Introduction to Algorithms

Topic 2: Asymptotic Mark and Recursive Equation

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Outline of Topics

- **1** Asymptotic Notation: O-, Ω- and Θ-otation
 - O-otation
 - \circ Ω -otation
 - Θ-otation
 - Other Asymptotic Notations
 - Comparing Functions
- Standard Notations and Common Functions
- 3 Recurrences
 - Substitution Method
 - Recursion Tree
 - Master Method

Ω-otation Θ-otation Other Asymptotic Notations Comparing Functions

O-otation

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O-otation Ω-otation Θ-otation Other Asymptotic Notations

Asymptotic Notation: *O*—notation

O-notation: upper bounds

We write f(n) = O(g(n)) if there exist constants $c > 0, n_0 > 0$ such that $0 \le f(n) \le cg(n)$ for all $n \ge n_0$.

O-otation

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Example:
$$2n^2 = O(n^3)$$
 $(c = 1, n_0 = 2)$

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Example:
$$2n^2 = O(n^3)$$
 $(c = 1, n_0 = 2)$ functions, not values

O-otation Ω-otation

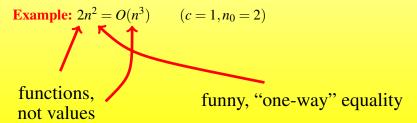
Θ-otation

Other Asymptotic Notations

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We write f(n) = O(g(n)) if there exist constants $c > 0, n_0 > 0$ such that $0 \le f(n) \le cg(n)$ for all $n > n_0$.



Set Definition of *O*-notation

$$O(g(n)) = \{f(n) : \text{there exist constants } c > 0, n_0 > 0 \text{ such that } 0 \le f(n) \le cg(n) \text{ for all } n \ge n_0\}.$$

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Example: $2n^2 \in O(n^3)$

O-otation

Macro Substitution

Convention: A set in a formula represents an anonymous function in the set.

Example:
$$f(n) = n^3 + O(n^2)$$

means
 $f(n) = n^3 + h(n)$
for some $h(n) \in O(n^2)$.

Asymptotic Notation: Ω -notation

O-notation is an upper-bound notation. The Ω -notation provides a lower bound.

Set definition of Ω -notation

$$\Omega(g(n)) = \{f(n) : \text{there exist constants } c > 0, n_0 > 0 \text{ such that}$$

$$0 \le c \cdot g(n) \le f(n) \text{ for all } n \ge n_0\}$$

Asymptotic Notation: Ω -notation

O-notation is an upper-bound notation. The Ω -notation provides a lower bound.

Set definition of Ω -notation

$$\Omega(g(n))=\{f(n):$$
 there exist constants $c>0,n_0>0$ such that
$$0\leq c\cdot g(n)\leq f(n) \text{ for all } n\geq n_0\}$$

Example:
$$\sqrt{n} = \Omega(\lg n)$$

Θ-notation: tight bounds

We write $f(n) = \Theta(g(n))$ if there exist constants $c_1 > 0, c_2 > 0, n_0 > 0$ such that $c_2g(n) \ge f(n) \ge c_1g(n) \ge 0$ for all $n \ge n_0$.

$$\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$$

Θ-notation: tight bounds

We write $f(n) = \Theta(g(n))$ if there exist constants $c_1 > 0, c_2 > 0, n_0 > 0$ such that $c_2g(n) \ge f(n) \ge c_1g(n) \ge 0$ for all $n \ge n_0$.

$$\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$$

$$\frac{1}{2}n^2 - 2n = \Theta\left(n^2\right)$$

Θ-notation: tight bounds

We write $f(n) = \Theta(g(n))$ if there exist constants $c_1 > 0, c_2 > 0, n_0 > 0$ such that $c_2g(n) \ge f(n) \ge c_1g(n) \ge 0$ for all $n \ge n_0$.

