

BST RUNNING TIME ANALYSIS

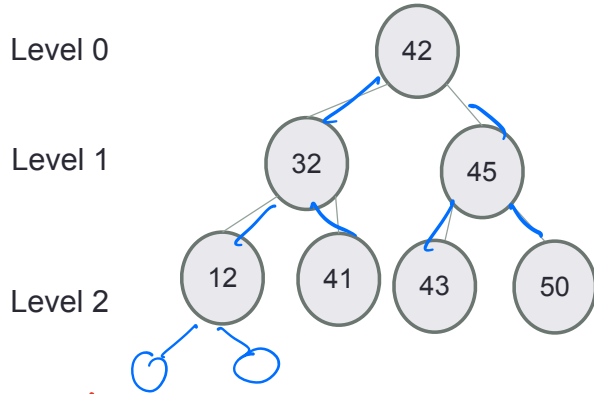
Problem Solving with Computers-II

C++

```
#include <iostream>
using namespace std;

int main(){
    cout<<"Hola Facebook!";
    return 0;
}
```

Types of BSTs



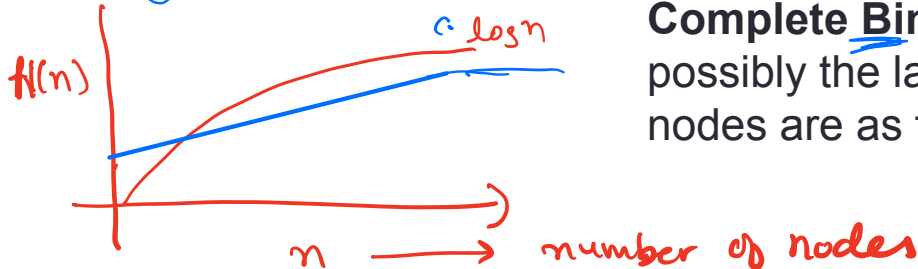
n : is the number of nodes in a BST
 H : is the height of the BST

Balanced BST: $H(n) = O(\log n)$

AVL, Red-Black Trees, Full BST

Full Binary Tree: Every node other than the leaves has two children.

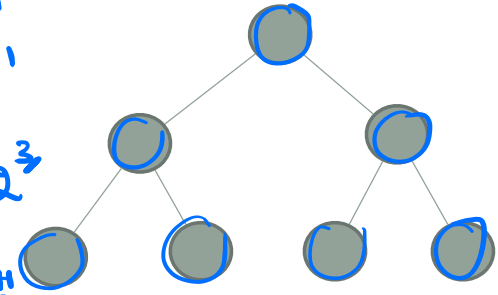
Complete Binary Tree: Every level, except possibly the last, is completely filled, and all nodes are as far left as possible



Relating H (height) and n (#nodes) for a full binary tree

Height
 0
 1
 2
 3
 ...
 H

n
 1 = $2^{0+1} - 1$
 1 + 2 = $2^{1+1} - 1$
 1 + 2 + 2^2 = $2^{2+1} - 1$
 1 + 2 + 2^2 + 2^3 = $2^{3+1} - 1$



