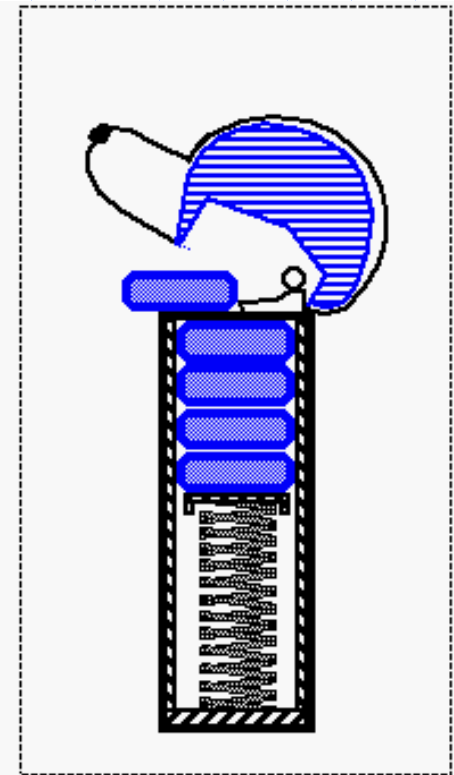


# Elementary Data Structures

Stacks, Queues, & Lists

Amortized analysis

Trees



# The Stack ADT (§2.1.1)



- ◆ The **Stack** ADT stores arbitrary objects
- ◆ Insertions and deletions follow the last-in first-out scheme
- ◆ Think of a spring-loaded plate dispenser
- ◆ Main stack operations:
  - **push**(object): inserts an element
  - object **pop**(): removes and returns the last inserted element
- ◆ Auxiliary stack operations:
  - object **top**(): returns the last inserted element without removing it
  - integer **size**(): returns the number of elements stored
  - boolean **isEmpty**(): indicates whether no elements are stored

# Applications of Stacks



## ◆ Direct applications

- Page-visited history in a Web browser
- Undo sequence in a text editor
- Chain of method calls in the Java Virtual Machine or C++ runtime environment

## ◆ Indirect applications

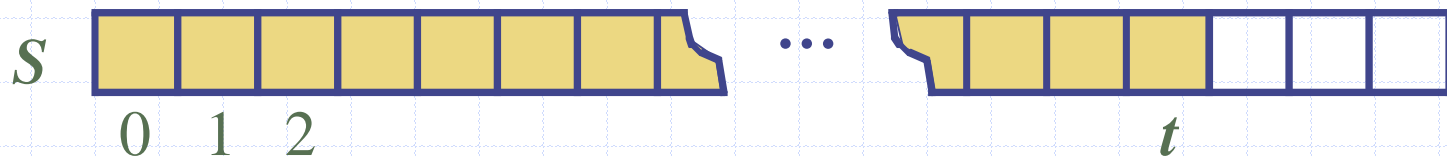
- Auxiliary data structure for algorithms
- Component of other data structures

# Array-based Stack (§2.1.1)

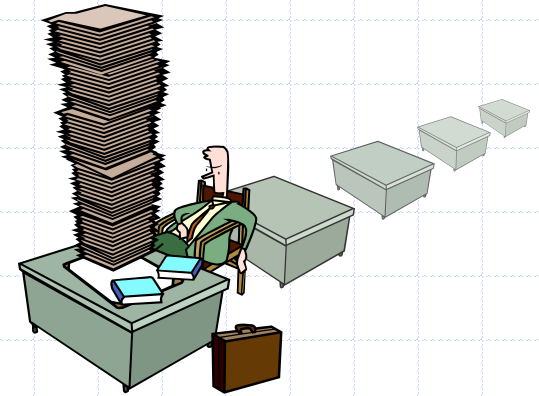
- ◆ A simple way of implementing the Stack ADT uses an array
- ◆ We add elements from left to right
- ◆ A variable  $t$  keeps track of the index of the top element (size is  $t+1$ )

```
Algorithm pop():  
  if isEmpty() then  
    throw EmptyStackException  
  else  
     $t \leftarrow t - 1$   
    return  $S[t + 1]$ 
```

```
Algorithm push(o)  
  if  $t = S.length - 1$  then  
    throw FullStackException  
  else  
     $t \leftarrow t + 1$   
     $S[t] \leftarrow o$ 
```



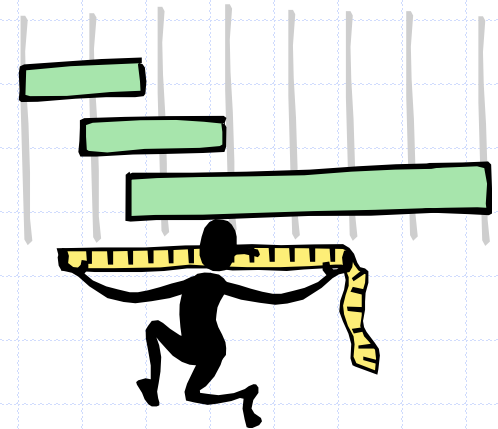
# Growable Array-based Stack (§1.5)



- ◆ In a push operation, when the array is full, instead of throwing an exception, we can replace the array with a larger one
- ◆ How large should the new array be?
  - incremental strategy: increase the size by a constant  $c$
  - doubling strategy: double the size

```
Algorithm push(o)  
  if  $t = S.length - 1$  then  
     $A \leftarrow$  new array of  
      size ...  
    for  $i \leftarrow 0$  to  $t$  do  
       $A[i] \leftarrow S[i]$   
       $S \leftarrow A$   
   $t \leftarrow t + 1$   
   $S[t] \leftarrow o$ 
```

# Comparison of the Strategies



- ◆ We compare the incremental strategy and the doubling strategy by analyzing the total time  $T(n)$  needed to perform a series of  $n$  push operations
- ◆ We assume that we start with an empty stack represented by an array of size 1
- ◆ We call **amortized time** of a push operation the average time taken by a push over the series of operations, i.e.,  $T(n)/n$

# Analysis of the Incremental Strategy

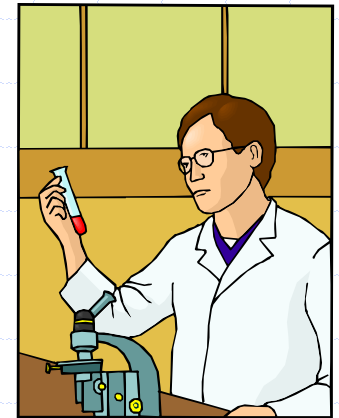


- ◆ We replace the array  $k = n/c$  times
- ◆ The total time  $T(n)$  of a series of  $n$  push operations is proportional to

$$\begin{aligned}n + c + 2c + 3c + 4c + \dots + kc &= \\n + c(1 + 2 + 3 + \dots + k) &= \\n + ck(k + 1)/2 &\end{aligned}$$

- ◆ Since  $c$  is a constant,  $T(n)$  is  $O(n + k^2)$ , i.e.,  $O(n^2)$
- ◆ The amortized time of a push operation is  $O(n)$

