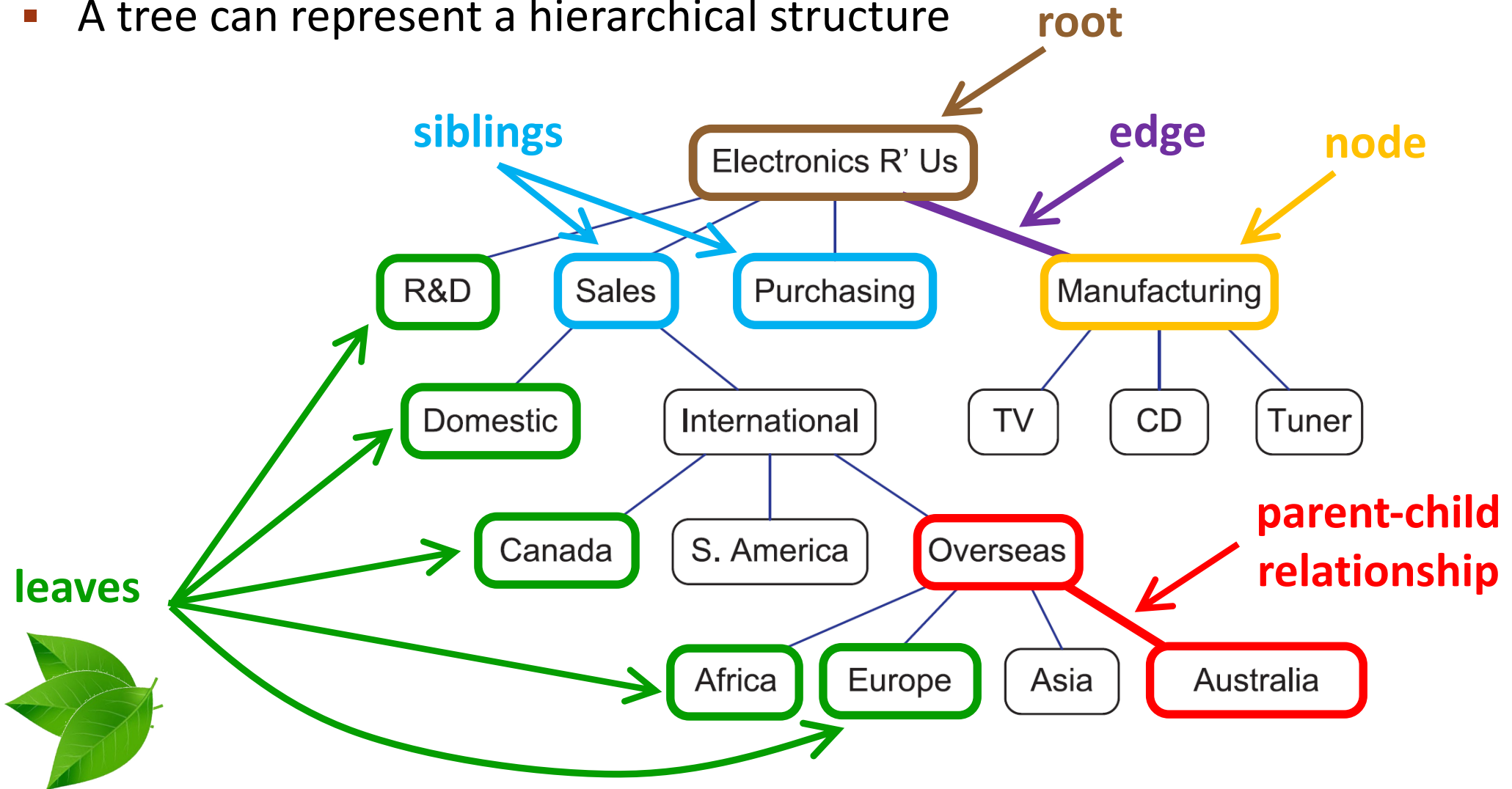


CHAPTER 7: TREES

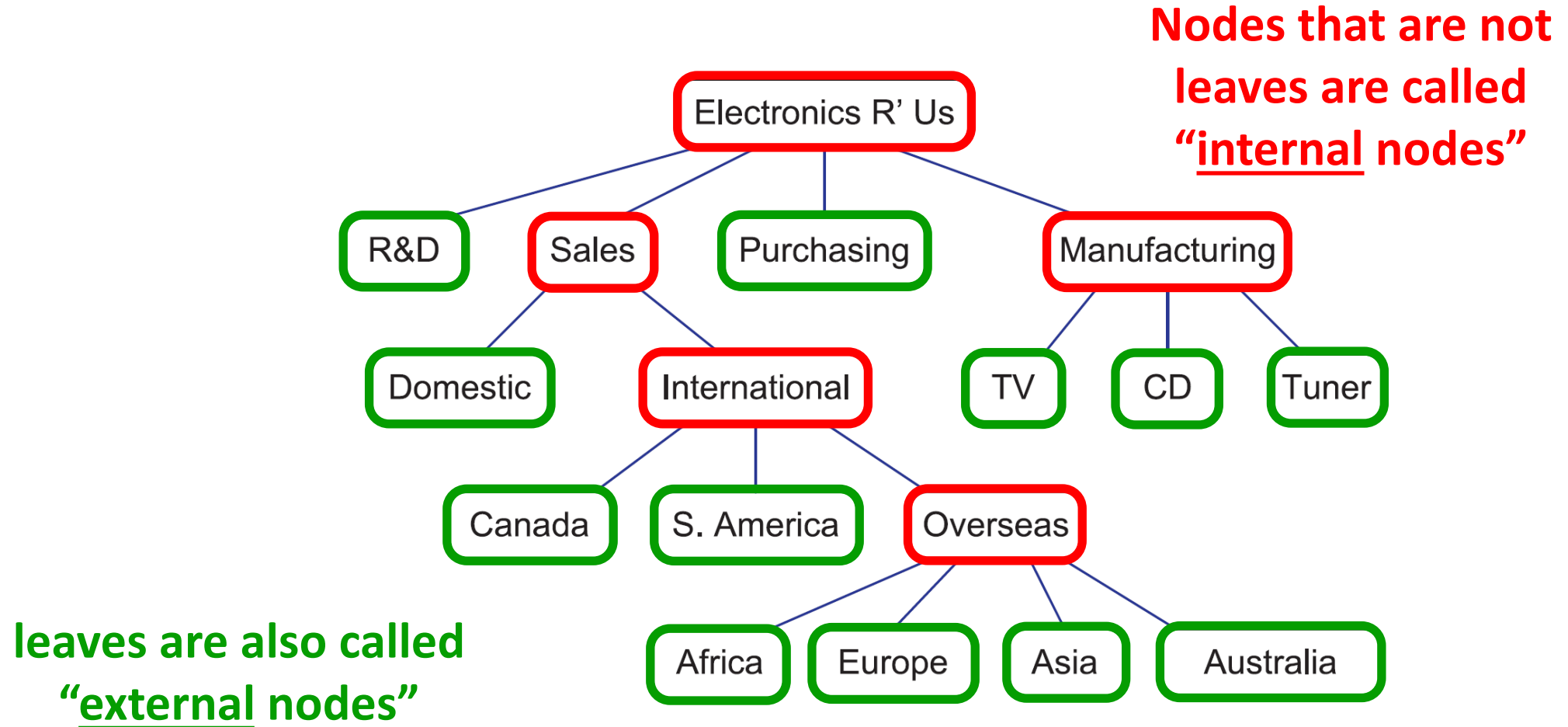
TREE EXAMPLE

- A tree can represent a hierarchical structure



