

CS32 Discussion Week 7

Muhao Chen

muhaochen@ucla.edu

<http://yellowstone.cs.ucla.edu/~muhao/>

Outline

- Big-O Notation
- Sorting
- Tree

Complexity

- Quantify the efficiency of a program.
- The magnitude of **time** and **space** cost for an algorithm given certain size of input.
 - Time complexity: quantifies the run time.
 - Space complexity: quantifies the usage of the memory (or sometimes hard disk drives, cloud disk drives, etc.).

Size of input vs. Running time

- A program gets some kind of **input**, does something meaningful (hopefully), and produces some **output**.
- Naturally, the size of input determines how long a program runs.
 - Sorting an array of 1000 items should run a lot longer than sorting an array of 10 items.
 - But how much longer?
- Sometimes, the size of input doesn't matter.
 - Figuring out the size of a C++ string (`s.size()`): always the same running time.

Big Question

- Given an input of size **n**, approximately how long does the algorithm take to finish the task, in terms of **n**?
→ Big-O notation

Formal Definition of Big-O

- Let $T(n)$ be the function that measures the runtime of the program given n size of input
- Let $g(n)$ be another function defined on the real number field.
 - $T(n) = O(g(n))$ iff $\exists M > 0$ and $\exists N > 0$ s.t. $\forall n > N : T(n) \leq M * g(n)$