# Direct Analysis of the Doubling Strategy

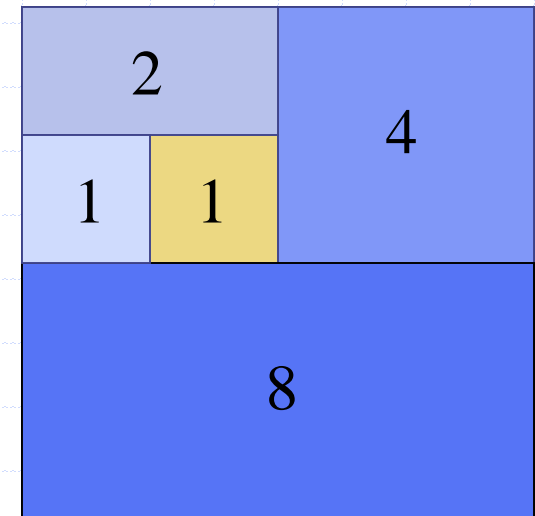


- ◆ We replace the array  $k = \log_2 n$  times
- ◆ The total time  $T(n)$  of a series of  $n$  push operations is proportional to

$$n + 1 + 2 + 4 + 8 + \dots + 2^k = \\ n + 2^{k+1} - 1 = 2n - 1$$

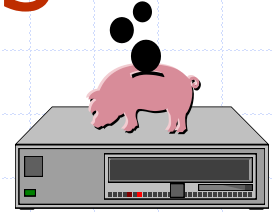
- ◆  $T(n)$  is  $O(n)$
- ◆ The amortized time of a push operation is  $O(1)$

geometric series





# Accounting Method Analysis of the Doubling Strategy

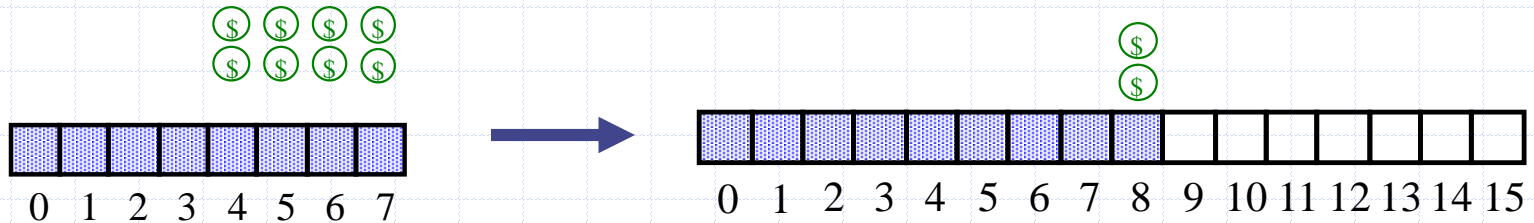


- ◆ The **accounting method** determines the amortized running time with a system of credits and debits
- ◆ We view a computer as a **coin-operated device** requiring 1 cyber-dollar for a constant amount of computing.
  - We set up a scheme for charging operations. This is known as an **amortization scheme**.
  - The scheme must give us always enough money to pay for the actual cost of the operation.
  - The total cost of the series of operations is no more than the total amount charged.
- ◆  $(\text{amortized time}) \leq (\text{total \$ charged}) / (\# \text{ operations})$

# Amortization Scheme for the Doubling Strategy



- ◆ Consider again the  $k$  phases, where each phase consisting of twice as many pushes as the one before.
- ◆ At the end of a phase we must have saved enough to pay for the array-growing push of the next phase.
- ◆ At the end of phase  $i$  we want to have saved  $i$  cyber-dollars, to pay for the array growth for the beginning of the next phase.



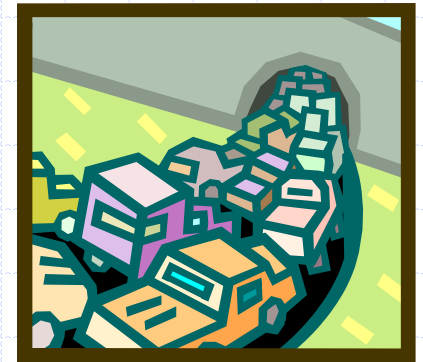
- We charge **\$3** for a push. The **\$2** saved for a regular push are “stored” in the second half of the array. Thus, we will have  $2(i/2) = i$  cyber-dollars saved at the end of phase  $i$ .
- Therefore, each push runs in  $O(1)$  amortized time;  $n$  pushes run in  $O(n)$  time.

# The Queue ADT (§2.1.2)



- ◆ The **Queue** ADT stores arbitrary objects
- ◆ Insertions and deletions follow the first-in first-out scheme
- ◆ Insertions are at the rear of the queue and removals are at the front of the queue
- ◆ Main queue operations:
  - **enqueue**(object): inserts an element at the end of the queue
  - object **dequeue**(): removes and returns the element at the front of the queue
- ◆ Auxiliary queue operations:
  - object **front**(): returns the element at the front without removing it
  - integer **size**(): returns the number of elements stored
  - boolean **isEmpty**(): indicates whether no elements are stored
- ◆ Exceptions
  - Attempting the execution of dequeue or front on an empty queue throws an **EmptyQueueException**

# Applications of Queues



## ◆ Direct applications

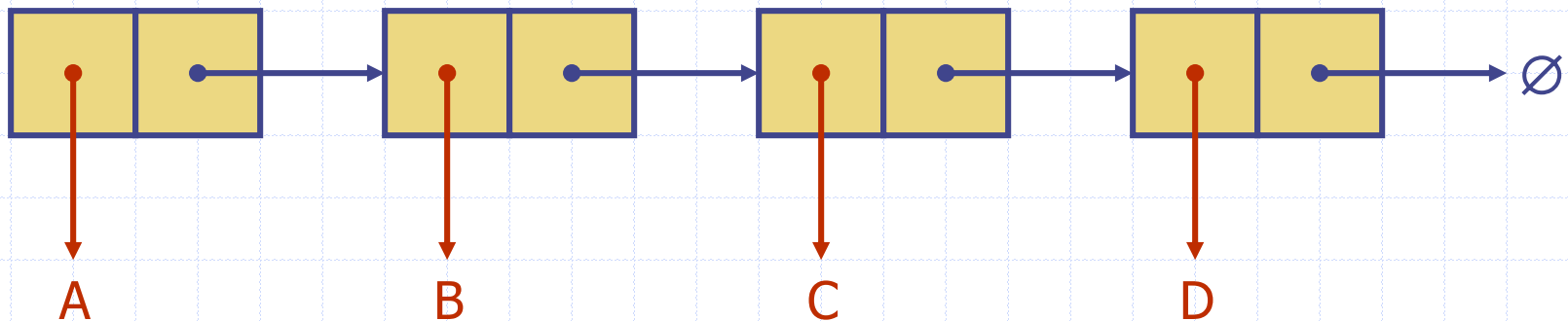
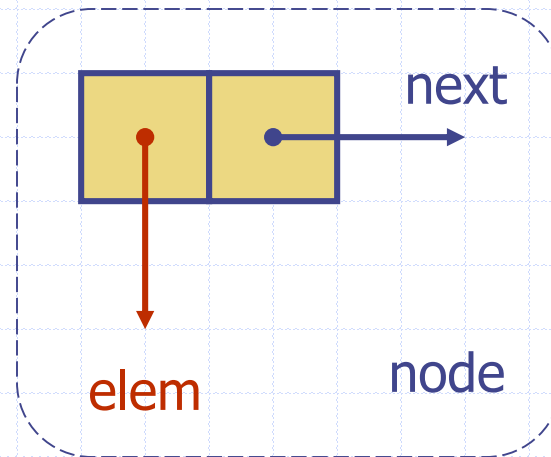
- Waiting lines
- Access to shared resources (e.g., printer)
- Multiprogramming

