

# BEST & WORST CASE ANALYSIS

## RUNNING TIME OF BST OPERATIONS

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Problem Solving with Computers-II

The image shows the C++ logo in blue, followed by a snippet of C++ code. The code is: 

```
#include <iostream>
using namespace std;
int main(){
    cout<<"Hola Facebook!n";
    return 0;
}
```

# Definition of Big-O

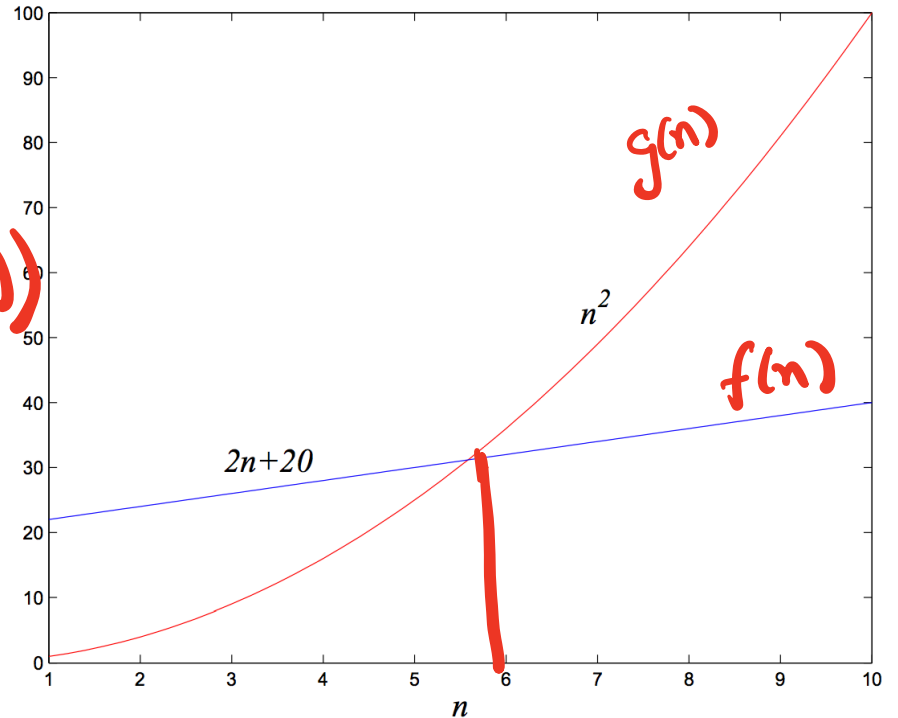
$f(n)$  and  $g(n)$  map positive integer inputs to positive reals.

We say  $f = O(g)$  if there is a constant  $c > 0$  and  $k > 0$  such that  $f(n) \leq c \cdot g(n)$  for all  $n \geq k$ .

$f = O(g)$

means that “f grows no faster than g”

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



# Big-Omega

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

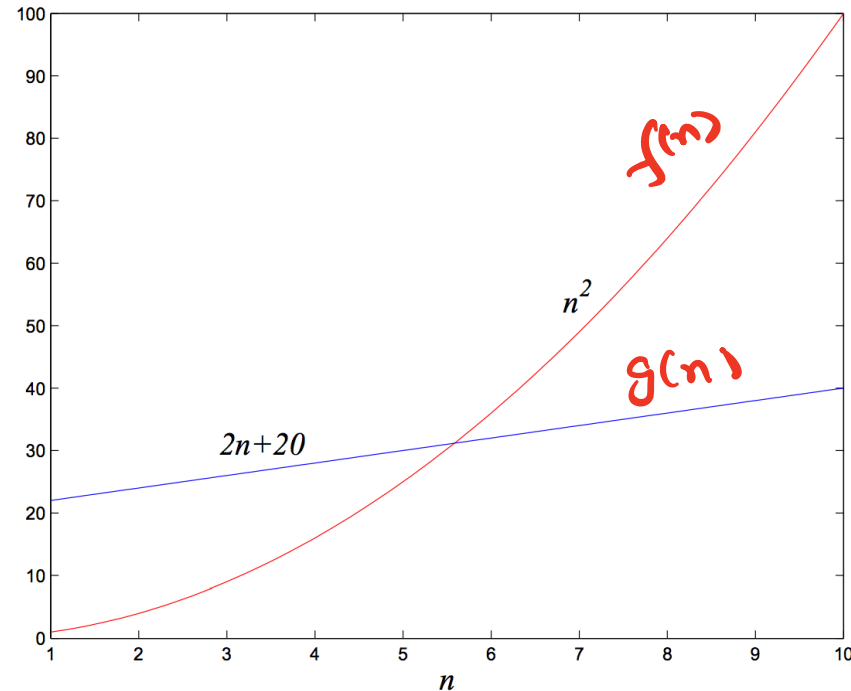
We say  $f = \Omega(g)$  if there are constants  $c > 0, k > 0$  such that

