

# CS2040S

## Data Structures and Algorithms

### Augmented Trees!

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#### Puzzle of the Week:

100 prisoners. Every so often, one is chosen at random to enter a room with a light bulb. You can turn the light bulb on or off.

- **WIN** if one prisoner announces correctly that all have visited the room.
- **LOSE** if announcement is incorrect.

What if, initially, the state of the light is unknown, either on or off?

# Today's Plan

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## On the importance of being balanced

- Height-balanced binary search trees
- AVL trees
- Rotations

## Tries

- How to handle text?

## Data structure design

- How to build new structures on existing ideas?



# Recap: Dictionary Interface

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A collection of (key, value) pairs:

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## **interface IDictionary**

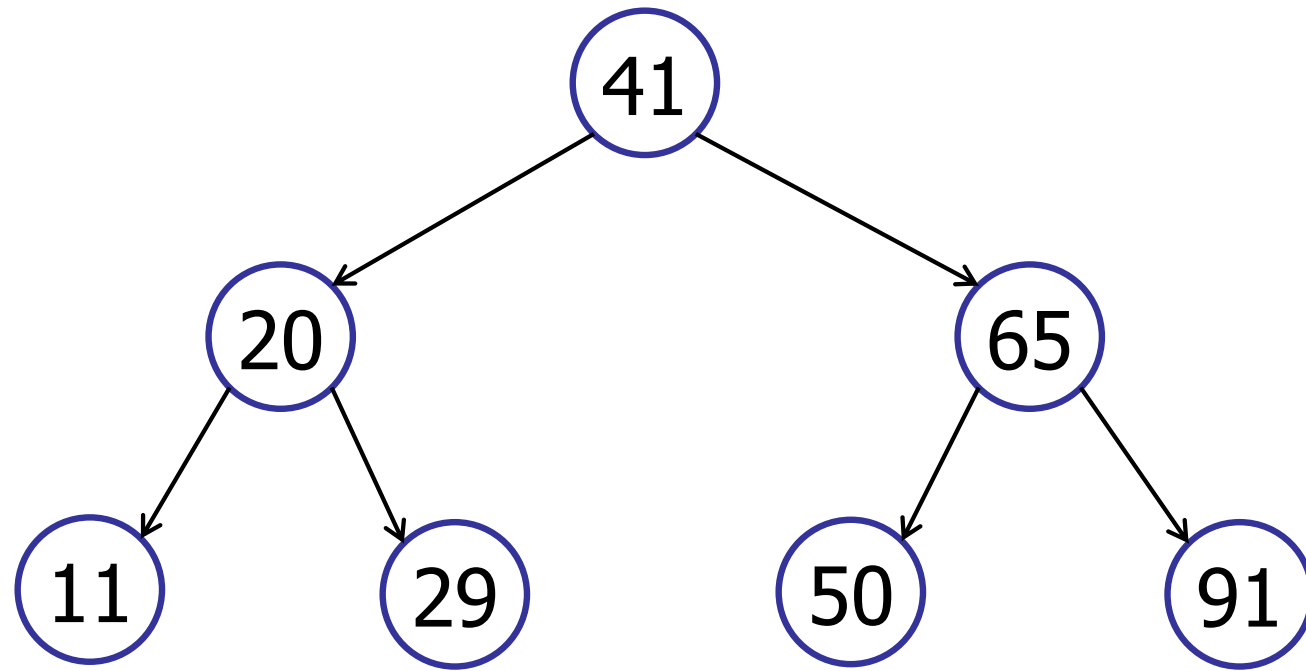
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void	insert(Key k, Value v)	<i>insert (k,v) into table</i>
Value	search(Key k)	<i>get value paired with k</i>
Key	successor(Key k)	<i>find next key &gt; k</i>
Key	predecessor(Key k)	<i>find next key &lt; k</i>
void	delete(Key k)	<i>remove key k (and value)</i>
boolean	contains(Key k)	<i>is there a value for k?</i>
int	size()	<i>number of (k,v) pairs</i>

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# Recap: Binary Search Trees

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- Two children:  $v.\text{left}$ ,  $v.\text{right}$
- Key:  $v.\text{key}$
- **BST Property**: all in left sub-tree  $<$  key  $<$  all in right sub-right

# Binary Search Tree

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Modifying Operations:  $O(h)$

- insert
- delete

Query Operations:  $O(h)$

- search
- predecessor, successor
- findMax, findMin

Traversals:  $O(n)$

# The Importance of Being Balanced

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Operations take  $O(h)$  time

$$\log(n) - 1 \leq h \leq n$$

*Key definition*

A BST is balanced if  $h = O(\log n)$

On a balanced BST: all operations run in  $O(\log n)$  time.

# The Importance of Being Balanced

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How to get a balanced tree:

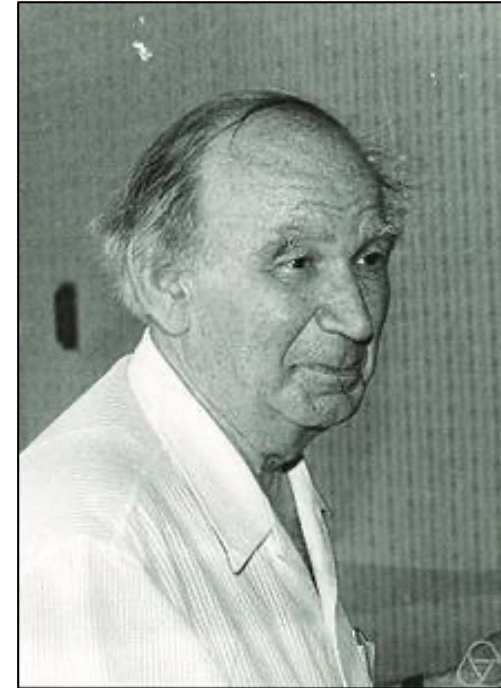
- Define a good property of a tree.
- Show that if the good property holds, then the tree is **balanced**.
- After every insert/delete, make sure the good property still holds. If not, fix it.



Invariant

# AVL Trees [Adelson-Velskii & Landis 1962]

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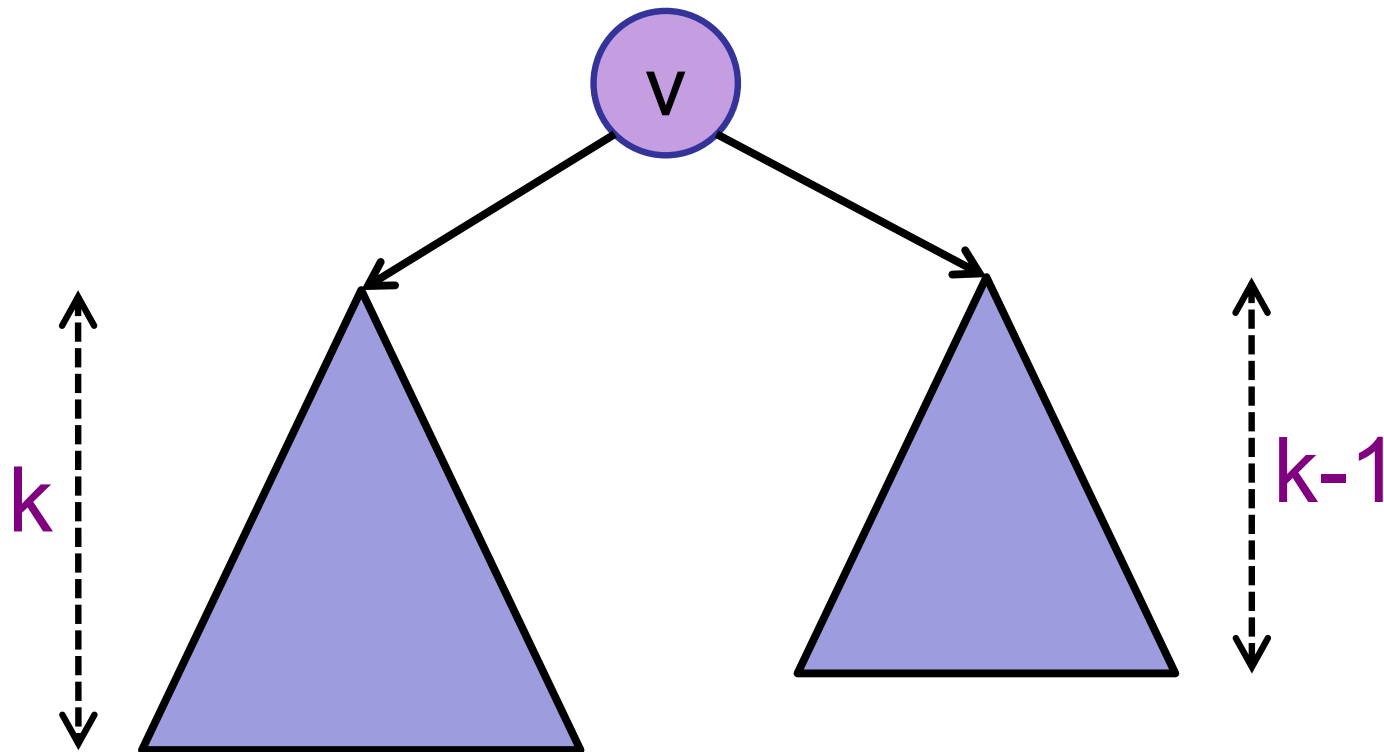
# AVL Trees [Adelson-Velskii & Landis 1962]

## Step 1: Define Invariant

- A node  $v$  is height-balanced if:

$$|v.\text{left.height} - v.\text{right.height}| \leq 1$$

Key definition



# Height-Balanced Trees

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Theorem:

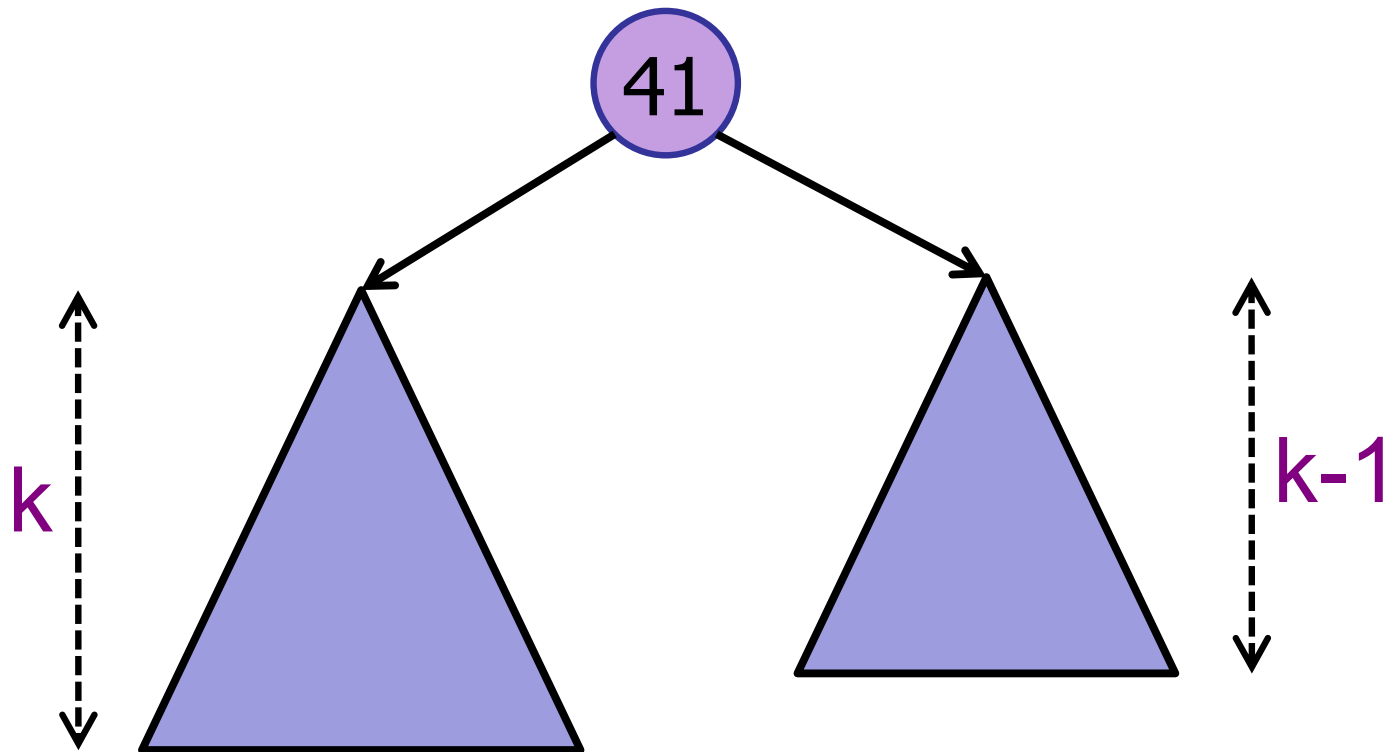
A height-balanced tree with  $n$  nodes has at most height  $h < 2\log(n)$ .