## ◆ Indirect applications

- Auxiliary data structure for algorithms
- Component of other data structures

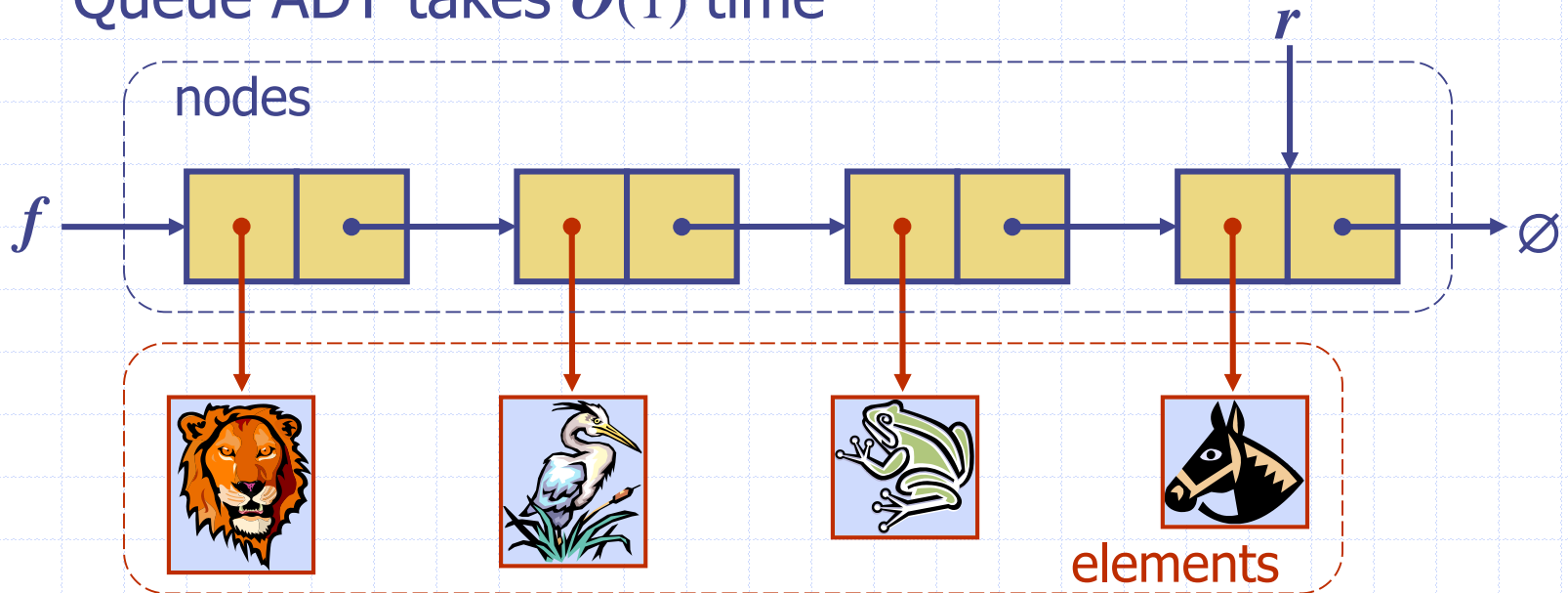
# Singly Linked List

- ◆ A singly linked list is a concrete data structure consisting of a sequence of nodes
- ◆ Each node stores
  - element
  - link to the next node

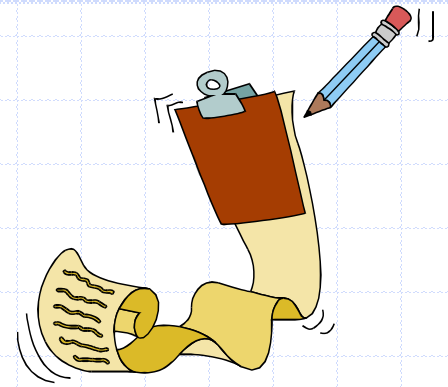


# Queue with a Singly Linked List

- ◆ We can implement a queue with a singly linked list
  - The front element is stored at the first node
  - The rear element is stored at the last node
- ◆ The space used is  $O(n)$  and each operation of the Queue ADT takes  $O(1)$  time



# List ADT (§2.2.2)



- ◆ The **List** ADT models a sequence of **positions** storing arbitrary objects
- ◆ It allows for insertion and removal in the “middle”
- ◆ Query methods:
  - **isFirst(p)**, **isLast(p)**

## Accessor methods:

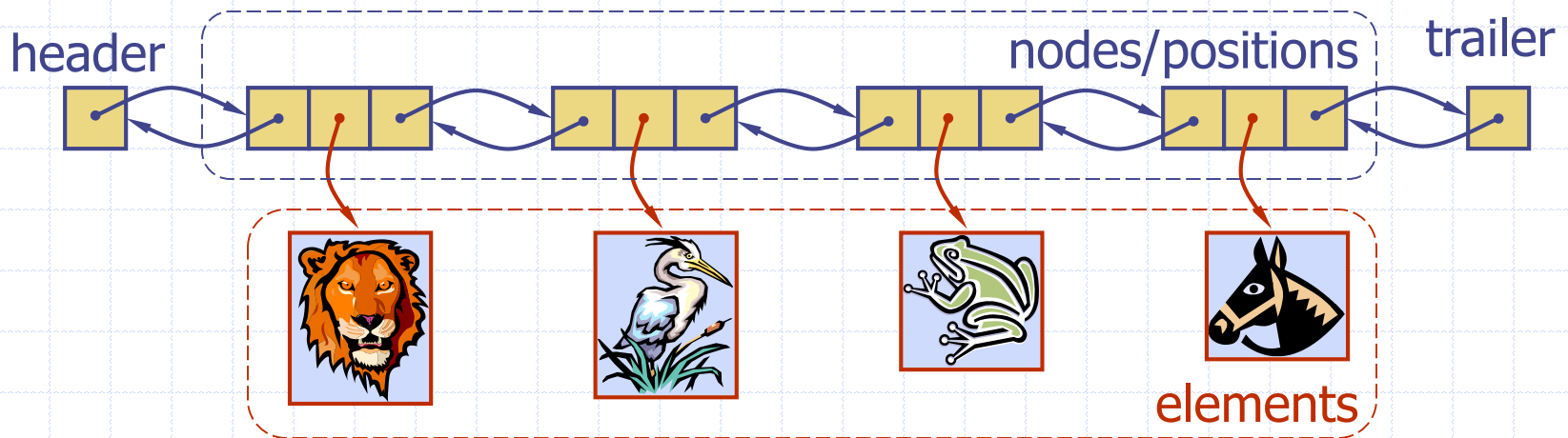
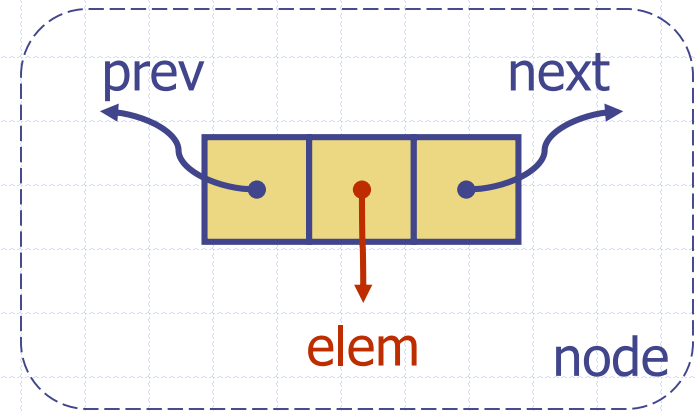
- **first()**, **last()**
- **before(p)**, **after(p)**