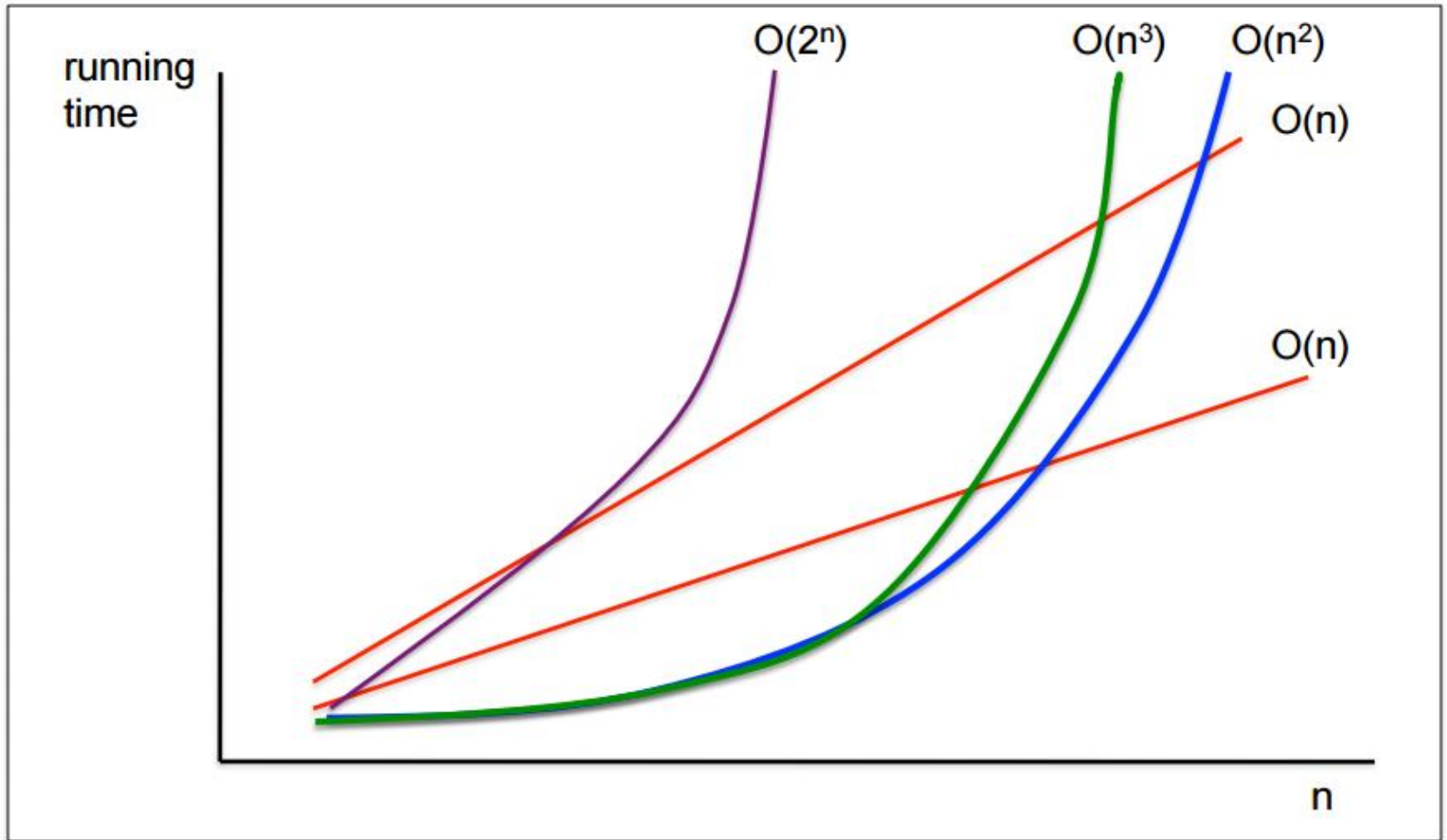
Upper bound

When the input size n grows above certain scale N , the runtime $T(n)$ is of the same or less magnitude of the function g inside $O()$

Big-O

- If your algorithm takes ...
 - about n steps: $O(n)$
 - about $2n$ steps: $O(n)$
 - about n^2 steps: $O(n^2)$
 - about $3n^2 + n$ steps: $O(n^2)$
 - about 2^n steps: $O(2^n)$
- Which one grows faster in the long term?
 - $10000n$ vs. $0.00001n^2$

Big-O



Efficiency

- Algorithms are considered efficient if it runs reasonably fast even with a large input.
- Algorithms are considered inefficient if it runs slow even with a small input.
- For more precise definitions, take CS180 and 181. In this class, we will use these simple intuitions to analyze algorithms.

Linear Search

- Unsorted array – look for an item

linear_search(array arr, size n, value v)

```
{  
  for (i=0 to n-1)           Best case?  
  {  
    if (arr[i] == v)         Average case?  
      return i;  
  }  
  return -1;                 Worst case?  
}
```

Linear Search

- Unsorted array – look for an item

linear_search(array arr, size n, value v)

{

 for (i=0 to n-1)

 {

 if (arr[i] == v)

 return i;

 }

 return -1;

}

Best case?

v is found in the first slot (a[0]) – takes 1 step

Average case?

takes $n/2 = \frac{1}{2}n$ steps (assuming v can appear at any location in the array with an equal probability)

Worst case?

not found – n steps

Linear Search

- Unsorted array – look for an item

linear_search(array arr, size n, value v)

{

 for (i=0 to n-1)

 {

 if (arr[i] == v)

 return i;

 }

 return -1;

}

Best case?

v is found in the first slot (a[0]) – **O(1)**

Average case?

O(n) (assuming v can appear at any location in the array with an equal probability)

Worst case?

not found – **O(n)**

All Pairs

- Find all ordered pairs

all_pairs(array arr, size n)

```
{  
  for (i=0 to n-1)  
    for (j=0 to n-1)  
      if (i ≠ j)  
        print “{arr[i] arr[j]}”;  
}
```

Best case?

Average case?

Worst case?

All Pairs

- Find all ordered pairs

all_pairs(array arr, size n)

{

for (i=0 to n-1)

for (j=0 to n-1)

if (i ≠ j)

print “{arr[i] arr[j]}”;

}

Best case?

Average case?

Worst case?

All $O(n^2)$

Unit Operations ($O(1)$ operations)

- Addition, Subtraction, Multiplication, Division
- Comparison, Assignment
- Input, Output of a small value (e.g. short string, an integer, etc.)

- If $O(1)$ operations are repeatedly done in a loop for n times, then that loop is $O(n)$.
- If this loop is within a loop that repeats n times, then this outer loop takes $O(n^2)$.

Big-O Arithmetic

- More generally:
- If things happen sequentially, we add Big-Os.
 - $O(1)$ operation followed by $O(1) = O(1) + O(1) = O(1)$
- If one thing happens within another, then we multiply Big-Os.
 - $O(1)$ operation within a $O(n)$ loop = $O(1) \times O(n) = O(n)$
- $O(f(n)) + O(g(n)) = O(\max(f(n), g(n)))$
- $O(f(n)) \times O(g(n)) = O(f(n) \times g(n))$

Linear Search

- Unsorted array – look for an item

linear_search(array arr, size n, value v)

{

for (i=0 to n-1) $O(n)$

{

if (arr[i] == v)

return i; $O(1)$

}

return -1;

}

Best case?

v is found in the first slot (a[0]) – $O(1)$

Average case?

$O(n)$ (assuming v can appear at any location in the array with an equal probability)

Worst case?

not found – $O(n)$

Usually, assessing complexity involves counting nested loops – only one loop means it's likely to be $O(n)$

All Pairs

- Find all ordered pairs

all_pairs(array arr, size n)

{

for (i=0 to n-1) $O(n)$

for (j=0 to n-1) $O(n)$

if (i \neq j)

print "{arr[i] arr[j]}";

$O(1)$

}

Best case?