$$\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$$

$$\frac{1}{2}n^2 - 2n = \Theta(n^2)$$

$$\Theta(n^0) \text{ or } \Theta(1)$$

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We write $f(n) = \Theta(g(n))$ if there exist constants $c_1 > 0, c_2 > 0, n_0 > 0$ such that $c_2g(n) \ge f(n) \ge c_1g(n) \ge 0$ for all $n \ge n_0$.

$$\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$$

$$\frac{1}{2}n^2 - 2n = \Theta(n^2)$$

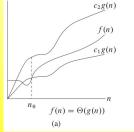
$$\Theta(n^0) \text{ or } \Theta(1)$$

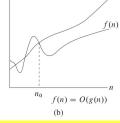
Theorem:

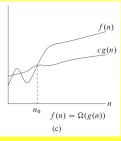
The leading constant and low order terms do not matter.

cg(n)

Graphic Examples of the Θ , O, Ω







Ω-otation
Θ-otation
Other Asymptotic Notations
Comparing Functions

O-otation

Other Asymptotic Notations

o-notation

 $o(g(n)) = \{f(n): \text{ for all } c > 0, \text{ there exist constants } n_0 > 0 \text{ such that } 0 \le f(n) < cg(n) \text{ for all } n \ge n_0\}.$

Other equivalent definition $\lim_{n\to\infty} \frac{f(n)}{g(n)} = 0$.

ω -notation

 $\omega(g(n)) = \{f(n): \text{ for all } c > 0, \text{ there exist constants } n_0 > 0 \text{ such that } 0 \le cg(n) < f(n) \text{ for all } n \ge n_0\}.$

Other equivalent definition $\lim_{n\to\infty} \frac{f(n)}{g(n)} = \infty$

A Helpful Analogy

$$f(n) = O(g(n))$$
 is similar to $f(n) \le g(n)$.

$$f(n) = o(g(n))$$
 is similar to $f(n) < g(n)$.

$$f(n) = \Theta(g(n))$$
 is similar to $f(n) = g(n)$.

$$f(n) = \Omega(g(n))$$
 is similar to $f(n) \ge g(n)$.

$$f(n) = \omega(g(n))$$
 is similar to $f(n) > g(n)$.

O-otation

Transitivity

$$f(n) = \Theta(g(n))$$
 and $g(n) = \Theta(h(n))$ imply $f(n) = \Theta(h(n))$.
 $f(n) = O(g(n))$ and $g(n) = O(h(n))$ imply $f(n) = O(h(n))$.
 $f(n) = \Omega(g(n))$ and $g(n) = \Omega(h(n))$ imply $f(n) = \Omega(h(n))$.
 $f(n) = o(g(n))$ and $g(n) = o(h(n))$ imply $f(n) = o(h(n))$.
 $f(n) = \omega(g(n))$ and $g(n) = \omega(h(n))$ imply $f(n) = \omega(h(n))$.

O-otation Ω-otation Θ-otation Other Asymptotic Notations Comparing Functions

Reflexivity

$$f(n) = \Theta(f(n))$$

$$f(n) = O(f(n))$$

$$f(n) = \Omega(f(n))$$

Ω-otation Θ-otation Other Asymptotic Notations Comparing Functions

O-otation

Symmetry & Transpose Symmetry

Symmetry

$$f(n) = \Theta(g(n))$$
 if and only if $g(n) = \Theta(f(n))$.

Transpose Symmetry

$$f(n) = O(g(n))$$
 if and only if $g(n) = \Omega(f(n))$.
 $f(n) = o(g(n))$ if and only if $g(n) = \omega(f(n))$.

Ω-otation Θ-otation Other Asymptotic Notations Comparing Functions

O-otation

Non-completeness

Non-completeness of O, Ω , and Θ notations

For real numbers a and b, we know that either a < b, or a = b, or a > b is true.

However, for two functions f(n) and g(n), it is possible that neither of the following is true: f(n) = O(g(n)), or $f(n) = \Theta(g(n))$, or f(n) = O(g(n)). For example, f(n) = n, and $g(n) = n^{1-\sin(n\pi/2)}$.