→ A height-balanced tree is balanced.

# AVL Trees [Adelson-Velskii & Landis 1962]

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Step 2: Show how to maintain height-balance



# Inserting in an AVL Tree

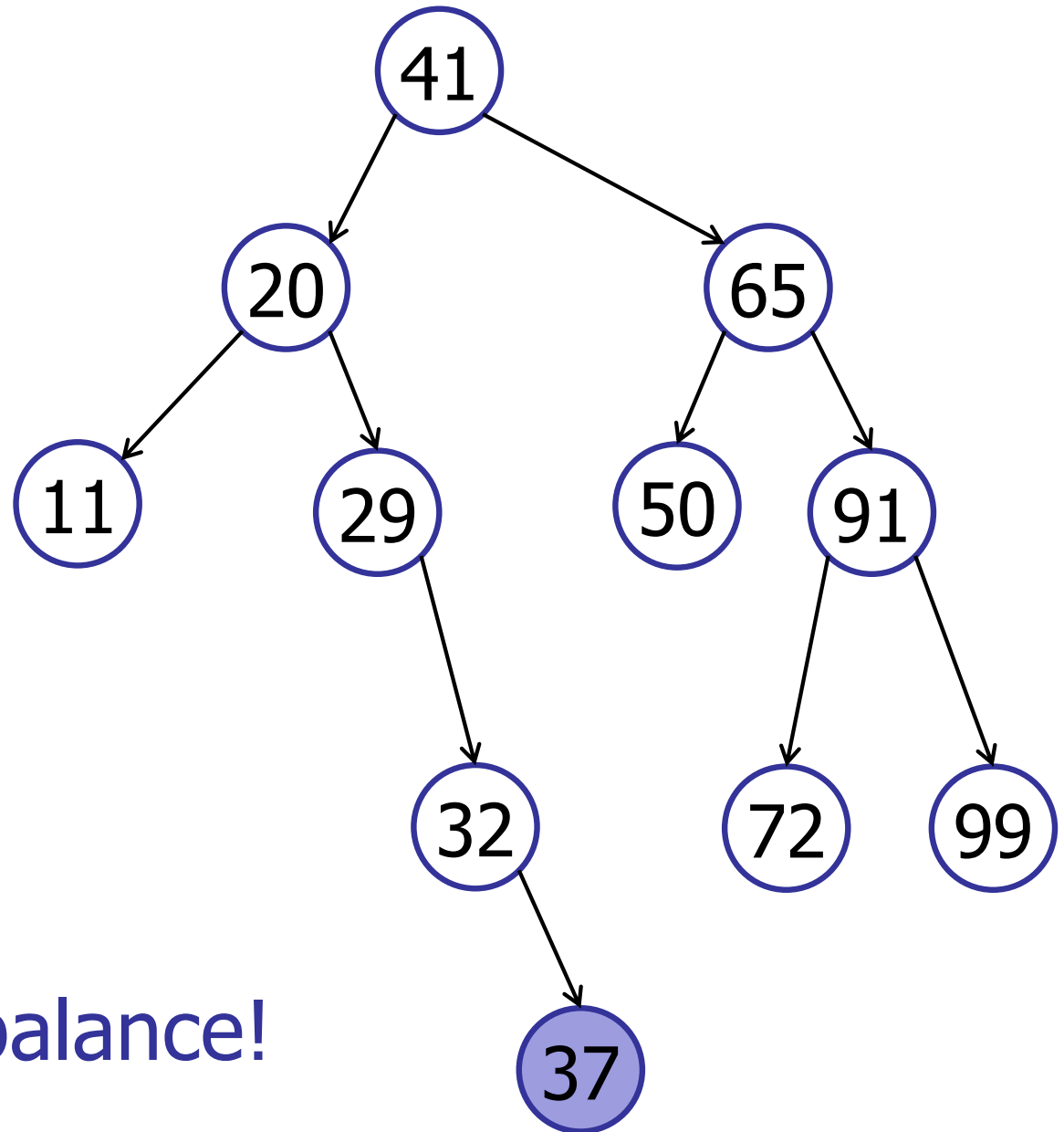
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Initially balanced

insert(37)

No longer balanced  
after insertion!

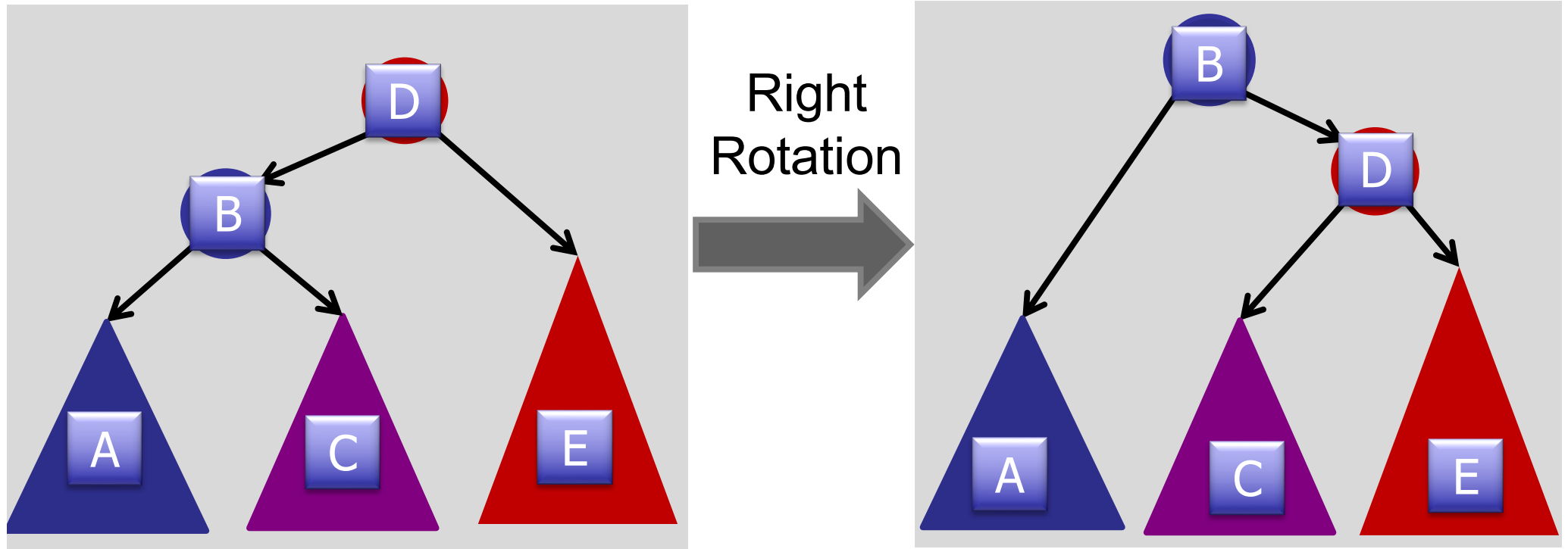
Use rotations to rebalance!



Quick review: a rotation costs:

- ✓ 1.  $O(1)$
- 2.  $O(\log n)$
- 3.  $O(n)$
- 4.  $O(n^2)$
- 5.  $O(2^n)$

# Tree Rotations



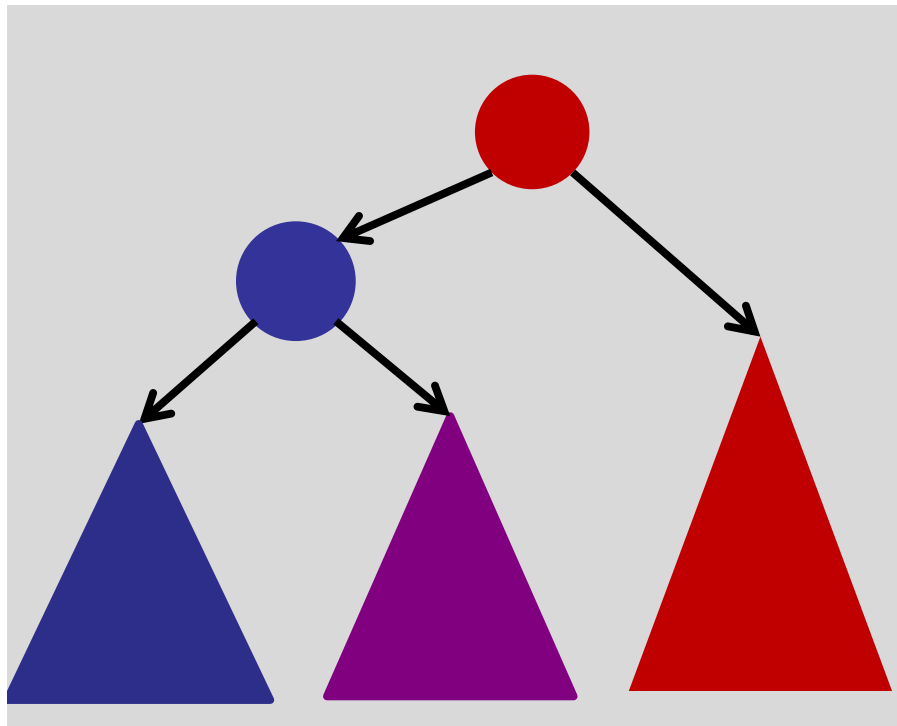
$A < B < C < D < E$

Rotations maintain ordering of keys.

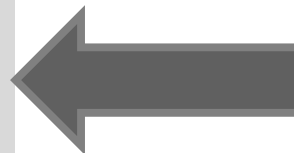
⇒ Maintains BST property.

# Tree Rotations

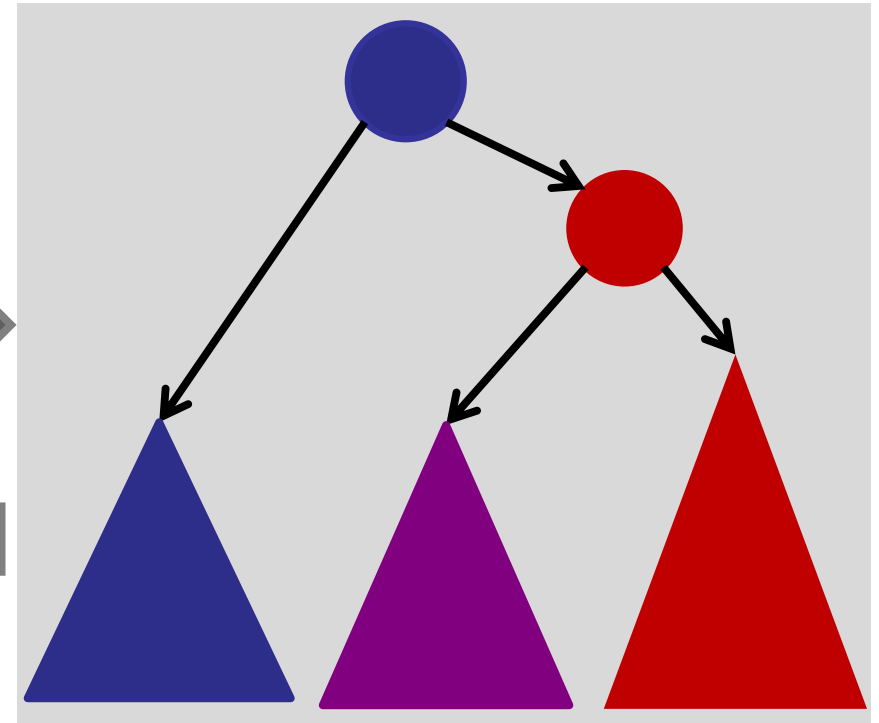
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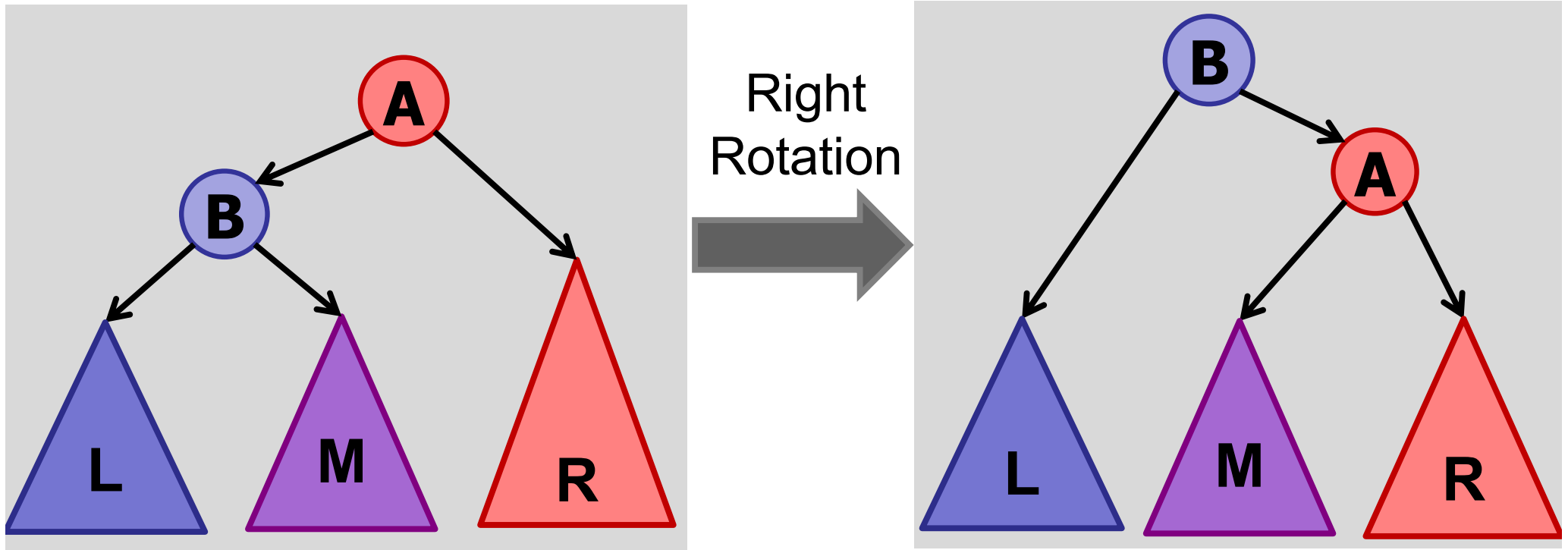
Right  
Rotation



