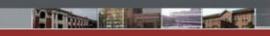


Chapter 4: Recurrences

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Overview



A recurrence is a function is defined in terms of

- one or more base cases, and
- itself, with smaller arguments



Examples:

$$T(n) = \begin{bmatrix} 1 & \text{if } n = 1 \\ T(n-1) + 1 & \text{if } n > 1 \end{bmatrix}$$

Solution: T(n) = n.

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

Solution: $T(n) = n \lg n + n$.

$$T(n) = \begin{bmatrix} 0 & \text{if } n = 2 \\ T(\sqrt{n}) + 1 & \text{if } n > 2 \end{bmatrix}$$

Solution: $T(n) = \lg \lg n$.

$$T(n) = \begin{bmatrix} 1 & \text{if } n = 1 \\ T(n/3) + T(2n/3) + n & \text{if } n > 1 \end{bmatrix}$$

Solution: $T(n) = \Theta(n \lg n)$

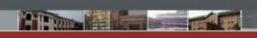




Many technical issues:

- Floors and ceilings
 [Floors and ceilings can easily be removed and don't affect the solution to the recurrence.]
- Exact vs. asymptotic functions
- Boundary condition



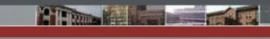


In algorithm analysis, we usually **express both the recurrence and its solution using asymptotic notation**.

- **E.g.** , with solution
- The boundary conditions are usually expressed as "T(n) = O(1) for sufficiently small n."
- ▶ (代表常數)
- When we desire an exact, rather than an asymptotic, solution, we need to deal with boundary conditions.
- In practice, we just use asymptotic most of the time, and we ignore boundary conditions.

Substitution method





- 1. Guess the solution.
- 2. Use induction to find the constants and show that the solution works.

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

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.]

Induction:

Bass:
$$n = 1 \rightarrow n \lg n + n = 1 = T(n)$$

Inductive step: Inductive hypothesis is that $T(k) = k \lg k + k$ for all k < kn.

We'll use this inductive hypothesis for
$$T(n/2)$$
. $T(n) = 2T \begin{bmatrix} -1 \\ -1 \end{bmatrix} + n$

$$T(n) = 2T \begin{bmatrix} -n \\ -1 \end{bmatrix} + n$$

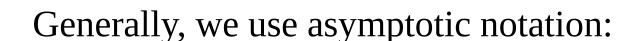
$$=2\left[\frac{n}{2}\lg\frac{n}{2}+\frac{n}{2}\right]+n$$

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

$$= n(\lg n - \lg 2) + n + n$$

$$=$$
 $n \lg n - n + n + n$.

$$=n \lg n + n$$



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

- Assume T(n) = O(1) for sufficiently small n
- Express the solution by asymptotic notation:
- $T(n) = \Theta(n \lg n)$. Don't worry about boundary cases, nor do we show base cases in the substitution proof.



- T(n) is always constant for any constant n.
- Since we are ultimately interested in asymptotic solution to a recurrence, it will always be possible to choose base cases that work
- When we want an asymptotic solution to a recurrence, we don't worry about the base cases in our proofs.
- When we want an exact solution, then we have to deal with base cases.

Subs

Substitution method (cont'd)

For the substitution method:

- Name the constant in the additive term
- ▶ Show the upper (O) and lower (Ω) bounds separately. Might need to use different constants for each notation

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





1. Upper bound:

Guess: $T^{(n)} \le dn \lg n$ 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:
$$T(n) \leq 2T(n/2) + cn$$

$$= 2 \left[\frac{1}{2} d \frac{n}{2} \lg \frac{n}{2} \right] + cn$$

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

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

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

Therefore, $T(n) = O(n \lg n)$

 $d \ge c$



2. Lower bound:

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



Consider the recurrence

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

For an upper bound:

Guess:
$$T(n) \le dn^3$$

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$$T(n) \leq 8T(n/2) + cn^2$$

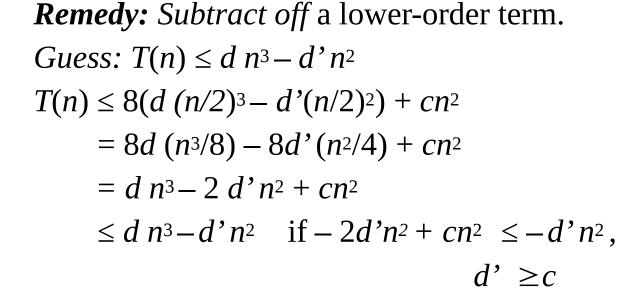
$$T(n) \leq 8d(n/2)^{3} + cn^{2}$$

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

$$= dn^{3} + cn^{2}$$

$$\leq dn^{3}$$

doesn't work!







Be careful when using asymptotic notation.

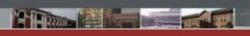
The false proof for the recurrence T(n) = 4T(n/4) + n, that T(n) = O(n):

$$T(n) \le 4(c(n/4)) + n$$

 $\le cn + n$
 $= O(n)$ wrong!

Because we haven't proven the *exact from* of our inductive hypothesis (which is that $T(n) \le cn$), this proof is false.



