

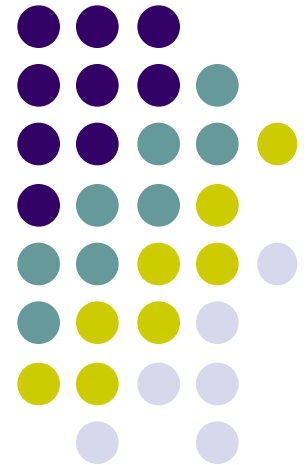
# COMP20010: Algorithms and Imperative Programming

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## Lecture 2

Data structures for binary trees

Priority queues



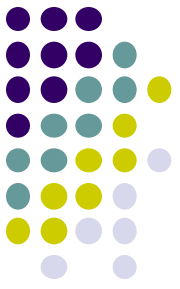
# Lecture outline



- Different data structures for representing binary trees (vector-based, linked), linked structure for general trees;
- Priority queues (PQs), sorting using PQs;

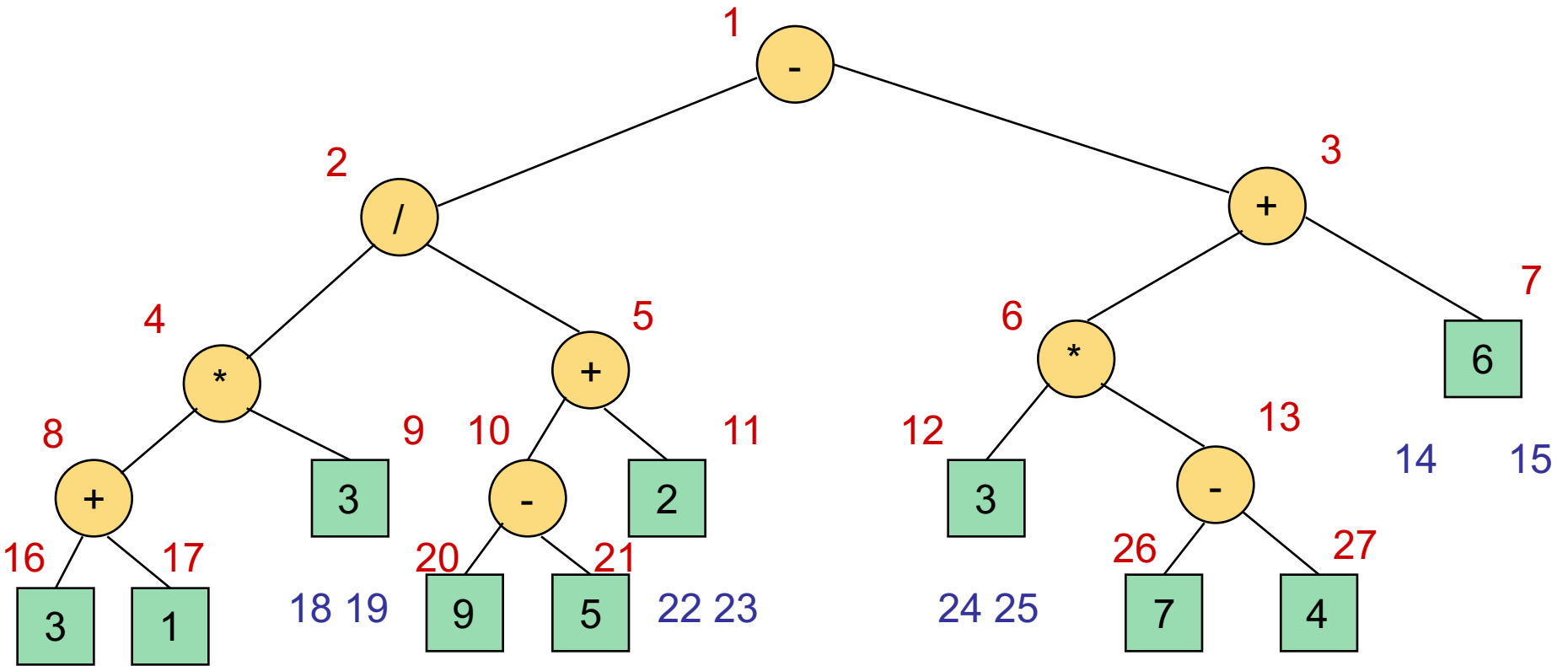
# Data structures for representing trees

## A vector-based data structure



- A vector-based structure for binary trees is based on a simple way of numbering the nodes of  $T$ .
- For every node  $v$  of  $T$  define an integer  $p(v)$ :
  - If  $v$  is the root, then  $p(v)=1$ ;
  - If  $v$  is the left child of the node  $u$ , then  $p(v)=2p(u)$ ;
  - If  $v$  is the right child of the node  $u$ , then  $p(v)=2p(u)+1$ ;
- The numbering function  $p(.)$  is known as a **level numbering** of the nodes in a binary tree  $T$ .

