

Asymptotic Efficiency of Recurrences

- Find the asymptotic bounds of recursive equations.
 - Substitution method
 - domain transformation
 - Changing variable
 - Recursive tree method
 - Master method (master theorem)
 - Provides bounds for: $T(n) = aT(n/b) + f(n)$ where
 - $a \geq 1$ (the number of subproblems).
 - $b > 1$, (n/b is the size of each subproblem).
 - $f(n)$ is a given function.

Recurrences

- MERGE-SORT

- Contains details:

- $T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + \Theta(n) & \text{if } n>1 \end{cases}$

- Ignore details, $T(n) = 2T(n/2) + \Theta(n)$.

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

The Substitution Method

- Two steps:
 1. Guess the form of the solution.
 - By experience, and creativity.
 - By some heuristics.
 - If a recurrence is similar to one you have seen before.
 - » $T(n)=2T(\lfloor n/2 \rfloor)+17)+n$, similar to $T(n)=2T(\lfloor n/2 \rfloor)+n$, , guess $O(n \lg n)$.
 - Prove loose upper and lower bounds on the recurrence and then reduce the range of uncertainty.
 - » For $T(n)=2T(\lfloor n/2 \rfloor)+n$, prove lower bound $T(n)=\Omega(n)$, and prove upper bound $T(n)=O(n^2)$, then guess the tight bound is $T(n)=O(n \lg n)$.
 - By recursion tree.
 2. Use mathematical induction to find the constants and show that the solution works.

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

- Guess the solution: $T(n)=O(n \lg n)$,
 - i.e., $T(n) \leq cn \lg n$ for **some** c .
- Prove the solution **by induction**:
 - Suppose this bound holds for $\lfloor n/2 \rfloor$, i.e.,
 - $T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg (\lfloor n/2 \rfloor)$.
 - $T(n) \leq 2(c \lfloor n/2 \rfloor \lg (\lfloor n/2 \rfloor)) + n$
 - $\leq cn \lg (n/2) + n$
 - $= cn \lg n - cn \lg 2 + n$
 - $= cn \lg n - cn + n$
 - $\leq cn \lg n$ (as long as $c \geq 1$)

Question: Is the above proof complete? Why?

Boundary (base) Condition

- In fact, $T(n) = 1$ if $n=1$, i.e., $T(1)=1$.
- However, $c \lg n = c \times 1 \times \lg 1 = 0$, which is odd with $T(1)=1$.
- Take advantage of asymptotic notation: it is required $T(n) \leq c \lg n$ hold for $n \geq n_0$ where n_0 is a constant of our choosing.
- Select $n_0 = 2$, thus, $n=2$ and $n=3$ as our induction bases. It turns out any $c \geq 2$ suffices for base cases of $n=2$ and $n=3$ to hold.

Subtleties

- Guess is correct, but induction proof not work.
- Problem is that inductive assumption not strong enough.
- Solution: revise the guess by subtracting a lower-order term.
- Example: $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + 1$.
 - Guess $T(n) = O(n)$, i.e., $T(n) \leq cn$ for some c .
 - However, $T(n) \leq c \lfloor n/2 \rfloor + c \lceil n/2 \rceil + 1 = cn + 1$, which does not imply $T(n) \leq cn$ for any c .
 - Attempting $T(n) = O(n^2)$ will work, but overkill.
 - New guess $T(n) \leq cn - b$ will work as long as $b \geq 1$.
 - (Prove it in an exact way).

Avoiding Pitfall

- It is easy to guess $T(n)=O(n)$ (i.e., $T(n) \leq cn$) for $T(n)=2T(\lfloor n/2 \rfloor)+n$.
- And wrongly prove:
 - $T(n) \leq 2(c \lfloor n/2 \rfloor)+n$
 - $\leq cn+n$
 - $=O(n)$. ← wrongly !!!!
- Problem is that it does not prove the *exact form* of $T(n) \leq cn$.

Find bound, ceiling, floor, lower term— domain transformation

- Find the bound: $T(n)=2T(n/2)+n$ ($O(n\log n)$)
- How about $T(n)=2T(\lfloor n/2 \rfloor)+n$?
- How about $T(n)=2T(\lceil n/2 \rceil)+n$?
 - $T(n)\leq 2T(n/2+1)+n$
 - Domain transformation
 - Set $S(n)=T(n+a)$ and assume $S(n)\leq 2S(n/2)+O(n)$ (so $S(n)=O(n\log n)$)
 - $S(n)\leq 2S(n/2)+O(n) \rightarrow T(n+a)\leq 2T(n/2+a)+O(n)$
 - $T(n)\leq 2T(n/2+1)+n \rightarrow T(n+a)\leq 2T((n+a)/2+1)+n+a$
 - Thus, set $n/2+a=(n+a)/2+1$, get $a=2$.
 - so $T(n)=S(n-2)=O((n-2)\log(n-2)) = O(n\log n)$.
- How about $T(n)=2T(n/2+19)+n$?
 - Set $S(n)=T(n+a)$ and get $a=38$.
- As a result, **ceiling, floor, and lower terms will not affect.**
 - Moreover, the master theorem also provides proof for this.