Average case?

Worst case?

All $O(n^2)$

Order of Complexity

Big O	Name	n = 128
$O(1)$	constant	1
$O(\log n)$	logarithmic	7
$O(n)$	linear	128
$O(n \log n)$	"n log n"	896
$O(n^2)$	quadratic	16192
$O(n^k), k \geq 1$	polynomial	
$O(2^n)$	exponential	10^{40}
$O(n!)$	factorial	10^{214}

Binary Search

- Find an item v in a **sorted** array

binary_search(array arr, value v , start index s , end index e)

{

 if ($s > e$)

 return -1

Best case?

 find the middle point $i = (s + e) / 2$

 if ($arr[i] == v$)

 return i

Average case?

 else if ($arr[i] < v$)

 return `binary_search(arr, v, i+1, e)`

Worst case?

 else

 return `binary_search(arr, v, s, i-1)`

}

Binary Search

- Find an item v in a **sorted** array

binary_search(array arr, value v , start index s , end index e)

```
{  
  if ( $s > e$ )  
    return -1  
  find the middle point  $i = (s + e) / 2$   
  if ( $arr[i] == v$ )  
    return  $i$   
  else if ( $arr[i] < v$ )  
    return binary_search(arr,  $v$ ,  $i+1$ ,  $e$ )  
  else  
    return binary_search(arr,  $v$ ,  $s$ ,  $i-1$ )  
}
```

Best case?

$O(1)$ – by now you can see the best case analysis doesn't help much

Average case?

$O(\log n)$

Worst case?

$O(\log n)$

Binary Search

- At every iteration, we divide the search space in half.
- You keep dividing the size by 2 until it becomes 1.
- It takes $\sim \log_2 n$ steps to get to 1.
- $\log_{10} n = (1 / \log_2 10) \log_2 n$
- So the base does not matter.

Why Big-O's are important

- You'll be asked about it in job interviews!!!!

Sorting Algorithms

- We now switch gears and discuss some well known sorting algorithms.

Selection Sort

4	3	1	5	2
---	---	---	---	---

1	3	4	5	2
----------	---	---	---	---

1	2	4	5	3
----------	----------	---	---	---

1	2	3	5	4
----------	----------	----------	---	---

1	2	3	4	5
----------	----------	----------	----------	---

- Find the smallest item in the unsorted portion, and place it in front.
- What is the running time (complexity) of this algorithm?

Insertion Sort

4	3	1	5	2
3	4	1	5	2
1	3	4	5	2
1	3	4	5	2
1	2	3	4	5

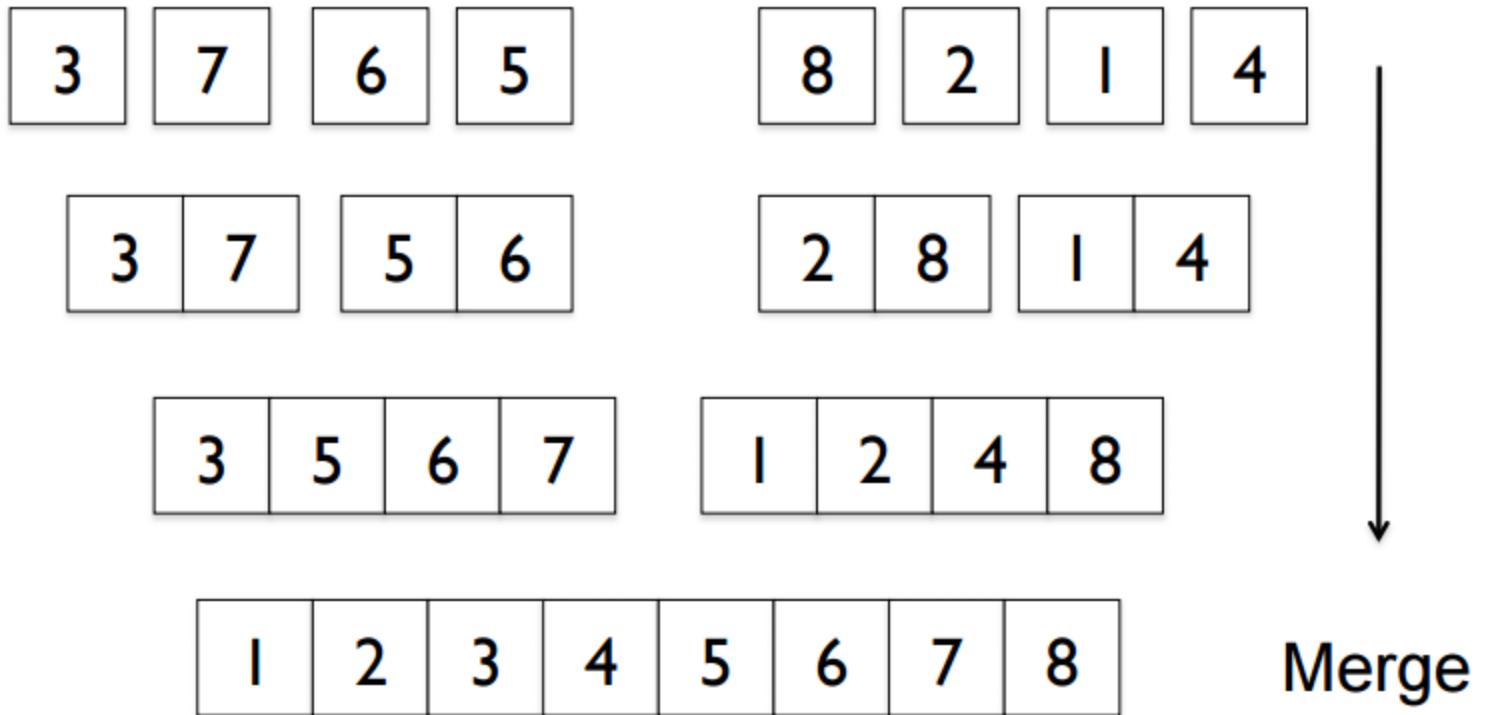
- Pick one from the unsorted part, and place it in the “right” position in the sorted part.
- Best case?
- Avg. case?
- Worst case?