Left  
Rotation



# Tree Rotations



After insert:

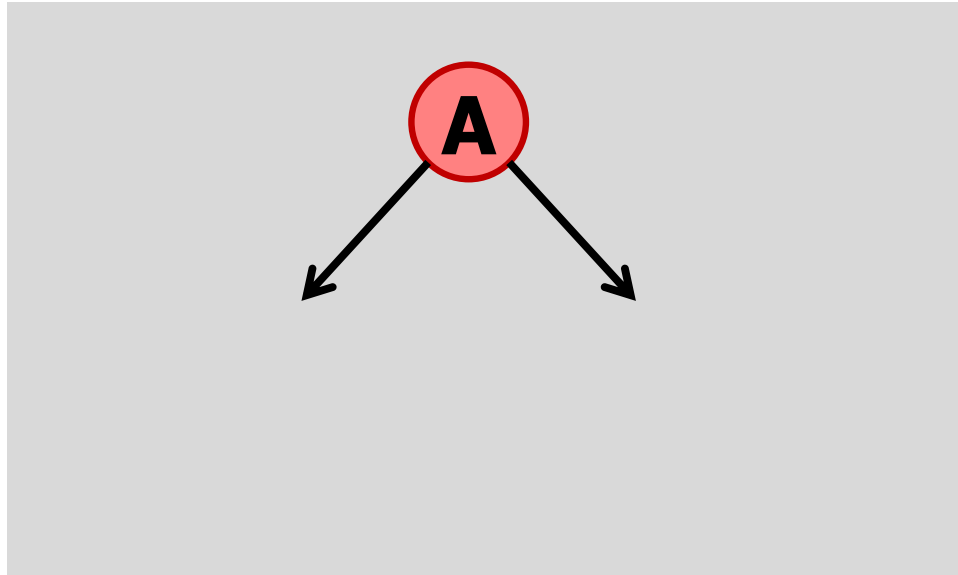
Use tree rotations to restore balance.

Height is out-of-balance by 1



# Tree Rotations

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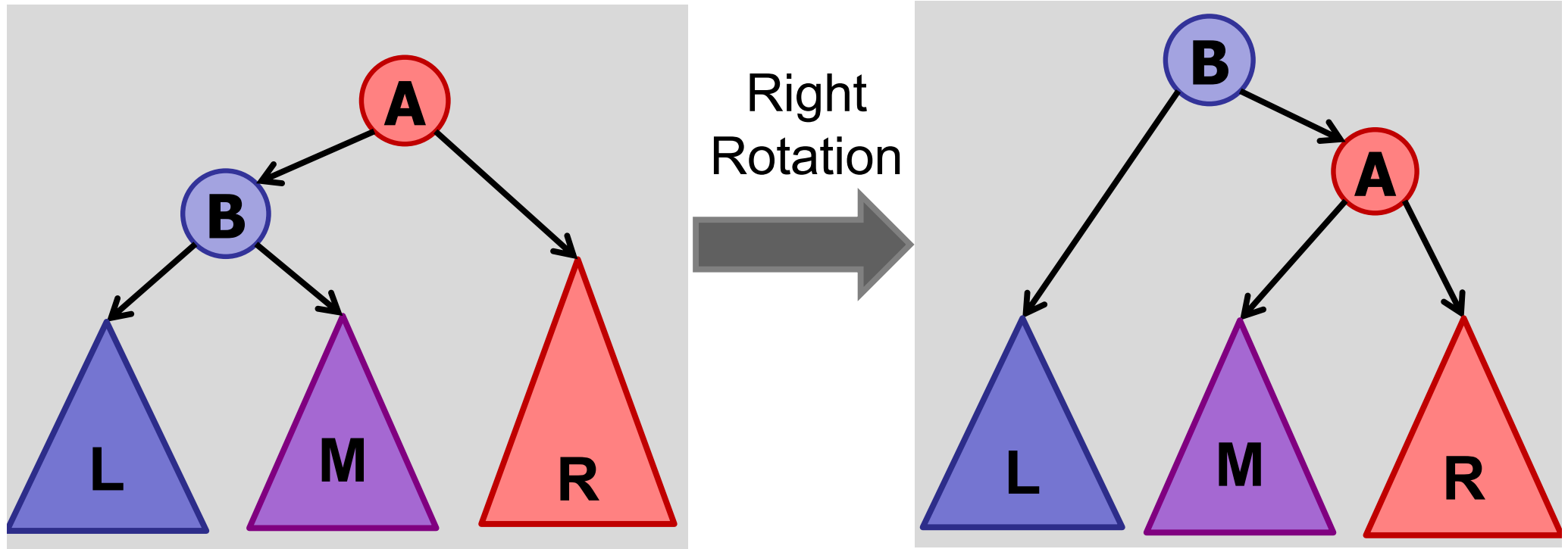


A is **LEFT-heavy** if left sub-tree has larger height than right sub-tree.

A is **RIGHT-heavy** if right sub-tree has larger height than left sub-tree.

# Tree Rotations

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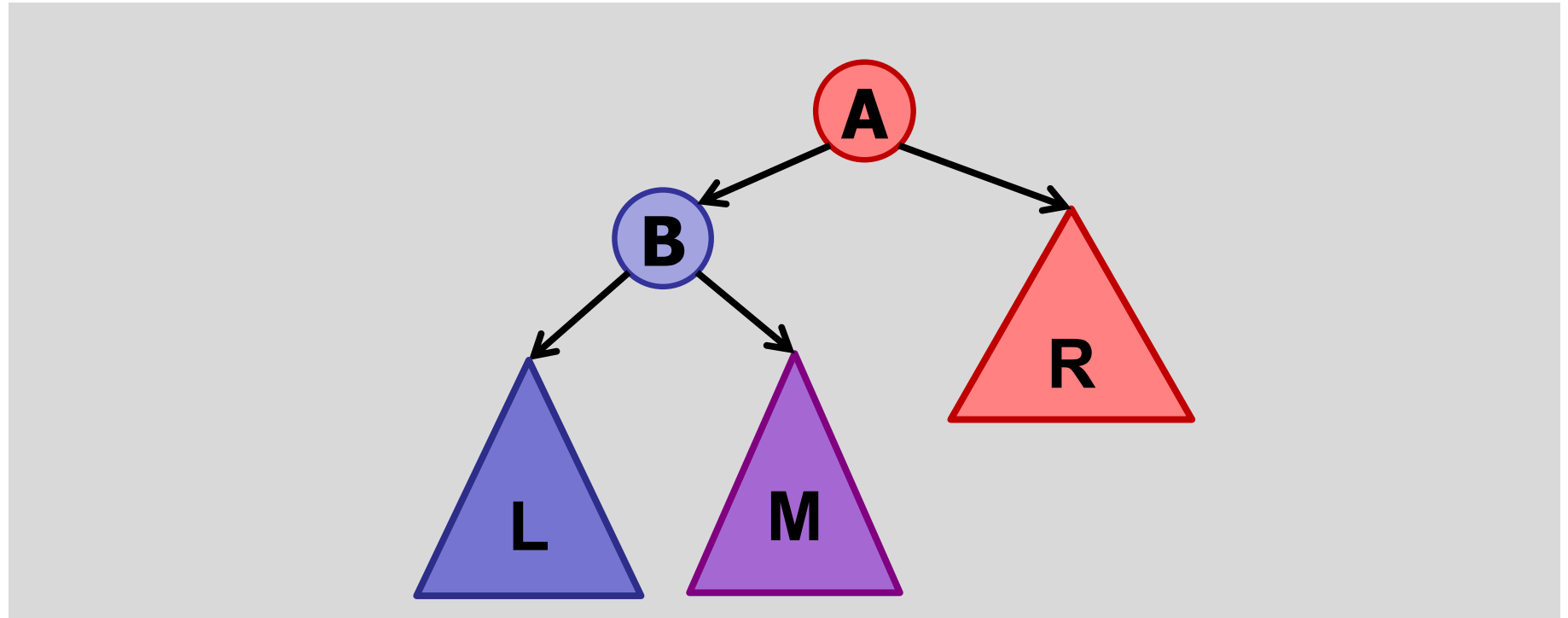


Use tree rotations to restore balance.

After insert, start at bottom, work your way up.

# Tree Rotations

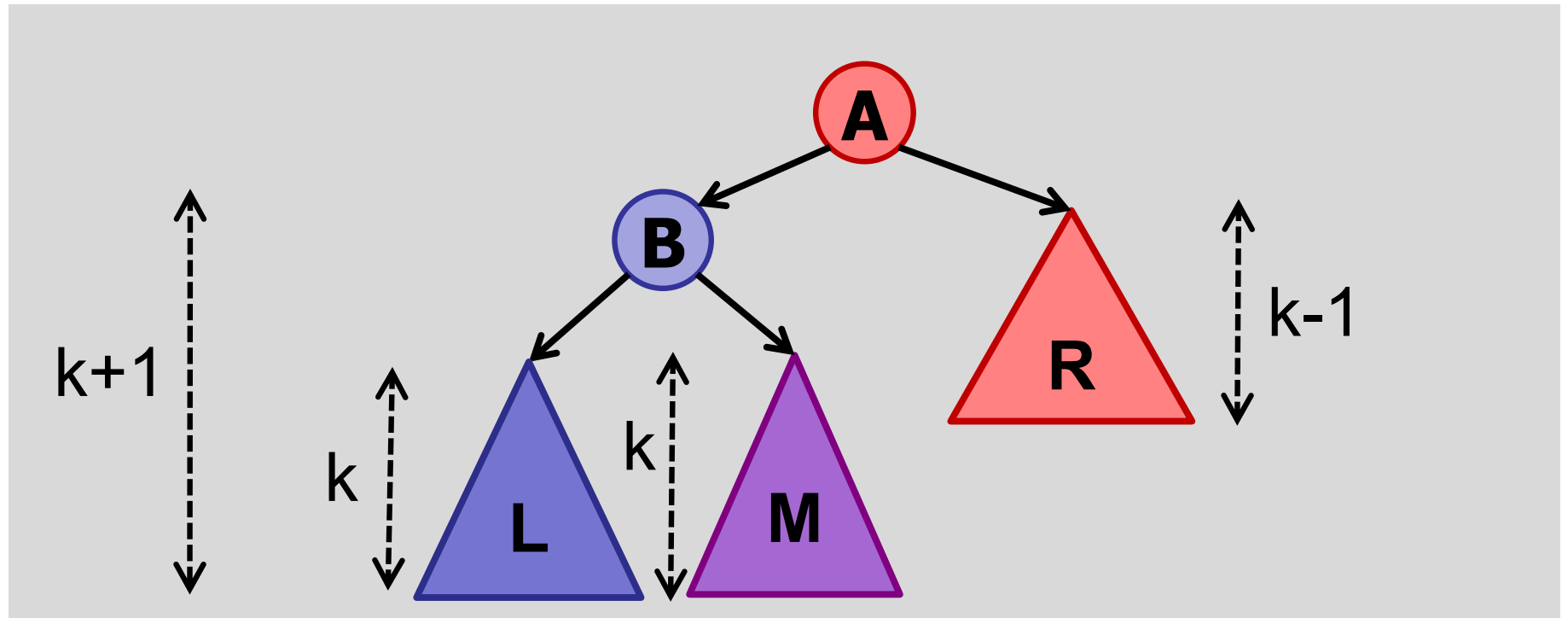
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Assume **A** is the lowest node in the tree violating balance property.

Assume A is **LEFT-heavy**.