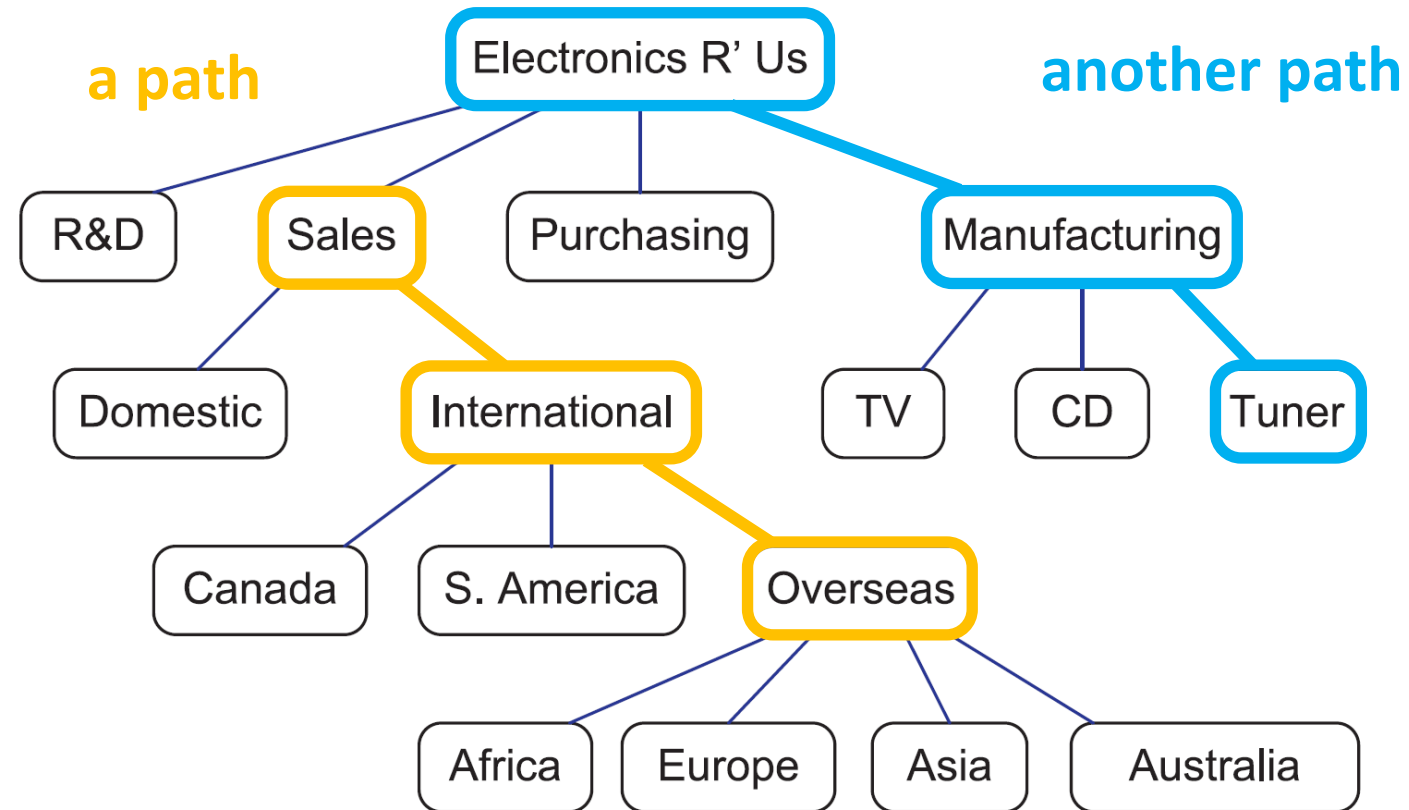
TREE EXAMPLE

- A tree can represent a hierarchical structure



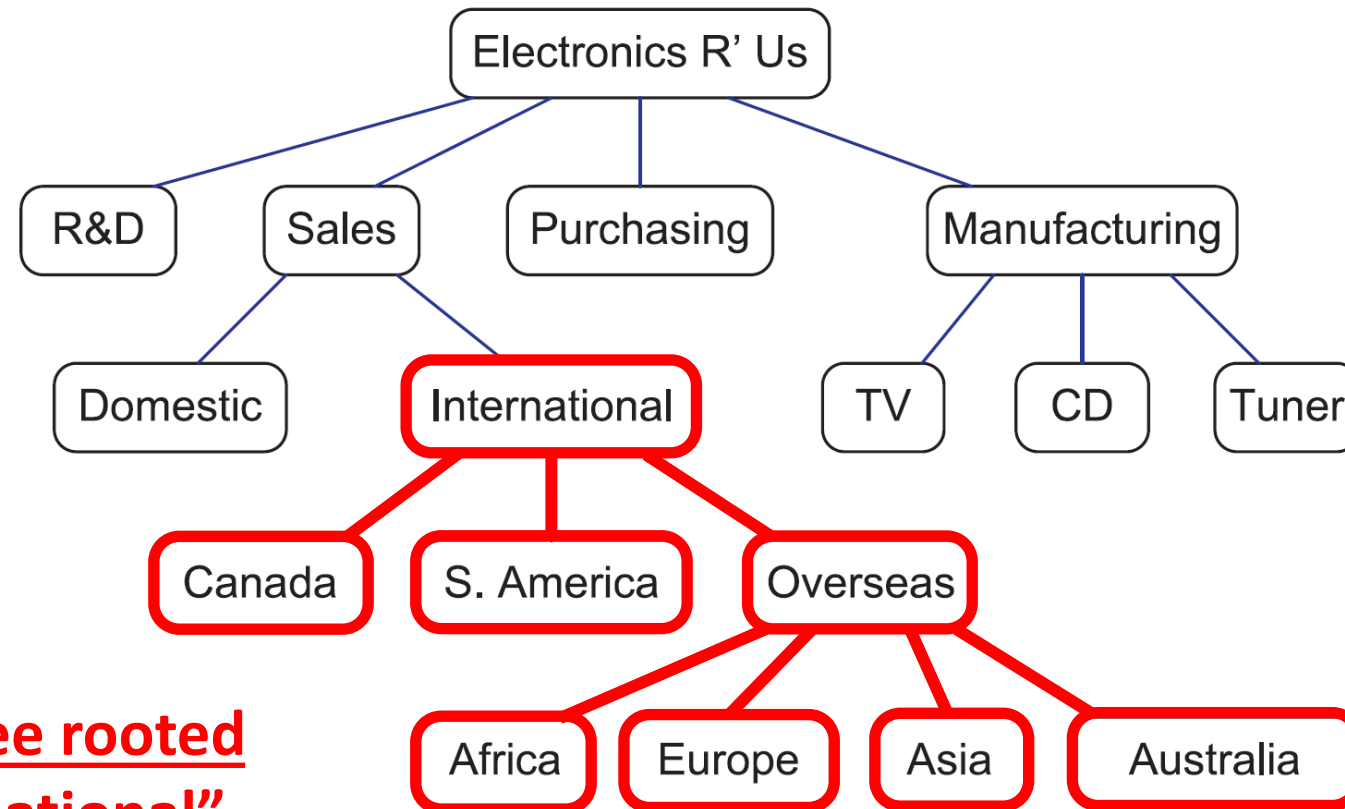
TREE EXAMPLE