$$c \cdot g(n) \leq f(n) \text{ for } n \geq k$$

$$f = \Omega(g)$$

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

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



# 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

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

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

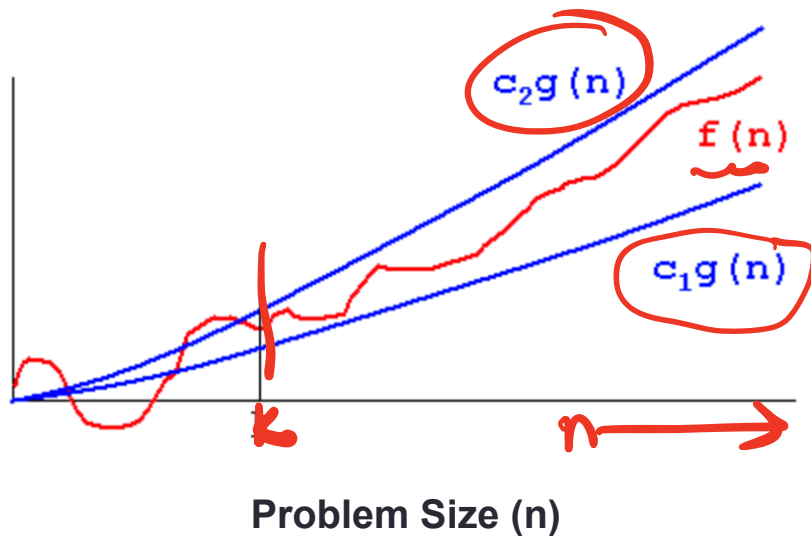
$$f(n) = 5n \log(n)$$

$$= O(n^2)$$

Running time

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

"tightest"  $\uparrow$  upper bound



# Best case and worst case analysis

What is the Big-O running time of search in a sorted array of size n?

*n is very large*

...using linear search?

*Best case: looking for the min key value :  $O(1)$*

*Worst case: looking for the max key value :  $O(n)$*

...using binary search?

*Best case: looking for the mid value :  $O(1)$*

*Worst case: looking for a key that doesn't exist:  $O(\log n)$*

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

# Worst case analysis of binary search

```
bool binarySearch(int arr[], int element, int n){
//Precondition: input array arr is sorted in ascending order
```

```

C1 {
int begin = 0;
int end = n-1;
int mid;
while (begin <= end){
mid = (end + begin)/2;
if(arr[mid]==element){
return true;
} else if (arr[mid] < element){
begin = mid + 1;
} else{
end = mid - 1;
}
}
return false;
}

```

Handwritten annotations:  $C_1$  is written next to the initialization of `begin`, `end`, and `mid`.  $C_2$  is written next to the `while` loop. Red circles highlight `begin + 1` and `end - 1`. Red arrows point to the `return true` and the `begin = mid + 1` line.

Iteration no.	end-begin
1	$n-1$
2	$\left(\frac{n-1}{2}\right) - 1$
3	$\frac{1}{2} \left(\frac{n-1}{2} - 1\right) - 1$ $= \frac{n-1}{2^2} - \frac{1}{2} - 1$
4	$\frac{n-1}{2^3} - \frac{1}{2^2} - \frac{1}{2} - 1$
:	

Handwritten annotations: The iteration numbers 1, 2, 3, and 4 are circled in red. The expression for iteration 2 is highlighted in pink. The expressions for iterations 3 and 4 are highlighted in blue.

Iteration No

$k$

end-begin

$$\frac{n-1}{2^{k-1}} - \left(1 + \frac{1}{2} + \frac{1}{2^2} + \dots + \frac{1}{2^{k-2}}\right)$$