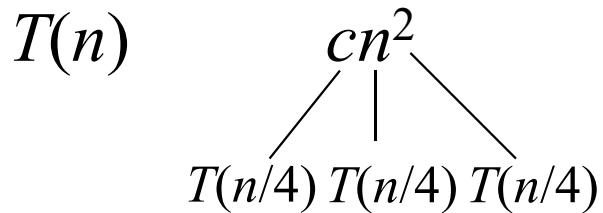
Changing Variables

- Suppose $T(n)=2T(\sqrt{n})+\lg n$.
- Rename $m=\lg n$. so $T(2^m)=2T(2^{m/2})+m$.
- Domain transformation:
 - $S(m)=T(2^m)$, so $S(m)=2S(m/2)+m$.
 - Which is similar to $T(n)=2T(n/2)+n$.
 - So the solution is $S(m)=O(m \lg m)$.
 - Changing back to $T(n)$ from $S(m)$, the solution is $T(n)=T(2^m)=S(m)=O(m \lg m)=O(\lg n \lg \lg n)$.

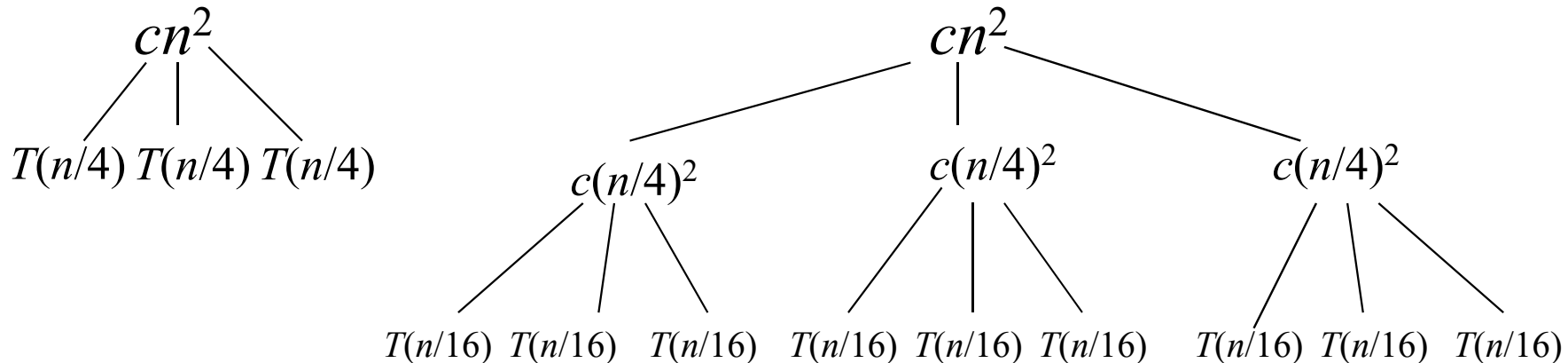
The Recursion-tree Method

- Idea:
 - Each node represents the cost of a single subproblem.
 - Sum up the costs with each level to get level cost.
 - Sum up all the level costs to get total cost.
- Particularly suitable for divide-and-conquer recurrence.
- Best used to generate a good guess, tolerating “sloppiness”.
- If trying carefully to draw the recursion-tree and compute cost, then used as direct proof.