# Tree Rotations (Left Heavy)

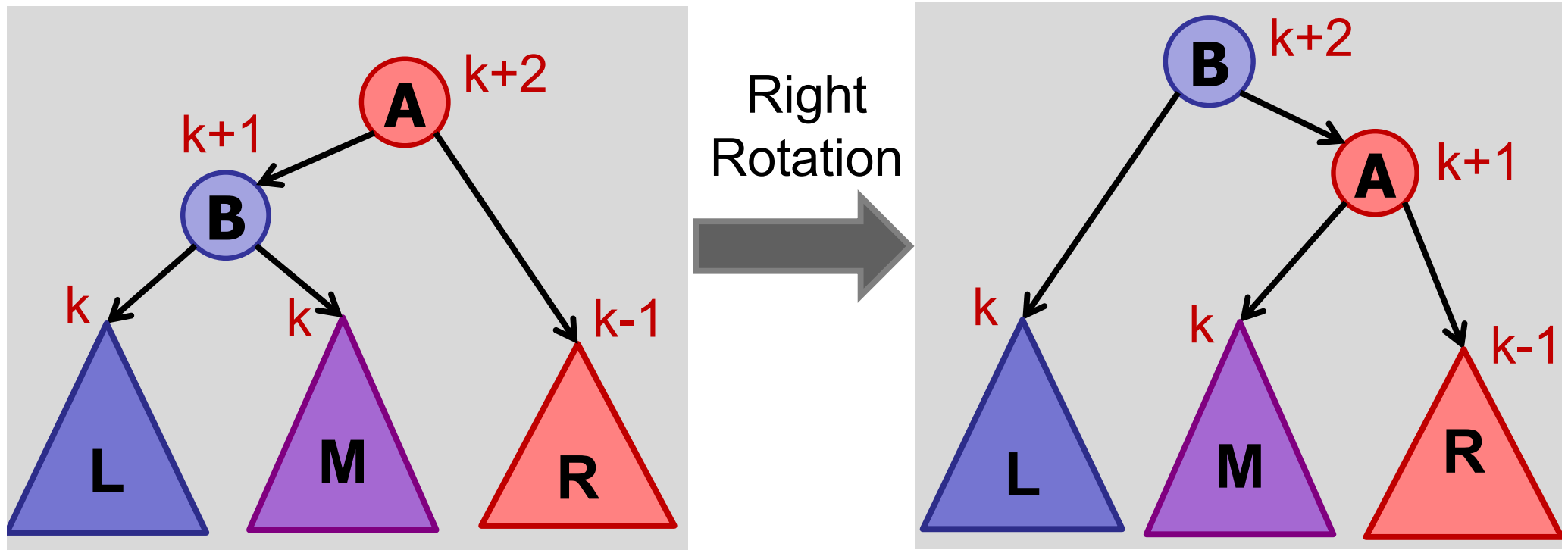


Assume **A** is the lowest node in the tree violating balance property.

Case 1: **B** is equi-height :  $h(\text{L}) = h(\text{M})$

$$h(\text{R}) = h(\text{M}) - 1$$

# Tree Rotations

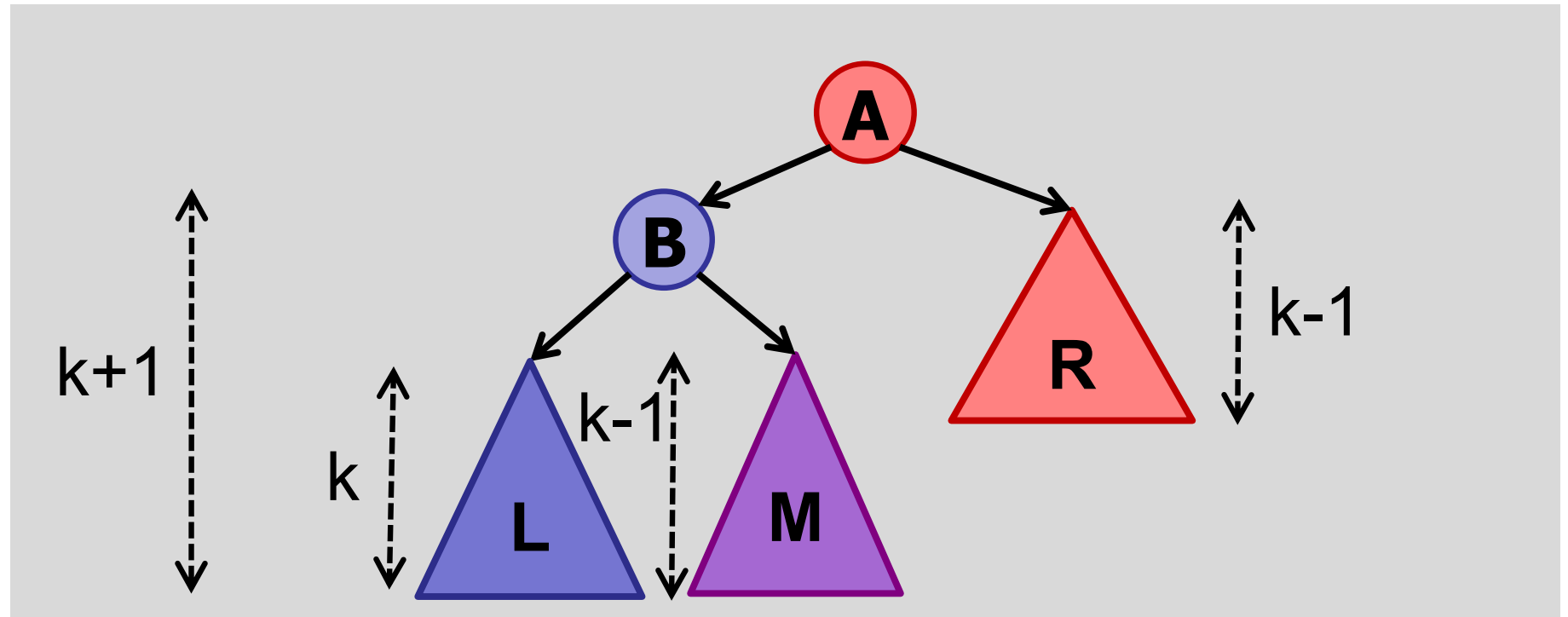


right-rotate:

Case 1: **B** is equi-height :  $h(\mathbf{L}) = h(\mathbf{M})$

$$h(\mathbf{R}) = h(\mathbf{M}) - 1$$

# Tree Rotations (Left Heavy)

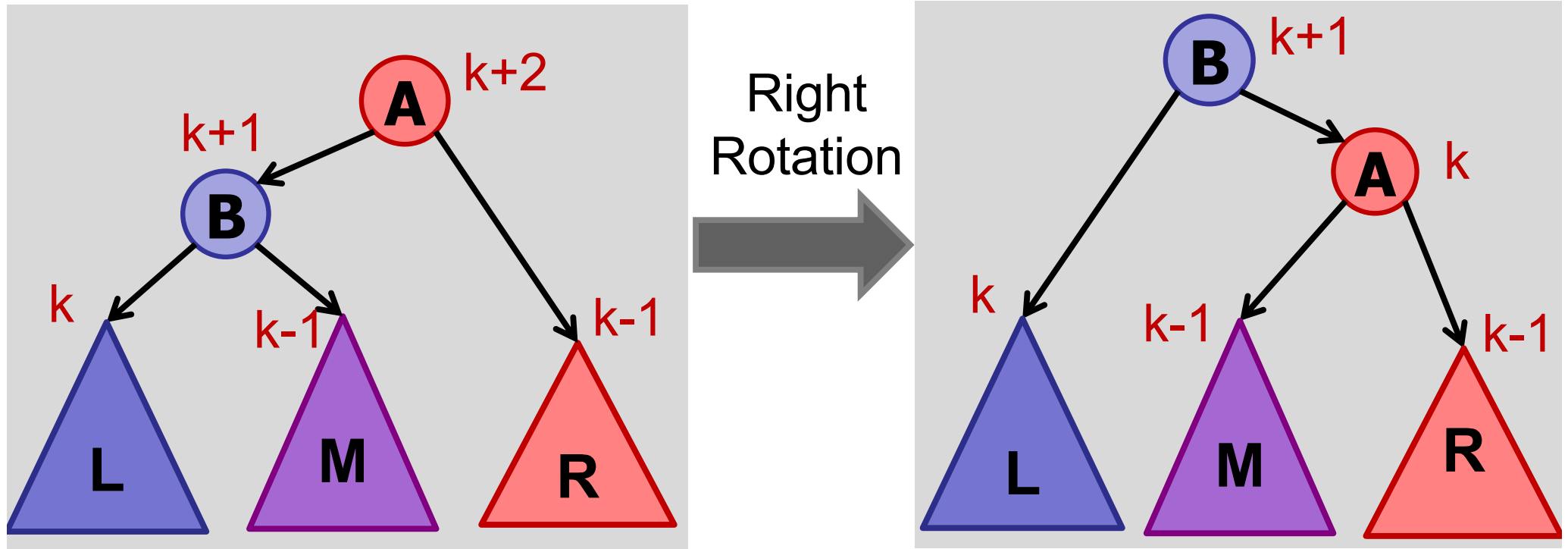


Assume **A** is the lowest node in the tree violating balance property.

Case 2: **B** is left-heavy :  $h(\mathbf{L}) = h(\mathbf{M}) + 1$

$$h(\mathbf{R}) = h(\mathbf{M})$$

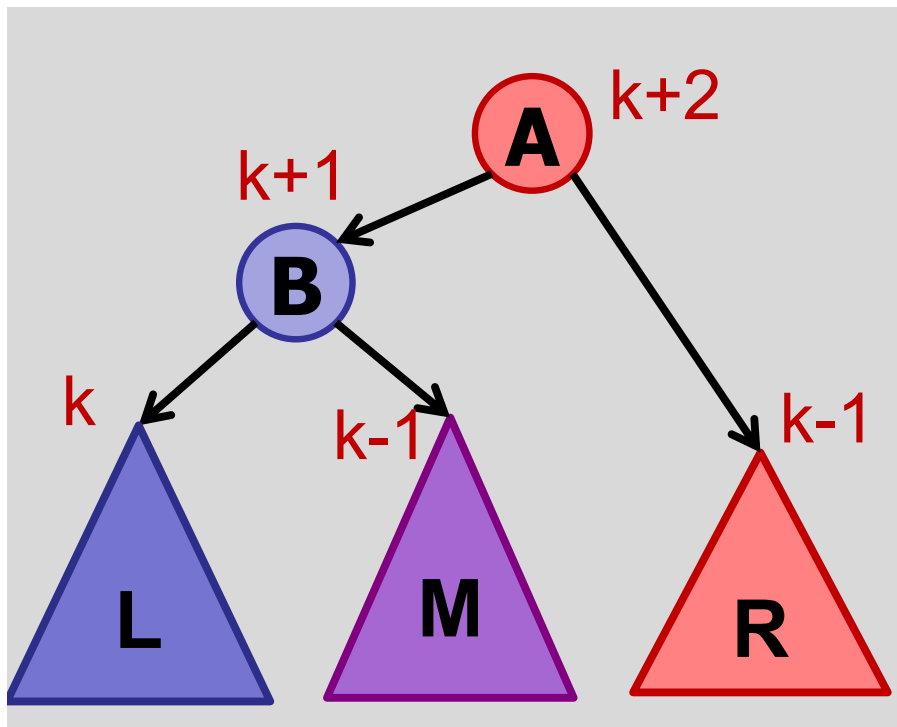
# Tree Rotations



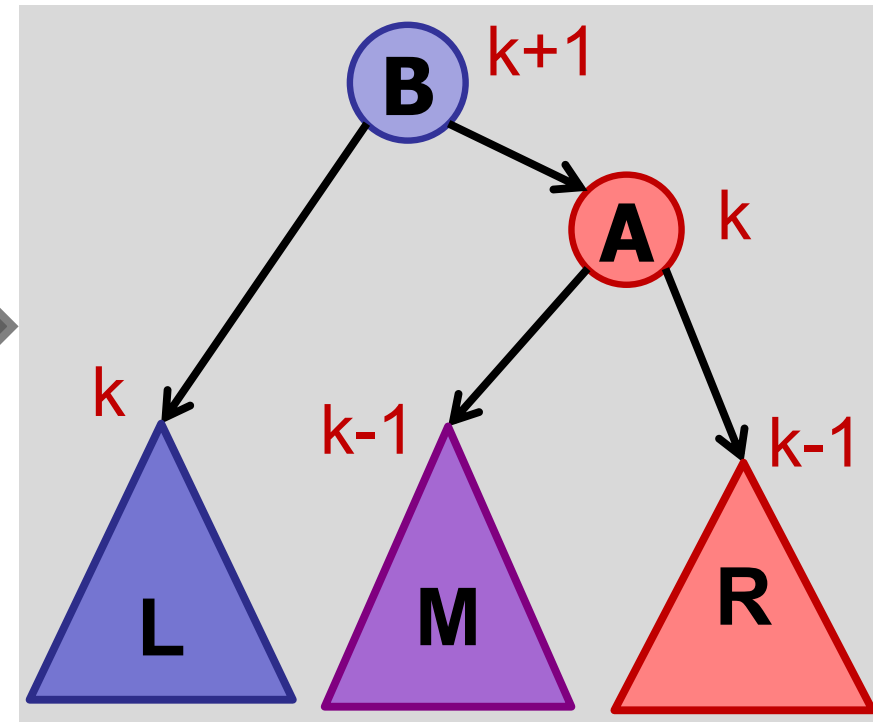
right-rotate:

Case 2: **B** is left-heavy:  $h(\mathbf{L}) = h(\mathbf{M}) + 1$

$$h(\mathbf{R}) = h(\mathbf{M})$$

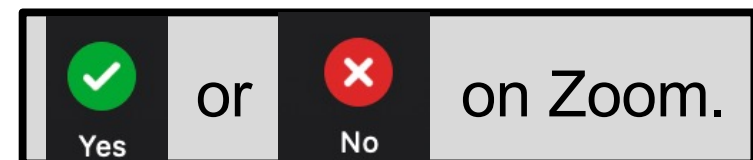


Right  
Rotation



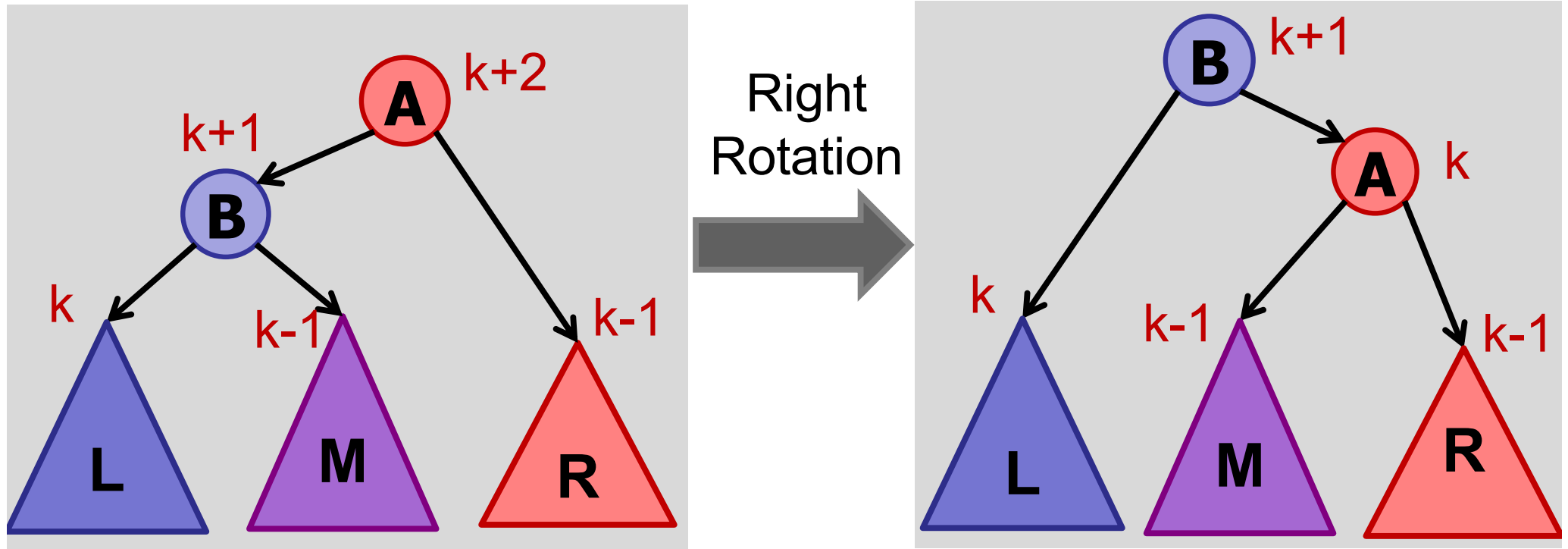
Is it balanced?

- ✓ 1. Yes.
- 2. No.
- 3. Maybe.





# Tree Rotations

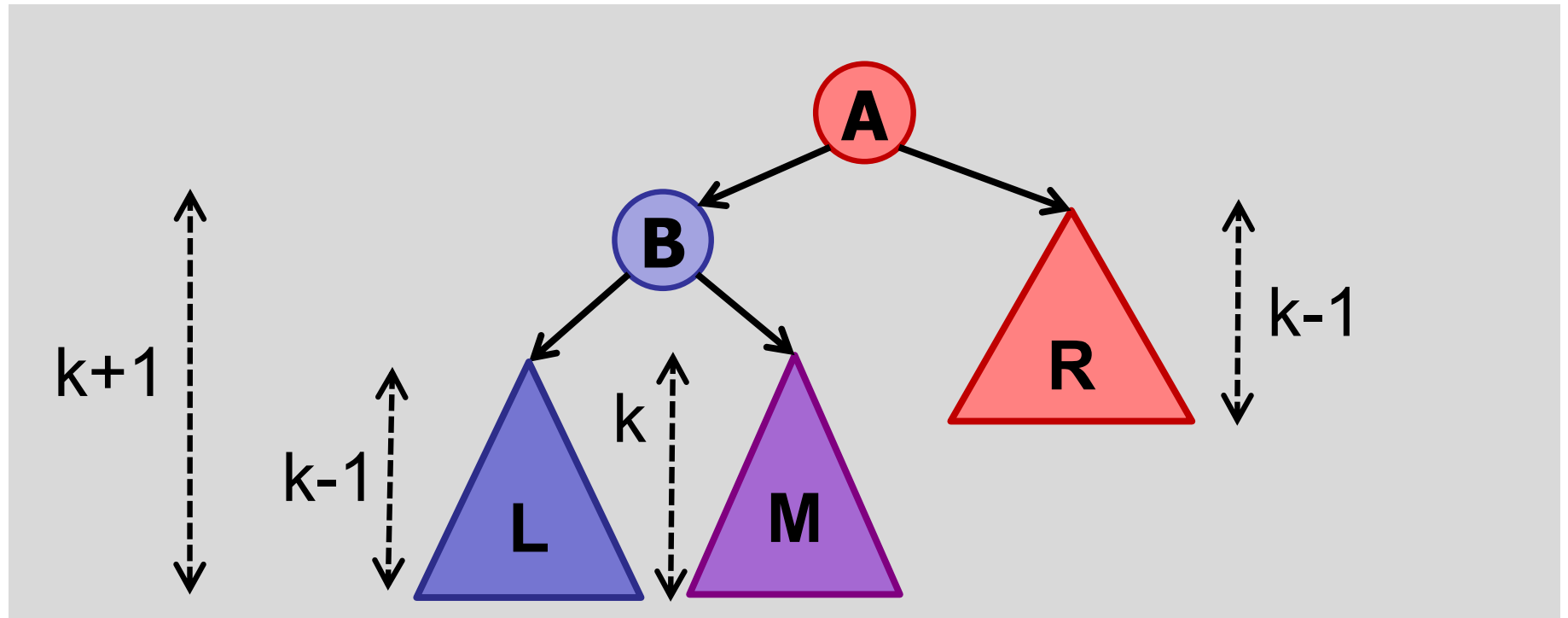


right-rotate:

Case 2: **B** is left-heavy:  $h(\mathbf{L}) = h(\mathbf{M}) + 1$

$$h(\mathbf{R}) = h(\mathbf{M})$$

# Tree Rotations (Left Heavy)

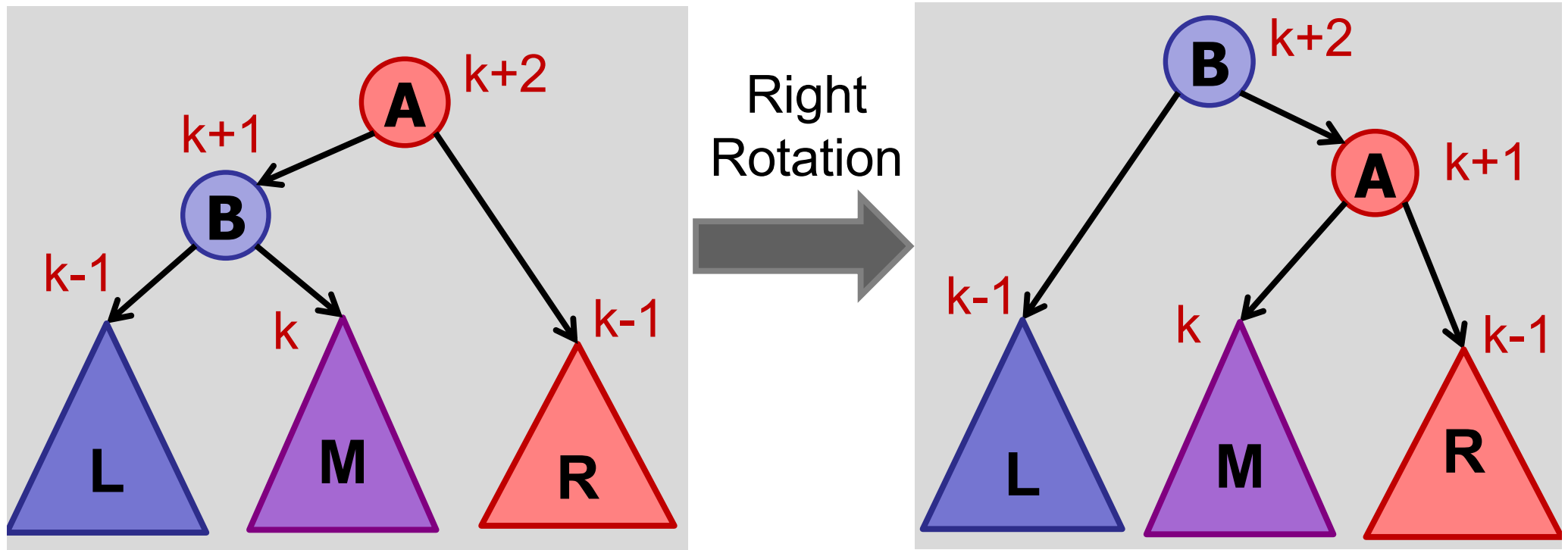


Assume **A** is the lowest node in the tree violating balance property.

Case 3: **B** is right-heavy :  $h(\text{L}) = h(\text{M}) - 1$

$$h(\text{R}) = h(\text{M})$$

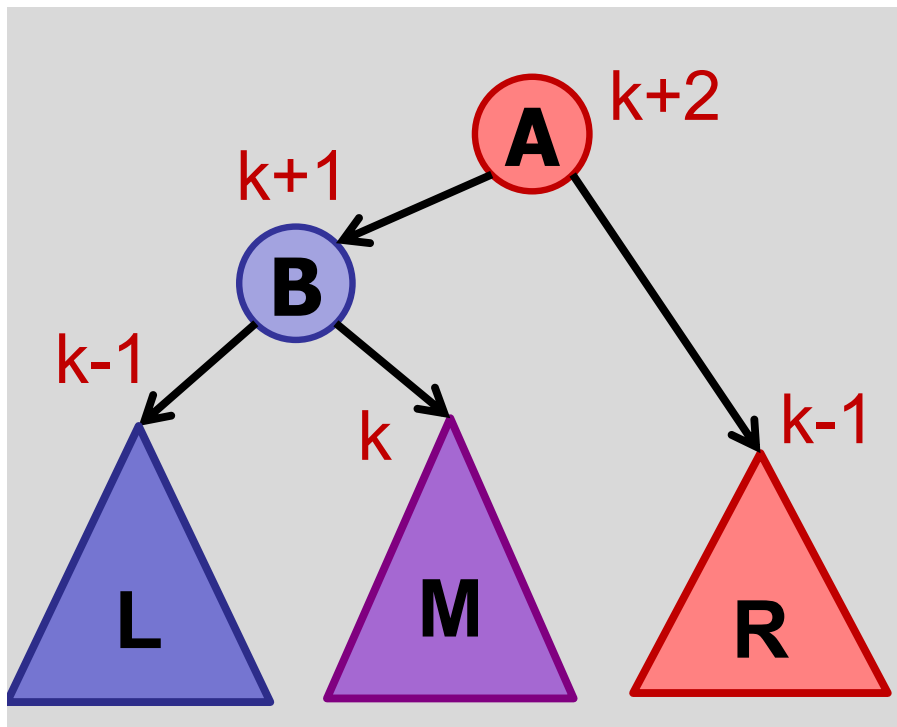
# Tree Rotations



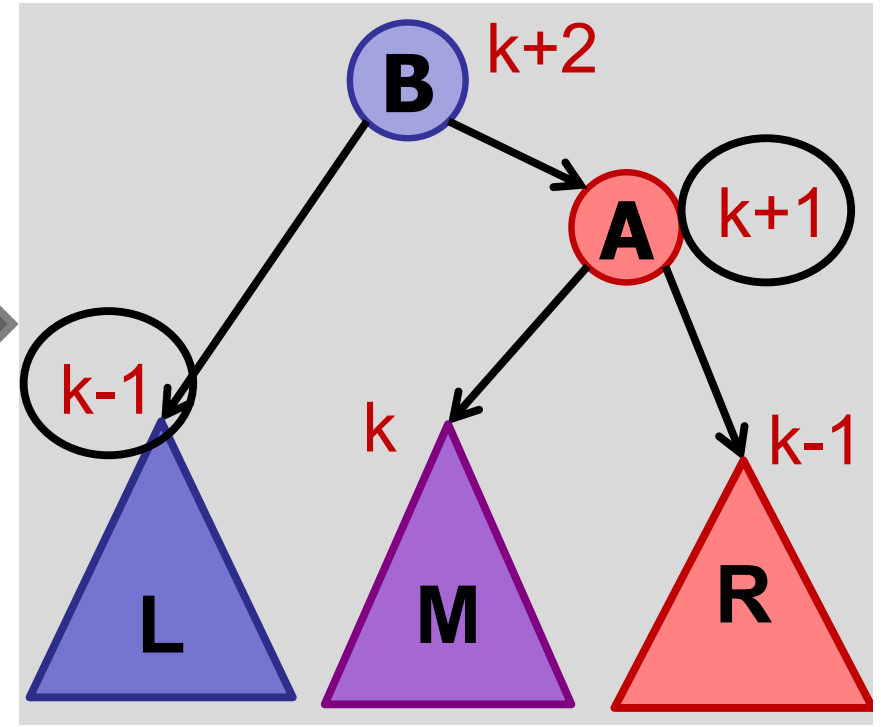
right-rotate:

Case 3: **B** is right-heavy:  $h(\mathbf{L}) = h(\mathbf{M}) - 1$

$$h(\mathbf{R}) = h(\mathbf{L})$$

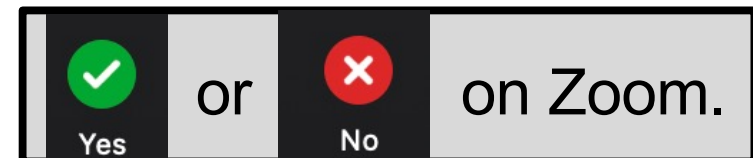


Right  
Rotation

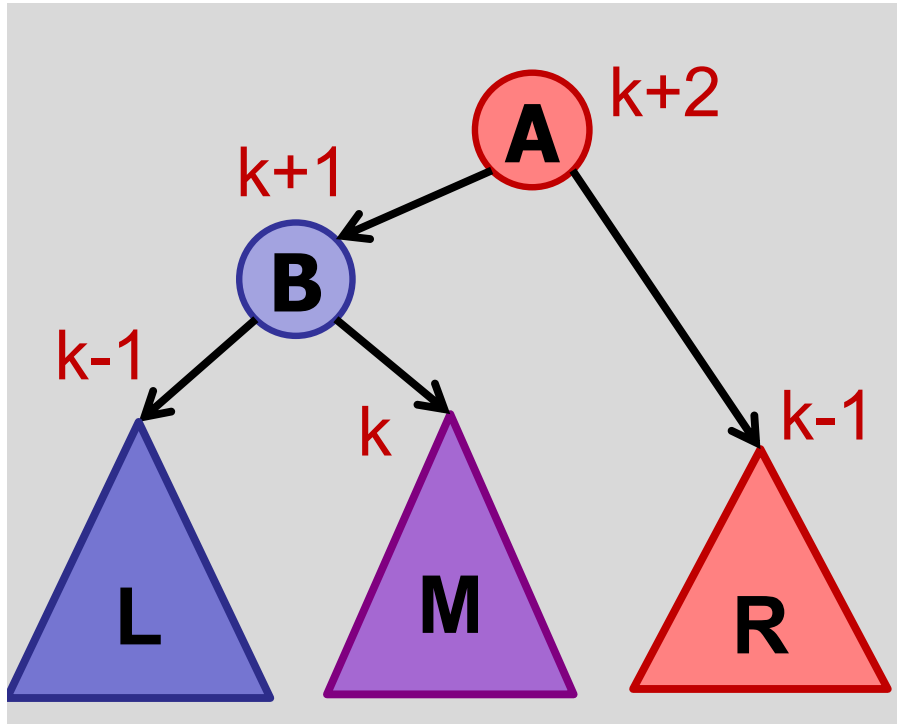


Is it balanced?

1. Yes.
- ✓ 2. No.
3. Maybe.



# Tree Rotations



Let's do something  
first before we  
right-rotate(A)

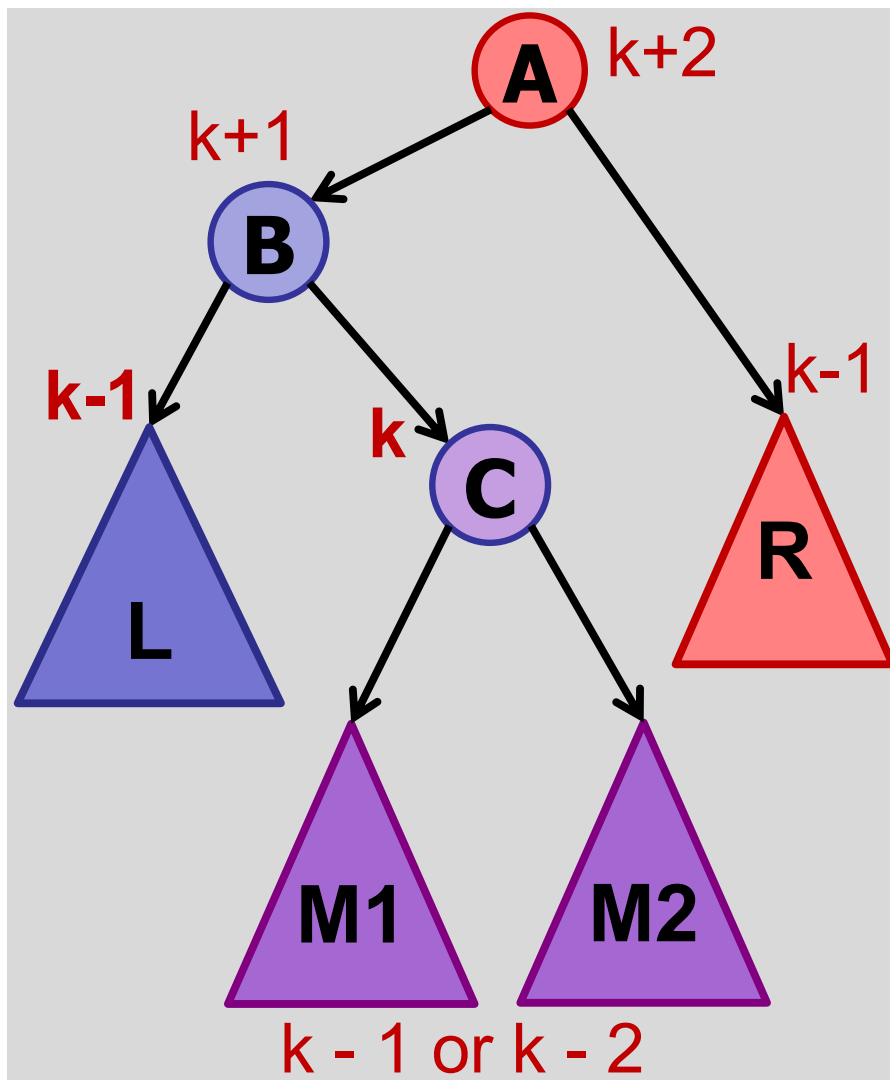
(Reduce it to a  
problem we have  
already solved!)

right-rotate:

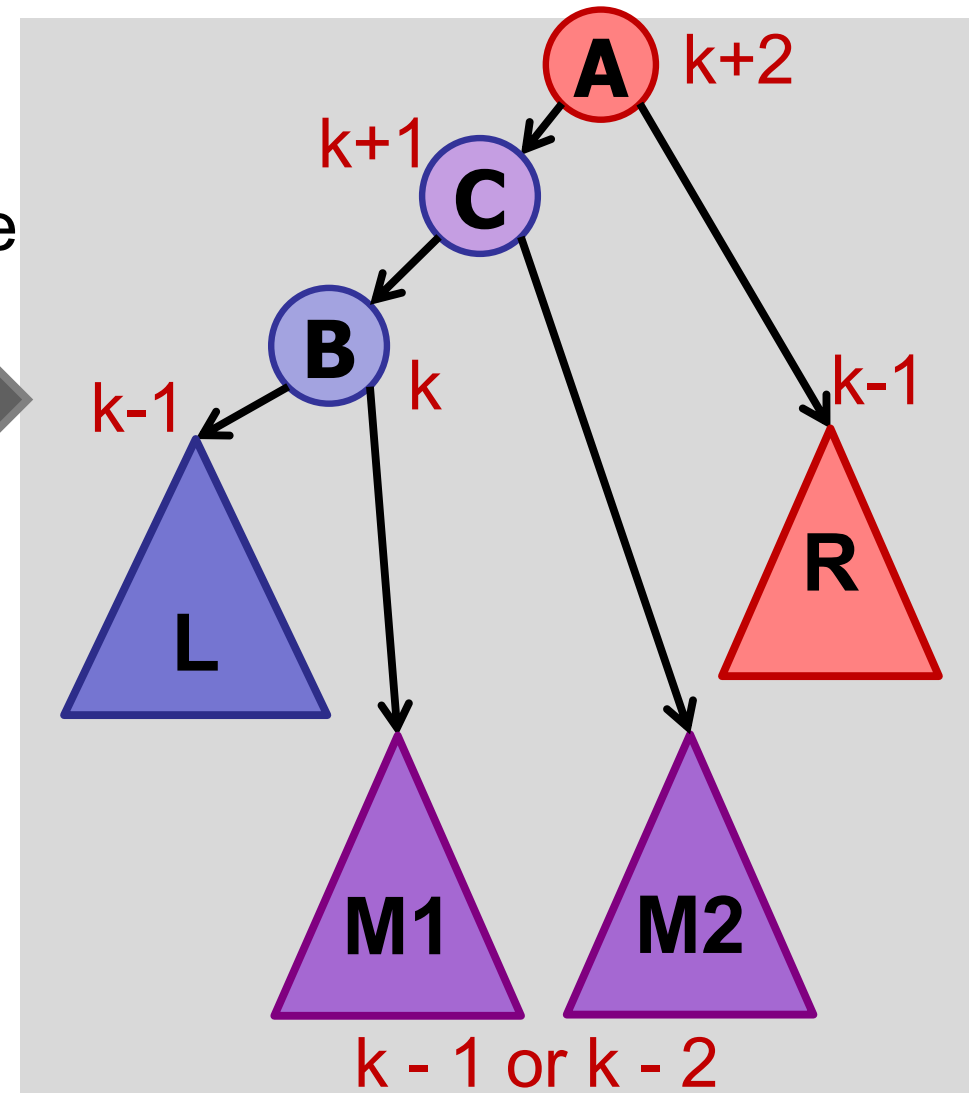
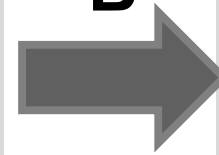
Case 3: **B** is right-heavy:  $h(\text{L}) = h(\text{M}) - 1$