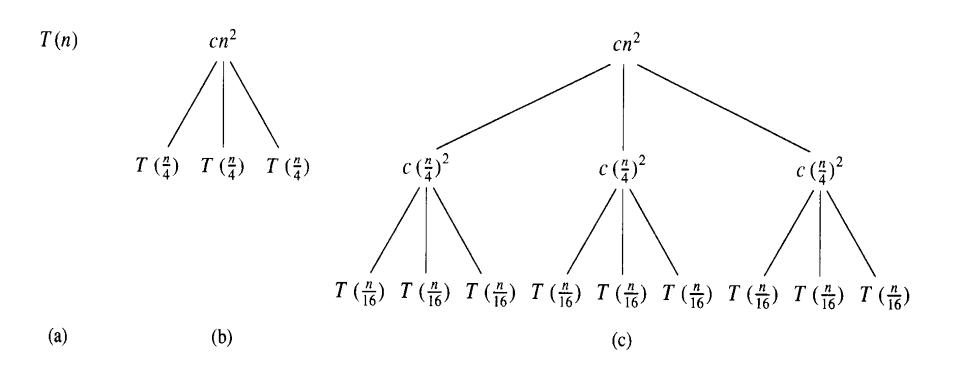


- Goal of the recursion-tree method
 - a good guess for the substitution method
 - ▷ a direct proof of a solution to a recurrence (provided by carefully drawing a recursion tree)

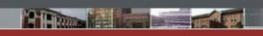


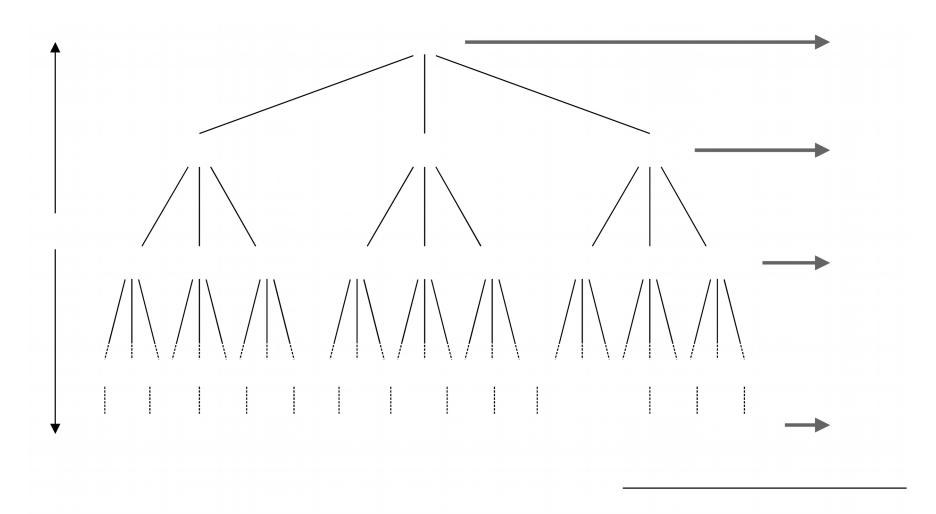


$$T(n) = 3T([n/4]) + \Theta(n^2)$$









Recurrence trees



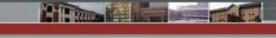
▶ The cost of the entire tree

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

$$= \sum_{i=0}^{\log_{4}n-1} \left[\frac{3}{16}\right]^{i}cn^{2} + \Theta(n^{\log_{4}3})$$

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





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

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

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

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

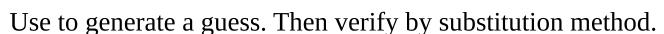
$$= O(n^2)$$



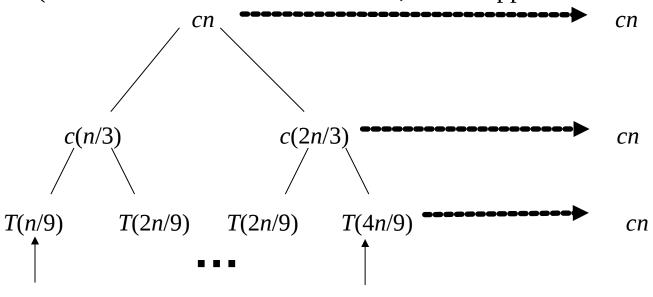


- Verify by the substitution method
 - Show that $T(n) \le dn^2$ for some constant d > 0 $T(n) \le 3T([n/4]) + cn^2$ $\le 3d[n/4]^2 + cn^2$ $\le 3d(n/4)^2 + cn^2$ $= \frac{3}{16}dn^2 + cn^2$ $\le dn^2,$

where the last step holds as long as $d \ge (16/13) c$



E.g.: $T(n) = T(n/3) + T(2n/3) + \Theta(n)$. For upper bound, rewrite as $T(n) \le T(n/3) + T(2n/3) + cn$; for lower bound, as $T(n) \ge T(n/3) + T(2n/3) + cn$. By summing across each level, the recursion tree shows the cost at each level of recursion (minus the costs of recursive calls, which appear in subtrees):



leftmost branch peters out after $log_3 n$ levels

rightmost branch peters out after $\log_{3/2} n$ levels

Recurrence trees (cont'd)

- There are $\log_3 n$ full levels, and after $\log_{3/2} n$ levels, the problem size is down to 1.
- \blacktriangleright Each level contributes $\leq cn$.
- Lower bound guess: $\geq d n \log_3 n = \Omega(n \log n)$ for some positive constant d.
- Upper bound guess: $\leq d n \log_{3/2} n = O(n \log n)$ for some positive constant d.
- Then *prove* by substitution.