- A tree can represent a hierarchical structure



TREE EXAMPLE

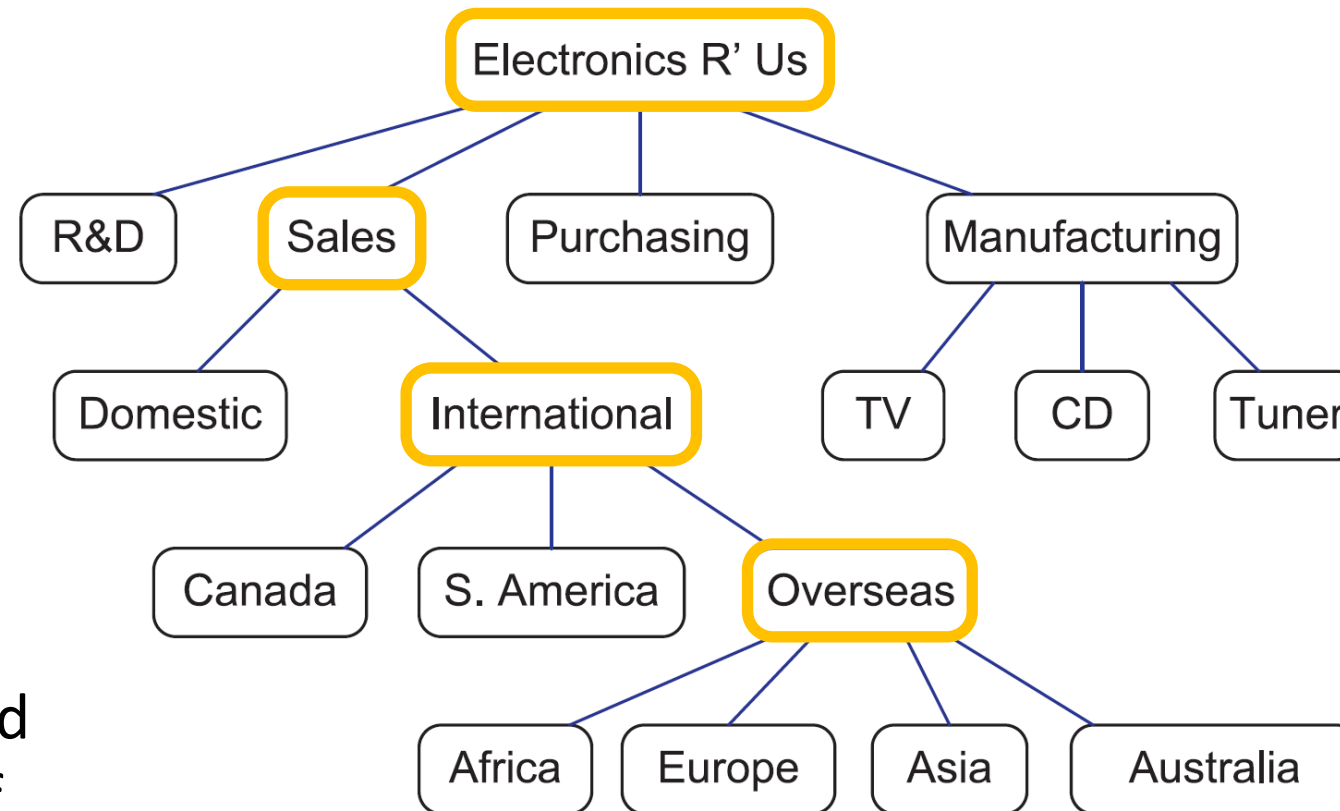
- A tree can represent a hierarchical structure