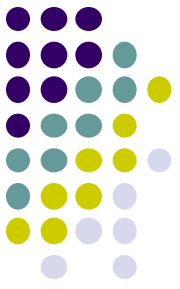
- $((((3+1)^*3)/((9-5)+2))-((3*(7-4))+6))$



## Binary tree level numbering

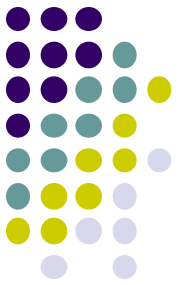
## A decorative graphic in the bottom right corner consisting of a grid of colored dots. The dots are arranged in a roughly rectangular shape, with colors including dark purple, teal, yellow, and light purple. The dots are of varying sizes and are scattered across the bottom right area of the slide.

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# Data structures for representing trees

## A vector-based data structure



Operation	Time
<i>positions(), elements()</i>	$O(n)$
swapElements(), replaceElement()	$O(1)$
root(), parent(), children()	$O(1)$
leftChild(), rightChild(), sibling()	$O(1)$
isInternal(), isExternal(), isRoot()	$O(1)$

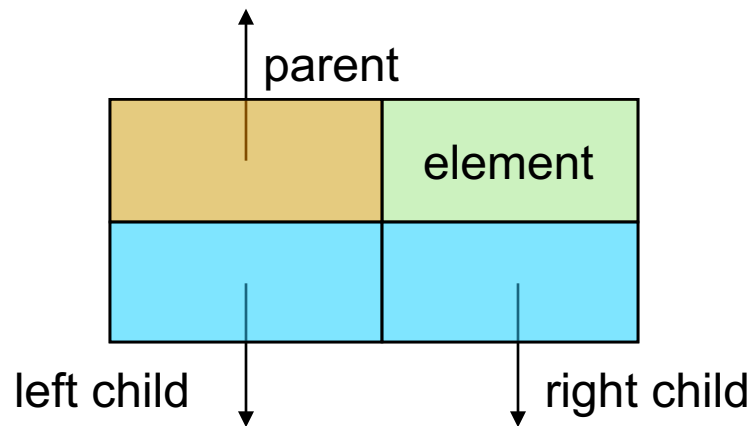
Running times of the methods when a binary tree  $T$  is implemented as a vector

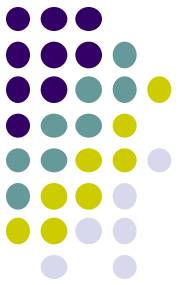
# Data structures for representing trees