## ◆ Update methods:

- **replaceElement(p, o)**, **swapElements(p, q)**
- **insertBefore(p, o)**, **insertAfter(p, o)**,
- **insertFirst(o)**, **insertLast(o)**
- **remove(p)**

# Doubly Linked List

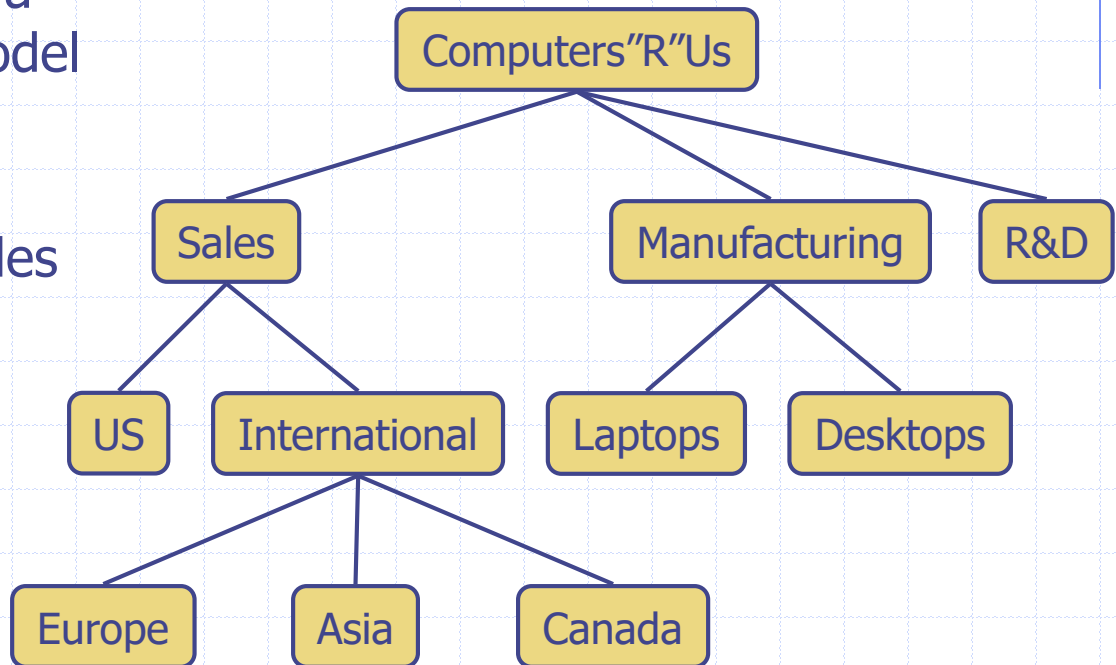
- ◆ A doubly linked list provides a natural implementation of the List ADT
- ◆ Nodes implement Position and store:
  - element
  - link to the previous node
  - link to the next node
- ◆ Special trailer and header nodes

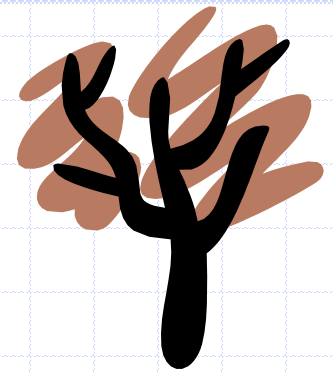




# Trees (§2.3)

- ◆ In computer science, a tree is an abstract model of a hierarchical structure
- ◆ A tree consists of nodes with a parent-child relation
- ◆ Applications:
  - Organization charts
  - File systems
  - Programming environments





# Tree ADT (§2.3.1)

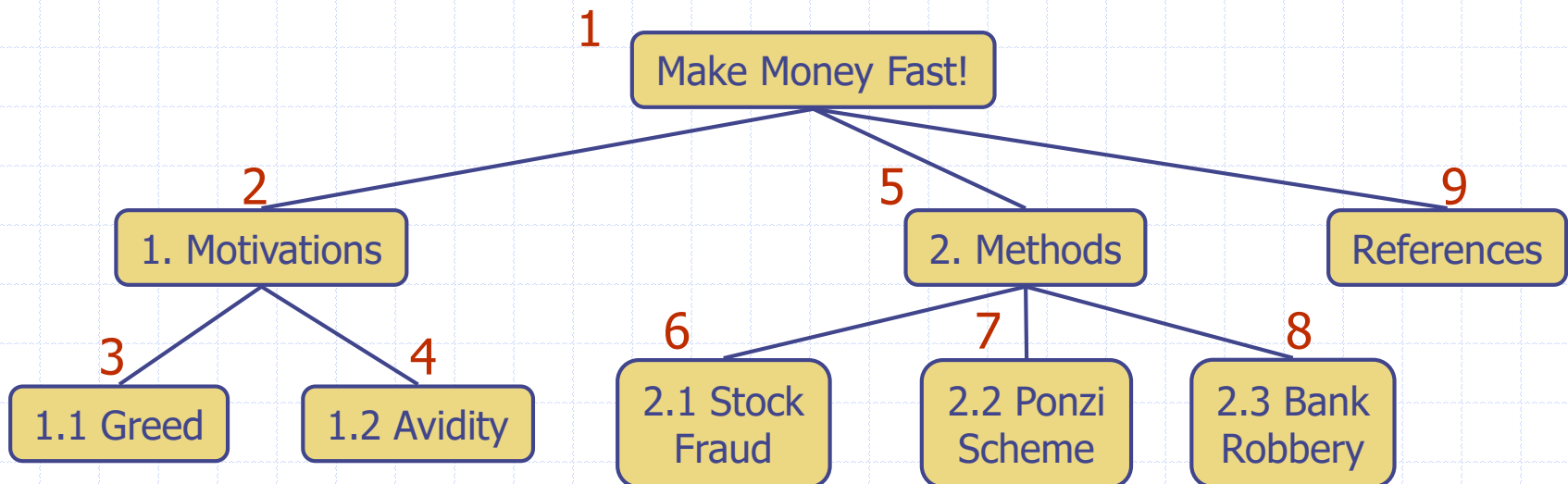
- ◆ We use positions to abstract nodes
- ◆ Generic methods:
  - integer `size()`
  - boolean `isEmpty()`
  - objectIterator `elements()`
  - positionIterator `positions()`
- ◆ Accessor methods:
  - position `root()`
  - position `parent(p)`
  - positionIterator `children(p)`
- ◆ Query methods:
  - boolean `isInternal(p)`
  - boolean `isExternal(p)`
  - boolean `isRoot(p)`
- ◆ Update methods:
  - `swapElements(p, q)`
  - object `replaceElement(p, o)`
- ◆ Additional update methods may be defined by data structures implementing the Tree ADT