**The subtree rooted
at “International”**

TREE EXAMPLE

- A tree can represent a hierarchical structure

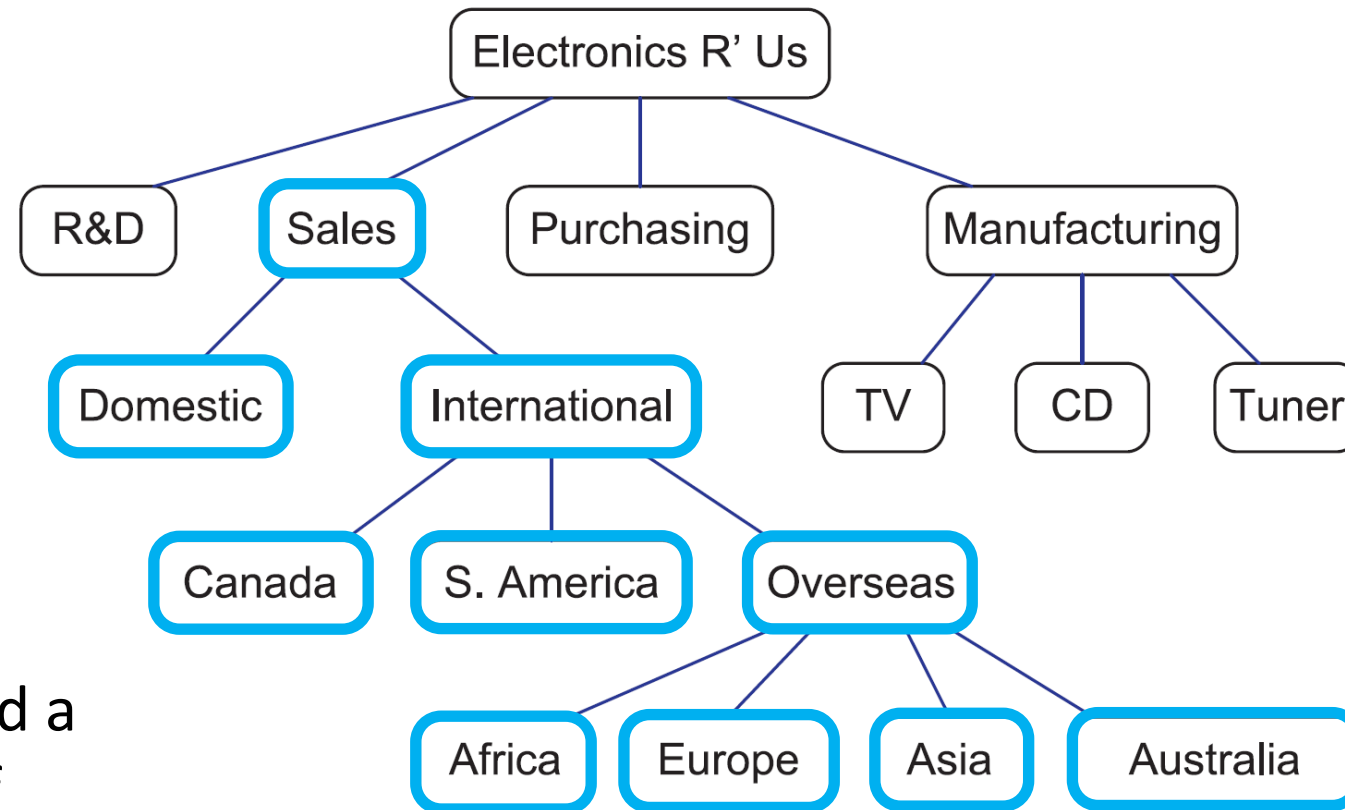


The ancestors
of “Overseas”

A node is considered
an ancestor of itself

TREE EXAMPLE

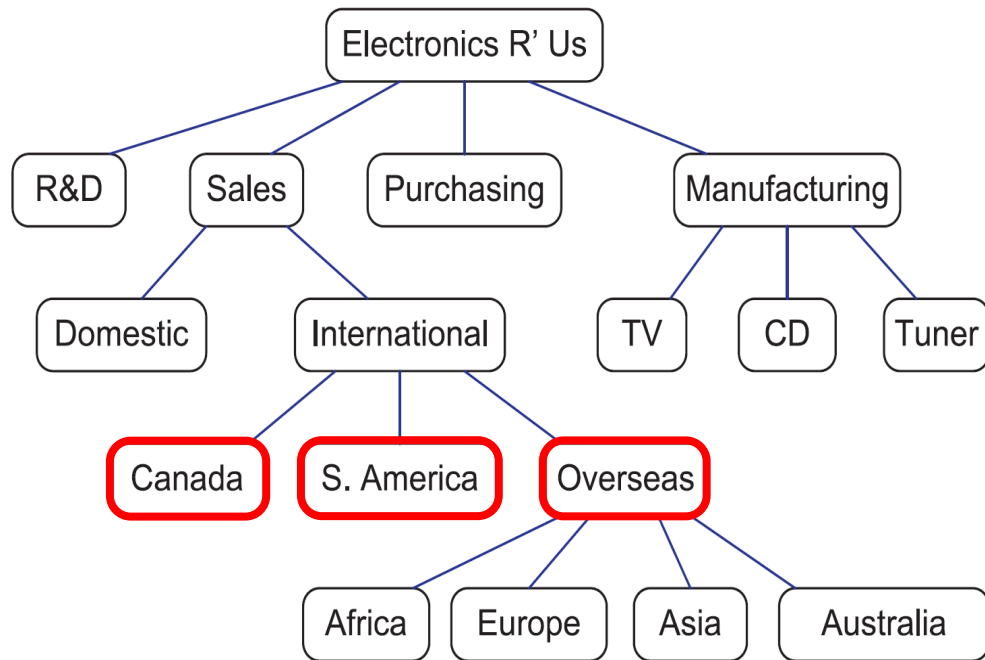
- A tree can represent a hierarchical structure



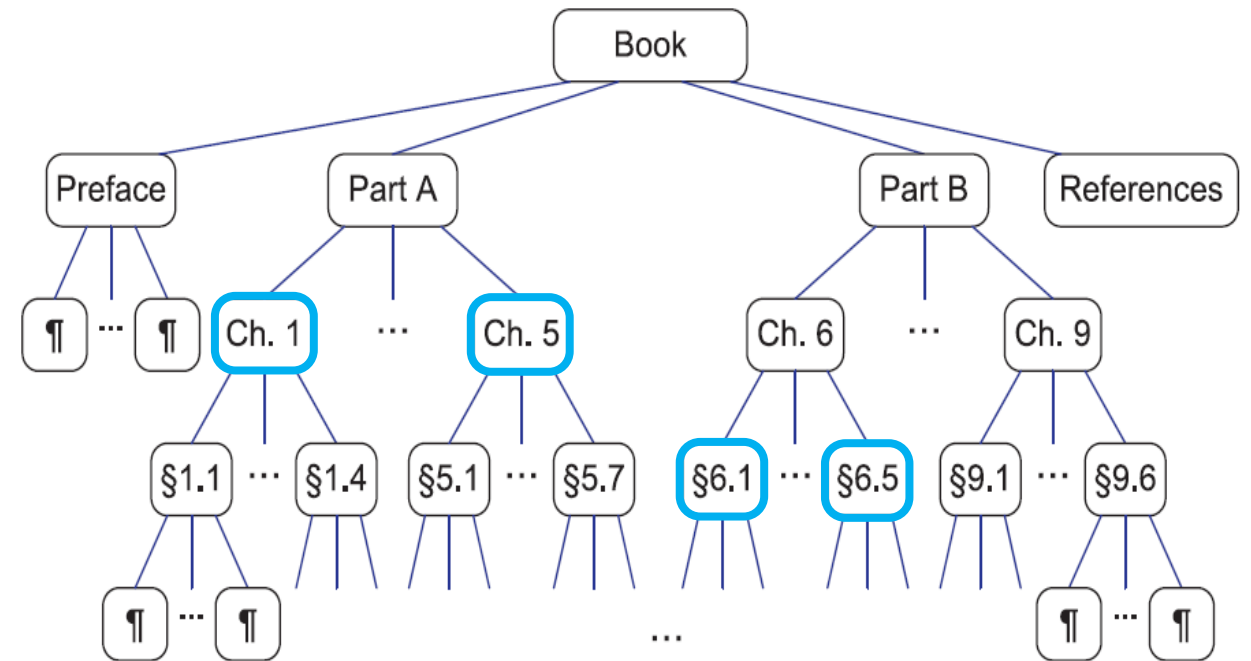
The descendants
of “Sales”

A node is considered a descendant of itself

TREE EXAMPLE



This is **not an ordered tree**, because there is no order between siblings



This tree is **ordered**, because the book parts are ordered, and in each part the chapters are ordered, etc.