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

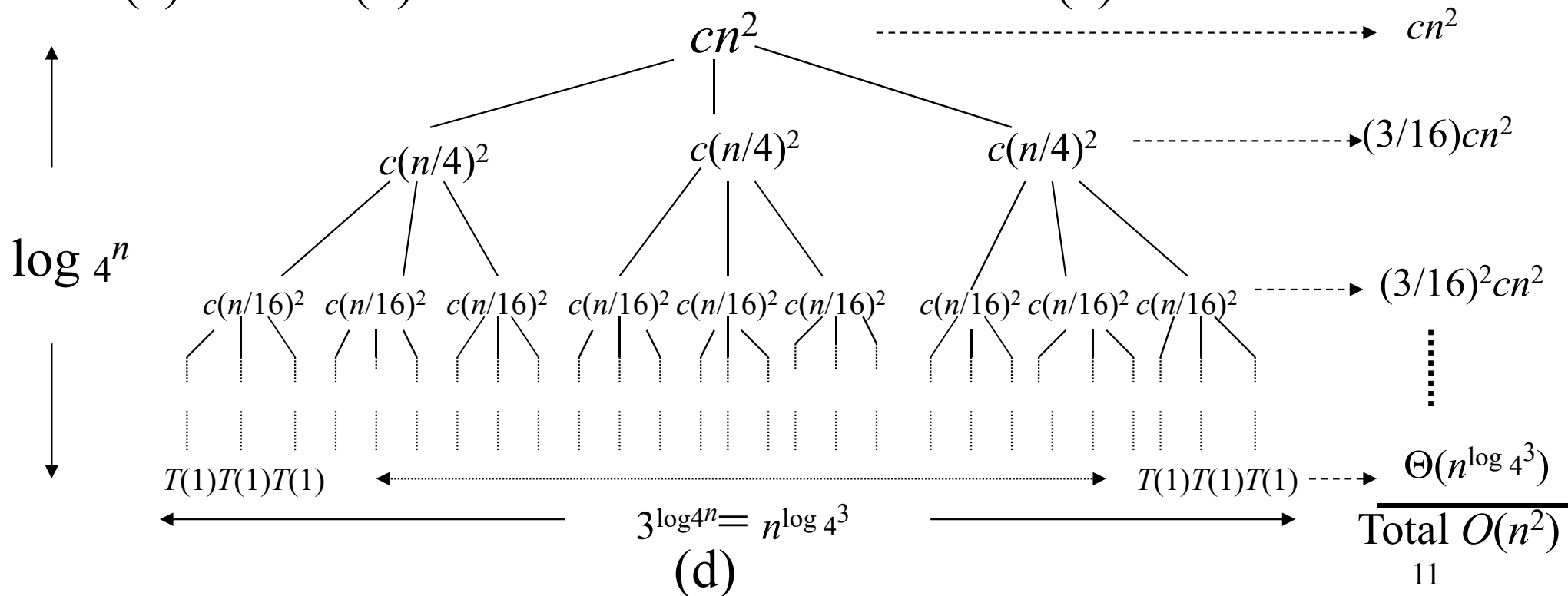


(a)



(b)

(c)



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

- The height is $\log_4 n$,
- #leaf nodes = $3^{\log_4 n} = n^{\log_4 3}$. Leaf node cost: $T(1)$.
- Total cost $T(n)=cn^2+(3/16)cn^2+(3/16)^2cn^2+\dots+(3/16)^{\log_4 n-1}cn^2+\Theta(n^{\log_4 3})$
 $= (1+3/16+(3/16)^2+\dots+(3/16)^{\log_4 n-1})cn^2+\Theta(n^{\log_4 3})$
 $< (1+3/16+(3/16)^2+\dots+(3/16)^m+\dots)cn^2+\Theta(n^{\log_4 3})$
 $= (1/(1-3/16))cn^2+\Theta(n^{\log_4 3})$
 $= 16/13cn^2+\Theta(n^{\log_4 3})$
 $= O(n^2)$.