$$h(\text{R}) = h(\text{L})$$

# Tree Rotations



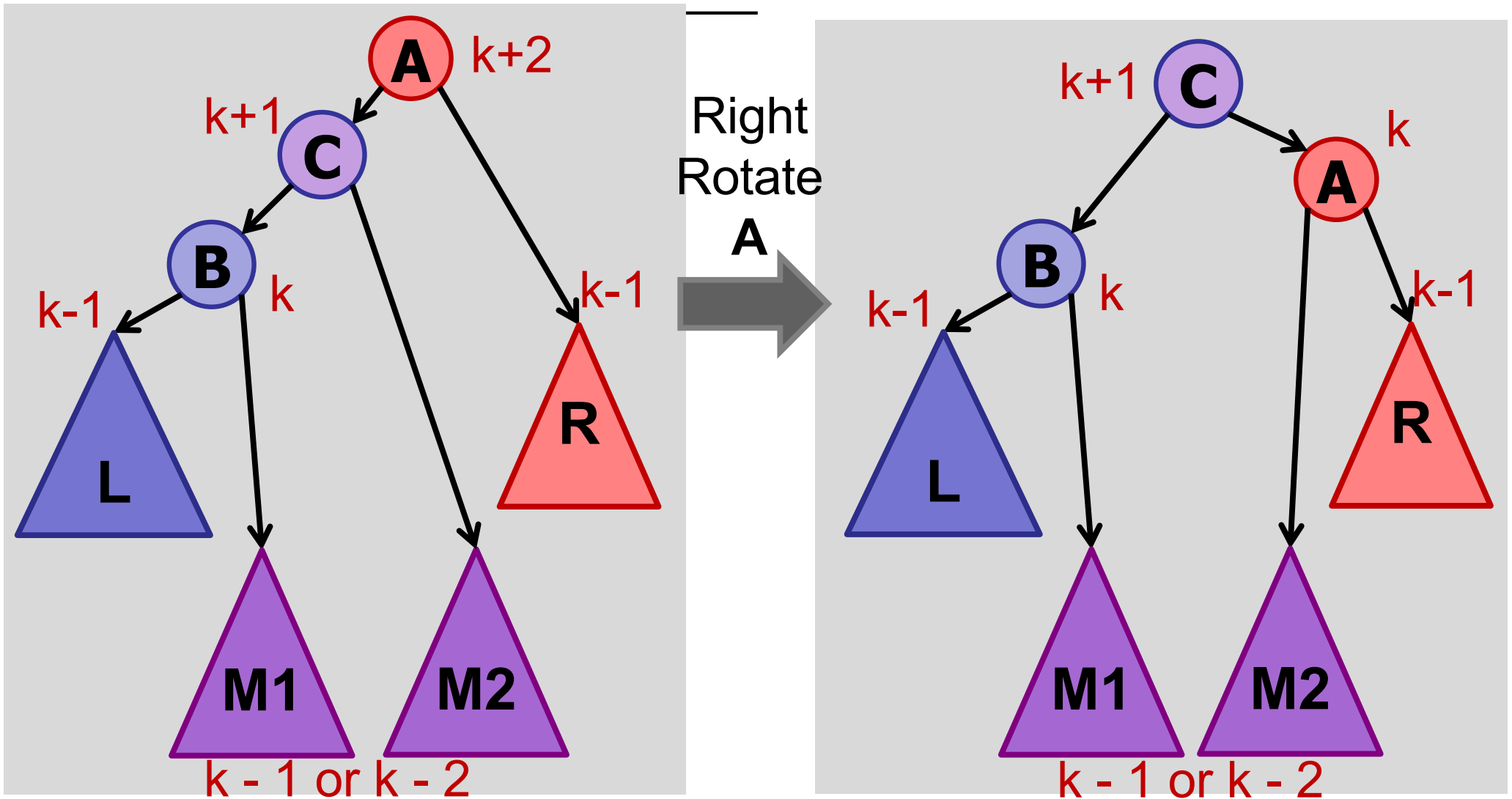
Left  
Rotate  
**B**

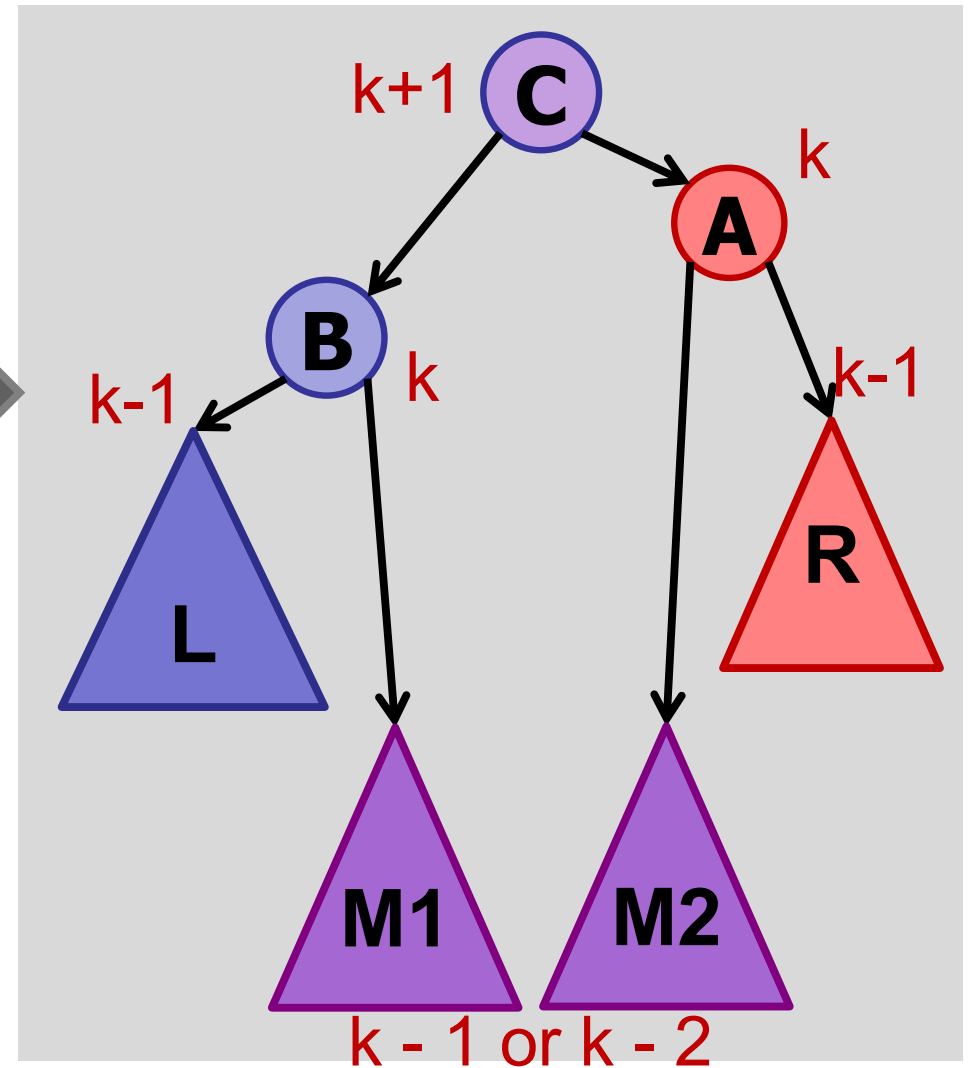
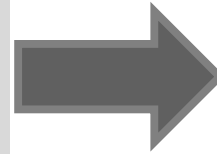
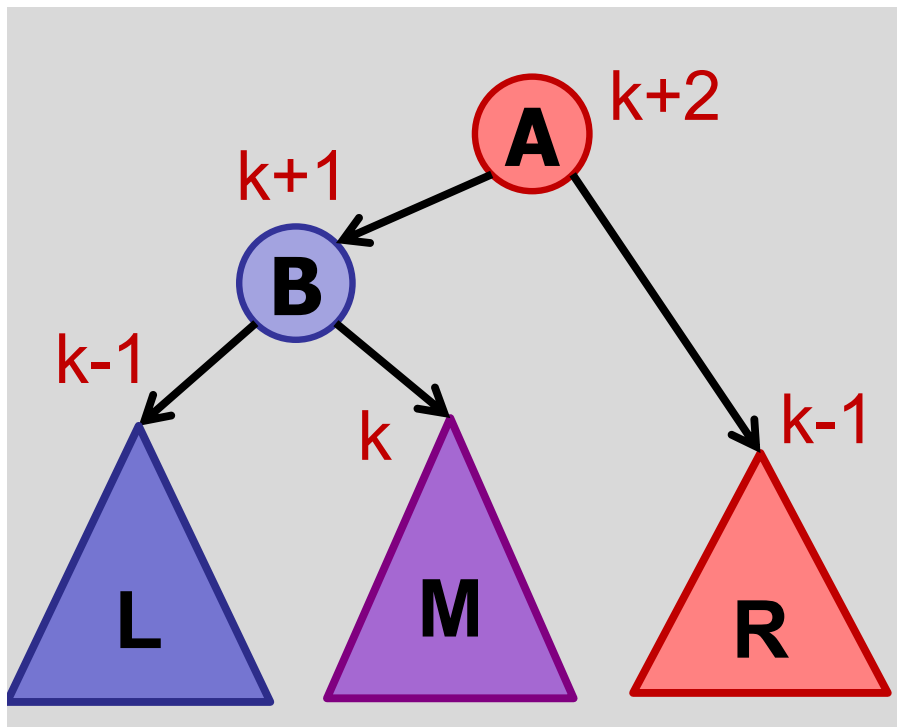


Left-rotate B

After left-rotate B: **A** and **C** still out of balance.

# Tree Rotations

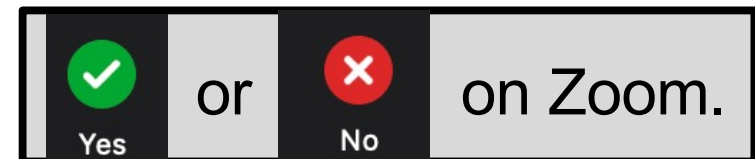




Double Rotate

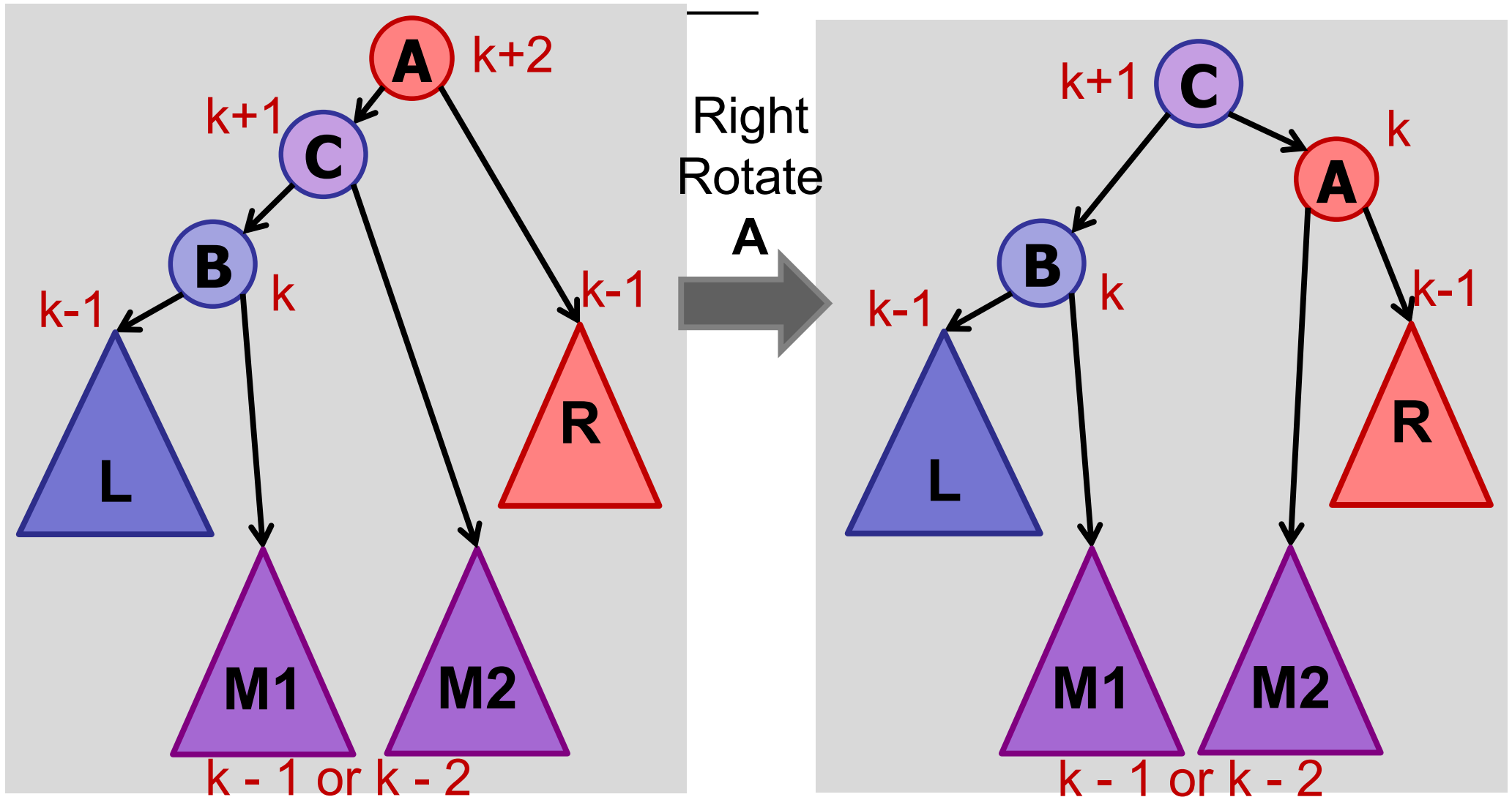
Is it balanced?

- ✓ 1. Yes.
- 2. No.
- 3. Maybe.





# Tree Rotations



After right-rotate A: all in balance.

# Rotations

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## Summary:

If  $v$  is out of balance and left heavy:

1.  $v.left$  is balanced:  $right\text{-}rotate(v)$
2.  $v.left$  is left-heavy:  $right\text{-}rotate(v)$
3.  $v.left$  is right-heavy:  $left\text{-}rotate(v.left)$   
 $right\text{-}rotate(v)$

If  $v$  is out of balance and right heavy:

Symmetric three cases....

How many rotations do you need after an insertion (in the worst case)?