NODE-RELATED METHODS

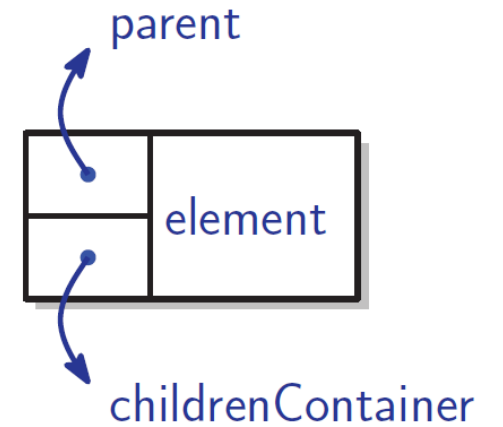
- Each node can be associated with a **position**. Given a position p , its node is accessed by $*p$ to get the element. We will often use the terms “position” and “node” interchangeably
- The position, p , is an **object** that not only points to a node, but also describes its position in the tree via its methods:
 - `parent()`: returns the parent of p (error, if p is the root)
 - `children()`: Returns a list of positions of all p 's children. If p is external, the returned list will be empty
 - `isRoot()`: Returns true, if p is the root
 - `isExternal()`: Returns true, if p is external

TREE-RELATED METHODS

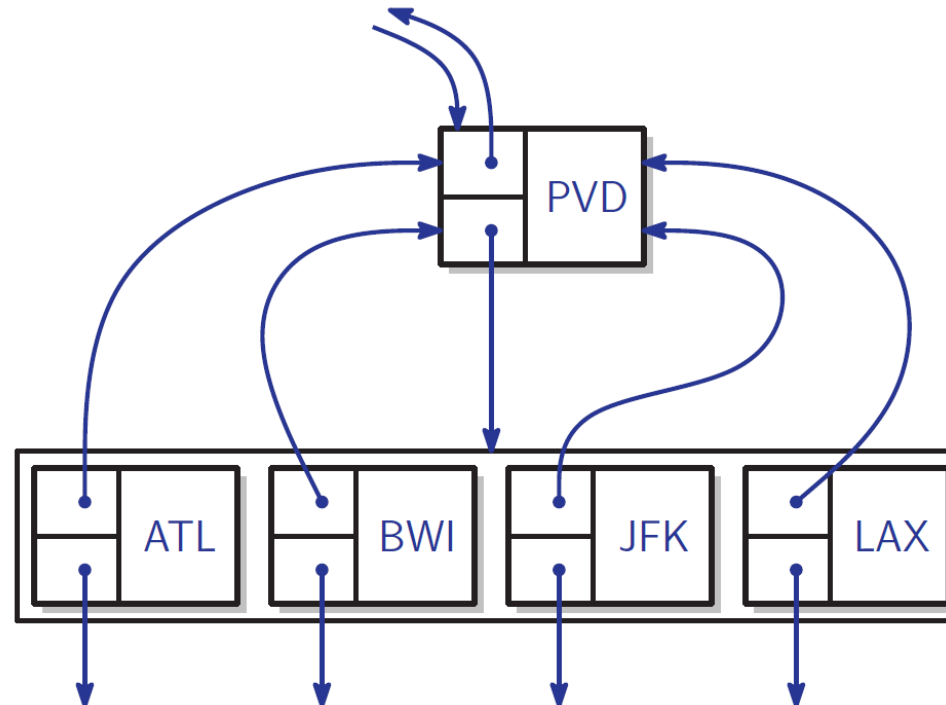
- The tree itself mainly supports, but not limited to, the following functionalities:
 - `size()`: Returns the number of nodes in the tree
 - `empty()`: Returns true, if the tree is empty
 - `root()`: Returns a position for the root (error, if tree is empty)
 - `positions()`: Returns a list of positions of all the nodes of the tree

TREE – LINKED STRUCTURE

- Here is an illustration of a **node**, which has an **element**, a **parent** (which is NULL if the node is the root) and a container of all its **children**



- Here is an illustration of the **tree** data structure



TREE – COMPLEXITY

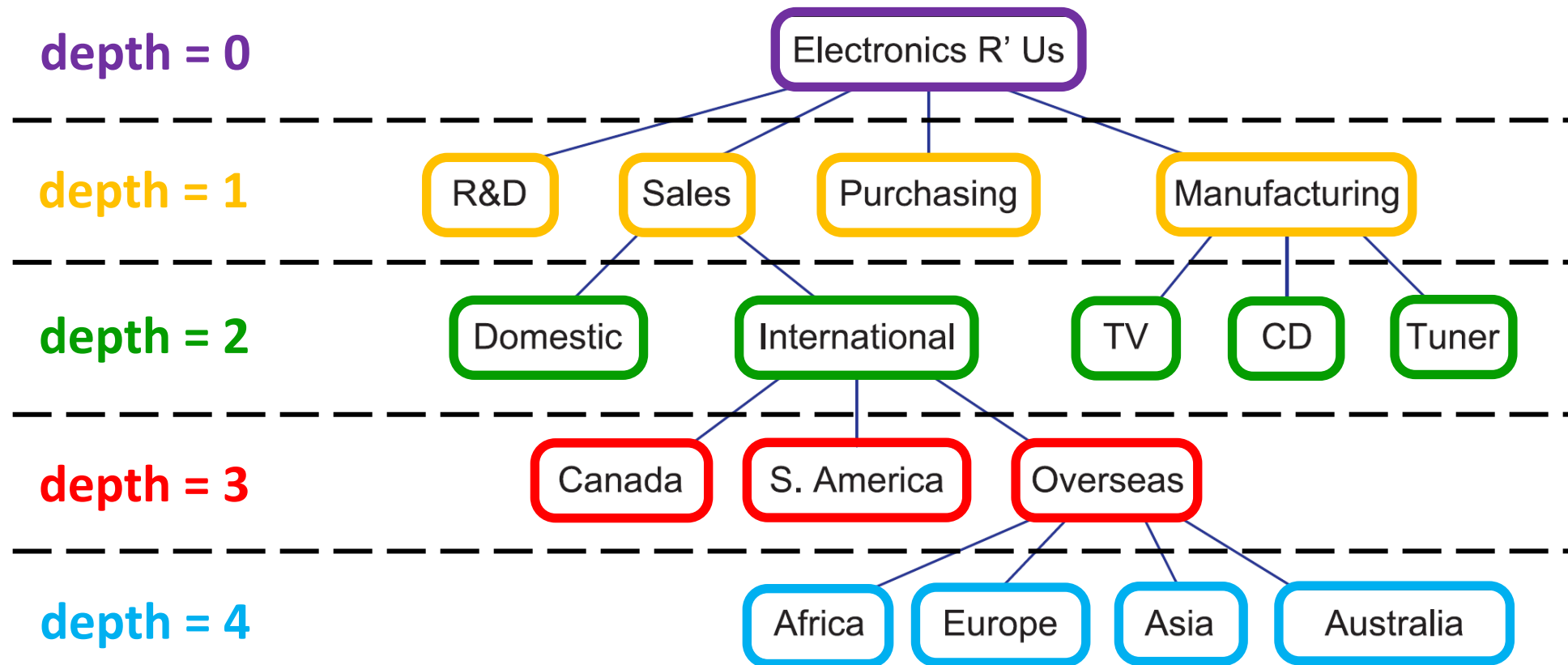
- The running time, where c_p is the number of children of p :