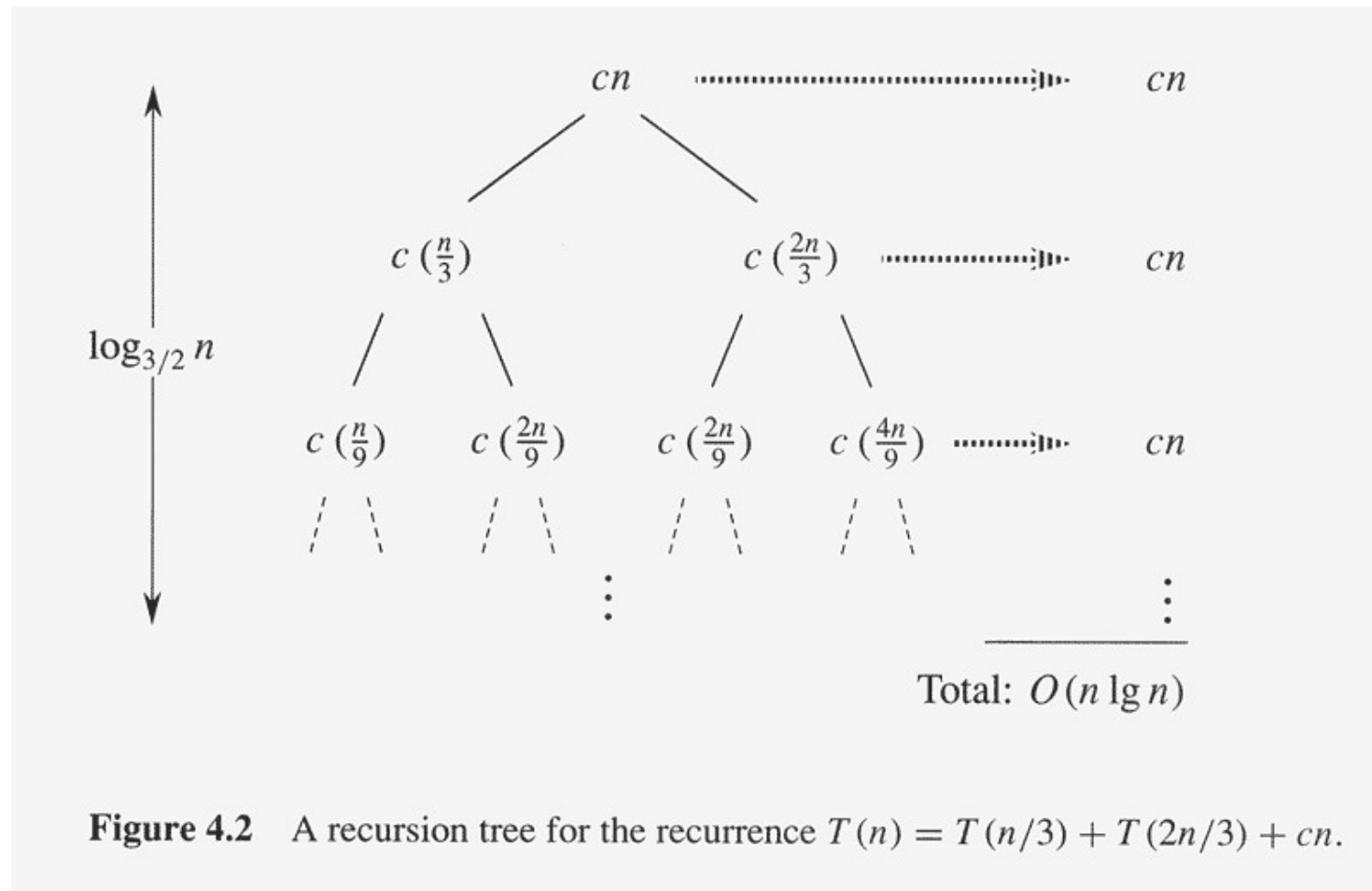
Prove the above Guess

- $T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2) = O(n^2)$.
- Show $T(n) \leq dn^2$ for some d .
- $$\begin{aligned} T(n) &\leq 3(d \lfloor n/4 \rfloor^2) + cn^2 \\ &\leq 3(d (n/4)^2) + cn^2 \\ &= 3/16(dn^2) + cn^2 \\ &\leq dn^2, \quad \text{as long as } d \geq (16/13)c. \end{aligned}$$

One more example

- $T(n) = T(n/3) + T(2n/3) + O(n)$.
- Construct its recursive tree ([Figure 4.2, page 71](#)).
- $T(n) = O(n \lg_{3/2} n) = O(n \lg n)$.
- Prove $T(n) \leq dn \lg n$.

Recursion Tree of $T(n) = T(n/3) + T(2n/3) + O(n)$



Master Method/Theorem

- Theorem 4.1 (page 73)
 - for $T(n) = aT(n/b) + f(n)$, n/b may be $\lceil n/b \rceil$ or $\lfloor n/b \rfloor$.
 - where $a \geq 1$, $b > 1$ are positive integers, $f(n)$ be a non-negative function.
 1. If $f(n) = O(n^{\log_b a - \varepsilon})$ for some $\varepsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
 2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
 3. If $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some $\varepsilon > 0$, and if $af(n/b) \leq cf(n)$ for some $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$.

Implications of Master Theorem

- Comparison between $f(n)$ and $n^{\log_b a}$ ($<, =, >$)
- Must be asymptotically smaller (or larger) by a polynomial, i.e., n^ϵ for some $\epsilon > 0$.
- In case 3, the “regularity” must be satisfied, i.e., $af(n/b) \leq cf(n)$ for some $c < 1$.
- There are gaps
 - between 1 and 2: $f(n)$ is smaller than $n^{\log_b a}$, but not polynomially smaller.
 - between 2 and 3: $f(n)$ is larger than $n^{\log_b a}$, but not polynomially larger.
 - in case 3, if the “regularity” fails to hold.

Application of Master Theorem

- $T(n) = 9T(n/3) + n$;
 - $a=9, b=3, f(n)=n$
 - $n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$
 - $f(n) = O(n^{\log_3 9 - \epsilon})$ for $\epsilon=1$
 - By case 1, $T(n) = \Theta(n^2)$.
- $T(n) = T(2n/3) + 1$
 - $a=1, b=3/2, f(n)=1$
 - $n^{\log_b a} = n^{\log_{3/2} 1} = \Theta(n^0) = \Theta(1)$
 - By case 2, $T(n) = \Theta(\lg n)$.