# Preorder Traversal (§2.3.2)



- ◆ A traversal visits the nodes of a tree in a systematic manner
- ◆ In a preorder traversal, a node is visited before its descendants
- ◆ Application: print a structured document

```
Algorithm preOrder(v)  
  visit(v)  
  for each child w of v  
    preorder (w)
```

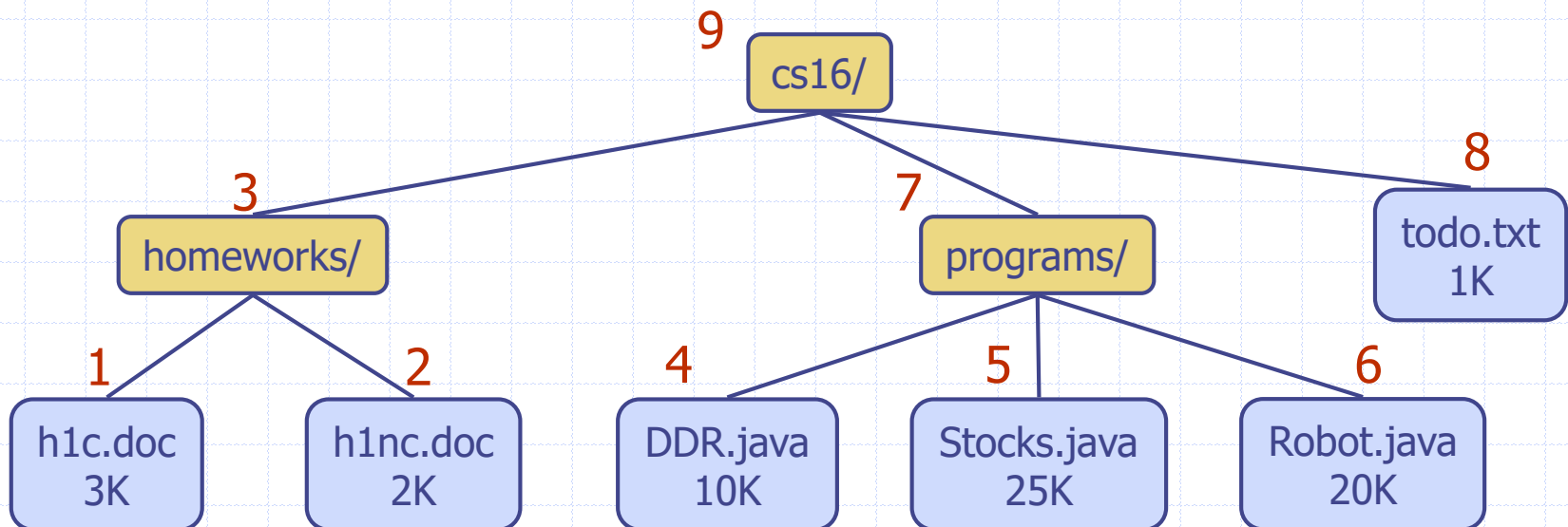


# Postorder Traversal (§2.3.2)



- ◆ In a postorder traversal, a node is visited after its descendants
- ◆ Application: compute space used by files in a directory and its subdirectories

**Algorithm** *postOrder*(*v*)  
for each child *w* of *v*  
    *postOrder* (*w*)  
*visit*(*v*)



# Amortized Analysis of Tree Traversal

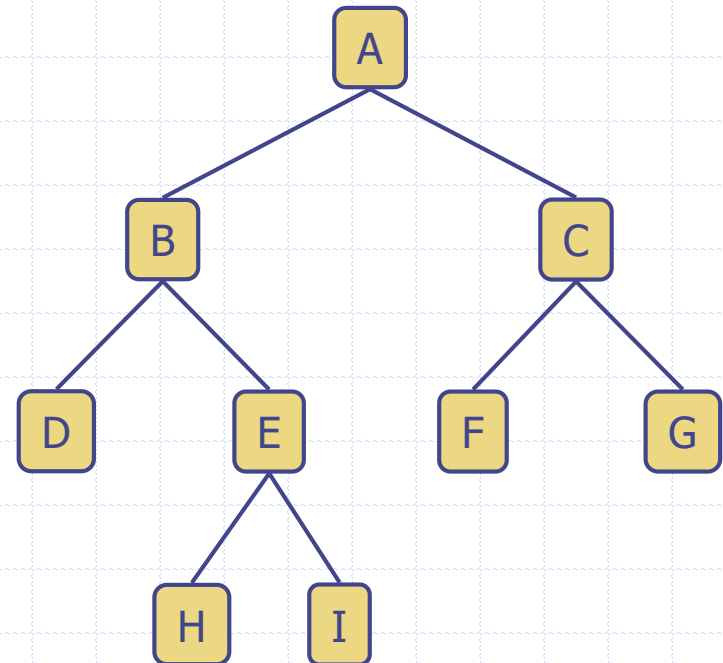


- ◆ Time taken in preorder or postorder traversal of an  $n$ -node tree is proportional to the sum, taken over each node  $v$  in the tree, of the time needed for the recursive call for  $v$ .
  - The call for  $v$  costs  $\$(c_v + 1)$ , where  $c_v$  is the number of children of  $v$
  - For the call for  $v$ , charge one cyber-dollar to  $v$  and charge one cyber-dollar to each child of  $v$ .
  - Each node (except the root) gets charged twice: once for its own call and once for its parent's call.
  - Therefore, traversal time is  **$O(n)$** .