<i>Operation</i>	<i>Time</i>	
isRoot, isExternal	$O(1)$	} Node-related functions
parent	$O(1)$	
children(p)	$O(c_p)$	
size, empty	$O(1)$	} Tree-related functions
root	$O(1)$	
positions	$O(n)$	

- **children(p)** iterates through the container containing the children, which takes $O(c_p)$
- With a similar reasoning, "**positions**" runs in $O(n)$ time

NODE DEPTH

- The “**depth**” of a node represents its **distance from the root**



NODE DEPTH

- The “**depth**” of a node represents its **distance from the root**

How would you compute the depth of a node, p , recursively?

Algorithm $\text{depth}(T, p)$:

Input: Tree T and a position p

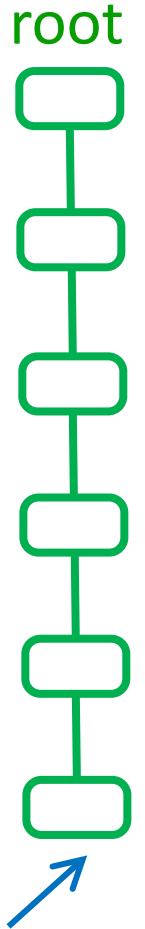
Output: The depth of the node referred to by p

if $p.\text{isRoot}()$ **then**

return 0

else return $1 + \text{depth}(T, p.\text{parent}())$

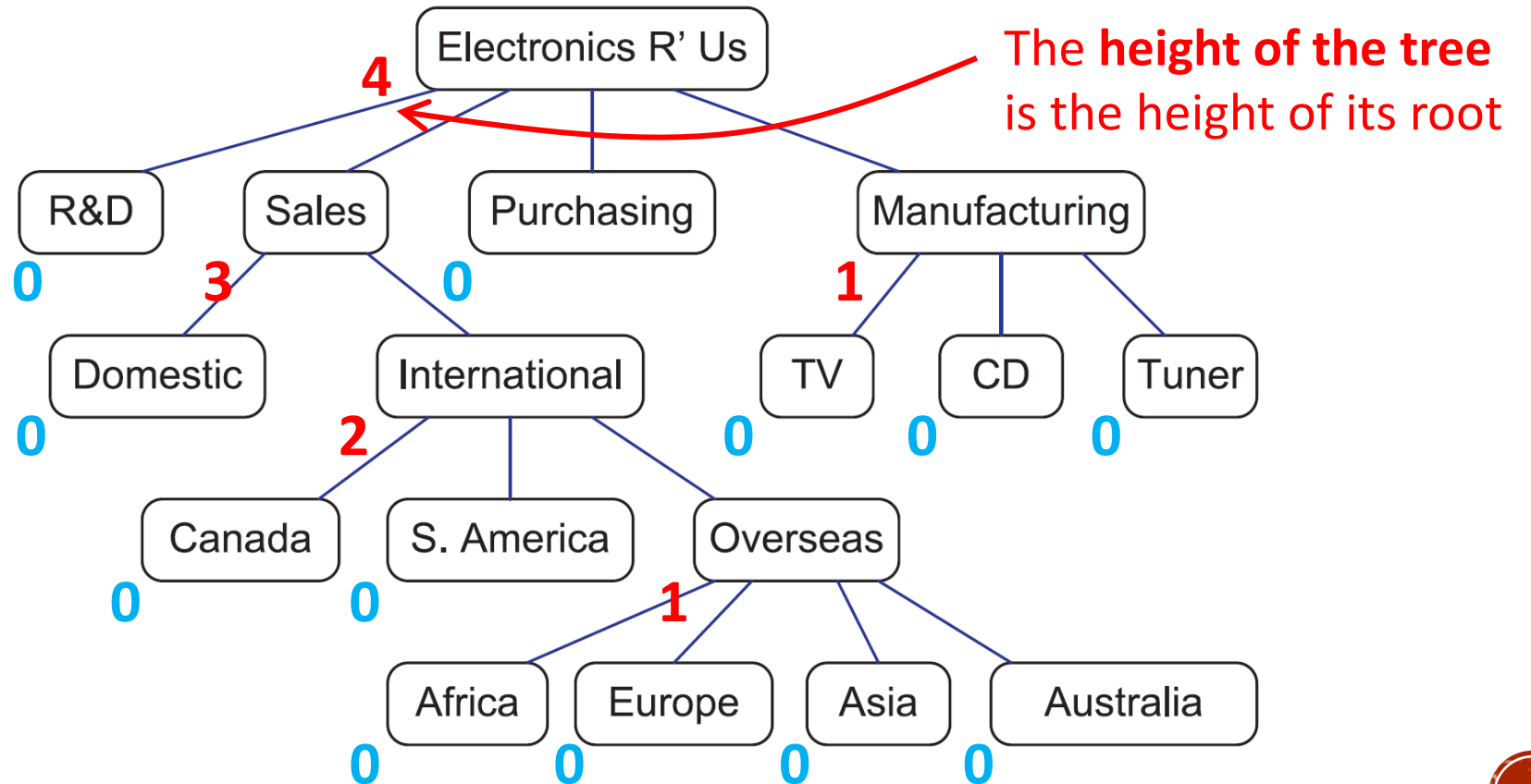
- The running time is $O(d_p)$, where d_p is the depth of node p in tree T .
- In the worst case, the depth algorithm runs in $O(n)$ time. What does the tree look like in the worst case? Like a chain!
- Still, it's more accurate to characterize the running time in terms of d_p rather than n , because d_p is often much smaller than n



For this node,
 $d_p = n$

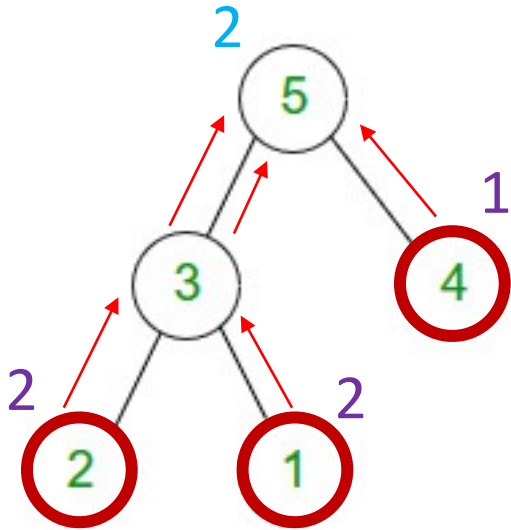
NODE HEIGHT

- The “**height**” of a node represents its **distance from the farthest descendant**
- **Proposition:** The height of a tree is equal to **the maximum depth of its leaves!**



NODE HEIGHT

- The “**height**” of a node represents its **distance from the farthest descendant**
- **Proposition:** The height of a tree is equal to **the maximum depth of its leaves!**
- Based on this proposition, how can we use the function “ $\text{depth}(T, q)$ ” to compute the height of a tree?