Insertion Sort

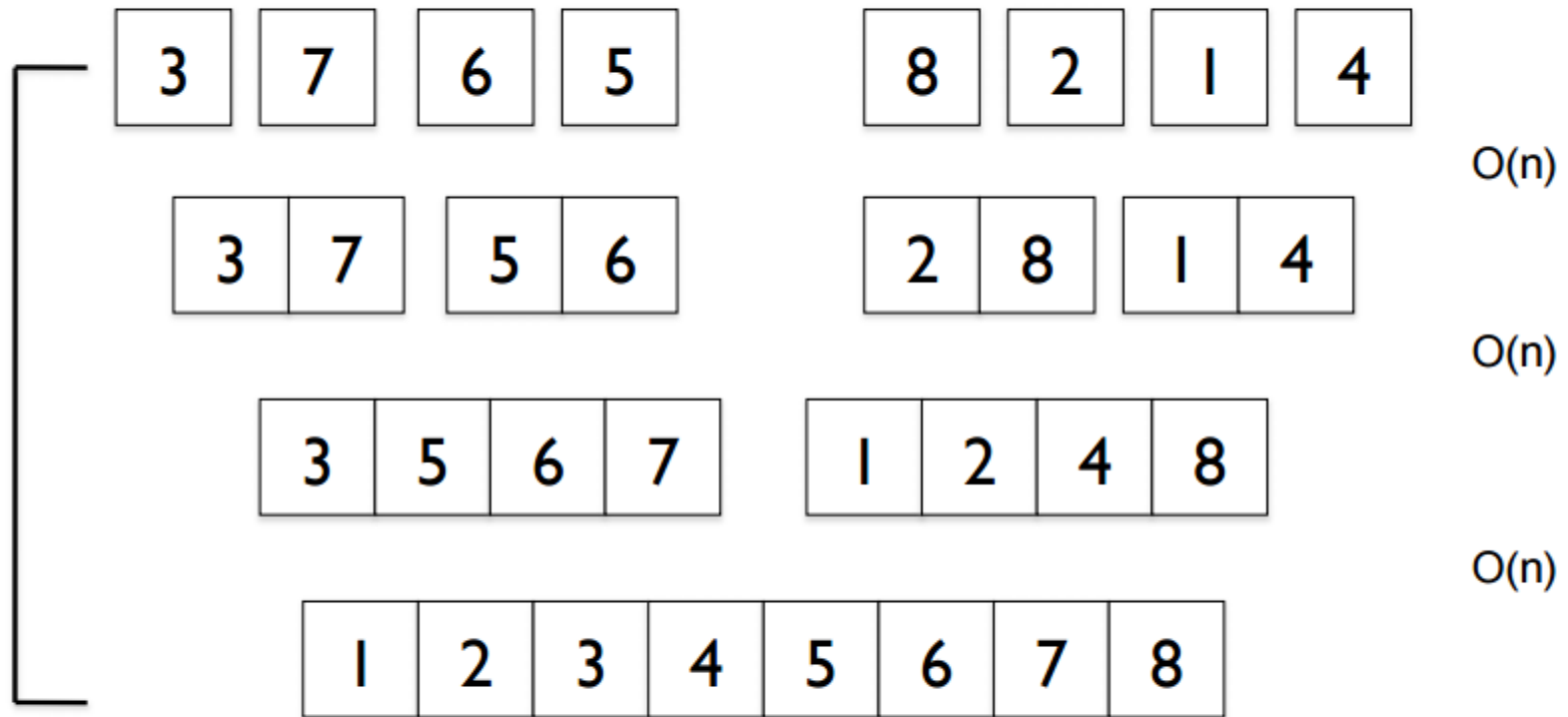
4	3	1	5	2
3	4	1	5	2
1	3	4	5	2
1	3	4	5	2
1	2	3	4	5