# Binary Trees (§2.3.3)

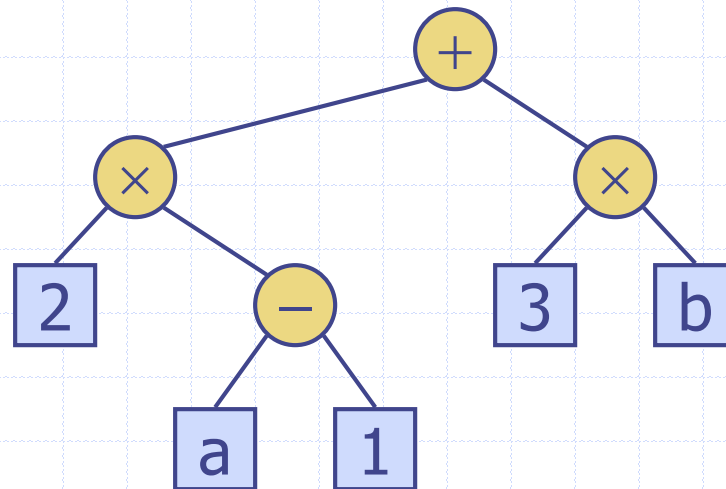
- ◆ A binary tree is a tree with the following properties:
  - Each internal node has two children
  - The children of a node are an ordered pair
- ◆ We call the children of an internal node left child and right child
- ◆ Alternative recursive definition: a binary tree is either
  - a tree consisting of a single node, or
  - a tree whose root has an ordered pair of children, each of which is a binary tree

- ◆ Applications:
  - arithmetic expressions
  - decision processes
  - searching



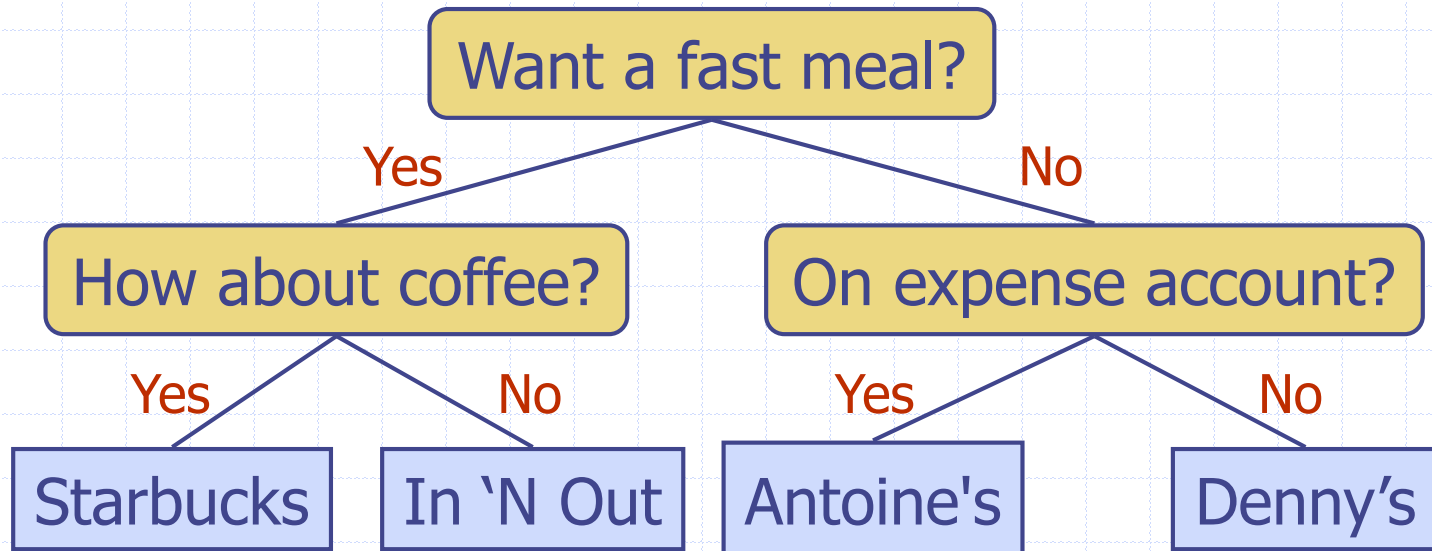
# Arithmetic Expression Tree

- ◆ Binary tree associated with an arithmetic expression
  - internal nodes: operators
  - external nodes: operands
- ◆ Example: arithmetic expression tree for the expression  $(2 \times (a - 1) + (3 \times b))$



# Decision Tree

- ◆ Binary tree associated with a decision process
  - internal nodes: questions with yes/no answer
  - external nodes: decisions
- ◆ Example: dining decision





# Properties of Binary Trees

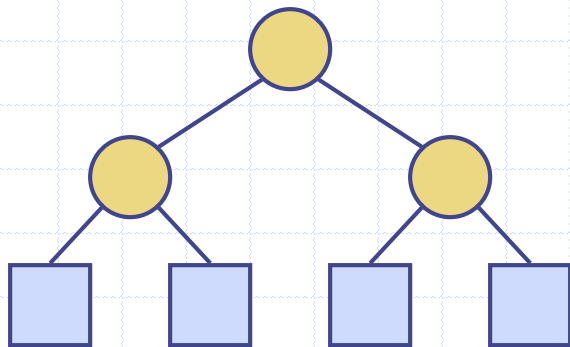
## ◆ Notation

$n$  number of nodes

$e$  number of external nodes

$i$  number of internal nodes

$h$  height



## ◆ Properties:

■  $e = i + 1$

■  $n = 2e - 1$

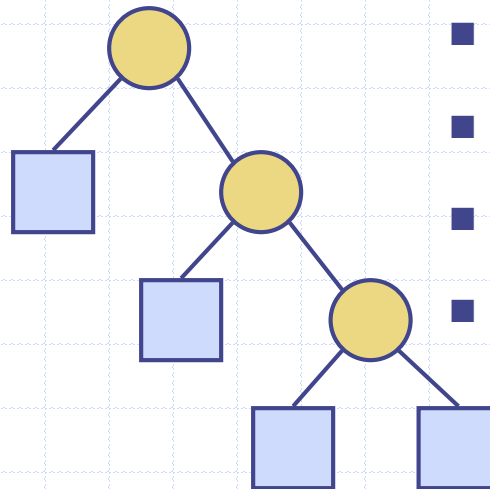
■  $h \leq i$

■  $h \leq (n - 1)/2$

■  $e \leq 2^h$

■  $h \geq \log_2 e$

■  $h \geq \log_2 (n + 1) - 1$



# Inorder Traversal

- ◆ In an inorder traversal a node is visited after its left subtree and before its right subtree
- ◆ Application: draw a binary tree
  - $x(v)$  = inorder rank of  $v$
  - $y(v)$  = depth of  $v$

**Algorithm** *inOrder*( $v$ )

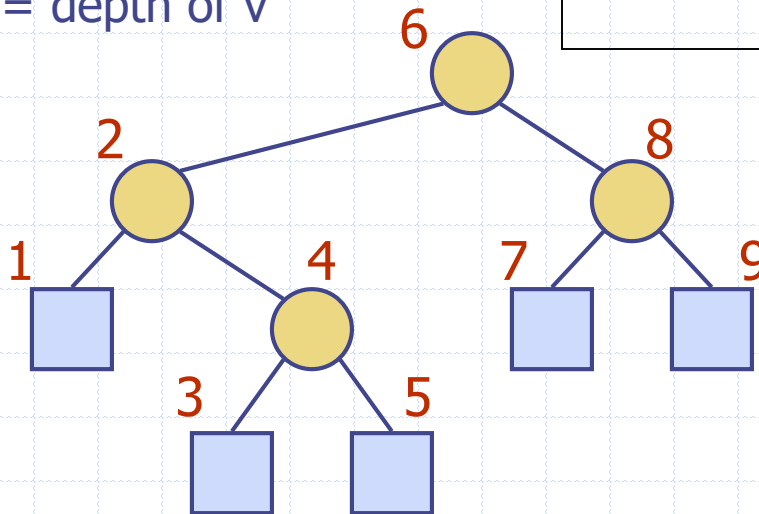
**if** *isInternal* ( $v$ )