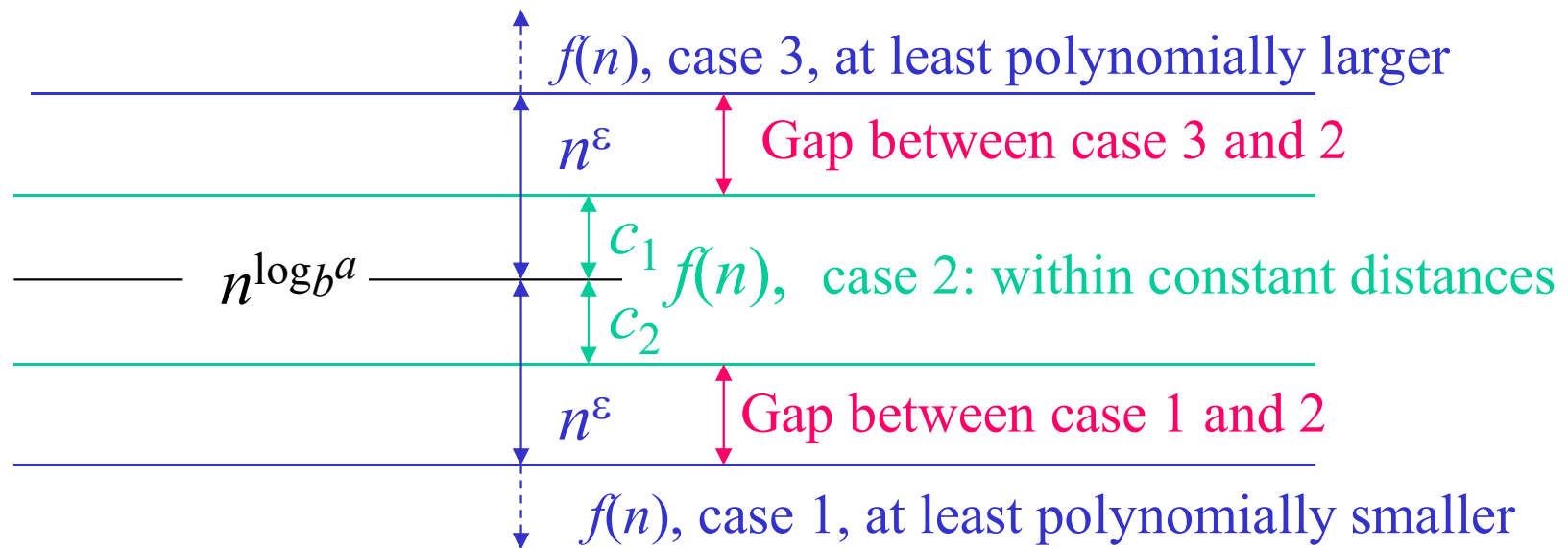
Application of Master Theorem

- $T(n) = 3T(n/4) + n \lg n$;
 - $a=3, b=4, f(n) = n \lg n$
 - $n^{\log_b a} = n^{\log_4 3} = \Theta(n^{0.793})$
 - $f(n) = \Omega(n^{\log_4 3 + \epsilon})$ for $\epsilon \approx 0.2$
 - Moreover, for large n , the “regularity” holds for $c=3/4$.
 - $af(n/b) = 3(n/4) \lg(n/4) \leq (3/4)n \lg n = cf(n)$
 - By case 3, $T(n) = \Theta(f(n)) = \Theta(n \lg n)$.

Exception to Master Theorem

- $T(n) = 2T(n/2) + n \lg n$;
 - $a=2, b=2, f(n) = n \lg n$
 - $n^{\log_b a} = n^{\log_2 2} = \Theta(n)$
 - $f(n)$ is asymptotically larger than $n^{\log_b a}$, but not polynomially larger because
 - $f(n)/n^{\log_b a} = \lg n$, which is asymptotically less than n^ε for any $\varepsilon > 0$.
 - Therefore, this is a gap between 2 and 3.

Where Are the Gaps



- Note:
1. for case 3, the **regularity** also must hold.
 2. if $f(n)$ is **lg n** smaller, then fall in gap in 1 and 2
 3. if $f(n)$ is **lg n** larger, then fall in gap in 3 and 2
 4. if $f(n) = \Theta(n^{\log b^a} \lg^k n)$, then $T(n) = \Theta(n^{\log b^a} \lg^{k+1} n)$. (as exercise)

Proof of Master Theorem

- The proof for the exact powers, $n=b^k$ for $k \geq 1$.
- Lemma 4.2
 - for $T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ aT(n/b)+f(n) & \text{if } n=b^k \text{ for } k \geq 1 \end{cases}$
 - where $a \geq 1$, $b > 1$, $f(n)$ be a nonnegative function,
 - Then
 - $T(n) = \Theta(n^{\log_b a}) + \sum_{j=0}^{\log_b n - 1} a^j f(n/b^j)$
- Proof:
 - By iterating the recurrence
 - By recursion tree ([See figure 4.3](#))