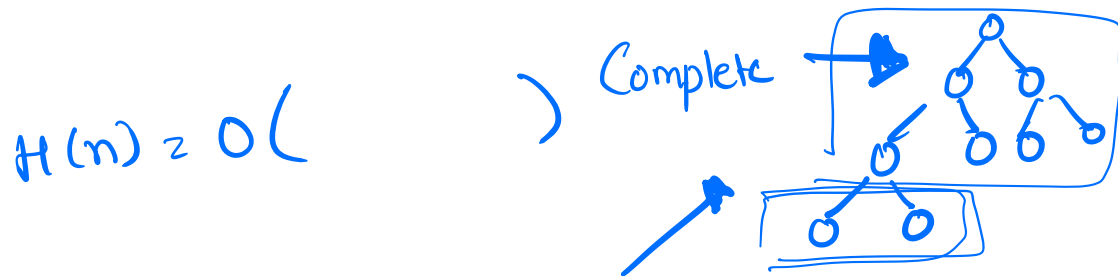
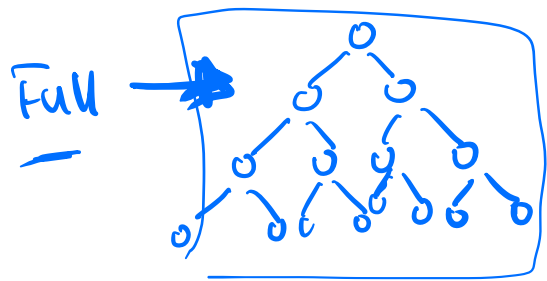
Level 0
 Level 1
 Level 2
 ...
 Level h

Claim: $1 + 2 + 2^2 + \dots + 2^H = 2^{H+1} - 1$

$n = 2^{H(n)+1} - 1$
 $n + 1 = 2^{H(n)+1}$
 $\log_2(n+1) = H(n) + 1$
 Therefore $H(n) = \log_2(n+1) - 1$,
 a tree with n nodes

$H(n) = O(\log_2 n)$

$$H(n) = O(\log n)$$



$$H_c(n) = \log_2(n+1) - 1 \pm 1$$
$$= O(\log_2(n))$$

Big-Omega

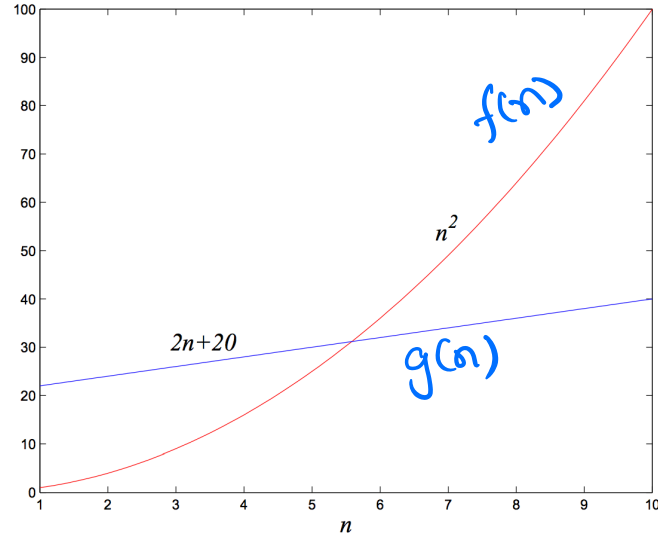
- $f(n)$ and $g(n)$ map positive integer inputs to positive reals.

We say $f = \Omega(g)$ if there are constants $c > 0, k > 0$ such that $c \cdot g(n) \leq f(n)$ for $n \geq k$

$$f = \Omega(g)$$

means that “ f grows at least as fast as g ”

$$f(n) = \Omega(g(n))$$



Big-Theta



In Practice

Big-Oh is really asking Big Theta

- $f(n)$ and $g(n)$ map positive integer inputs to positive reals.

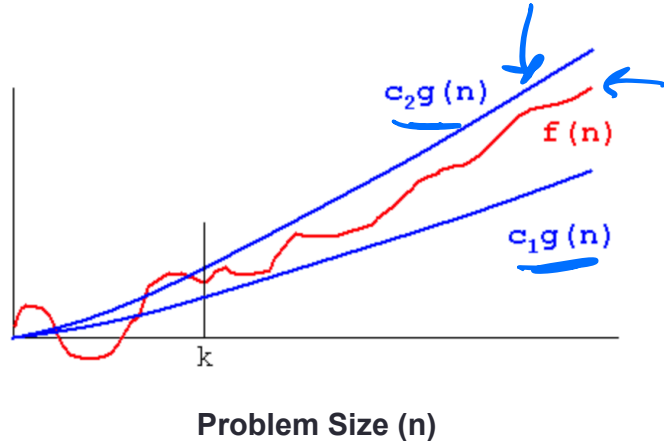
We say $f = \Theta(g)$ if there are constants c_1, c_2, k such that

