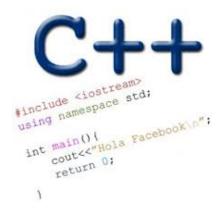
# RUNNING TIME ANALYSIS - PART 2 BINARY SEARCH TREES

Problem Solving with Computers-II



# How is PA01 going?

- A. Done!
- B. On track to finish
- C. On track to finish but my code is a mess
- D. Stuck and struggling
- E. Haven't started

#### Midterm 2

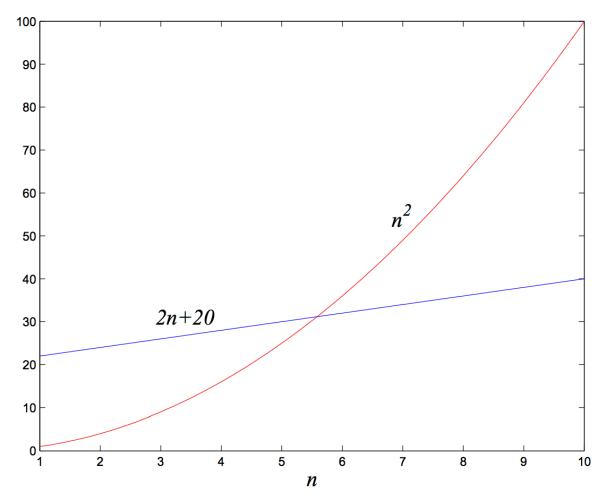
- Cumulative but the focus will be on
  - BST
  - Running time analysis

## A more precise definition of Big-O

- f(n) and g(n): running times of two algorithms on inputs of size n.
- 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  $f(n) \le c \cdot g(n)$  for all n > = k.

f = O(g)means that "f grows no faster than g"



# What is the Big-O running time of algoX?

 Assume dataA is some data structure that supports the following operations with the given running times, where N is the number of keys stored in the data structure:

```
insert: O(log N)
 • min: O(1)
 delete: O(log N)
void algoX(int arr[], int N)
       dataA ds;//ds contains no keys
        for (int i=0; i < N; i=i++)
               ds.insert(arr[i]);
        for (int i=0; i < N; i=i++)
               arr[i] = ds.min();
               ds.delete(arr[i]);
```

- A.  $O(N^2)$
- B. O(N logN)
- C. O(N)
- D. O(log N)
- E. Not enough information to compute

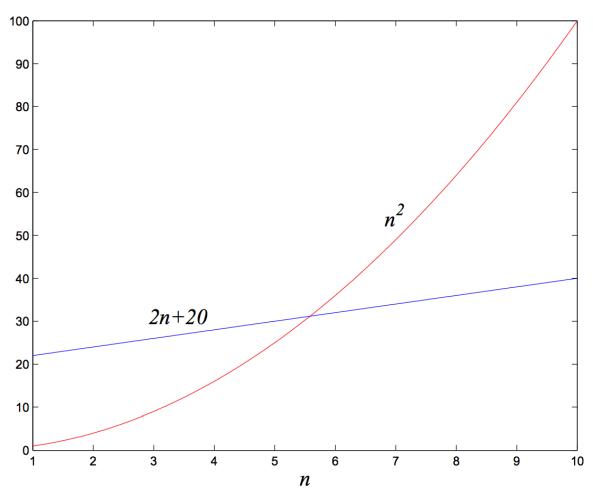
# Big-Omega

- f(n) and g(n): running times of two algorithms on inputs of size n.
- 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  $c \cdot g(n) \le f(n)$  for n >= k

 $f = \Omega(g)$ means that "f grows at least as fast as g"

g is a lower bound



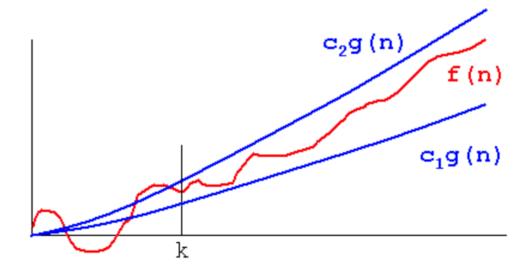
### Big-Theta

- f(n) and g(n): running times of two algorithms on inputs of size n.
- 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  $0 \le c_1 g(n) \le f(n) \le c_2 g(n)$ , for  $n \ge k$ 