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Floors and Ceilings

Floor

For any real number x, we denote the greatest integer less than or equal to x by |x| (read "the floor of x")

Ceiling

For any real number x, we denote the least integer greater than or equal to x by $\lceil x \rceil$ (read "the ceiling of x")

$$x - 1 < |x| \le x \le \lceil x \rceil \le x + 1.$$

For any integer n, $\lceil n/2 \rceil + \lfloor n/2 \rfloor = n$.

For any real number $x \ge 0$ and integers a, b > 0,

$$\lceil \frac{\lceil x/a \rceil}{b} \rceil = \lceil \frac{x}{ab} \rceil, \lfloor \frac{\lfloor x/a \rfloor}{b} \rfloor = \lfloor \frac{x}{ab} \rfloor, \lceil \frac{a}{b} \rceil \le \frac{a + (b-1)}{b}, \lfloor \frac{a}{b} \rfloor \ge \frac{a - (b-1)}{b},$$

Modular Arithmetic

Mod

For any integer a and any positive integer n, the value $a \mod n$ is the remainder (or residue) of the quotient a/n:

$$a \mod n = a - n \lfloor a/n \rfloor$$
.

Equivalent

If $(a \mod n) = (b \mod n)$, we write $(a \equiv b) \mod n$ and say that a is equivalent to b, modulo n.

Exponentials

$$\forall a > 0, \quad a^0 = 1; \quad (a^m)^n = (a^n)^m = a^{mn}; \quad a^m a^n = a^{m+n}$$

When
$$a > 1$$
, $\lim_{n \to \infty} \frac{n^b}{a^n} = 0$. That is, $n^b = o(a^n)$.

For all real
$$x$$
, $e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + ... = \sum_{i=0}^{\infty} \frac{x^i}{i!}$
When $|x| \le 1$, $1 + x \le e^x \le 1 + x + x^2$
When $x \to 0$, $e^x = 1 + x + \Theta(x^2)$
For all x , $\lim_{n \to \infty} (1 + \frac{x}{n})^n = e^x$

Logarithms

$$\lg n = \log_2 n; \quad \ln n = \log_e n; \quad \lg^k n = (\lg n)^k; \quad \lg\lg n = \lg(\lg n)$$

For all real
$$a,b,c>0$$
, and n , $a=b^{\log_b a}$; $\log_c(ab)=\log_c a+\log_c b$; $\log_b a^n=n\log_b a$; $\log_b a=\frac{\lg a}{\lg b}$; $a^{\log_b c}=c^{\log_b a}$

When
$$a > 0$$
, $\lim_{n \to \infty} \frac{\lg^b n}{(2^a)^{\lg n}} = \lim_{n \to \infty} \frac{\lg^b n}{n^a} = 0$. That is, $\lg^b n = o(n^a)$.

When
$$|x| \le 1$$
, $\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \frac{x^5}{5} - \dots$
For $x > -1$, $\frac{x}{1+x} \le \ln(1+x) \le x$

Factorials

$$n! = \begin{cases} 1 & \text{if} & n = 0 \\ n \cdot (n-1)! & \text{if} & n > 0 \end{cases}$$

 $n! \le n^n$. A better bound:

Stirling's approximation

$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)$$

Functional iteration

functional iteration

We use the notation $f^{(i)}(n)$ to denote the function f(n) iteratively applied i times to an initial value of n. Formally, let f(n) be a function over the reals. For non-negative integers i, we recursively define

$$f^{(i)}(n) = \begin{cases} n & \text{if } i = 0, \\ f(f^{(i-1)}(n)) & \text{if } i > 0, \end{cases}$$

if
$$f(n) = 2n$$
, then $f^{(i)}(n) = 2^{i}n$.