## A linked data structure



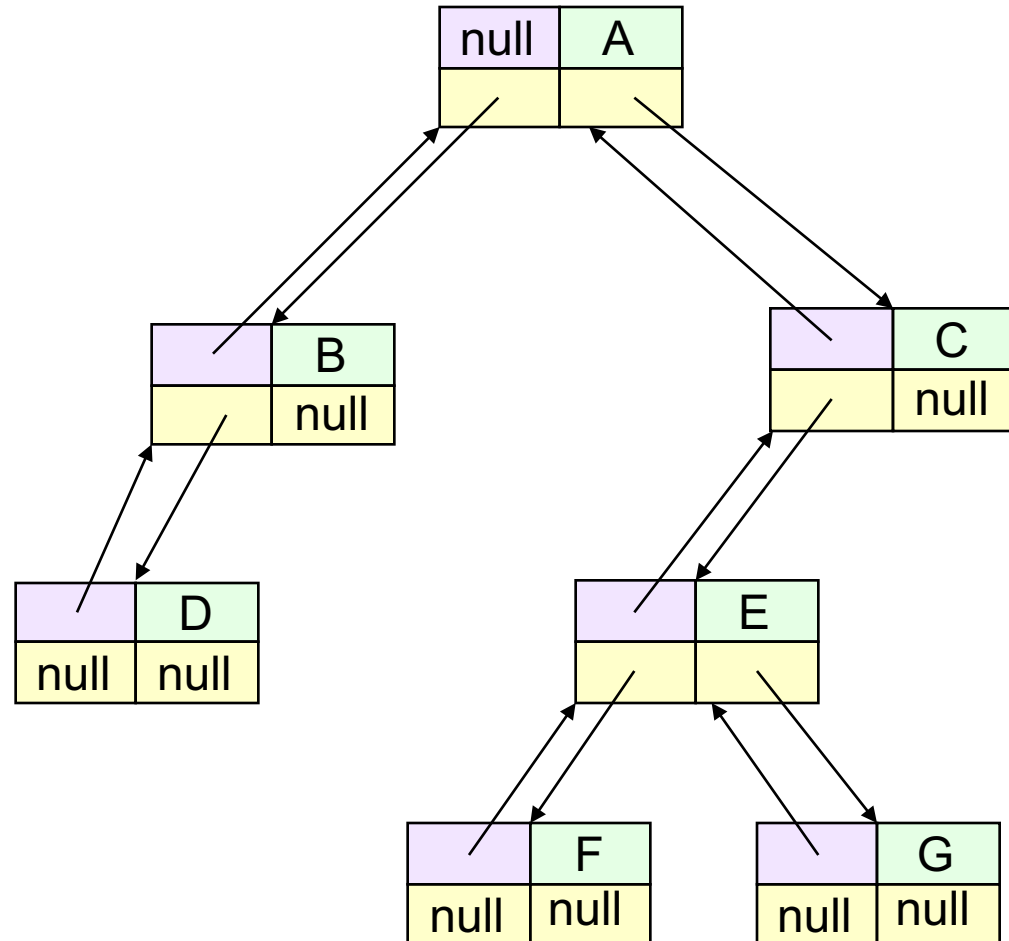
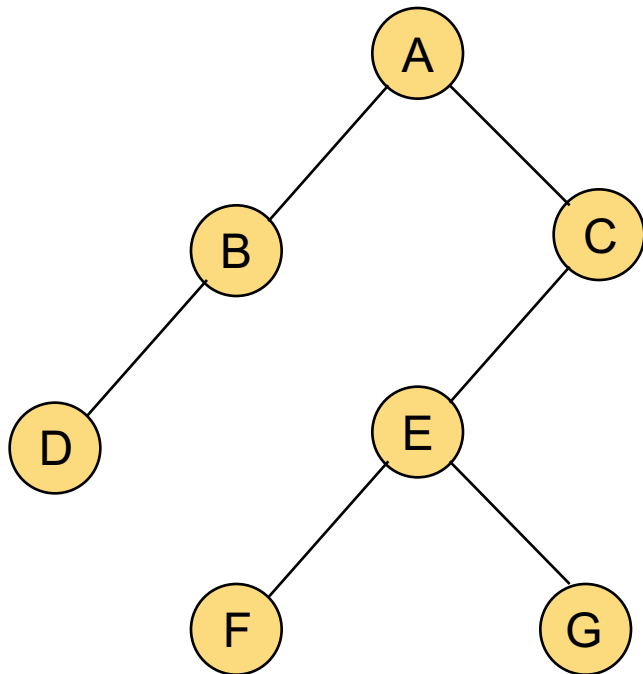
- The vector implementation of a binary tree is fast and simple, but it may be space inefficient when the tree height is large (**why?**);
- A natural way of representing a binary tree is to use a **linked structure**.
- Each node of  $T$  is represented by an object that references to the element  $v$  and the positions associated with its parent and children.