Recurrence trees (cont'd)





1. Upper bound:

Guess: $T(n) \leq dn \lg n$.

Substitution:

$$T(n) \leq T(n/3) + T(2n/3) + cn$$

$$\leq d(n/3) \lg(n/3) + d(2n/3) \lg(2n/3) + cn$$

$$= (d(n/3) \lg n - d(n/3) \lg 3) + (d(2n/3) \lg n - d(2n/3) \lg(3/2)) + cn$$

$$= d n \lg n - d((n/3) \lg 3 + (2n/3) \lg(3/2)) + cn$$

$$= d n \lg n - d((n/3) \lg 3 + (2n/3) \lg 3 - (2n/3) \lg 2) + cn$$

$$= d n \lg n - d n (\lg 3 - 2/3) + cn$$

$$\leq d n \lg n \qquad \text{if } -d n (\lg 3 - 2/3) + cn \leq 0, \quad c$$

$$d \geq \frac{1}{\lg 3 - 2/3}$$
Therefore, $T(n) = O(n \lg n)$.

Note: Make sure that symbolic constants used in the recurrence (e.g.,c) and the guess (e.g.,d) are different.

Recurrence trees (cont'd)





2. Lower bound:

Guess: $T(n) \ge dn \lg n$.

Substitution: Same as for the upper bound, but replacing \leq by \geq . End up

needing

$$0 < d \le \frac{c}{\lg 3 - 2/3}$$

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

Since $T(n) = O(n \lg n)$ and $T(n) = \Omega(n \lg n)$, we conclude that $T(n) = \Theta(n \lg n)$



Used for many divide-and-conquer recurrences of the form

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

Where $a \ge 1$, b > 1, and f(n) > 0.

Based on the *master theorem* (Theorem 4.1).

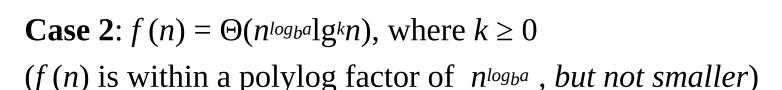
Compare n^{logba} vs. f(n):

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

(f(n) is polynomially smaller than n^{logba})

Solution: $T(n) = \Theta(n^{\log_{ba}})$

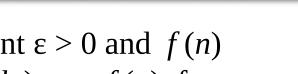




Solution: $T(n) = \Theta(n^{\log_{b^a}} \lg^{k+1} n)$

Simple case:

$$k = 0 \rightarrow f(n) = \Theta(n^{\log_b a}) \rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$$



Case 3: $f(n) = \Omega$ ($n^{\log_b a + \varepsilon}$) for some constant $\varepsilon > 0$ and f(n)satisfies the regularity condition $a f(n/b) \le c f(n)$ for some constant c < 1 and all sufficiently large n.

(f(n) is polynomially greater than n^{logba})

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

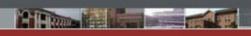




What's with the Case 3 regularity condition?

- Generally not a problem.
- It always holds whenever $f(n) = n^k$ and $f(n) = \Omega(n^{\log_{b^{a+\epsilon}}})$ for constant $\epsilon > 0$. So you don't need to check it when f(n) is a polynomial.

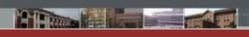
Master method (cond't)



Examples:

- ► $T(n) = 5T(n/2) + \Theta(n^2)$ $n^{\log_2 5}$ vs. n^2 Since $\log_2 5 - \varepsilon = 2$ for some constant $\varepsilon > 0$, use Case $1 \rightarrow T(n) = \Theta(n^{\lg 5})$
- ► $T(n) = 27T(n/3) + \Theta(n^3 \lg n)$ $n^{\log_3 27} = n^3 \text{ vs. } n^3 \lg n$ Use Case 2 with $k = 1 \rightarrow T(n) = \Theta(n^3 \lg^2 n)$





T(n) = 5T(n/2) + Θ(n³) $n^{\log_2 5}$ vs. n^3

Now $\lg 5 + \varepsilon = 3$ for some constant $\varepsilon > 0$

Check regularity condition (don't really need to since f(n) is a polynomial):

$$a f(n/b) = 5(n/2)^3 = 5n^3/8 \le cn^3$$
 for c = 5/8 <1
Use Case 3 → $T(n) = \Theta(n^3)$

► $T(n) = 27T(n/3) + \Theta(n^3/\lg n)$ $n^{\log_3 27} = n^3 \text{ vs. } n^3/\lg n = n^3 \lg^{-1} n \neq \Theta(n^3 \lg^k n) \text{ for any } k \geq 0$ Cannot use the master method.