The iterated logarithm function

We use the notation $\lg^* n$ to denote the iterated logarithm.

$$\lg^* n = min\{i \ge 0 : \lg^{(i)} n \le 1\}.$$

Example:

$$lg^* 2 = 1,$$

$$lg^* 4 = 2,$$

$$lg^* 16 = 3,$$

$$lg^* (2^{65536}) = 5.$$

Fibonacci Numbers

Fibonacci numbers

We define the Fibonacci numbers by the following recurrence:

$$F_0 = 0,$$

 $F_1 = 1,$
 $F_i = F_{i-1} + F_{i-2}, \quad for \ i \ge 2.$

Each Fibonacci number is the sum of the two previous ones, yielding the sequence

$$0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \dots$$

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Solving Recurrences

Recurrences go hand in hand with the divide-and-conquer paradigm. A recurrence is an equation or inequality that describes a function in terms of its value on smaller inputs.

Three methods for solving recurrences

- substitution method: guess a bound and use mathematical induction to prove the guess correct.
- recursion-tree method: converts the recurrence into a tree and use techniques for bounding summations.
- master method: provides bounds of the form $T(n) = a \cdot T(\frac{n}{h}) + f(n)$.

Substitution Method

The most general method

- 1. Guess the form of the solution.
- 2. Verify by induction.
- 3. Solve for constants.
 - This method only works if we can guess the form of the answer.
 - The method can be used to establish either upper or lower bounds on a recurrence.

Example of Substitution

Example:
$$T(n) = 4T(n/2) + n$$

- Assume that $T(1) = \Theta(1)$.
- Guess $T(n) = O(n^3)$. (Note that if we guess Θ , we need prove O and Ω separately.)
- Assume that $T(k) \le ck^3$ for k < n and some constant c > 0.
- Prove $T(n) \le cn^3$ by induction.

Example of Substitution

$$T(n) = 4T(n/2) + n$$

$$\leq 4c(n/2)^3 + n$$

$$= (c/2)n^3 + n$$

$$= cn^3 - ((c/2)n^3 - n) \qquad \text{desired - residual}$$

$$\leq cn^3 \qquad \text{desired}$$
whenever $(c/2)n^3 - n \geq 0$, for example, if $c \geq 2$ and $n \geq 1$.

Example (Continued)

- We must also handle the initial conditions, that is, ground the induction with base cases.
- Base: $T(n) = \Theta(1)$ for all $n < n_0$, where n_0 is a suitable constant.
- For $1 \le n < n_0$, we have " $\Theta(1)$ " $\le cn^3$, if we pick c big enough.

Example (Continued)

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- Base: $T(n) = \Theta(1)$ for all $n < n_0$, where n_0 is a suitable constant.
- For $1 \le n < n_0$, we have " $\Theta(1)$ " $\le cn^3$, if we pick *c* big enough.

This bound is not tight!

We shall prove that $T(n) = O(n^2)$.

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Assume that $T(k) \le ck^2$ for k < n:

$$T(n) = 4T(n/2) + n$$

$$\leq 4c(n/2)^{2} + n$$

$$= cn^{2} + n$$

$$= O(n^{2})$$

We shall prove that $T(n) = O(n^2)$.

Assume that $T(k) \le ck^2$ for k < n:

$$T(n) = 4T(n/2) + n$$

$$\leq 4c(n/2)^{2} + n$$

$$= cn^{2} + n$$



Wrong! We must prove the I.H.

We shall prove that $T(n) = O(n^2)$.

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$$T(n) = 4T(n/2) + n$$

$$\leq 4c(n/2)^{2} + n$$

$$= cn^{2} + n$$



Wrong! We must prove the I.H.

$$=cn^2-(-n)$$
 [desired – residual]

 $< cn^2$ for no choice of c > 0. Lose!

IDEA: Strengthen the inductive hypothesis.

• Subtract a low-order term.

Inductive hypothesis: $T(k) \le c_1 k^2 - c_2 k$ for k < n

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• Subtract a low-order term.

Inductive hypothesis: $T(k) \le c_1 k^2 - c_2 k$ for k < n

$$T(n) = 4T(n/2) + n$$

$$\leq 4(c_1(n/2)^2 - c_2(n/2)) + n$$

$$= c_1n^2 - 2c_2n + n$$

$$= c_1n^2 - c_2n - (c_2n - n)$$

$$\leq c_1n^2 - c_2n \text{ if } c_2 > 1$$

Pick c_1 big enough to handle the initial conditions.