- Pick one from the unsorted part, and place it in the “right” position in the sorted part.
- Best case? $O(n)$
- Avg. case? $O(n^2)$
- Worst case? $O(n^2)$

Merge Sort



Merge Sort: Running Time?



$O(\log n)$

$$O(n)O(\log n) = \mathbf{O(n \log n)}$$

General Sorting: Running Time

- $O(n \log n)$ is faster than $O(n^2)$ – merge sort is more efficient than selection sort or insertion sort.
- $O(n \log n)$ is the best average complexity that a general (comparison) sorting algorithm can get (assuming you know nothing about the data set).
- With more information about the data set provided, you can sometimes sort things almost linearly.

Quick Sort



- Pick a **pivot**, and move numbers that are less than the pivot to front, and ones that are greater than the pivot to end. (Does this sound familiar?)
- On average, $O(n \log n)$
- Depending on how you pick your pivots, it can be as bad as $O(n^2)$

Permutation

```
void permutation(vector<int>& nums, int start) {
    if (start == nums.size() - 1) {
        for(int i=0; i<nums.size(); ++i)
            cout << nums[i] << ' '; cout << endl;
    }
    permutation(nums, start + 1);
    for (int i=start+1; i<nums.size(); ++i) {
        swap(nums[start], nums[i]);
        permutation(nums, start + 1);
        swap(nums[start], nums[i]);
    }
}

permutation(nums, 0); //call this function
```