Recursion tree for $T(n) = aT(n/b) + f(n)$

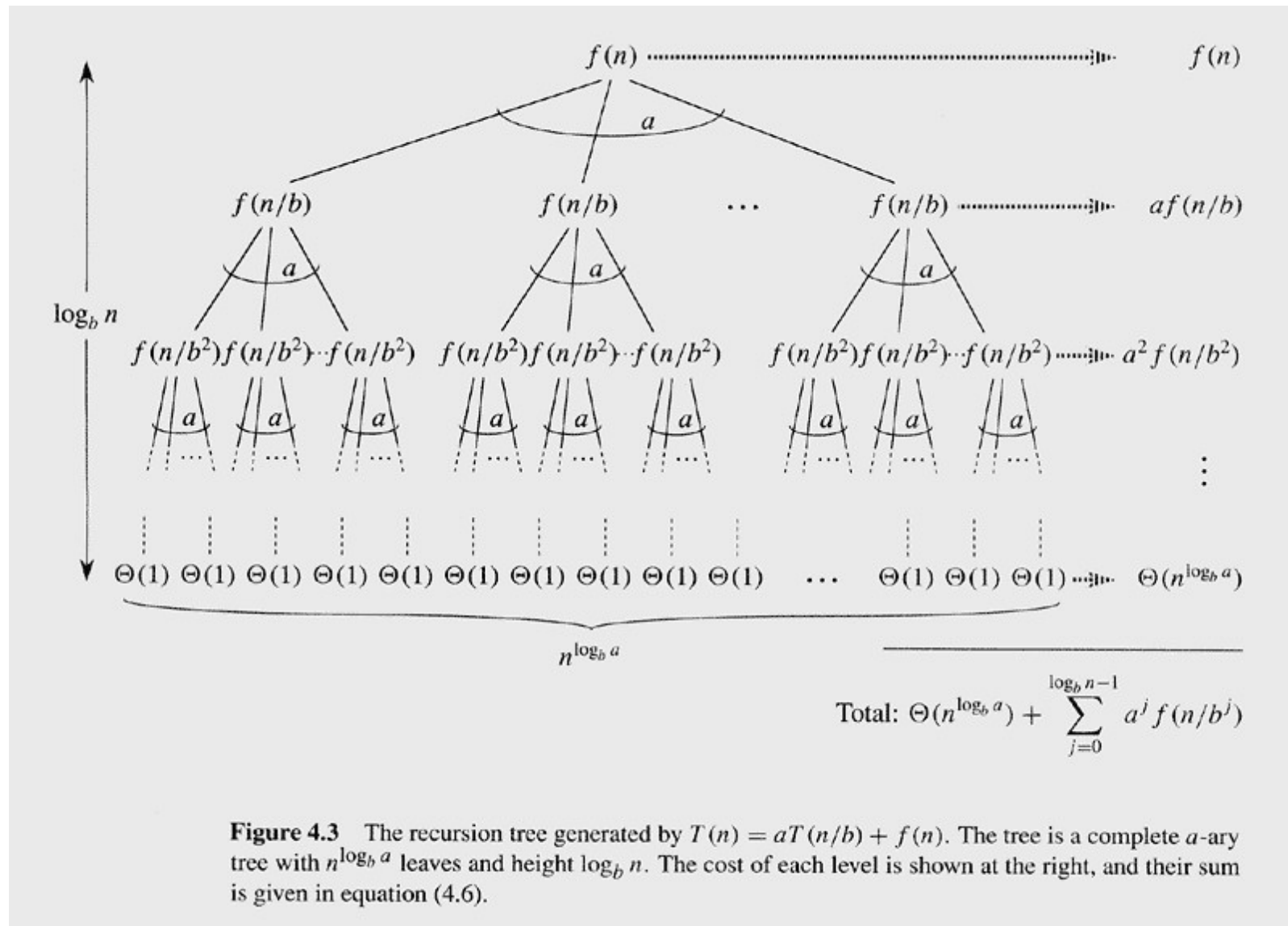


Figure 4.3 The recursion tree generated by $T(n) = aT(n/b) + f(n)$. The tree is a complete a -ary tree with $n^{\log_b a}$ leaves and height $\log_b n$. The cost of each level is shown at the right, and their sum is given in equation (4.6).

Proof of Master Theorem (cont.)

- Lemma 4.3:
 - Let $a \geq 1$, $b > 1$, $f(n)$ be a nonnegative function defined on exact power of b , then
 - $g(n) = \sum_{j=0}^{\log_b n - 1} a^j f(n/b^j)$ can be bounded for exact power of b as:
 1. If $f(n) = O(n^{\log_b a - \varepsilon})$ for some $\varepsilon > 0$, then $g(n) = O(n^{\log_b a})$.
 2. If $f(n) = \Theta(n^{\log_b a})$, then $g(n) = \Theta(n^{\log_b a} \lg n)$.
 3. If $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some $\varepsilon > 0$ and if $a f(n/b) \leq c f(n)$ for some $c < 1$ and all sufficiently large $n \geq b$, then $g(n) = \Theta(f(n))$.

