

filler

Master Theorem

Overview

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Theorem

Suppose a complexity function $T(n)$ is eventually non-decreasing and satisfies

$$\begin{aligned} T(n) &= aT\left(\frac{n}{b}\right) + O(n^d) && \text{for } n > 1, n \text{ a power of } b \\ T(1) &= c \end{aligned}$$

Where $b \geq 2$ and $d \geq 0$ are constant *integers*, and a and c are constant such that $a > 0$ and $c > 0$. Then

$$T(n) \in \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \lg n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Also, using the same theorem, if in the recurrence relation

$$T(n) = aT\left(\frac{n}{b}\right) + O(n^d)$$

the equality is replaced by an inequality, such as

$$T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d) \quad \text{or} \quad T(n) \geq aT\left(\frac{n}{b}\right) + O(n^d)$$

Then the results above hold true with “big O ” or Ω , respectively, replacing Θ

Proof






It is important to understand the different parts of the recurrence relation used in the theorem.

$$\underbrace{T(n)}_{\text{Recurrence relation}} = \underbrace{a}_{\text{Subproblems}} \underbrace{T\left(\frac{n}{b}\right)}_{\text{Divisor}} + \underbrace{O(n^d)}_{\text{Degree of work time complexity}}$$

The general solution to the recurrence relation above can be described as

$$T(n) = \sum_{i=0}^{\log_b n} O(n^d) \left(\frac{a}{b^d}\right)^i$$

This can be visually seen by drawing the recursion tree generated by the relation shown above. The tree has a *depth* of $\log_b n$ and a *branching factor* of a . At level i , there are a^i nodes each with a work of $O\left(\frac{n}{b}\right)^d$. The value of the relation is the sum of all the nodes on the tree. The sum is found by summing the work from all $\log_b n$ levels.

Level	Size of Problem	Subproblems	Work
0		a	$O(n^d)$
1		a^1	$a \cdot O\left(\frac{n}{b}\right)^d$ $= O(n^d) \cdot \frac{a}{b^d}$
i		a^i	$a^i \cdot O\left(\frac{n}{b^i}\right)^d$ $= O(n^d) \cdot \left(\frac{a}{b^d}\right)^i$
\vdots			
$\log_b n$		$a^{\log_b n}$	$a^{\log_b n} \cdot O(1)$ $= O(n^{\log_b a})$

$$\text{Total Work} = \sum_{i=0}^{\log_b n} O(n^d) \left(\frac{a}{b^d}\right)^i$$

Now that the solution is found, let's address the three cases shown in the theorem.

Case One

$$T(n) \in \Theta(n^d) \iff a < b^d, \text{ or } d > \log_b a.$$

For case one we will use substitution.

$$\sum_{i=0}^{\log_b n} \underbrace{O(n^d)}_a \underbrace{\left(\frac{a}{b^d}\right)^i}_r = O(n^d) \iff \left(\frac{a}{b^d}\right)^i < 1$$

Proof:

$$\begin{aligned}
\sum_{i=0}^{n-1} ar^i &= ar^0 + ar + ar^2 + ar^3 + \dots + ar^{n-1} \\
&= a \left(\frac{1 - r^n}{1 - r} \right) \quad r \neq 1 \\
r > 1 \text{ or } r < 1 &= \begin{cases} O(a), & r < 1 \\ O(ar^{n-1}), & r > 1 \end{cases} \\
\sum_{i=0}^{n-1} ar^i \leq O(a) &\iff r < 1 \\
\sum_{i=0}^{n-1} ar^i \leq O(ar^{n-1}) &\iff r > 1
\end{aligned}$$

Case Two

$$T(n) \in \Theta \left(n^d \lg n \right) \iff a = b^d, \text{ or } d = \log_b a$$

Proof:

$$\begin{aligned}
\frac{a}{b^d} &= 1 \iff d = \log_b a \\
\sum_{i=0}^{\log_b n} O(n^d) \left(\frac{a}{b^d} \right)^i &= \sum_{i=0}^{\log_b n} O(n^d) \\
&= (\log_b n + 1) \cdot O(n^d) \\
&= (\log_b n) \cdot O(n^d) + \underbrace{O(n^d)}_{\text{lower order term}} \\
&= \log_b n \cdot O(n^d) \\
&= O \left(n^d \cdot \frac{\log n}{\log b} \right) \\
&= O \left(n^d \lg n \right)
\end{aligned}$$

Case Three

$$T(n) \in \Theta \left(n^{\log_b n} \right) \iff a > b^d, \text{ or } d < \log_b a.$$