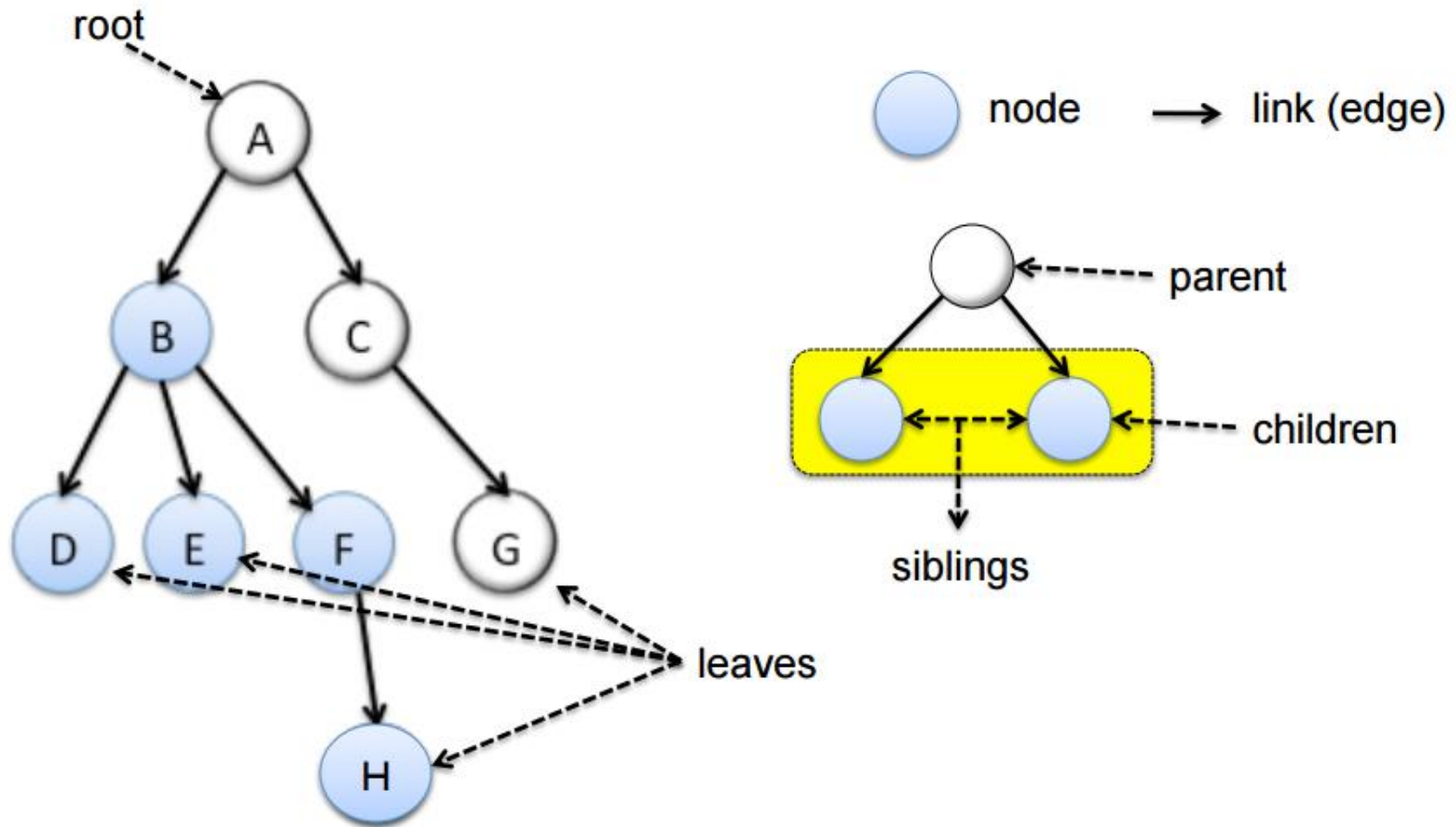
$O(n!)$

Quick Questions

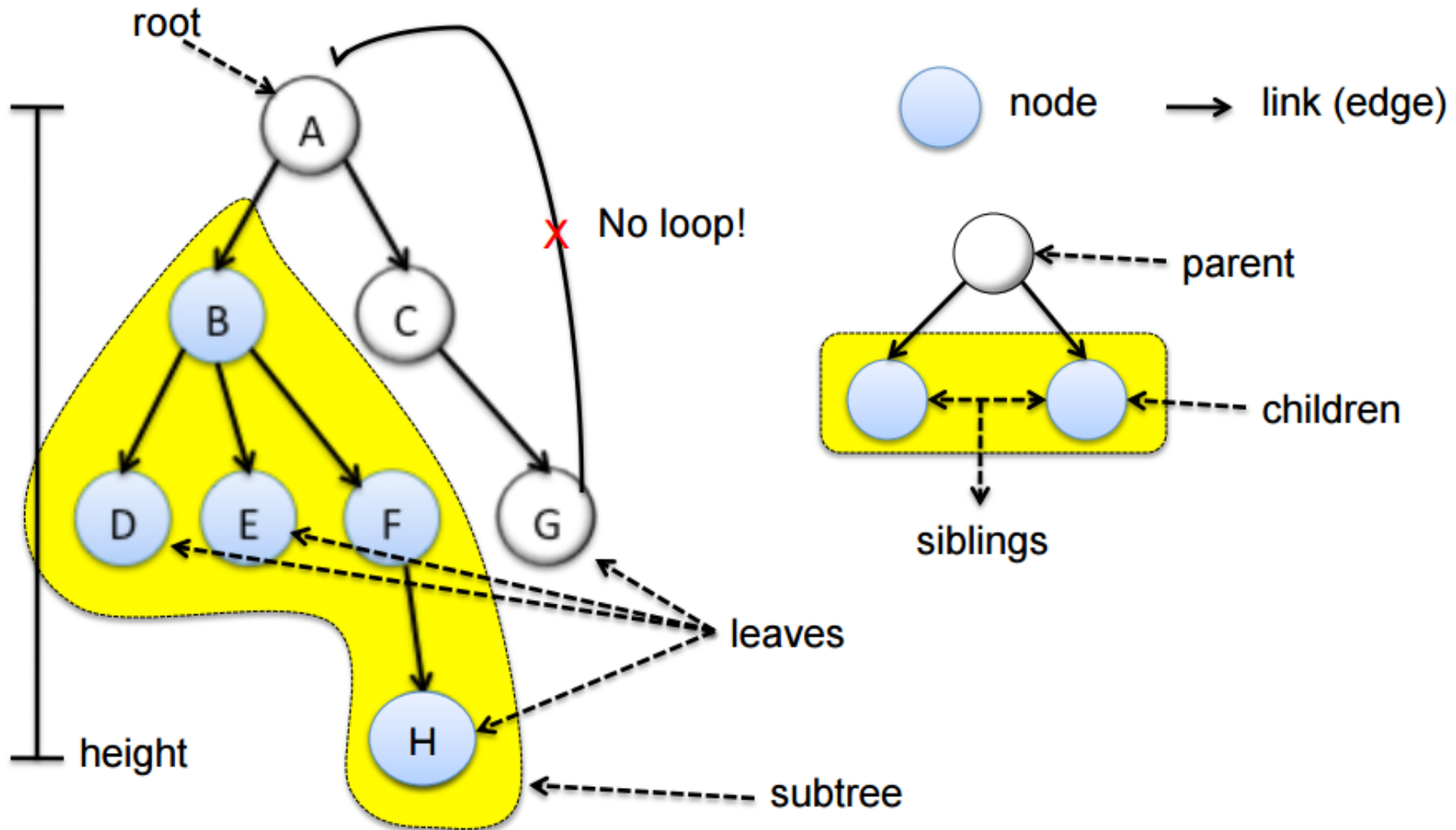
- Given an unsorted array of n items, what is the best you can do to search for an item, if you are to run this search only once?
- Given an unsorted array of n items, what is the best you can do to search for an item, if you are to run this search 100 times? (assume: $n \gg 100$)
- Given an unsorted array of n items, what is the best you can do to search for an item, if you are to run this search n times?