(Sum the geometric series)

$$= \frac{n-1}{2^{k-1}} - \frac{\left(1 - \frac{1}{2^{k-2+1}}\right)}{\left(1 - \frac{1}{2}\right)}$$

$$= \frac{n-1}{2^{k-1}} - 2 \cdot \left(1 - \frac{1}{2^{k-1}}\right)$$

$$= \frac{(n-1+2)}{2^{k-1}} - 2$$

At iteration  $k$ ,

$$(\text{end-begin}) \geq \frac{(n+1)}{2^{k-1}} - 2 \quad \text{--- (1)}$$

We stop when  $(\text{end-begin}) < 0$

(Substitute for end-begin from (1))

$$\frac{n+1}{2^{k-1}} < 2$$

$$2 \cdot 2^{k-1} > n+1$$

$$2^k > n+1$$

$$k > \log_2(n+1)$$

The max number of iterations of the while loop is  $\log_2(n+1)$

Running time of binary search on array of size  $n$

$$T(n) = c_1 + \log(n+1) \cdot c_2, \text{ for some constants } c_1 \geq c_2$$

$$= \underline{O(\log n)} \text{ (By def. of Big O)}$$



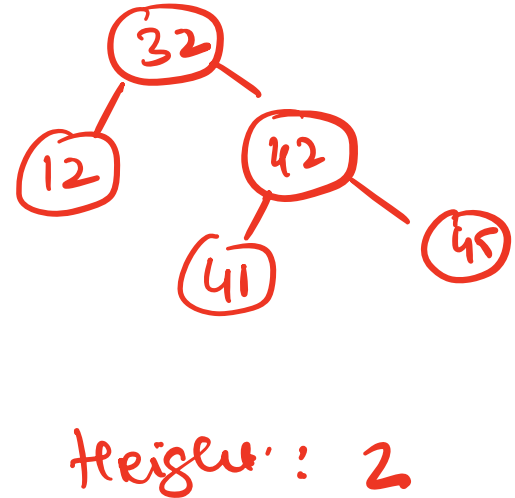
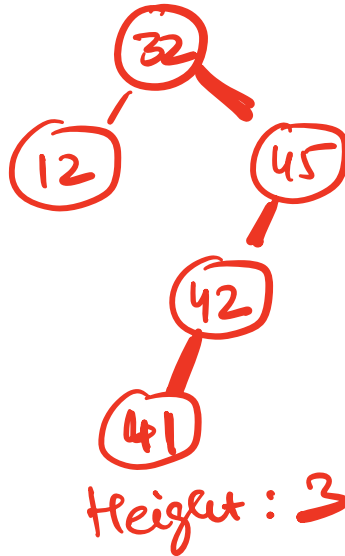
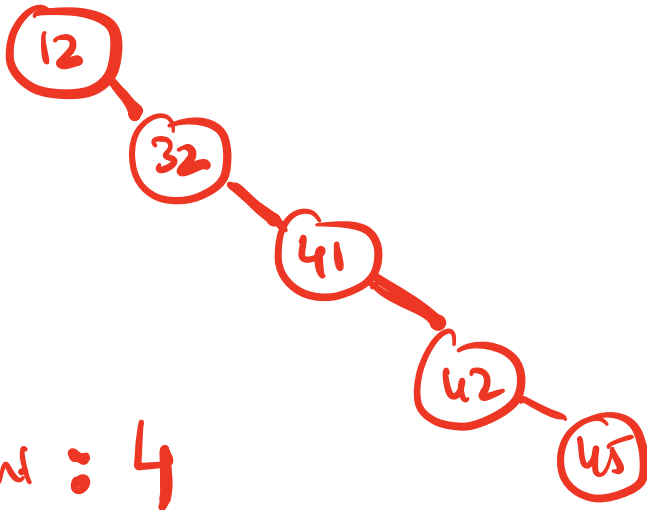
# Best case and worst case : sorted array

	Best case	Worst case
• Search (Binary search)	$O(1)$	$O(\log n)$
• Min/Max	$O(1)$	$O(1)$
• Median	$O(1)$	$O(1)$
• Successor/Predecessor	$O(1)$	$O(1)$
• Insert	$O(1)$	$O(n)$
• Delete	$O(1)$	$O(n)$

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14



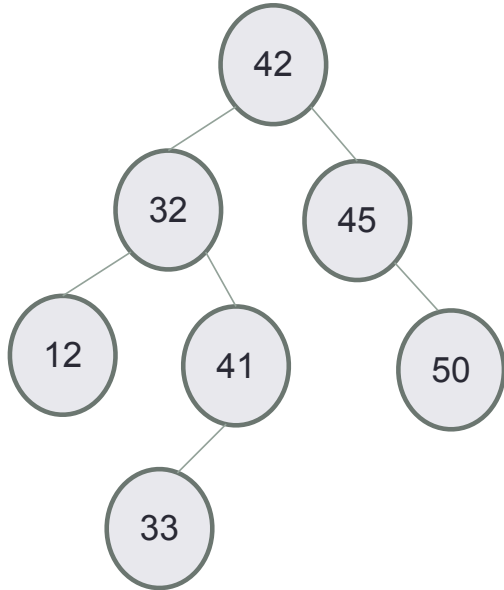
- Path – a sequence of (zero or more) connected nodes.
- Length of a path - number of edges traversed on the path
- Height of node – Length of the longest path from the node to a leaf node.
- **Height of the tree** - Length of the longest path from the **root** to a leaf node.