# Data structures for representing trees

## A linked data structure for binary trees



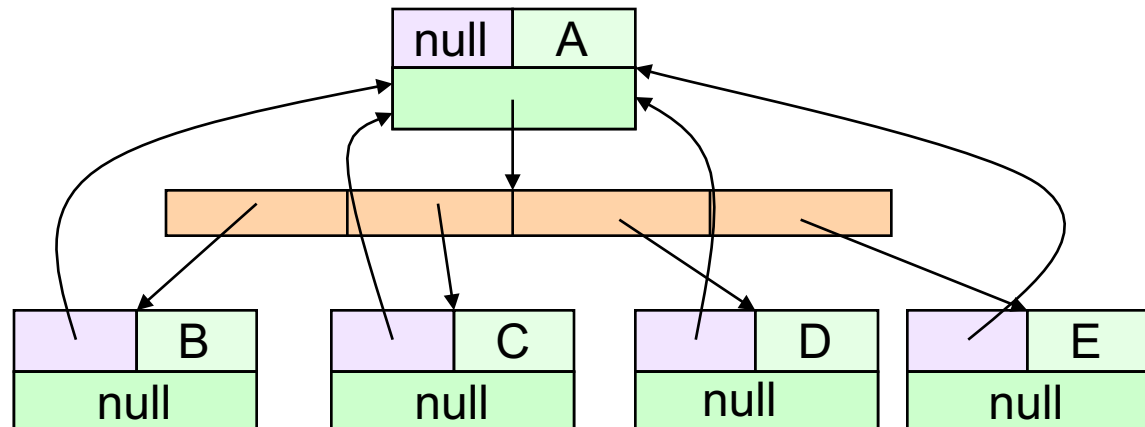
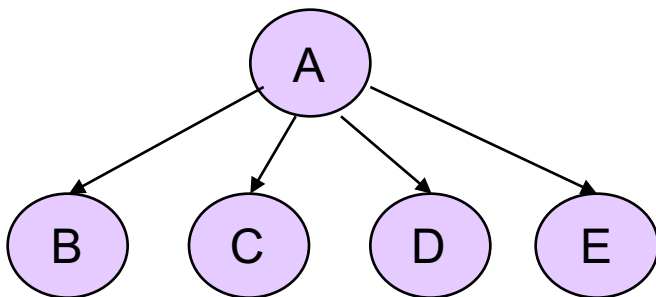


# Data structures for representing trees

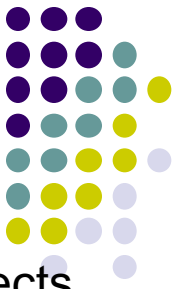
## A linked data structure for general trees



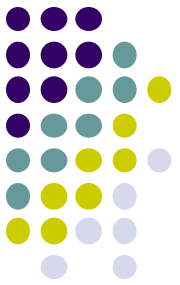
- In order to extend the previous data structure to the case of general trees;
- In order to register a potentially large number of children of a node, we need to use a container (a list or a vector) to store the children, instead of using instance variables;



# Keys and the total order relation

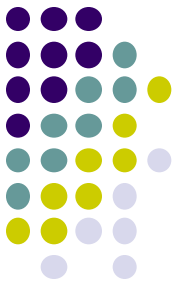


- In various applications it is frequently required to compare and rank objects according to some parameters or properties, called **keys** that are assigned to each object in a collection.
- A **key** is an object assigned to an element as a specific attribute that can be used to identify, rank or weight that element.
- A rule for comparing keys needs to be robustly defined (not contradicting).
- We need to define a **total order** relation, denoted by  $\leq$  with the following properties:
  - Reflexive property:  $k \leq k$  ;
  - Antisymmetric property: if  $k_1 \leq k_2$  and  $k_2 \leq k_1$  , then  $k_1 = k_2$  ;
  - Transitive property: if  $k_1 \leq k_2$  and  $k_2 \leq k_3$ , then  $k_1 \leq k_3$  ;
- The comparison rule that satisfies the above properties defines a linear ordering relationship among a set of keys.
- In a finite collection of elements with a defined total order relation we can define the **smallest key**  $k_{\min}$  as the key for which  $k_{\min} \leq k$  for any other key  $k$  in the collection.



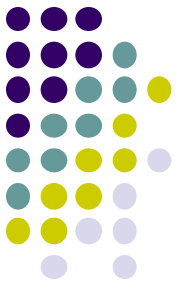
# Priority queues

- A **priority queue**  $P$  is a container of elements with keys associated to them at the time of insertion.
- Two fundamental methods of a priority queue  $P$  are:
  - `insertItem( $k, e$ )` – inserts an element  $e$  with a key  $k$  into  $P$ ;
  - `removeMin()` – returns and removes from  $P$  an element with a smallest key;
- The priority queue ADT is simpler than that of the sequence ADT. This simplicity originates from the fact that the elements in a PQ are inserted and removed based on their keys, while the elements are inserted and removed from a sequence based on their positions and ranks.



