) = \Theta(\lg n)$ .

$$T(n) = 9T(n/3) + n$$

We get  $n^{\log_b a} = \Theta(n^2)$  and  $f(n) = O(n^{\log_3 9 - \epsilon})$  with  $\epsilon = 1$ . Thus, we are in case (1) and the solution is  $T(n) = \Theta(n^2)$ .

$$T(n) = T(2n/3) + 1$$

Here  $n^{\log_b a} = 1$  and case (2) applies, hence  $T(n) = \Theta(\lg n)$ .

$$T(n) = 3T(n/4) + n \lg n$$

Here  $n^{\log_b a} = O(n^{0.793})$  and  $f(n) = \Omega(n^{\log_4 3 + \epsilon})$ . Further, we can show that the regularity condition holds for f(n). Case (3) applies, hence  $T(n) = \Theta(n \lg n)$ .

$$T(n) = 9T(n/3) + n$$

We get  $n^{\log_b a} = \Theta(n^2)$  and  $f(n) = O(n^{\log_3 9 - \epsilon})$  with  $\epsilon = 1$ . Thus, we are in case (1) and the solution is  $T(n) = \Theta(n^2)$ .

$$T(n) = T(2n/3) + 1$$

Here  $n^{\log_b a} = 1$  and case (2) applies, hence  $T(n) = \Theta(\lg n)$ .

$$T(n) = 3T(n/4) + n \lg n$$

Here  $n^{\log_b a} = O(n^{0.793})$  and  $f(n) = \Omega(n^{\log_4 3 + \epsilon})$ . Further, we can show that the regularity condition holds for f(n). Case (3) applies, hence  $T(n) = \Theta(n \lg n)$ .

$$T(n) = 2T(n/2) + n \lg n$$

The master theorem does not apply here, since  $f(n) = n \lg n$  is not polynomially larger than  $n^{\log_b a} = n$ . In fact,  $\frac{f(n)}{n^{\log_b a}} = \frac{n \lg n}{n} = \lg n < n^\epsilon$  asymptotically.

# Suggested reading

Chapters 4.1, 4.2, 4.3 of:

"Introduction to Algorithms – 2nd Ed.", Cormen et al.