Proof of Lemma 4.3

- For case 1: $f(n)=O(n^{\log_b a-\varepsilon})$ implies $f(n/b^j)=O((n/b^j)^{\log_b a-\varepsilon})$, so

- $g(n)=\sum_{j=0}^{\log_b n-1} a^j f(n/b^j) = O\left(\sum_{j=0}^{\log_b n-1} a^j (n/b^j)^{\log_b a-\varepsilon}\right)$
- $= O(n^{\log_b a-\varepsilon} \sum_{j=0}^{\log_b n-1} a^j / (b^{\log_b a-\varepsilon})^j) = O(n^{\log_b a-\varepsilon} \sum_{j=0}^{\log_b n-1} a^j / (a^j (b^{-\varepsilon})^j))$
- $= O(n^{\log_b a-\varepsilon} \sum_{j=0}^{\log_b n-1} (b^{\varepsilon})^j) = O(n^{\log_b a-\varepsilon} ((b^{\varepsilon})^{\log_b n}-1)/(b^{\varepsilon}-1))$
- $= O(n^{\log_b a-\varepsilon} (((b^{\log_b n})^{\varepsilon}-1)/(b^{\varepsilon}-1))) = O(n^{\log_b a} n^{-\varepsilon} (n^{\varepsilon}-1)/(b^{\varepsilon}-1))$
- $= O(n^{\log_b a})$

Proof of Lemma 4.3(cont.)

- For case 2: $f(n) = \Theta(n^{\log_b a})$ implies $f(n/b^j) = \Theta((n/b^j)^{\log_b a})$, so

- $$g(n) = \sum_{j=0}^{\log_b n-1} a^j f(n/b^j) = \Theta\left(\sum_{j=0}^{\log_b n-1} a^j (n/b^j)^{\log_b a}\right)$$
- $$= \Theta\left(n^{\log_b a} \sum_{j=0}^{\log_b n-1} a^j / (b^{\log_b a})^j\right) = \Theta\left(n^{\log_b a} \sum_{j=0}^{\log_b n-1} 1\right)$$
- $$= \Theta(n^{\log_b a} \log_b n) = \Theta(n^{\log_b a} \lg n)$$

Proof of Lemma 4.3(cont.)

- For case 3:
 - Since $g(n)$ contains $f(n)$, $g(n) = \Omega(f(n))$
 - Since $af(n/b) \leq cf(n)$, $a^j f(n/b^j) \leq c^j f(n)$, why???
 - $g(n) = \sum_{j=0}^{\log_b n-1} a^j f(n/b^j) \leq \sum_{j=0}^{\log_b n-1} c^j f(n) \leq f(n) \sum_{j=0}^{\infty} c^j$
 - $= f(n)(1/(1-c)) = O(f(n))$
 - Thus, $g(n) = \Theta(f(n))$

Proof of Master Theorem (cont.)

- Lemma 4.4:
 - for $T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ aT(n/b)+f(n) & \text{if } n=b^k \text{ for } k \geq 1 \end{cases}$
 - where $a \geq 1$, $b > 1$, $f(n)$ be a nonnegative function,
 1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
 2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
 3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$.

Proof of Lemma 4.4 (cont.)

- Combine Lemma 4.2 and 4.3,
 - For case 1:
 - $T(n) = \Theta(n^{\log_b a}) + O(n^{\log_b a}) = \Theta(n^{\log_b a})$.
 - For case 2:
 - $T(n) = \Theta(n^{\log_b a}) + \Theta(n^{\log_b a} \lg n) = \Theta(n^{\log_b a} \lg n)$.
 - For case 3:
 - $T(n) = \Theta(n^{\log_b a}) + \Theta(f(n)) = \Theta(f(n))$ because $f(n) = \Omega(n^{\log_b a + \varepsilon})$.

Floors and Ceilings

- $T(n)=aT(\lfloor n/b \rfloor)+f(n)$ and $T(n)=aT(\lceil n/b \rceil)+f(n)$
- Want to prove both equal to $T(n)=aT(n/b)+f(n)$
- Two results:
 - Master theorem applied to all integers n .
 - Floors and ceilings do not change the result.
 - (Note: we proved this by domain transformation too).
- Since $\lfloor n/b \rfloor \leq n/b$, and $\lceil n/b \rceil \geq n/b$, upper bound for floors and lower bound for ceiling is held.
- So prove upper bound for ceilings (similar for lower bound for floors).

