



COMPSCI 130B Discussion

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Objectives

- Time complexity analysis.
- Proof by induction.

Time Complexity Analysis

- Evaluate the running time of a procedure/algorithm.
- Expressed in term of a function of the input size.
- Asymptotic notations Big O, Big Ω , Big Θ .

Time Complexity Analysis – Big O

- Read: Asymptotic Upper Bound.
- $f(n) = O(g(n)) \Leftrightarrow \forall n > n_0, \exists c > 0, f(n) \leq c * g(n).$
- $f(n)$ grows no faster than $g(n)$.

Time Complexity Analysis – Big Ω

- Read: Asymptotic Lower Bound.
- $f(n) = \Omega(g(n)) \Leftrightarrow \forall n > n_0, \exists c > 0, c * g(n) \leq f(n)$.
- $f(n)$ grows no slower than $g(n)$

Time Complexity Analysis – Big Θ

- Read: Composite bound.
- $f(n) = \Theta(g(n)) \Leftrightarrow f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$.
- $f(n)$ grows at the same rate as $g(n)$.

Time Complexity Analysis

- Small o and small ω : very similar to Big O and Big Ω .
- $f(n) = o(g(n)) \Leftrightarrow \forall n > n_0, \exists c > 0, f(n) < c * g(n)$.
- $f(n) = \omega(g(n)) \Leftrightarrow \forall n > n_0, \exists c > 0, c * g(n) < f(n)$.

Time Complexity Analysis

$$f = O(g) \text{ and } f = \Omega(g) \Leftrightarrow f = \Theta(g)$$

$$f = O(g) \Leftrightarrow g = \Omega(f)$$

$$f = o(g) \Leftrightarrow g = \omega(f)$$

$$f = o(g) \Rightarrow f = O(g)$$

$$f = \omega(g) \Rightarrow f = \Omega(g)$$

$$f \sim g \Rightarrow f = \Theta(g)$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) \neq 0, \infty \Rightarrow f = \Theta(g)$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) \neq \infty \Rightarrow f = O(g)$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) \neq 0 \Rightarrow f = \Omega(g)$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) = 1 \Rightarrow f \sim g$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) = 0 \Rightarrow f = o(g)$$

$$\lim_{n \rightarrow \infty} f(n)/g(n) = \infty \Rightarrow f = \omega(g)$$

Time Complexity Analysis

- Generally, time complexity is determined by the most dominant term in the function.
 - $O(5) = O(1)$
 - $O(2n^2) = O(n^2)$
 - $O(4n^3 + 2n^2 + 1) = O(n^3)$
 - $O(2n + 5 \log(n)) = O(n)$
 - $O(n \log(n) + n) = O(n \log(n))$
- In terms of dominance:

$$a^n > n^a > \log_a n > a$$

Time Complexity Analysis - Example

- Prove that $n^2 + 17n + 2$ is $O(n^2)$.

$$\begin{aligned}n^2 + 17n + 2 &\leq n^2 + 17n^2 + 2 \text{ (when } n \geq 2\text{)} \\&\leq n^2 + 17n^2 + n^2 \text{ (when } n \geq 2\text{)} \\&= 19n^2\end{aligned}$$

So when $n \geq 2$, with $c = 19$, $n^2 + 17n + 2 \leq cn^2$. Therefore, $n^2 + 17n + 2 = O(n^2)$.

- Is $(n^2 + 17n + 2)$ $O(n^3)$? $O(n)$?

Time Complexity Analysis - Example

- Prove that $\log_{10}(x)$ is $O(\log_2(x))$.

$$\lim_{n \rightarrow \infty} \frac{\log_{10}(x)}{\log_2(x)} = \lim_{n \rightarrow \infty} \frac{\log_2(x)}{\log_2(10) \log_2(x)} = \lim_{n \rightarrow \infty} \frac{1}{\log_2(10)} = \frac{1}{\log_2(10)}$$

Therefore, $\log_{10}(x) = \Theta(\log_2(x))$, which implies $\log_{10}(x) = O(\log_2(x))$.

Proof by Induction

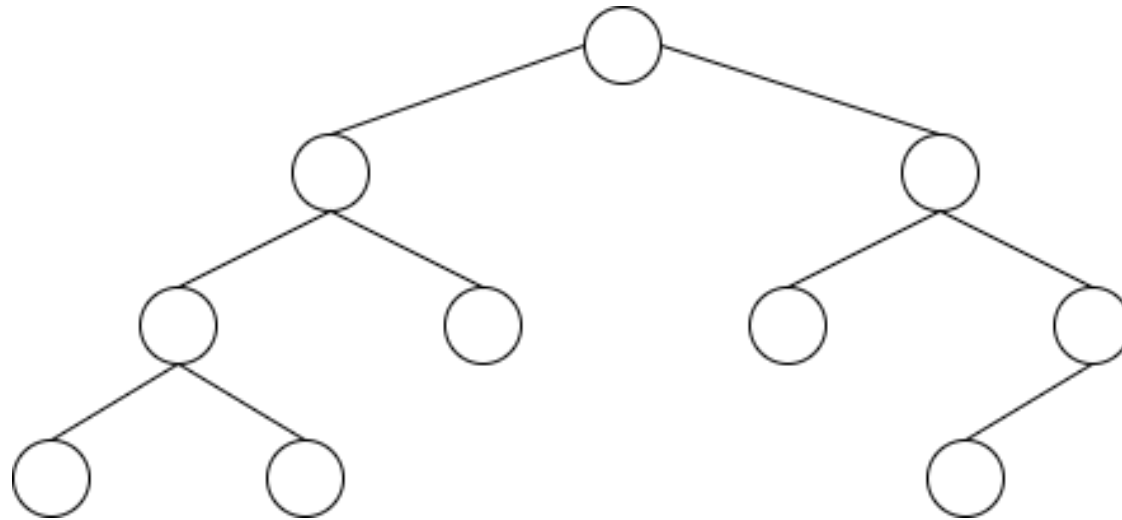
- Prove an assertion $P(n)$ is true for all $n \geq 0$.
- Base cases:
 - Check that $P(n)$ holds for $0 \leq n \leq k$.
- Induction step:
 - Prove that $P(n) \Rightarrow P(n + 1)$ for $n > k$.
- Conclusion: $P(n)$ is true for all $n \geq 0$.

Proof by Induction - Example

- $P(n)$: $7^n - 1 \div 6$
- Base case: $n = 1, 7^1 - 1 = 6 \div 6$
- Induction step:
 - Assume $P(k)$ holds.
 - $7^{k+1} - 1 = 7(7^k - 1) + 6 \div 6$.
 - Therefore, $P(k + 1)$ also holds.
- Conclusion: $P(n)$ holds for all $n \geq 1$.

Proof by Induction - Example

- $P(n)$: A binary tree with n nodes has $n - 1$ edges.



Proof by Induction - Example

- Base case: Binary tree with 1 node has 0 edge.
- Induction step:
 - Assume $P(k)$ is true.
 - Consider a tree with $k + 1$ nodes.
 - Hint: Removing a random leaf node.
 - ...
- Conclusion.