Proof:

$$\begin{aligned}
\frac{a}{b^d} > 1 &\iff d < \log_b a \\
\sum_{i=0}^{\log_b n} O(n^d) \left(\frac{a}{b^d}\right)^i &= O\left(O(n^d) \cdot \left(\frac{a}{b^d}\right)^{\log_b n}\right) \\
&= O\left(O(n^d) \cdot \frac{a^{\log_b n}}{(b^d)^{\log_b n}}\right) \\
&= O\left(O(n^d) \cdot \frac{n^{\log_b a}}{b^{\log_b n^d}}\right) \\
&= O\left(O(n^d) \cdot \frac{n^{\log_b a}}{n^d}\right) \\
&= O\left(\frac{Cn^d}{n^d} \cdot n^{\log_b a}\right) \\
&= O\left(n^{\log_b a}\right)
\end{aligned}$$

Examples

Merge Sort

Merge Sort is a recursive algorithm, the time complexity can be expressed with the following recurrence relation.

$$T(n) = \underbrace{2}_a T\left(\underbrace{\frac{n}{2}}_b\right) + \underbrace{O(n)}_{d=1}$$

Using the Master Theorem we get

$$a = 2$$

$$b = 2$$

$$d = 1$$

Which means the middle case holds true for Merge Sort

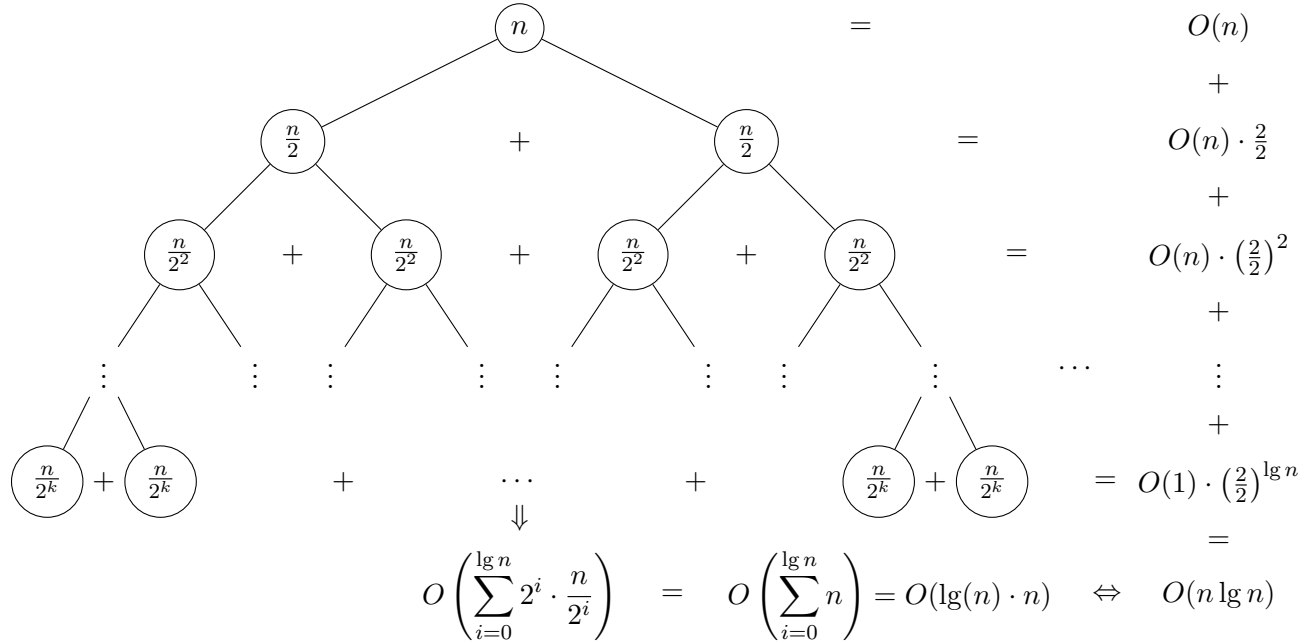
$$T(n) \in \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \lg n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

$$a = b^d$$

$$2 = 2^1$$

$$T(n) \in \Theta(n \lg n).$$

The proof of this can be seen visually with the diagram below



Other Examples

Consider the following recurrence relation

$$T(n) = \underbrace{9}_a T(n / \underbrace{3}_b) + 5n \underbrace{1}_d$$

By the Master Theorem, using case 3, because $9 > 3^1$

$$T(n) \in \Theta\left(n^{\log_3 9}\right) = \Theta(n^2).$$

Consider the following recurrence relation

$$T(n) = \underbrace{3}_a T(n / \underbrace{2}_b) + 3n \underbrace{3}_d + 2n^2.$$

By the Master Theorem, using case 1, because $3 < 2^3$

$$T(n) \in \Theta(n^3).$$