Upper bound of proof for $T(n)=aT(\lceil n/b \rceil)+f(n)$

- consider sequence $n, \lceil n/b \rceil, \lceil \lceil n/b \rceil / b \rceil, \lceil \lceil \lceil n/b \rceil / b \rceil / b \rceil, \dots$
- Let us define n_j as follows:
- $n_j = n$ if $j=0$
- $= \lceil n_{j-1}/b \rceil$ if $j>0$
- The sequence will be $n_0, n_1, \dots, n_{\lfloor \log_b n \rfloor}$
- Draw recursion tree:

Recursion tree of $T(n) = aT(\lceil n/b \rceil) + f(n)$

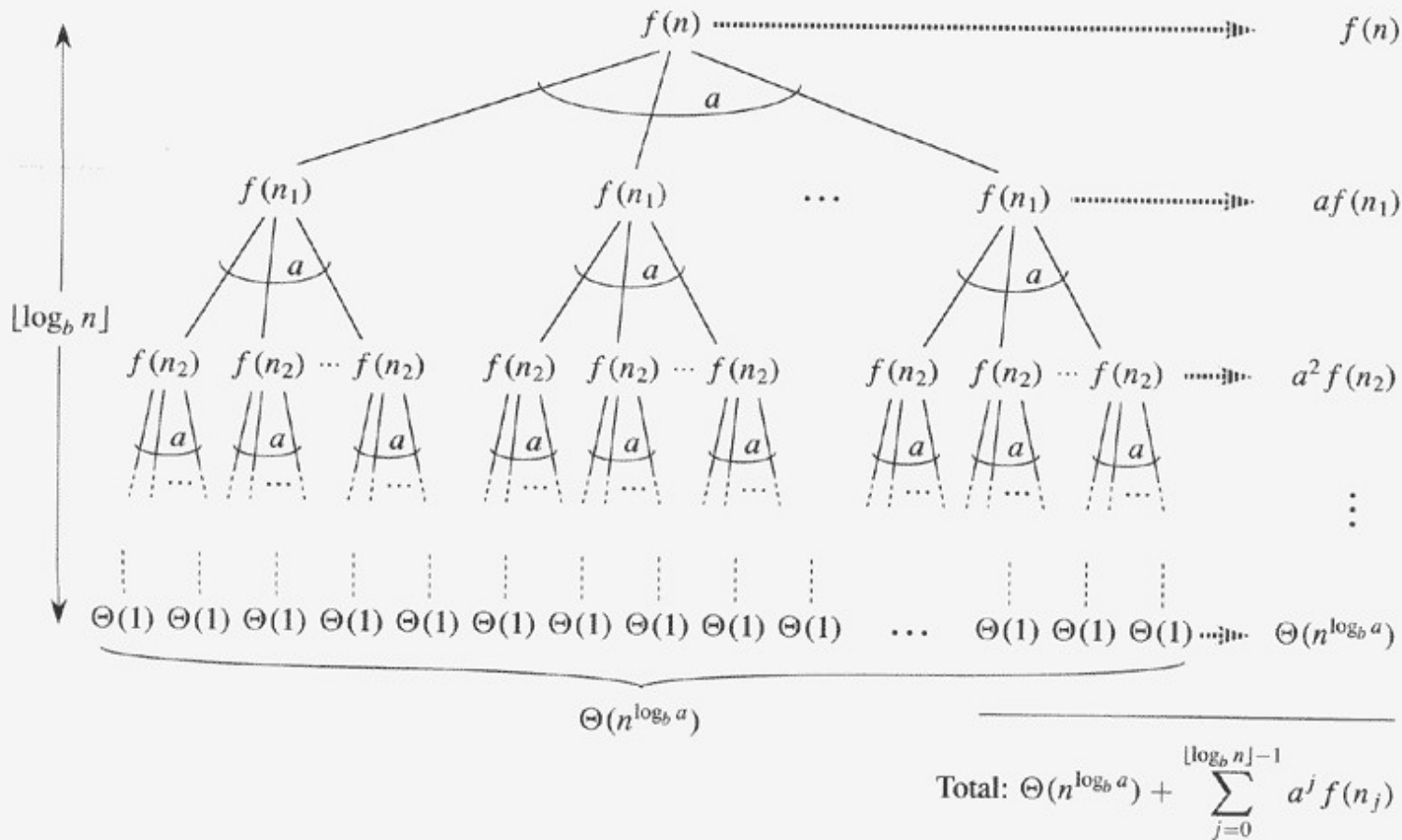


Figure 4.4 The recursion tree generated by $T(n) = aT(\lceil n/b \rceil) + f(n)$. The recursive argument n_j is given by equation (4.12).

The proof of upper bound for ceiling

$$- T(n) = \Theta(n^{\log_b a}) + \sum_{j=0}^{\lfloor \log_b n \rfloor - 1} a^j f(n_j)$$

- Thus similar to Lemma 4.3 and 4.4, the upper bound is proven.

The simple format of master theorem

- $T(n)=aT(n/b)+cn^k$, with a, b, c, k are positive constants, and $a \geq 1$ and $b \geq 2$,

- $$T(n) = \begin{cases} O(n^{\log_b a}), & \text{if } a > b^k. \\ O(n^k \log n), & \text{if } a = b^k. \\ O(n^k), & \text{if } a < b^k. \end{cases}$$

Summary

Recurrences and their bounds

- Substitution
- Recursion tree
- Master theorem.
- Proof of subtleties
- Recurrences that Master theorem does not apply to.