Algorithm treeHeight(T):

Input: Tree T

Output: The height of the tree T

$h \leftarrow 0$

$nodes \leftarrow T.positions()$

for $q \leftarrow nodes.begin()$ to $nodes.end()$ **do**

if $q.isExternal()$ **then**

$h \leftarrow \max(h, \text{depth}(T, q))$

return h

NODE HEIGHT

What is the running time, knowing that $\text{depth}(T, p)$ runs in $O(d_p)$ time?

Algorithm treeHeight(T):

Input: Tree T

Output: The height of the tree T

$h \leftarrow 0$

$nodes \leftarrow T.positions()$

for $q \leftarrow nodes.begin()$ to $nodes.end()$ **do**

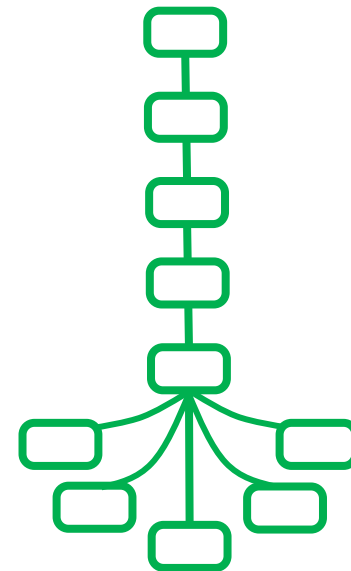
if $q.isExternal()$ **then**

$h \leftarrow \max(h, \text{depth}(T, q))$

return h

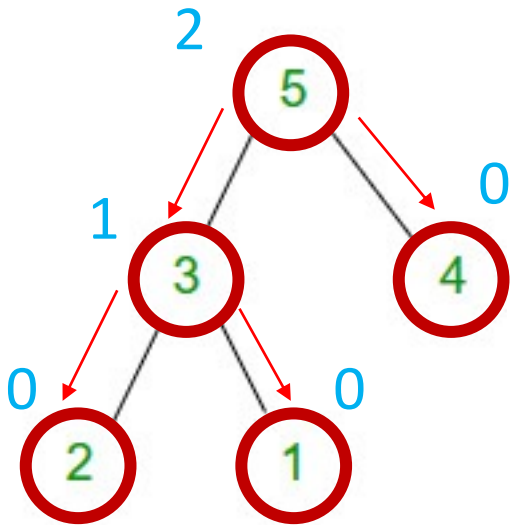
- For each **internal node**, we need to do some constant time operations, i.e., $O(1)$
- For each **external node**, q , we need to call $\text{depth}(T, q)$, which runs in $O(d_p)$
- The total is then $\approx O(n + \sum_{p \in \text{ExternalNodes}} d_p)$
- The **worst case** is when:
 - $n/2$ nodes form a chain from the root
 - The remaining $n/2$ nodes are the leaves

This way, runtime is $O\left(n + \sum_{i=1}^{n/2} n/2\right)$ which is $O(n^2)$



NODE HEIGHT

- Here is another recursive algorithm, which computes the height of any given node, p , in a tree, T (to compute the height of T , set p to be the root of T).



Algorithm height(T, p):

Input: Tree T and position p

Output: The height of the node referred to by p

if $q.isExternal()$ **then**

return 0

$h \leftarrow 0$

$childrenList \leftarrow p.children()$

for $q \leftarrow childrenList.begin()$ to $childrenList.end()$ **do**

$h \leftarrow \max(h, \text{height}(T, q))$

return $h + 1$

NODE HEIGHT

What is the running time of this algorithm?

```
if  $q.isExternal()$  then
    return 0
 $h \leftarrow 0$ 
 $childrenList \leftarrow p.children()$ 
for  $q \leftarrow childrenList.begin()$  to  $childrenList.end()$  do
     $h \leftarrow \max(h, height(T, q))$ 
return  $h + 1$ 
```

- The algorithm is **recursive**, and, if it is initially called on the root of T , it will eventually be called on each node of T .
- Thus, we can determine the running time by summing, over all the nodes, the amount of time spent at each node (on the non-recursive part).
- For each node, p , this amount is $\approx O(c_p)$ because of iterating over p 's children
- Thus, the algorithm takes $\approx O(\sum_p c_p)$ time.
- Finally, note that $\sum_p c_p = n - 1$. This because each node of T (except the root) is a child of another node, and thus contributes one unit to the sum.
- Thus, this algorithm runs in $O(n)$

NODE HEIGHT

Algorithm treeHeight(T):

Input: Tree T

Output: The height of the tree T

$h \leftarrow 0$

$nodes \leftarrow T.positions()$

for $q \leftarrow nodes.begin()$ to $nodes.end()$ **do**

if $q.isExternal()$ **then**

$h \leftarrow \max(h, \text{depth}(T, q))$

$O(n^2)$

return h

Algorithm height(T, p):

Input: Tree T and position p

Output: The height of the node referred to by p

if $q.isExternal()$ **then**

return 0

$h \leftarrow 0$

$childrenList \leftarrow p.children()$

for $q \leftarrow childrenList.begin()$ to $childrenList.end()$ **do**

$h \leftarrow \max(h, \text{height}(T, q))$

$O(n)$

return $h + 1$

What happened?