$$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n), \text{ for } n \geq k$$

$$f(n) = \Theta(g(n))$$

$$f(n) = O(g(n))$$

$$f(n) \approx \Omega(g(n)) \text{ Running time}$$



Big-O analysis of iterative fibonacci

```
function F(n){  
  { Create an array fib[1..n] }  $O(1)$   
  fib[1] = 1  
  fib[2] = 1  
  for i = 3 to n:  $n-3$   
    fib[i] = fib[i-1] + fib[i-2]  $\rightarrow O(1)$   
  return fib[n]  
}
```

$T(n)$ is the running time of calculating $F(n)$

$$T(n) = O(1) + (n-3) \cdot O(1)$$

$$= O(n)$$

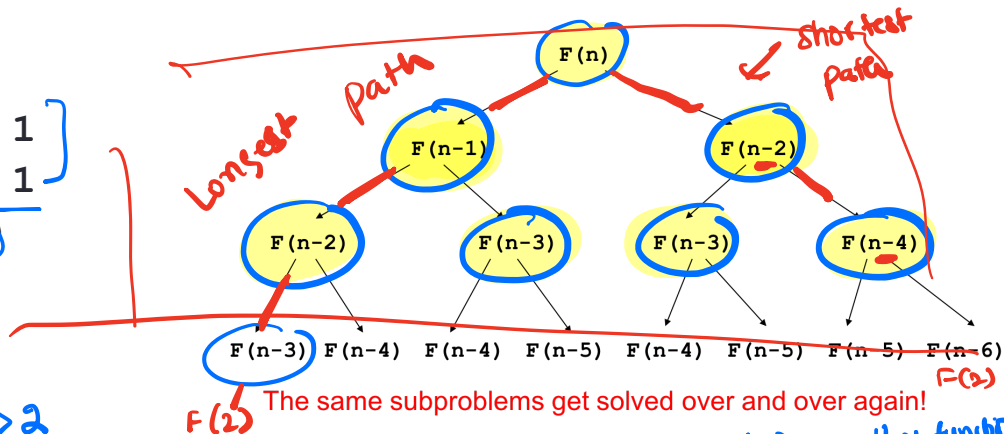
Big-O analysis of recursive fibonacci

What takes so long? Let's unravel the recursion...

```
function F(n) {  
  if (n == 1) return 1  
  if (n == 2) return 1  
  return F(n-1) + F(n-2)  
}
```

$$T(1) = 2$$
$$T(2) = 3$$

$$T(n) = 1 + 1 + 2 + 1 + T(n-1) + T(n-2)$$
$$= T(n-1) + T(n-2) + 5$$



$$T(n) = c \cdot \text{\# of function calls}$$

Length of the shortest path $\approx \frac{n}{2}$

Length of the longest path = $n-2$

min number of function calls.
 $= 2^{\frac{n}{2}+1} - 1$

max number of function calls
 $= 2^{n+1} - 1$

$\rightarrow T(n) = \Omega\left(2^{\frac{n}{2}}\right) \leftarrow \text{exp.}$

$\rightarrow T(n) = O\left(2^n\right) \leftarrow \text{exp.}$

$T(n) = \Theta\left(2^{0.69n}\right)$

Balanced trees

- Balanced trees by definition have a height of $O(\log N)$
- A completely filled tree is one example of a balanced tree
- Other Balanced BSTs include AVL trees, red black trees and so on
- Visualize operations on an AVL tree: <https://visualgo.net/bn/bst>

Summary of operations

Operation	Sorted Array	Binary Search Tree	Linked List
Min			
Max			
Median			
Successor			
Predecessor			
Search			
Insert			
Delete			