*inOrder* (*leftChild* ( $v$ ))

*visit*( $v$ )

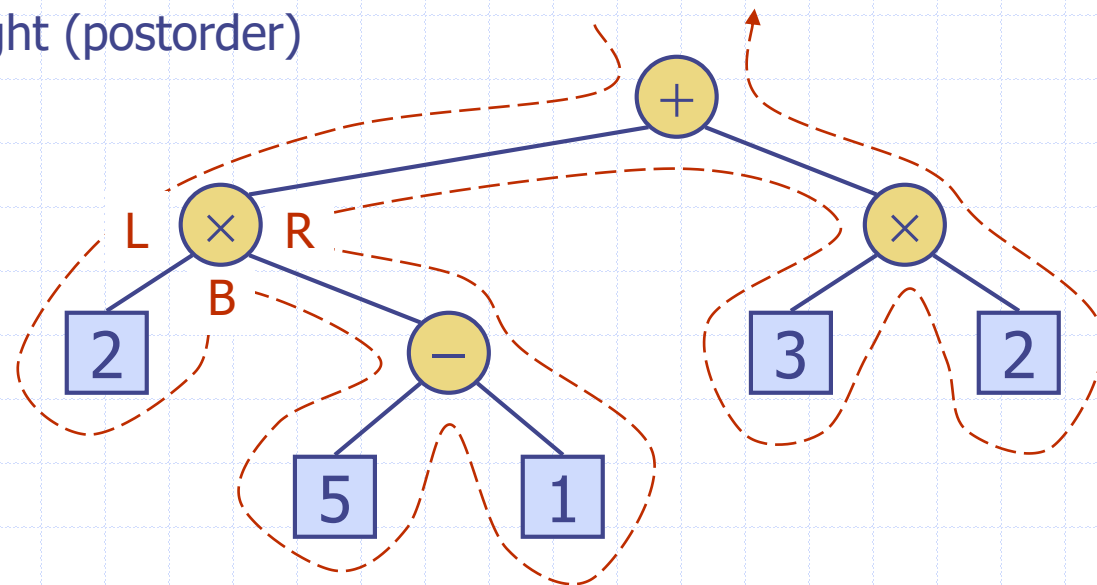
**if** *isInternal* ( $v$ )

*inOrder* (*rightChild* ( $v$ ))



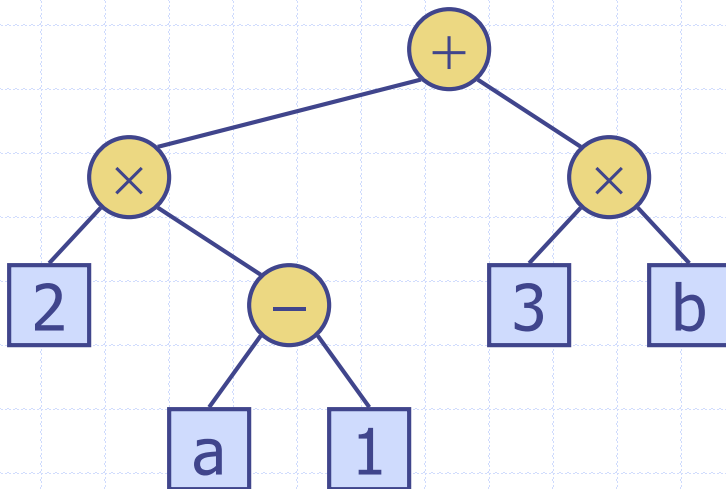
# Euler Tour Traversal

- ◆ Generic traversal of a binary tree
- ◆ Includes a special cases the preorder, postorder and inorder traversals
- ◆ Walk around the tree and visit each node three times:
  - on the left (preorder)
  - from below (inorder)
  - on the right (postorder)



# Printing Arithmetic Expressions

- ◆ Specialization of an inorder traversal
  - print operand or operator when visiting node
  - print "(" before traversing left subtree
  - print ")" after traversing right subtree



**Algorithm** *printExpression(v)*

**if** *isInternal* (v)

*print*("(")

*inOrder* (*leftChild* (v))

*print*(v.*element* ())

**if** *isInternal* (v)

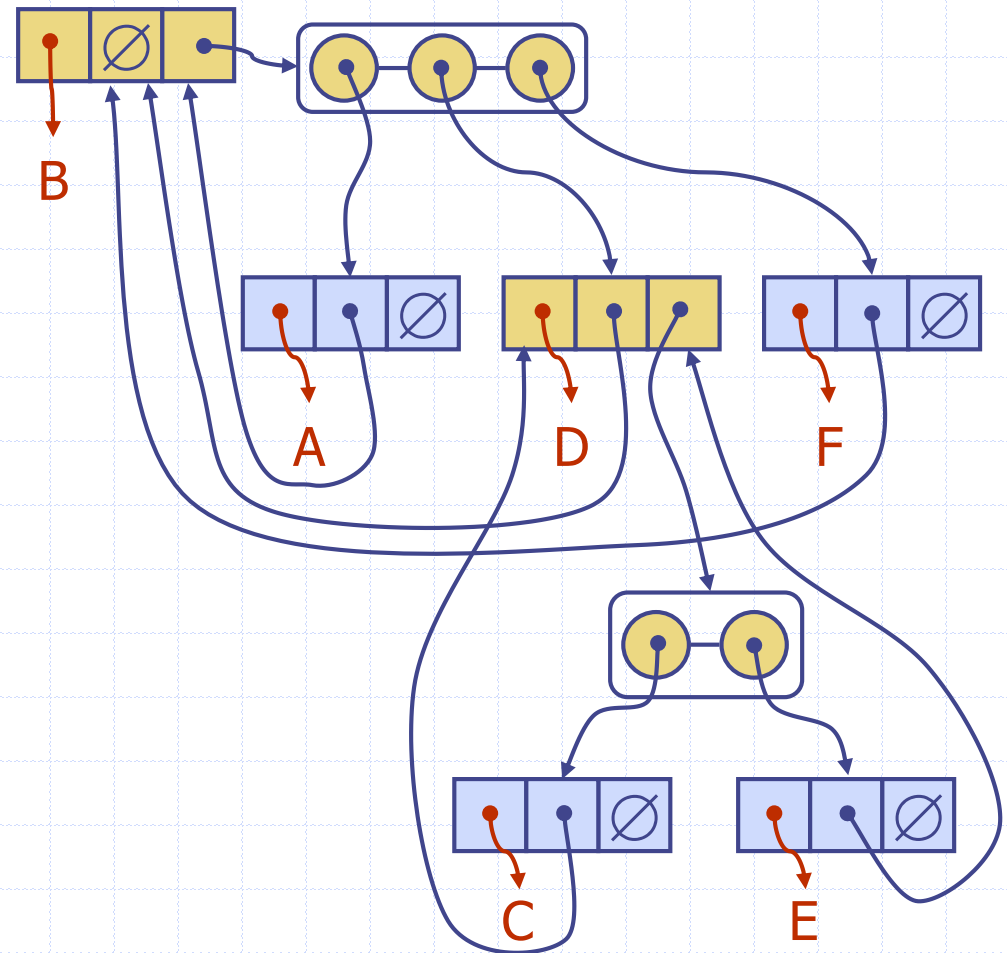
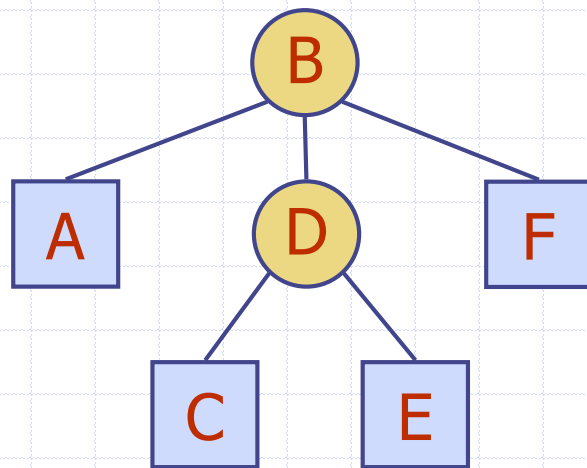
*inOrder* (*rightChild* (v))