Tree

Tree: Definitions

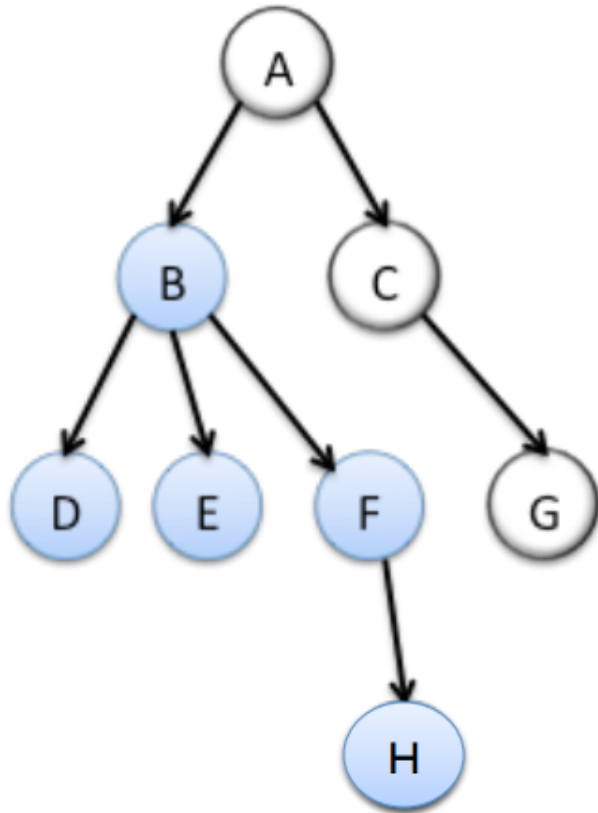


Tree: Definitions



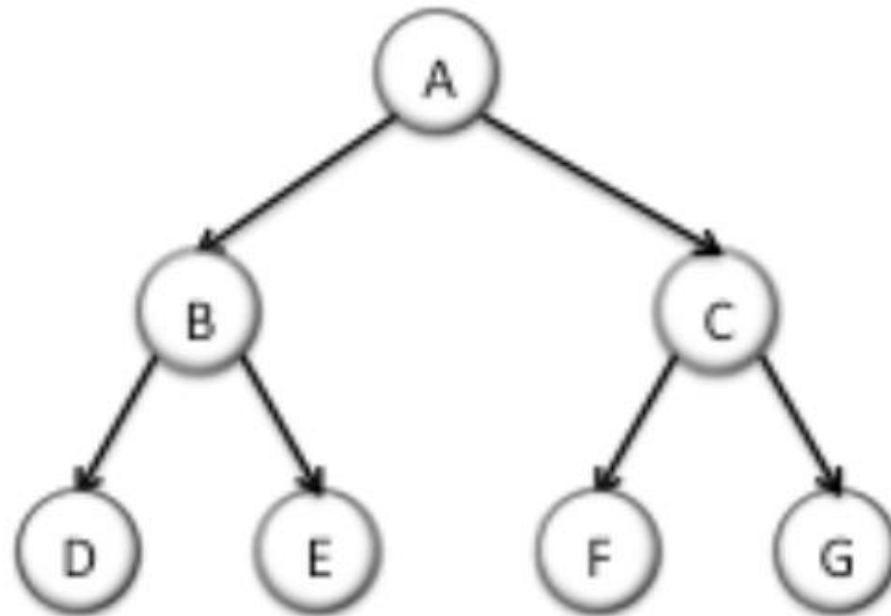
Bound on # of edges

How many edges should there be in a tree of n nodes?



