

CS 341

Background Information

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1 Exponent Laws

1. $\log_b a = \frac{\ln a}{\ln b}$

Proof. Let $x = \log_b a \Leftrightarrow b^x = a$. Then we have

$$\begin{aligned} b^x &= a \\ \ln(b^x) &= \ln a \\ x \ln b &= \ln a \\ x &= \frac{\ln a}{\ln b} \\ \log_b a &= \frac{\ln a}{\ln b}. \end{aligned}$$

□

2 Geometric Series

Reproduced from CS 240 Module 01, Slide 42:

$$\sum_{i=0}^{n-1} a r^i = \begin{cases} a \frac{r^n - 1}{r - 1} \in \Theta(r^{n-1}) & \text{if } r > 1 \\ na \in \Theta(n) & \text{if } r = 1 \\ a \frac{1 - r^n}{1 - r} \in \Theta(1) & \text{if } 0 < r < 1. \end{cases}$$

3 Analysis

Definition 3.1. $f(n) \in O(g(n))$ if there exist constants $c > 0$ and $n_0 > 0$ such that $0 \leq f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

Definition 3.2. $f(n) \in \Omega(g(n))$ if there exist constants $c > 0$ and $n_0 > 0$ such that $0 \leq c \cdot g(n) \leq f(n)$ for all $n \geq n_0$.

Remarks:

1. $f(n) \in \Omega(g(n))$ if and only if $g(n) \in O(f(n))$ (just take the reciprocal of the constant, and the same n_0).

Definition 3.3. $f(n) \in \Theta(g(n))$ if there exist constants $c_1, c_2 > 0$ and $n_0 > 0$ such that $0 \leq c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$ for all $n \geq n_0$.

Remarks:

1. $f(n) \in \Theta(g(n)) \Leftrightarrow f(n) \in O(g(n))$ and $f(n) \in \Omega(g(n))$.

Useful Facts:

1. $\log_b(n) \in \Theta(\log n)$ for all $b > 1$. (Our convention will be that $\log n$ will mean $\log_2 n$.) Proved in CS 240 Lecture Notes and, more elegantly, in the Beidl book.

Definition 3.4. $f(n) \in o(g(n))$ if for all constants $c > 0$, there exists $n_0 > 0$ such that $0 \leq f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

Definition 3.5. $f(n) \in \omega(g(n))$ if $g(n) \in o(f(n))$.

Relationships between Order Notations

1. $f(n) \in \Theta(g(n)) \Leftrightarrow g(n) \in \Theta(f(n))$
2. $f(n) \in O(g(n)) \Leftrightarrow g(n) \in \Omega(f(n))$
3. $f(n) \in o(g(n)) \Leftrightarrow g(n) \in \omega(f(n))$
4. $f(n) \in o(g(n)) \Rightarrow f(n) \in O(g(n))$
5. $f(n) \in o(g(n)) \Rightarrow f(n) \notin \Omega(g(n))$
6. $f(n) \in \omega(g(n)) \Rightarrow f(n) \in \Omega(g(n))$
7. $f(n) \in \omega(g(n)) \Rightarrow f(n) \notin O(g(n))$

Algebra of Order Notations

1. Identity Rule: $f(n) \in \Theta(f(n))$
2. Maximum Rules: Suppose that $f(n) > 0$ and $g(n) > 0$ for all $n \geq n_0$.
Then
 - (a) $O(f(n) + g(n)) = O(\max\{f(n), g(n)\})$.
 - (b) $\Omega(f(n) + g(n)) = \Omega(\max\{f(n), g(n)\})$.
3. Transitivity:
 - (a) If $f(n) \in O(g(n))$ and $g(n) \in O(h(n))$, then $f(n) \in O(h(n))$.
 - (b) If $f(n) \in \Omega(g(n))$ and $g(n) \in \Omega(h(n))$, then $f(n) \in \Omega(h(n))$.

Techniques of Order Notations

1. Limit Rule: Suppose that $f(n) > 0$ and $g(n) > 0$ for all $n > n_0$.
Suppose that

$$L = \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}, \text{ in particular, the limit exists.}$$

Then

$$f(n) \in \begin{cases} o(g(n)) & \text{if } L = 0 \\ \Theta(g(n)) & \text{if } 0 < L < \infty \\ \omega(g(n)) & \text{if } L = \infty. \end{cases}$$

Note: sufficient, not necessary.

Growth Rates

1. If $f(n) \in \Theta(g(n))$, then the growth rates of $f(n)$ and $g(n)$ are the same.
2. If $f(n) \in o(g(n))$, then the growth rate of $f(n)$ is less than the growth rate of $g(n)$.
3. If $f(n) \in \omega(g(n))$, then the growth rate of $f(n)$ is greater than the growth rate of $g(n)$.

Useful Facts:

1. The growth rate of $\log n$ is less than the growth rate of n . Proved in CS 240 Lecture Notes.
2. The growth rate of $(\log n)^c$ is less than the growth rate of n^d , where $c > 0$ and $d > 0$ are arbitrary real numbers. Proved in CS 240 Lecture Notes.