A Tighter Lower Bound

We shall prove that $T(n) = \Omega(n^2)$.

A Tighter Lower Bound

We shall prove that $T(n) = \Omega(n^2)$.

Assume that $T(k) \ge ck^2$ for k < n, and for some chosen constant c.

$$T(n) = 4T(n/2) + n$$

$$\geq 4c(n/2)^{2} + n$$

$$= cn^{2} + n$$

$$\geq cn^{2}$$

Recursion-tree Method

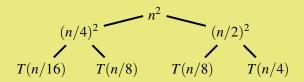
- A recursion tree models the costs (time) of a recursive execution of an algorithm.
- The recursion-tree method can be unreliable.
- The recursion tree method is good for generating guesses for the substitution method.

Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

$$T(n/4) \xrightarrow{n^2} T(n/2)$$

Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:



Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

$$(n/4)^{2} \qquad n^{2} \qquad (n/2)^{2}$$

$$(n/16)^{2} \qquad (n/8)^{2} \qquad (n/8)^{2} \qquad (n/4)^{2}$$

$$\Theta(1)$$

Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

$$(n/4)^{2} \qquad n^{2} \qquad (n/2)^{2}$$

$$(n/16)^{2} \qquad (n/8)^{2} \qquad (n/8)^{2} \qquad (n/4)^{2}$$

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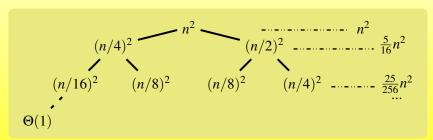
Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

$$(n/4)^{2} \qquad n^{2} \qquad (n/2)^{2} \qquad n^{2} \qquad \frac{5}{16}n^{2}$$

$$(n/16)^{2} \qquad (n/8)^{2} \qquad (n/8)^{2} \qquad (n/4)^{2}$$

$$\Theta(1)$$

Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:



Solve
$$T(n) = T(n/4) + T(n/2) + n^2$$
:

$$(n/4)^{2} \qquad n^{2} \qquad (n/2)^{2} \qquad n^{2} \qquad \frac{5}{16}n^{2}$$

$$(n/16)^{2} \qquad (n/8)^{2} \qquad (n/8)^{2} \qquad (n/4)^{2} \qquad \frac{25}{256}n^{2}$$

$$\Theta(1)$$

Total=
$$n^2 (1 + \frac{5}{16} + (\frac{5}{16})^2 + (\frac{5}{16})^3 + \cdots) = \Theta(n^2)$$
(geometric series)

The Master Method

Master method

The master method applies to recurrences of the form

$$T(n) = aT(\frac{n}{b}) + f(n)$$

where $a \ge 1$, b > 1, and f is asymptotically positive.

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 $n^{\log_b a}$ (by an n^{ε} factor). Solution: $T(n) = \Theta(n^{\log_b a})$.

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 $n^{\log_b a}$ (by an n^{ε} factor). Solution: $T(n) = \Theta(n^{\log_b a})$.
- 2. $f(n) = \Theta(n^{\log_b a} \lg^k n)$ for some constant $k \ge 0$
 - f(n) and $n^{\log_b a} \lg^k n$ grow at similar rates. Solution: $T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$.

Three Common Cases

Compare f(n) with $n^{\log_b a}$:

- 3. $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some constant $\varepsilon > 0$.
 - f(n) grows polynomially faster than $n^{\log_b a}$ (by an n^{ε} factor), and f(n) satisfies the **regularity condition** that $af(n/b) \le cf(n)$ for some constant c < 1 and all sufficiently large n.

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

Ex.
$$T(n) = 4T(n/2) + n$$

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n.$
Case 1: $f(n) = O(n^{2-\varepsilon})$ for $\varepsilon = 1$
 $\therefore T(n) = \Theta(n^2).$

Ex.
$$T(n) = 4T(n/2) + n$$

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n.$
Case 1: $f(n) = O(n^{2-\varepsilon})$ for $\varepsilon = 1$
 $\therefore T(n) = \Theta(n^2).$

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: $f(n) = \Theta(n^2 l g^0 n)$, that is, $k = 0$.
 $\therefore T(n) = \Theta(n^2 l g n)$.