BSTs of different heights are possible with the same set of keys  
 Examples for keys: 12, 32, 41, 42, 45

# BST search - best case

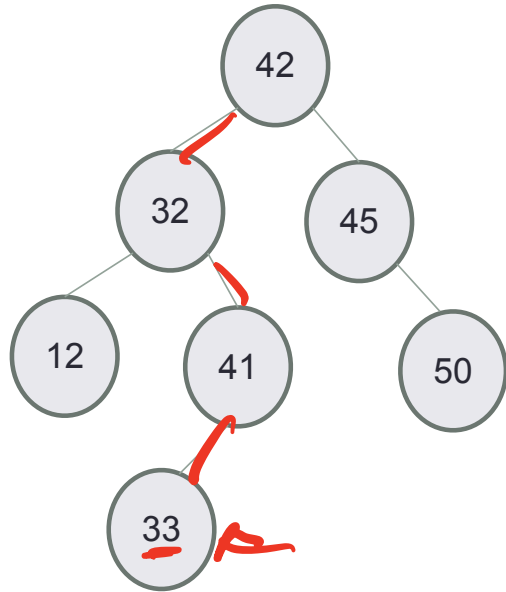
$O(1)$



Given a BST with N nodes, in the best case, which key would be searching for?

- A. root node (e.g. 42)
- B. any leaf node (e.g. 12 or 33 or 50)
- C. leaf node that is on the longest path from the root (e.g. 33)
- D. any key, there is no best or worst case

# BST search - worst case



Given a BST with N nodes, in the worst case, which key would be searching for?

A. root node (e.g. 42)

B. leaf node (e.g. 12 or 41 or 50)

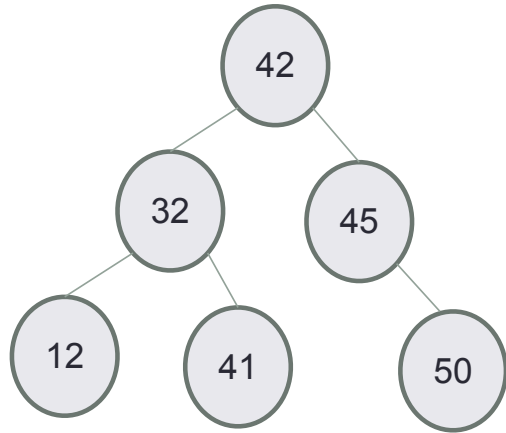
**C.** leaf node that is on the longest path from the root (e.g. 33)

D. a key that doesn't exist in the tree

$$T(n) = O(H)$$

H: Height of the tree

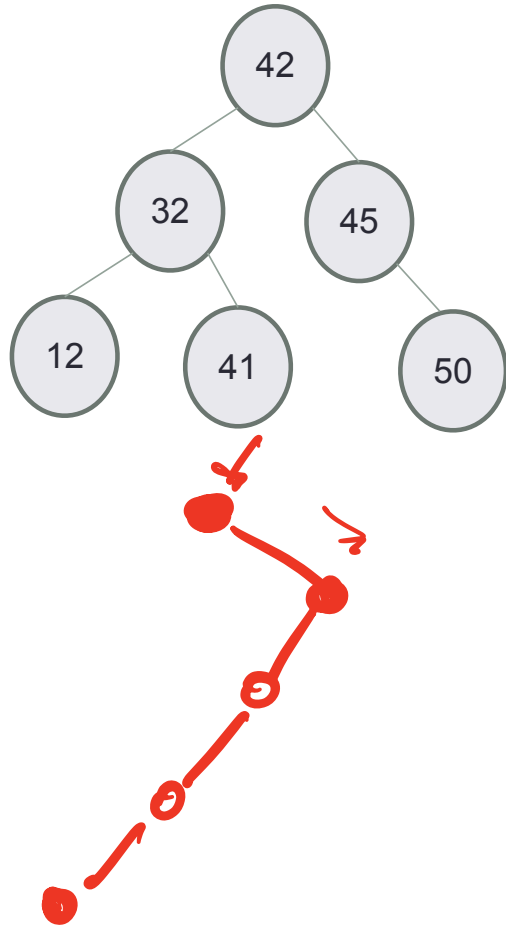
# Worst case Big-O of search, insert, min, max



Given a BST of height  $H$  with  $N$  nodes, what is the running time complexity of searching for a key (in the worst case)?