f and g grow at the same rate

**Running time** 

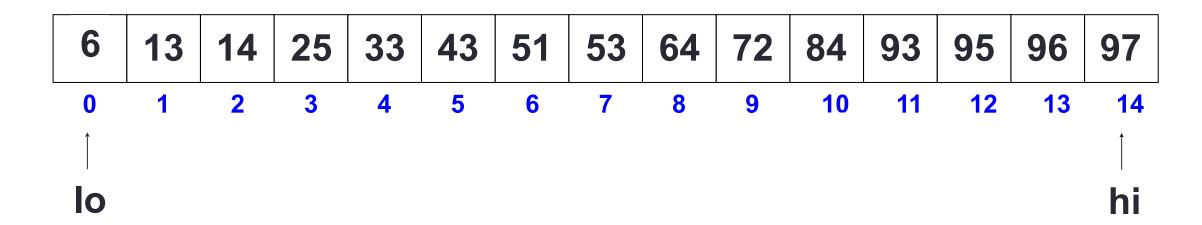


**Problem Size (n)** 

### Best case, worst case, average case running times

#### **Operations on sorted arrays**

- Min:
- Max:
- Median:
- Successor:
- Predecessor:
- Search:
- Insert :
- Delete:



#### Worst case analysis of binary search

```
bool binarySearch(int arr[], int element, int N){
//Precondition: input array arr is sorted in ascending order
  int begin = 0;
  int end = N-1;
  int mid;
  while (begin <= end){</pre>
    mid = (end + begin)/2;
    if(arr[mid] == element){
      return true;
    }else if (arr[mid] < element){</pre>
      begin = mid + 1;
    }else{
      end = mid - 1;
  return false;
```

# **Binary Search Trees**

- WHAT are the operations supported?
- HOW do we implement them?
- WHAT are the (worst case) running times of each operation?

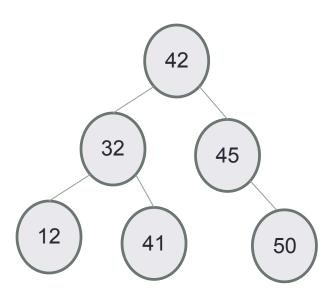
#### Height of the tree



- Path a sequence of nodes and edges connecting a node with a descendant.
- A path starts from a node and ends at another node or a leaf
- Height of node The height of a node is the number of edges on the longest downward path between that node and a leaf.

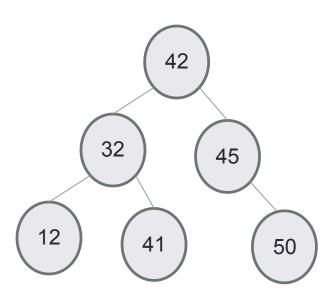
BSTs of different heights are possible with the same set of keys Examples for keys: 12, 32, 41, 42, 45

# Worst case Big-O of search



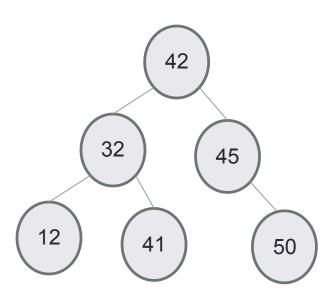
- Given a BST of height H with N nodes, what is the worst case complexity of searching for a key?
- A. O(1)
- B. O(log H)
- C. O(H)
- D. O(H\*log H)
- E. O(N)

# Worst case Big-O of insert



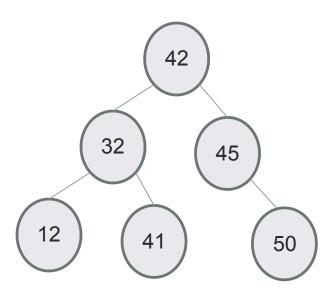
- Given a BST of height H and N nodes, what is the worst case complexity of inserting a key?
- A. O(1)
- B. O(log H)
- C. O(H)
- D. O(H\*log H)
- E. O(N)

# Worst case Big-O of min/max



- Given a BST of height H and N nodes, what is the worst case complexity of finding the minimum or maximum key?
- A. O(1)
- B. O(log H)
- C. O(H)
- D. O(H\*log H)
- E. O(N)

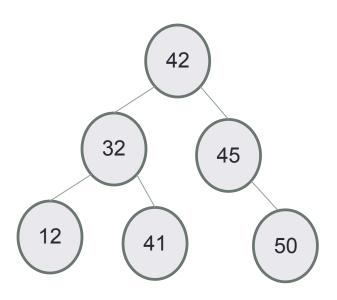
#### Worst case Big-O of predecessor/successor



• Given a BST of height H and N nodes, what is the worst case complexity of finding the predecessor or successor key?

