

CS 341

These notes are meant to be supplementary to lecture slides & the textbook, and so may not contain all covered materials. Here I've chosen content which might not be easy to remember and/or is helpful to look at when doing assignments.

Asymptotic Analysis

- **Order Notations:**

- $f(n) \in O(g(n))$ if $\exists c > 0$ and $n_0 > 0$ such that $0 \leq f(n) \leq cg(n) \forall n \geq n_0$
 - f “grows no faster than” g
- $f(n) \in \Omega(g(n))$ if $\exists c > 0$ and $n_0 > 0$ such that $0 \leq cg(n) \leq f(n) \forall n \geq n_0$
 - f “grows no slower than” g
- $f(n) \in \Theta(g(n))$ if $\exists c_1, c_2 > 0$ and $n_0 > 0$ such that $0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \forall n \geq n_0$
 - f and g have the same complexity
- $f(n) \in o(g(n))$ if $\forall c > 0, \exists n_0 > 0$ such that $0 \leq f(n) < cg(n) \forall n \geq n_0$
 - f has lower complexity than g
- $f(n) \in \omega(g(n))$ if $\forall c > 0, \exists n_0 > 0$ such that $0 \leq cg(n) < f(n) \forall n \geq n_0$
 - f has higher complexity than g
- $f \in O(g)$ and $f \in \Omega(g) \iff f \in \Theta(g)$

- **Limit method:** suppose $L = \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$

- $f \in o(g)$ if $L = 0$
- $f \in \Theta(g)$ if $0 < L < \infty$
- $f \in \omega(g)$ if $L = \infty$

- Useful facts for first-principles proofs:

- $\log n \geq 1 \forall n \geq 2$; i.e. $\log n$ grows faster than 1
- $\log n \geq n \forall n \geq 0$; i.e. $\log n$ grows faster than n

- Some math rules:

- $\sum_{i=1}^n i^k \in \Theta(n^{k+1})$
- $\sum_{i=1}^n c^i \in$
 - $\Theta(c^n)$ if $c > 1$
 - $\Theta(n)$ if $c = 1$
 - $\Theta(1)$ if $c < 1$
- $a^{\log_b n} = n^{\log_b a}$ (Useful for recursion trees)
- $\lim_{x \rightarrow c} \frac{f(x)}{g(x)} = \lim_{x \rightarrow c} \frac{f'(x)}{g'(x)}$ if it exists (L'Hopital's Rule)

- $\log(n!) \in \Theta(n \log n)$ (Stirling's Approximation)
- $\sum_{i=1}^n \frac{1}{i} \in \Theta(\log n)$ (Harmonic series)

Divide and Conquer

- **Recursion-tree method:**

- E.g. given the recurrence $T(n) = aT(n/b) + f(n), T(1) = c$:
 - a is the # of recursive calls made in the function
 - b is the # by which the input size n is divided in each recursive call
 - $f(n)$ is the runtime of the “work done outside of the recursive calls”
 - c is the constant-time work done in each recursive call in the base case
- Each node of the recursion tree represents the cost of the work done other than making recursive calls
- Each row represents the total cost of work done in all recursive calls at that recursion “level”
- The height of the tree depends on the factor that the input size is divided by; i.e. $\log_b n$
- E.g.: Picture this as a tree where each node (except for leaves) has a children;

Level 0:	$f(n)$			total = $f(n)$
Level 1:	$f(n/b)$...	$f(n/b)$	total = $a f(n/b)$
Level 2:	$f(n/b^2)$...	$f(n/b^2)$	total = $a^2 f(n/b^2)$
...				
Level k :	$f(n/b^k)$...	$f(n/b^k)$	total = $a^k f(n/b^k)$
...				
Level $\log_b n$:	c	...	c	total = $ca^{\log_b n} = cn^{\log_b a}$

- Total runtime of recursion tree is (summing every row total):

$$T(n) \in \Theta \left(\sum_{i=0}^{\log_b n - 1} a^i f(n/b^i) \right) + \Theta(n^{\log_b a})$$