- A.  $O(1)$
- B.  $O(\log H)$
- C.  $O(H)$
- D.  $O(H * \log H)$
- E.  $O(N)$

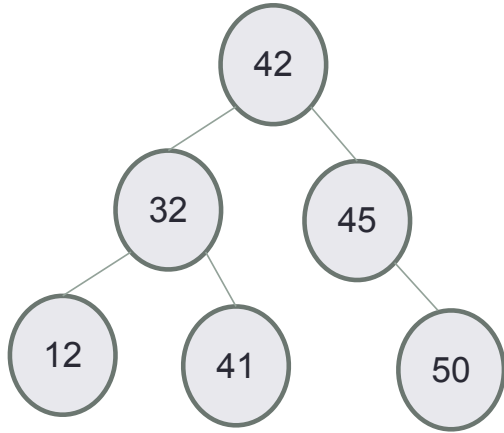
# BST operations (worst case)



Given a BST of height  $H$  and  $N$  nodes, which of the following operations has a complexity of  $O(H)$ ?

- A. min or max
- B. insert
- C. predecessor or successor
- D. delete
- E. All of the above

# Big O of traversals

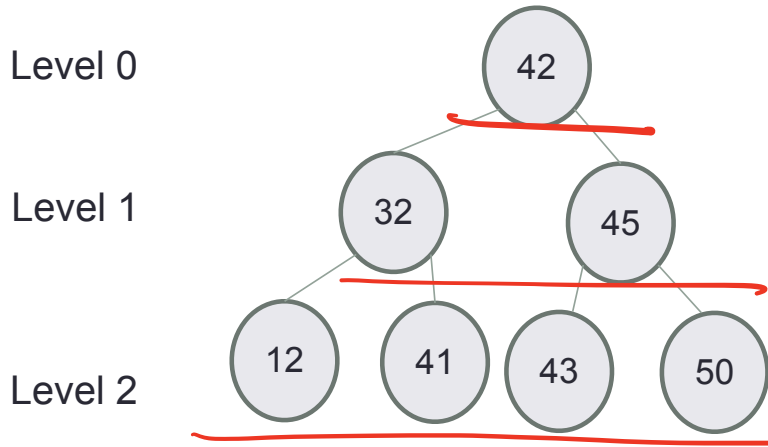


In Order:  $O(N)$

Pre Order:  $O(N)$

Post Order:  $O(N)$

# Types of BSTs



Example of a full  
binary tree

**Balanced BST:** Any BST where the height is  $O(\log n)$

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

**Full Binary Tree:** A complete binary tree whose last level is completely filled

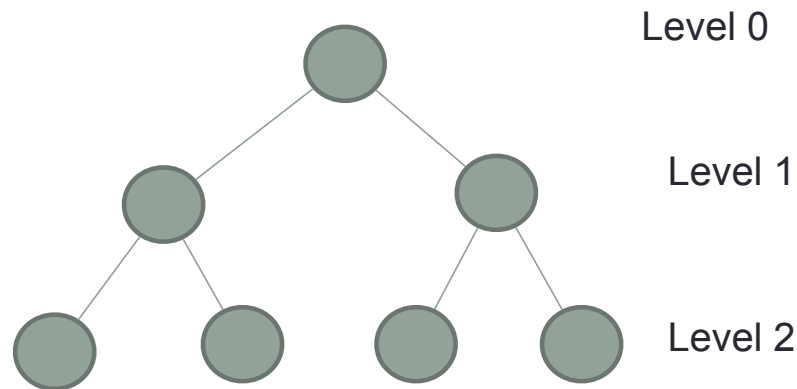


# Relating H (height) and n (#nodes) for a full binary tree

We derived this result in the previous lecture.

Height of full binary tree :  $H$   
 Number of nodes :  $N$

Sum of the number of nodes from level 0 to level  $H$  must be equal to  $N$



Sum of the number of nodes from level 0 to  $H$ ...

$$= 2^0 + 2^1 + 2^2 + \dots + 2^H$$

$$= 2^{H+1} - 1$$

Therefore,

$$\Rightarrow \frac{2^{H+1} - 1}{H+1} = \frac{N}{\log_2(N+1)} \Rightarrow$$

$$H = \log_2(N+1) - 1$$

$$= O(\log N)$$

A full binary tree is a balanced tree because its height is  $O(\log N)$

# 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>