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

# Worst case Big-O of delete

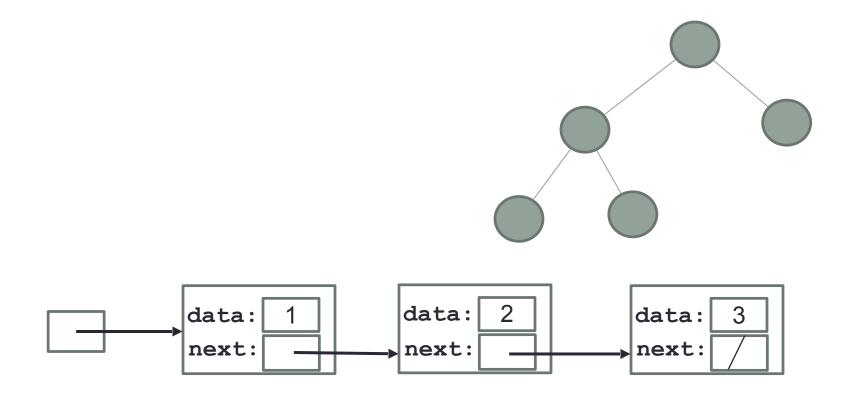


- Given a BST of height H and N nodes, what is the worst case complexity of deleting the key (assume no duplicates)?
- A. O(1)
- B. O(log H)
- C. O(H)
- D. O(H\*log H)
- E. O(N)

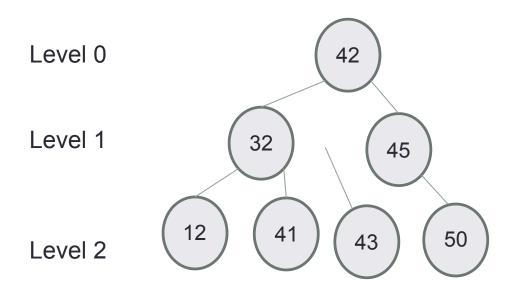
## Worst case analysis

Are binary search trees *really* faster than linked lists for finding elements?

- A. Yes
- B. No

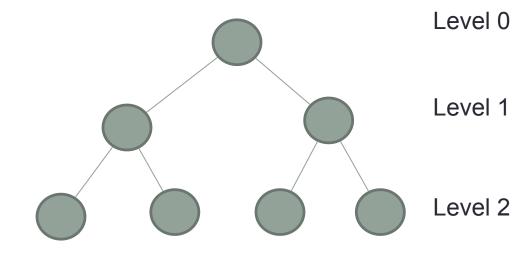


### Completely filled binary tree



Nodes at each level have exactly two children, except the nodes at the last level

# Relating H (height) and N (#nodes) find is O(H), we want to find a f(N) = H



How many nodes are on level L in a completely filled binary search tree?

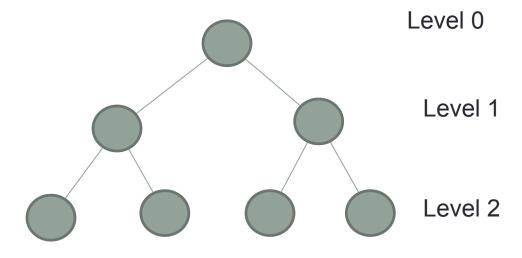
A.2

B.L

C.2\*L

D.2L

# Relating H (height) and N (#nodes) find is O(H), we want to find a f(N) = H

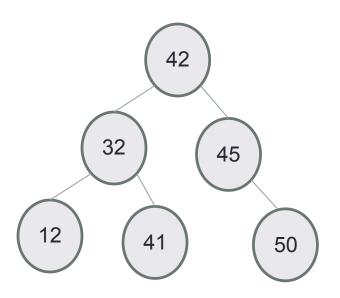


Finally, what is the height (exactly) of the tree in terms of N?

#### 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: <a href="https://visualgo.net/bn/bst">https://visualgo.net/bn/bst</a>

# Big O of traversals



In Order:

Pre Order:

Post Order:

# Summary of operations

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