# Priority queues

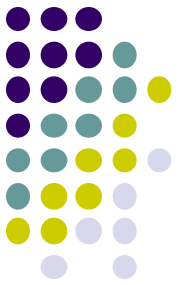
- A **comparator** is an object that compares two keys. It is associated with a priority queue at the time of construction.
- A comparator method provides the following objects, each taking two keys and comparing them:
  - `isLess(  $k_1, k_2$  )` – true if  $k_1 < k_2$ ;
  - `isLessOrEqualTo(  $k_1, k_2$  )` – true if  $k_1 \leq k_2$  ;
  - `isEqualTo(  $k_1, k_2$  )` – true if  $k_1 = k_2$ ;
  - `isGreater(  $k_1, k_2$  )` – true if  $k_1 > k_2$ ;
  - `isGreaterOrEqualTo(  $k_1, k_2$  )` – true if  $k_1 \geq k_2$  ;
  - `isComparable( $k$ )` – true if  $k$  can be compared;



# PQ-Sort

- Sorting problem is to transform a collection  $C$  of  $n$  elements that can be compared and ordered according to a total order relation.
- Algorithm outline:
  - Given a collection  $C$  of  $n$  elements;
  - In the first phase we put the elements of  $C$  into an initially empty priority queue  $P$  by applying  $n$  `insertItem(c)` operations;
  - In the second phase we extract the elements from  $P$  in non-decreasing order by applying  $n$  `removeMin` operations, and putting them back into  $C$ ;

# PQ-Sort



- **Algorithm** PQ-Sort( $C, P$ );
  - **Input:** A sequence  $C[1:n]$  and a priority queue  $P$  that compares keys (elements of  $C$ ) using a total order relation;
  - **Output:** A sequence  $C[1:n]$  sorted by the total order relation;  
    **while**  $C$  is not empty **do**
    - $e \leftarrow C.\text{removeFirst}()$ ;      {remove an element  $e$  from  $C$ }
    - $P.\text{insertItem}(e, e)$ ;      {the key is the element itself}
  - while**  $P$  is not empty **do**
    - $e \leftarrow P.\text{removeMin}()$       {remove the smallest element from  $P$ }
    - $C.\text{insertLast}(e)$       {add the element at the end of  $C$ }
- This algorithm does not specify how the priority queue  $P$  is implemented. Depending on that, several popular schemes can be obtained, such as selection-sort, insertion-sort and heap-sort.

# Priority queue implemented with an unordered sequence and Selection-Sort



- Assume that the elements of  $P$  and their keys are stored in a sequence  $S$ , which is implemented as either an array or a doubly-linked list.
- The elements of  $S$  are pairs  $(k, e)$ , where  $e$  is an element of  $P$  and  $k$  is the key.
- New element is added to  $S$  by appending it at the end (executing  $\text{insertLast}(k, e)$ ), which means that  $S$  will be unsorted.
- $\text{insertLast}()$  will take  $O(1)$  time, but finding the element in  $S$  with a minimal key will take  $O(n)$ .

# Priority queue implemented with an unordered sequence and Selection-Sort



- The first phase of the algorithm takes  $O(n)$  time, assuming that each insertion takes  $O(1)$  time.
- Assuming that two keys can be compared in  $O(1)$  time, the execution time of each removeMin operation is proportional to the number of elements currently in  $P$ .
- The main bottleneck of this algorithm is the repeated selection of a minimal element from an unsorted sequence in Phase 2. This is why the algorithm is referred to as **selection-sort**.
- The bottleneck of the selection-sort algorithm is the second phase. Total time needed for the second phase is

$$O(n) + O(n-1) + \cdots + O(1) = \sum_{i=1}^n O(i) = O(n^2)$$



# Priority queue implemented with a sorted sequence and Insertion-Sort



- An alternative approach is to sort the elements in the sequence  $S$  by their key values.
- In this case the method `removeMin` actually removes the first element from  $S$ , which takes  $O(1)$  time.
- However, the method `insertItem` requires to scan through the sequence  $S$  for an appropriate position to insert the new element and its key. This takes  $O(n)$  time.
- The main bottleneck of this algorithm is the repeated insertion of elements into a sorted priority queue. This is why the algorithm is referred to as **insertion-sort**.
- The total execution time of insertion-sort is dominated by the first phase and is  $O(n^2)$ .