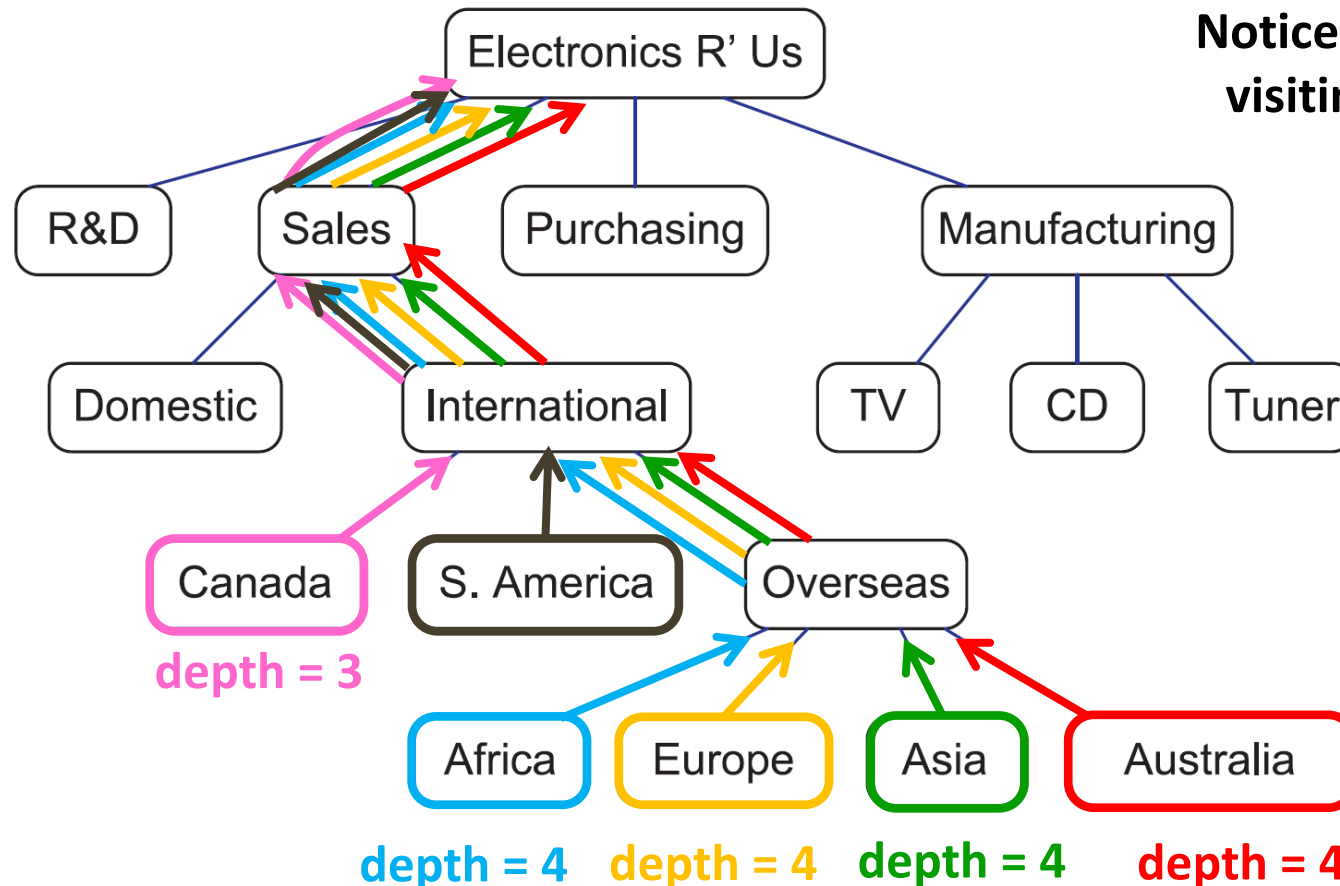
What went wrong with the left algorithm?

Can you spot the repetition here?



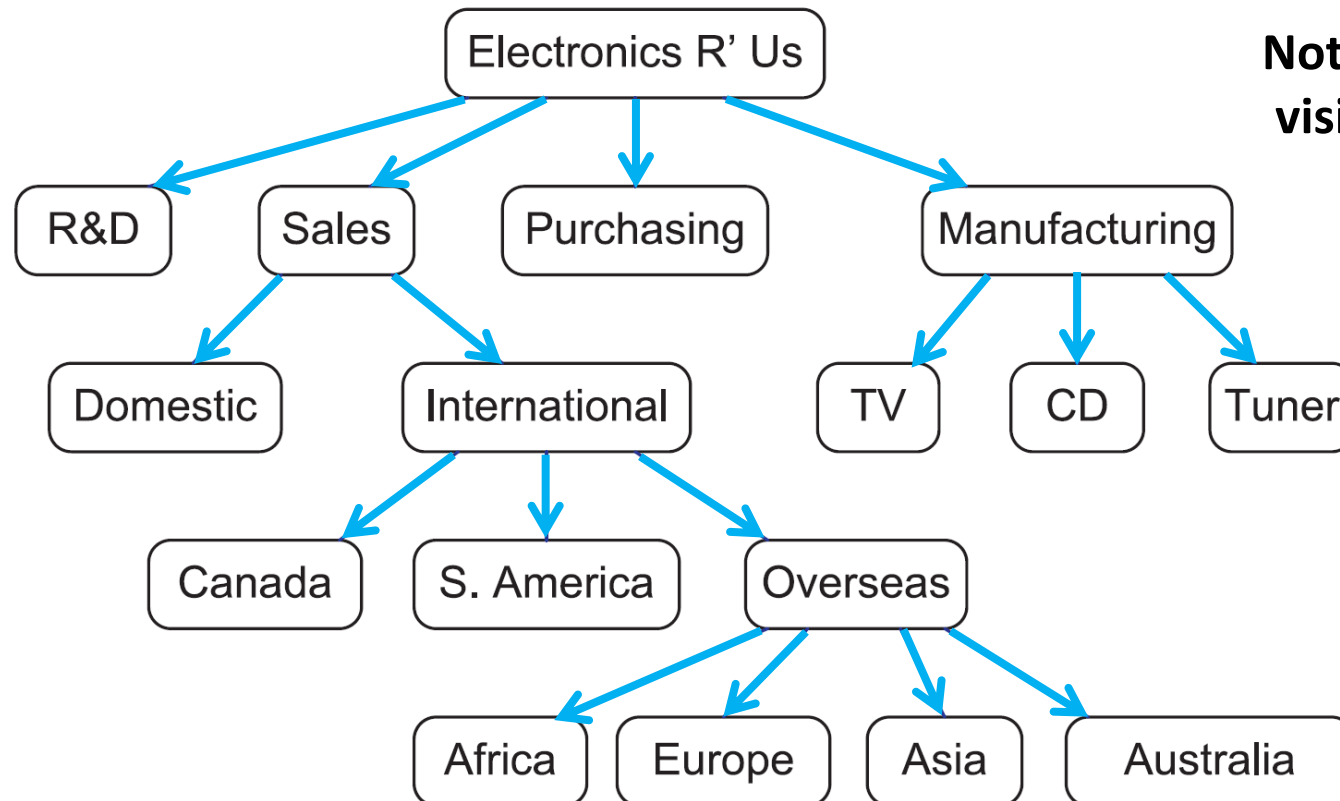
WHAT WENT WRONG

- The inefficient algorithm calculates the **depth of leaves**. Thus it starts **from the bottom** and traverses the tree **upward**!



WHAT WENT WRONG

- In contrast, the efficient algorithm calculates the height of the nodes, by starting **from the top** and **traversing the tree downward**!



Notice how the algorithm never visits an edge more than once!