Complexity of Algorithms

1. Worst-case complexity of an algorithm Add, if needed.
2. Average-case complexity of an algorithm Add, if needed.

Definition 3.6. $f(n, m) \in O(g(n, m))$ if there exist constants $c > 0$ and $n_0 > 0, m_0 > 0$ such that $0 \leq f(n, m) \leq c \cdot g(n, m)$ for all $n \geq n_0$ **or** $m \geq m_0$ (i.e. finitely many exceptions).

Remarks:

1. Weaker Definition: there exist constants $c > 0$ and $n_0 > 0, m_0 > 0$ such that $0 \leq f(n, m) \leq c \cdot g(n, m)$ for all $n \geq n_0$ **and** $m \geq m_0$.
2. It will not matter much which definition we use.

Recursive Relations (See CS 240, Module 01)

Recursion	resolves to	example
$T(n) = T(n/2) + \Theta(1)$	$T(n) \in \Theta(\log n)$	Binary search
$T(n) = 2T(n/2) + \Theta(n)$	$T(n) \in \Theta(n \log n)$	Mergesort
$T(n) = 2T(n/2) + \Theta(\log n)$	$T(n) \in \Theta(n)$	Heapify (\rightarrow later)
$T(n) = T(cn) + \Theta(n)$ for some $0 < c < 1$	$T(n) \in \Theta(n)$	Selection (\rightarrow later)
$T(n) = 2T(n/4) + \Theta(1)$	$T(n) \in \Theta(\sqrt{n})$	Range Search (\rightarrow later)
$T(n) = T(\sqrt{n}) + \Theta(1)$	$T(n) \in \Theta(\log \log n)$	Interpolation Search (\rightarrow later)

MergeSort Reference: See the Beidl book, CS 240E detailed analysis of MergeSort.

Lemma 3.7. For any constant $r > 1$, $f(n) \in \Theta(r^{n-1})$ if and only if $f(n) \in \Theta(r^n)$.

Proof. 1. Forward direction Assume $f(n) \in \Theta(r^{n-1})$.

(a) Since $f(n) \in \Theta(r^{n-1})$, we have

i. c_1, n_1 such that $f(n) \leq c_1 r^{n-1}$, for all $n \geq n_1$, and

ii. c_2, n_2 such that $c_2 r^{n-1} \leq f(n)$, for all $n \geq n_2$.

(b) From 1(a)i, $f(n) \leq \left(\frac{c_1}{r}\right) r^n$, for all $n \geq n_1$.

(c) From 1(a)ii, $\left(\frac{c_2}{r}\right) r^n \leq f(n)$, for all $n \geq n_2$.

(d) Hence $f(n) \in \Theta(r^n)$ as claimed.

2. Backward direction This is clear enough from the forward direction, I think.

□

4 Heaps

1. Refer to CS 240, Module 2, Section on Binary Heaps.
2. Quick Summary: A **heap** is a binary tree with the following properties:
 - (a) **Structural Property:** All the levels of a heap are completely filled, except (possibly) for the last level. The filled items in the last level are left-justified.
 - (b) **Heap Order Property:** For any node i , the key of the parent of i is larger than or equal to key of i .

The full name for this is **max-oriented binary heap**.

Task: Make this into a proper definition, when time permits.

5 Randomized Algorithms

1. Refer to CS 240, Module 3, Section on Randomized Algorithms.

6 Dictionary Using Ordered Linked List

1. Refer to CS 240, Module 4, Section on ADT Dictionary.
2. Quick Summary: Ordering the array improves search from $\Theta(n)$ to $\Theta(\log n)$, compared against the unordered option.

Task: Make this into a proper definition, when time permits.

7 AVL Trees

1. Refer to CS 240, Module 4, Section on AVL Trees.
2. Quick Summary: An **AVL** is a BST, with the additional balance property that the heights of the left and right subtrees can differ by at most 1.

Task: Make this into a proper definition, when time permits.

8 Tries

1. Refer to CS 240, Module 6, Section on Tries.
2. Quick Summary: A **Trie** is a radix tree (label each edge with the appropriate character).

Task: Make this into a proper definition, when time permits.

9 KD Trees

1. Refer to CS 240, Module 8, Section on KD Trees.
2. Quick Summary: A **KD Tree** is a binary tree, which has roughly half of its points in each subtree, at each level.

Task: Make this into a proper definition, when time permits.

10 Huffman Trees

1. Refer to CS 240, Module 10, Section on Huffman Trees.
2. Quick Summary: A **Huffman Tree** is a tree, to store an encoding, which will produce the minimum length of coded words, I think.

Task: Make this into a proper definition, when time permits.

11 Graphs

Notation: A **DAG** is a **directed acyclic graph**.

A **Hamiltonian Path** is a path that visits each vertex exactly once.

A **Hamiltonian Cycle** is a cycle that is also a Hamiltonian path.

12 Fibonacci Numbers

When time permits, copy/move the stuff from Lecture 07 about the Fibonacci numbers and the **Golden Ratio**.