

Often when analyzing a divide & conquer algorithm, we obtain a recurrence for its running time of the following form

$$T(n) = aT\left(\frac{n}{b}\right) + cn^k \tag{1}$$

In words, on input size n , the algorithm generates a subproblems, each of size n/b ; combining these subproblems to obtain the overall solution requires time polynomial in n , specifically cn^k .

Such recurrences appear frequently so it is useful to know asymptotic bounds for them in terms of a, b and k (as we will see, c does not affect the asymptotic solution). To this end, we will analyze the recursion tree for this recurrence (see Figure 1).

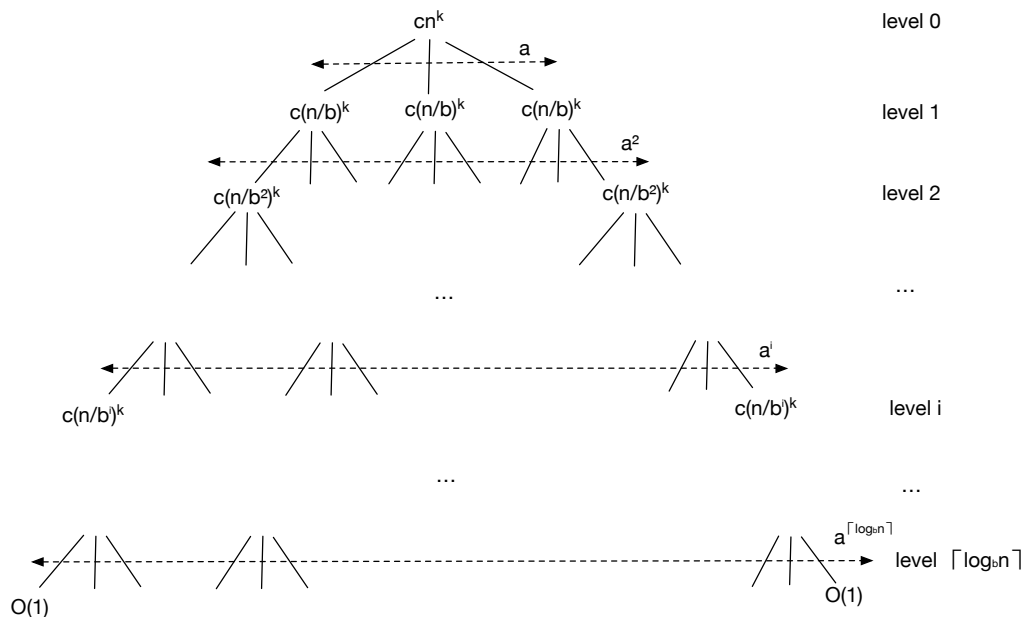


Figure 1: The recursion tree for recurrence (1). a is the branching factor, b is the factor by which the input size shrinks at every recursive call and cn^k is the time required to combine the solutions to the subproblems into the overall solution for input size is n . The smallest possible size of a subproblem is $O(1)$; typically, solving input instances of small constant size requires constant time c .

Note that

- a is the branching factor of the tree: every subproblem gives rise to a new subproblems at the next level of the tree; thus
 1. at level 1, we have a subproblems
 2. at level 2, **each** of the a subproblems in level 1 gives rise to a new subproblems; therefore there are a total of a^2 subproblems
 3. at level 3, **each** of the a^2 subproblems in level 2 generates a new subproblems; therefore there are a total of a^3 subproblems
 4. at the level i , there are a^i subproblems

- b is the factor by which the input size shrinks at every level; thus
 1. at level 1, the input size of each subproblem shrinks by a factor of b , that is, from n it now becomes n/b ;
 2. at level 2, the input size of each subproblem further shrinks by a factor of b , that is, from n/b it now becomes $(n/b)/b = n/b^2$;
 3. at level 3, the input size of each subproblem again shrinks by a factor of b , hence becomes $(n/b^2)/b = n/b^3$;
 4. at level i , the size of each subproblem is n/b^i

⇒ at level i , the amount of work spent on each subproblem of size n/b^i is ¹:

$$c \left(\frac{n}{b^i} \right)^k$$

⇒ at level i , the work spent on **all** subproblems is

$$a^i c \left(\frac{n}{b^i} \right)^k = cn^k \left(\frac{a}{b^k} \right)^i$$

We need one more observation before we can compute the total work spent on the recursion tree.

Fact 1 *The depth of the tree in Figure 1 is $\lceil \log_b n \rceil$ levels.*

Proof. The last level of the recursion tree, call it d , consists of subproblems of size 1. Since at level i subproblems have size n/b^i , we are looking for d such that

$$\frac{n}{b^d} = 1 \Rightarrow d = \log_b n$$

Since d is an integer, $d = \lceil \log_b n \rceil$. □

We are now ready to derive a bound for $T(n)$ by computing the total work spent on this recursion tree, which is given by the sum of the work spent at each level of the tree:

$$T(n) = \sum_{i=0}^{\lceil \log_b n \rceil} cn^k \left(\frac{a}{b^k} \right)^i = cn^k \sum_{i=0}^{\lceil \log_b n \rceil} \left(\frac{a}{b^k} \right)^i \quad (2)$$

Note that $T(n)$ depends on a sum over $\lceil \log_b n \rceil$ terms of a geometric progression with common ratio a/b^k and initial value $(a/b^k)^0 = 1$. Depending on the value of the common ratio a/b^k , this sum will exhibit the following behavior:

1. $\frac{a}{b^k} = 1$; in this case, we have

$$\sum_{i=0}^{\lceil \log_b n \rceil} \left(\frac{a}{b^k} \right)^i = \sum_{i=0}^{\lceil \log_b n \rceil} 1 = \lceil \log_b n \rceil + 1 = \Theta(\log_b n) \quad (3)$$

2. $\frac{a}{b^k} < 1$; in this case, you can show that the sum of the entire geometric progression is dominated by its initial value, that is,

$$\sum_{i=0}^{\lceil \log_b n \rceil} \left(\frac{a}{b^k} \right)^i = \Theta \left(\left(\frac{a}{b^k} \right)^0 \right) = \Theta(1) \quad (4)$$

¹Recall that the amount of work spent on combining the subproblems when the the input size is n is cn^k .

3. $\frac{a}{b^k} > 1$; again, you can show that the sum of the entire geometric progression is now dominated by its last term, that is,

$$\sum_{i=0}^{\lceil \log_b n \rceil} \left(\frac{a}{b^k}\right)^i = \Theta\left(\left(\frac{a}{b^k}\right)^{\log_b n}\right) = \Theta\left(\frac{a^{\log_b n}}{b^{k \log_b n}}\right) = \Theta\left(\frac{n^{\log_b a}}{n^k}\right) \quad (5)$$

Plugging back equations (3), (4), (5) into equation (2), we summarize our findings in the following theorem.

Theorem 1 (Master theorem) *If $T(n) = aT(\lceil n/b \rceil) + O(n^k)$ for some constants $a > 0$, $b > 1$, $k \geq 0$, then*

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