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: $f(n) = \Omega(n^{2+\varepsilon})$ for $\varepsilon = 1$
and $4(n/2)^3 \le cn^3$ (reg. cond.) for $c = 1/2$.
 $\therefore T(n) = \Theta(n^3).$

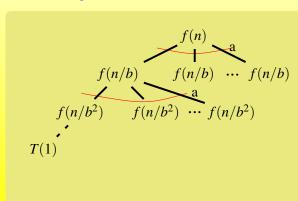
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 $\therefore T(n) = \Theta(n^3).$

Ex.
$$T(n) = 4T(n/2) + n^2/\lg n$$

 $a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2$; $f(n) = n^2/\lg n$.
Master method does not apply. In particular, for every constant $\varepsilon > 0$, we have $n^{\varepsilon} = \omega(\lg n)$.

$$T(n) = aT(\frac{n}{b}) + f(n)$$
. Recursion tree:



$$T(n) = aT(\frac{n}{h}) + f(n)$$
. Recursion tree:

$$f(n) \xrightarrow{a} f(n/b) \cdots f(n/b) \cdots f(n/b) \cdots af(n/b)$$

$$f(n/b^2) \xrightarrow{a} f(n/b^2) \cdots f(n/b^2) \cdots a^2 f(n/b^2$$

$$T(n) = aT(\frac{n}{b}) + f(n)$$
. Recursion tree:

$$f(n) \xrightarrow{a} f(n/b) \cdots f(n/b) \cdots f(n/b) \cdots af(n/b)$$

$$f(n/b^2) \xrightarrow{a} f(n/b^2) \cdots f(n/b^2) \cdots a^2 f(n/b^2$$

$$T(n) = aT(\frac{n}{b}) + f(n)$$
. Recursion tree:

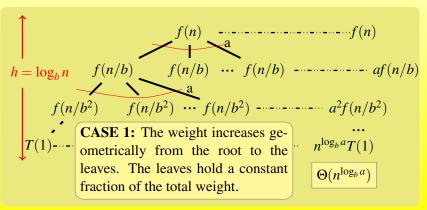
$$f(n) \xrightarrow{a} f(n/b) \cdots f(n/b) \cdots f(n/b) \cdots af(n/b) \cdots af(n/b^2) \cdots f(n/b^2) \cdots f(n/b$$

$$T(n) = aT(\frac{n}{b}) + f(n)$$
. Recursion tree:

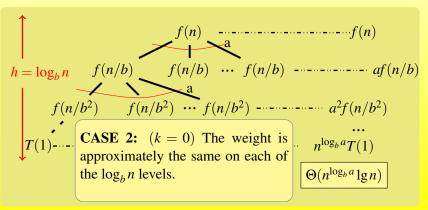
$$f(n) \xrightarrow{a} f(n/b) \cdots f(n/b) \cdots f(n/b) \cdots af(n/b)$$

$$f(n/b^2) \xrightarrow{a} f(n/b^2) \cdots f(n/b^2) \cdots a^2 f(n/b^2$$

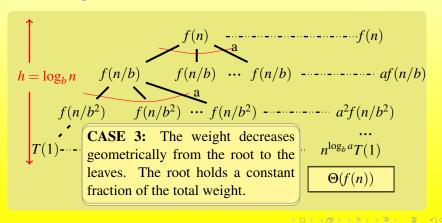
$$T(n) = aT(\frac{n}{h}) + f(n)$$
. Recursion tree:



$$T(n) = aT(\frac{n}{h}) + f(n)$$
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. Recursion tree:



Appendix: Geometric Series

$$1+x+x^2+\cdots+x^n=\frac{1-x^{n+1}}{1-x}$$
 for $x \neq 1$

$$1+x+x^2+\cdots = \frac{1}{1-x}$$
 for $|x| < 1$