*print* (")")

$((2 \times (a - 1)) + (3 \times b))$

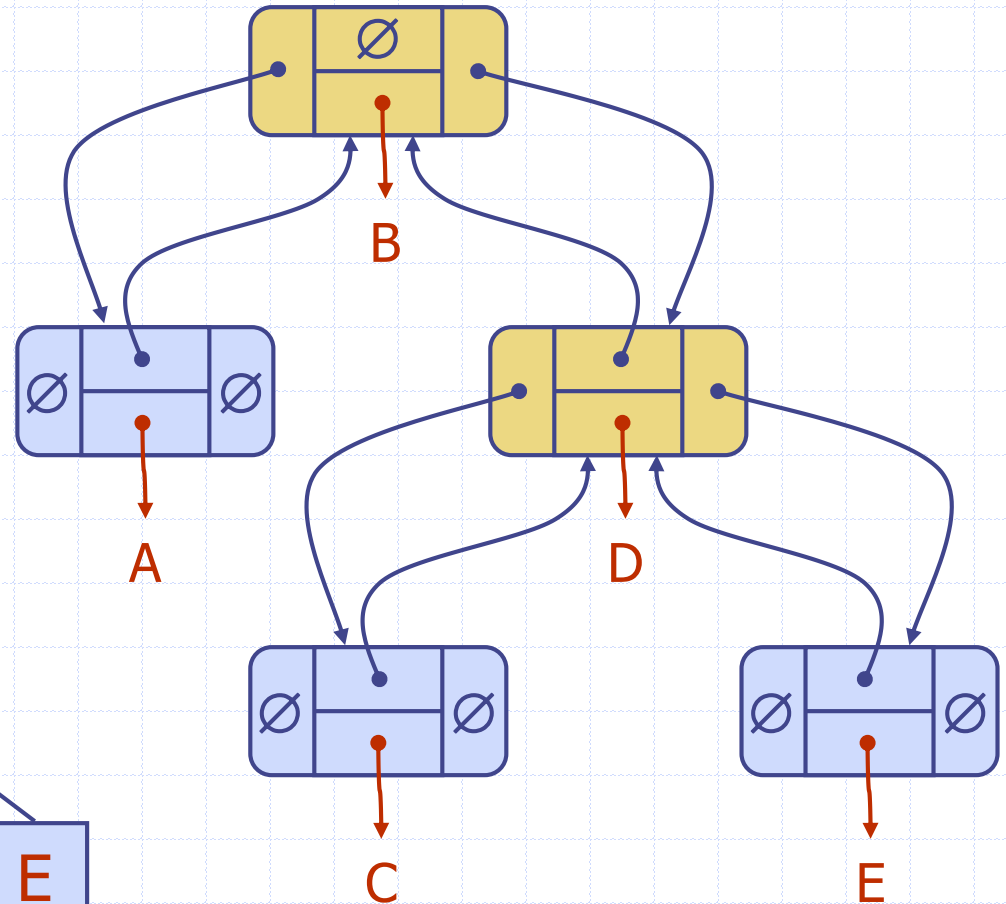
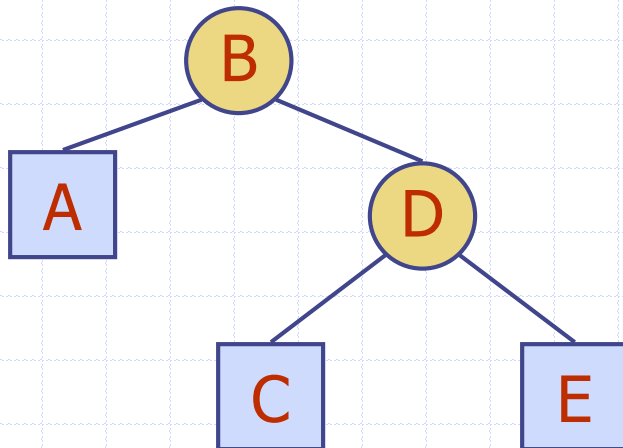
# Linked Data Structure for Representing Trees (§2.3.4)

- ◆ A node is represented by an object storing
  - Element
  - Parent node
  - Sequence of children nodes
- ◆ Node objects implement the Position ADT



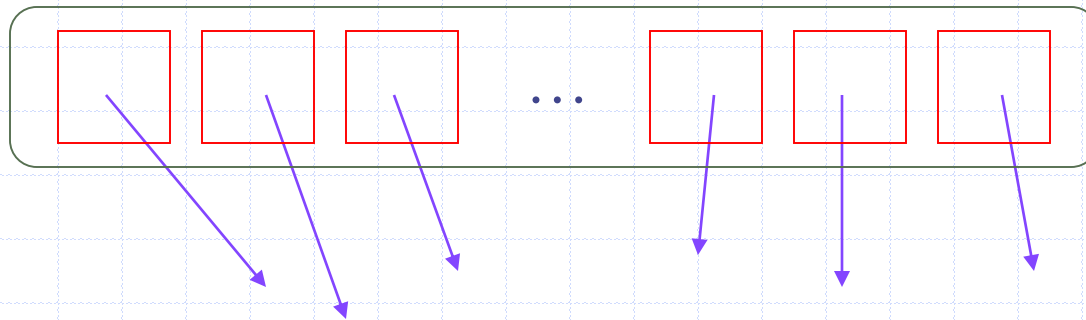
# Linked Data Structure for Binary Trees

- ◆ A node is represented by an object storing
  - Element
  - Parent node
  - Left child node
  - Right child node
- ◆ Node objects implement the Position ADT



# Array-Based Representation of Binary Trees

- ◆ nodes are stored in an array



- let  $\text{rank}(\text{node})$  be defined as follows:

- $\text{rank}(\text{root}) = 1$
- if node is the left child of  $\text{parent}(\text{node})$ ,  
 $\text{rank}(\text{node}) = 2 * \text{rank}(\text{parent}(\text{node}))$
- if node is the right child of  $\text{parent}(\text{node})$ ,  
 $\text{rank}(\text{node}) = 2 * \text{rank}(\text{parent}(\text{node})) + 1$