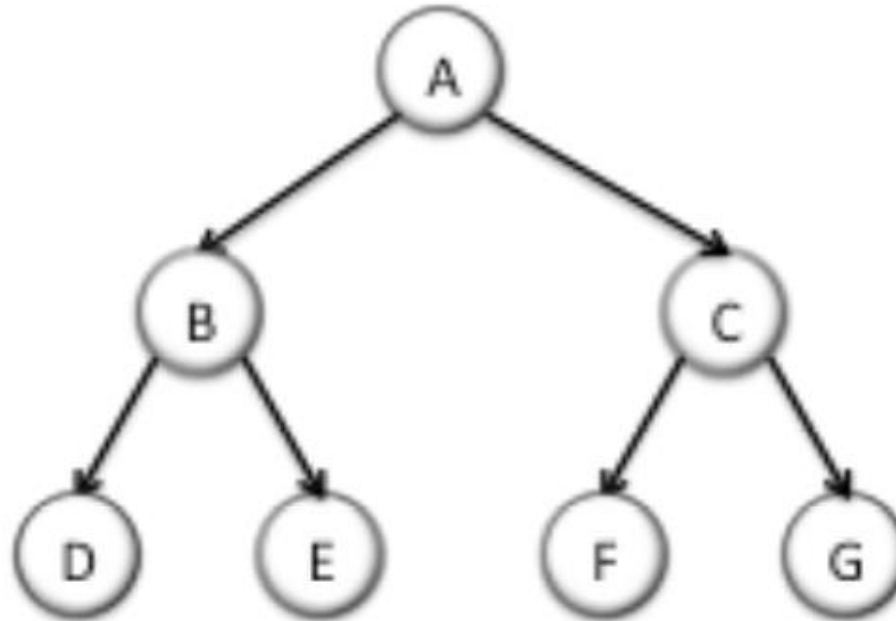
$n - 1$

Binary Trees



No node has more than 2 children (left child + right child).

Binary Trees



$$2^0 + 2^1 + \dots + 2^h = 2^{h+1} - 1$$

How many nodes can a binary tree of height ***h*** have?
(one with max. # of nodes == full binary tree)

Tree is a data structure!

- For every data structure we need to know:
 - how to insert a node,
 - how to remove a node,
 - search for a node
- and (for tree only)
 - how to traverse the tree

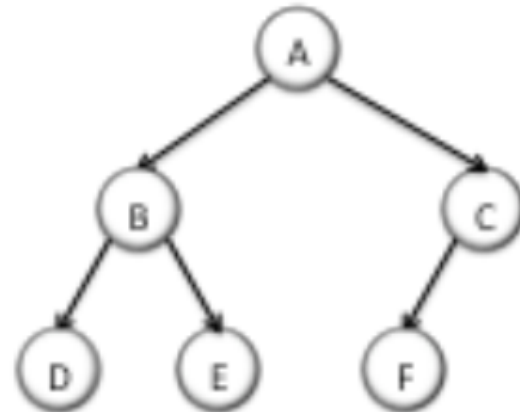
```
struct Node
{
    ItemType val;
    Node* left;
    Node* right;
};
```


Three Methods of Traversal

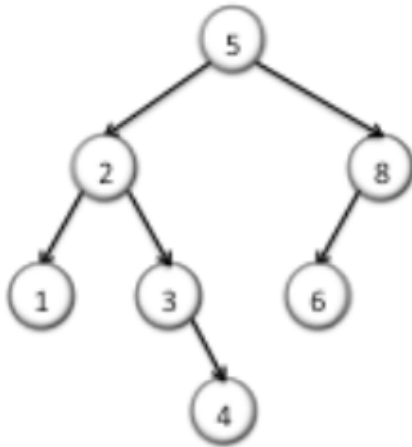
```
void preorder(const Node *node)
{
    if (node == NULL) return;
    cout << node->val << " ";
    preorder(node->left);
    preorder(node->right);
}
```

```
void inorder(const Node *node)
{
    if (node == NULL) return;
    inorder(node->left);
    cout << node->val << " ";
    inorder(node->right);
}
```

```
void postorder(const Node *node)
{
    if (node == NULL) return;
    postorder(node->left);
    postorder(node->right);
    cout << node->val << " ";
}
```



treeHeight



```
int treeHeight(const Node *node)
{
    if (node == NULL)
        return 0.;

    int leftHeight = treeHeight(node->left);
    int rightHeight = treeHeight(node->right);

    if (leftHeight > rightHeight)
        return leftHeight + 1;
    else
        return rightHeight + 1;
}
```

if (\$thirsty==TRUE)



Bugs in your software are actually special features :)