1. 1
2. 2
3. 4
4.  $\log(n)$
5.  $2\log(n)$
6.  $n$

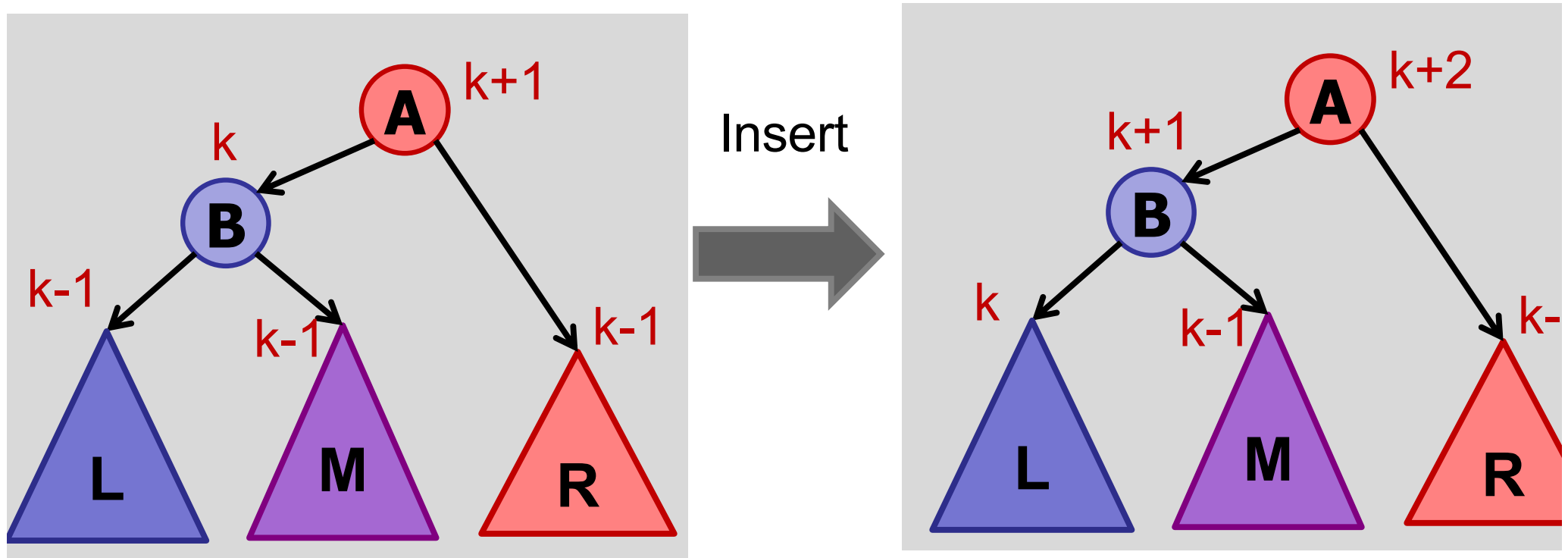
How many rotations do you need after an insertion (in the worst case)?

- 1. 1
- ✓ 2. 2
- 3. 4
- 4.  $\log(n)$
- 5.  $2\log(n)$
- 6.  $n$

Question:

Why isn't it  $2\log(n)$ ?

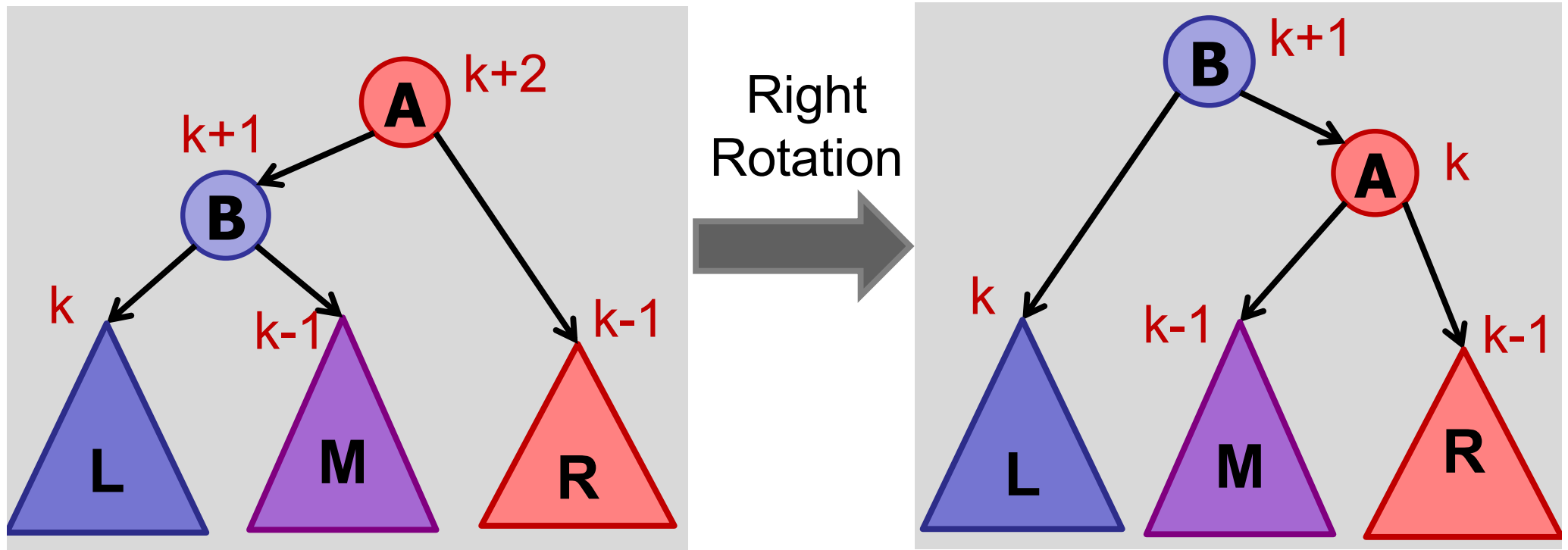
# How many rotations?



Case 2: **B** is left-heavy

Insert increased heights by 1.

# How many rotations?

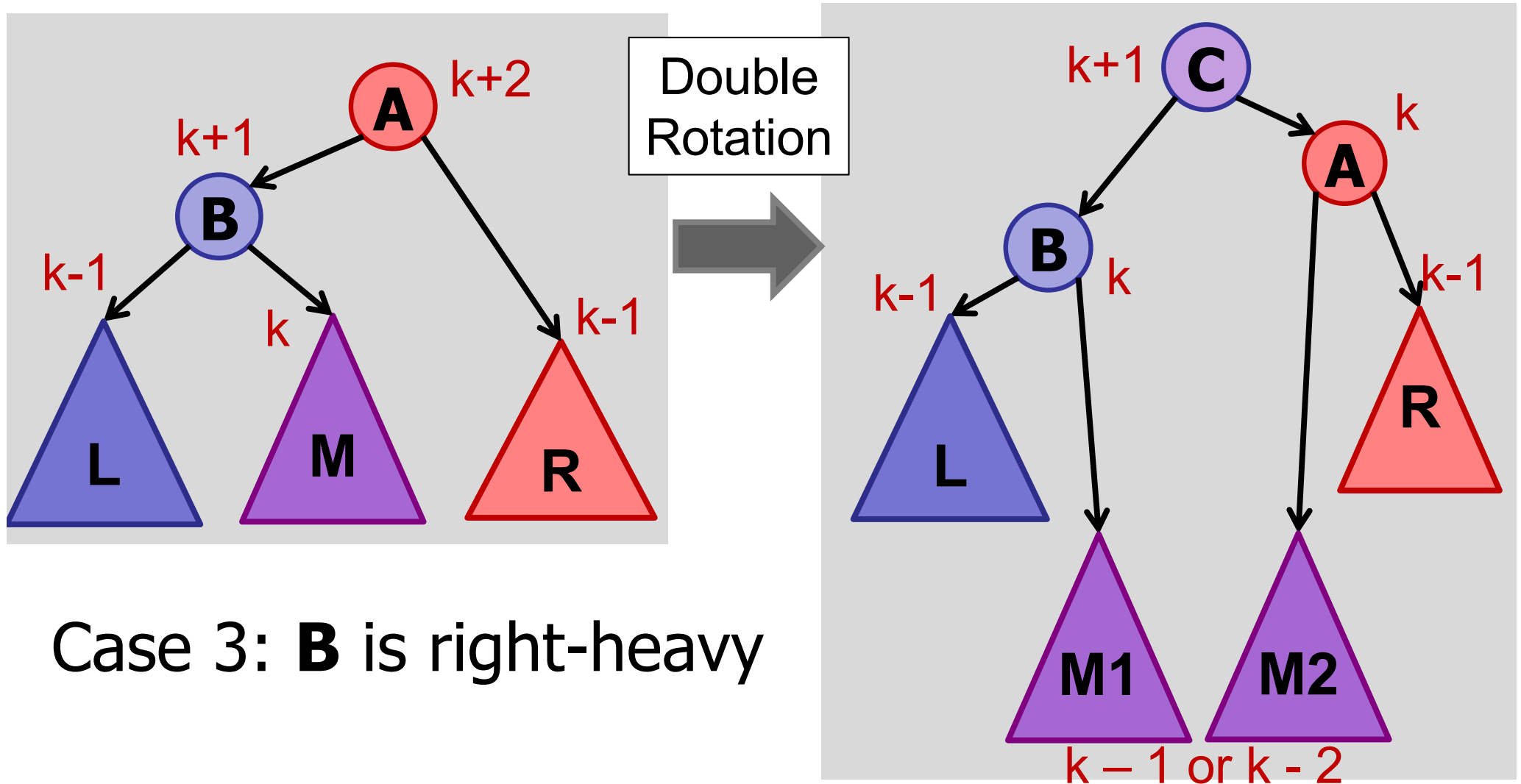


Case 2: **B** is left-heavy

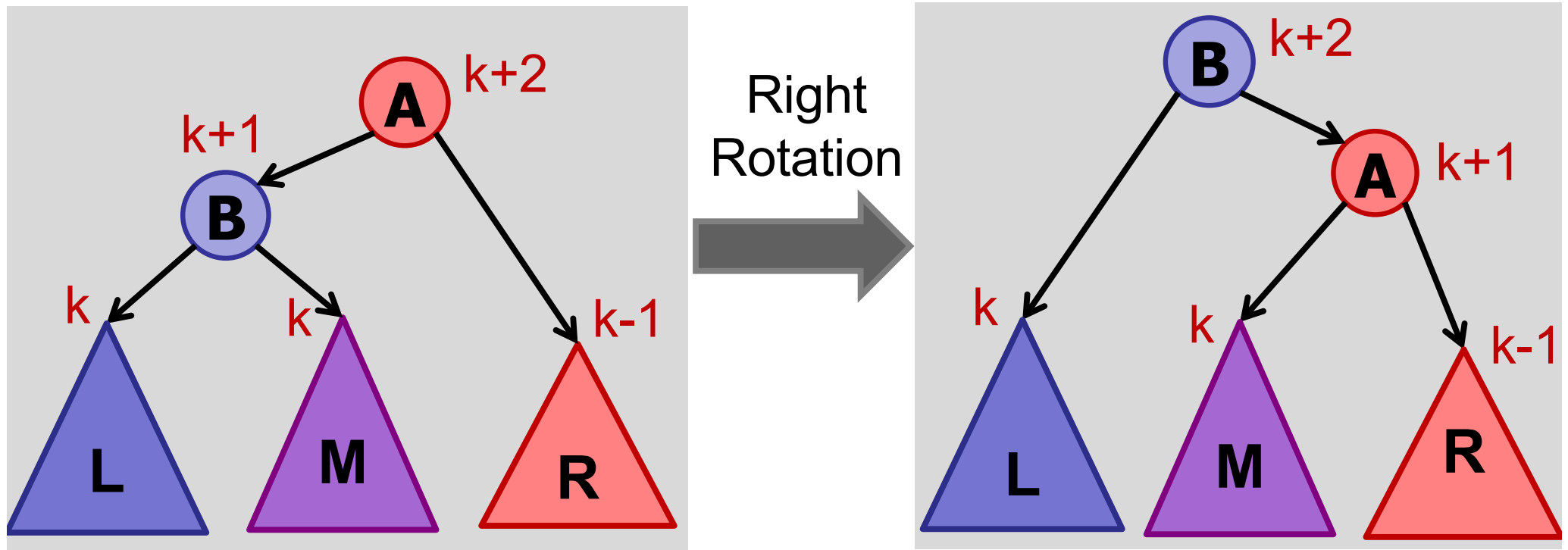
Rotation reduces root height by 1.

(Everything higher in tree is unchanged!)

# How many rotations?



# How many rotations?



Case 1: **B** is balanced

Rotation does *not* reduce height by 1.

Challenge: figure out why this is okay!



# Insert in AVL Tree

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## Summary:

- Insert key in BST.
- Walk up tree:
  - At every step, check for balance.
  - If out-of-balance, use rotations to rebalance and return.

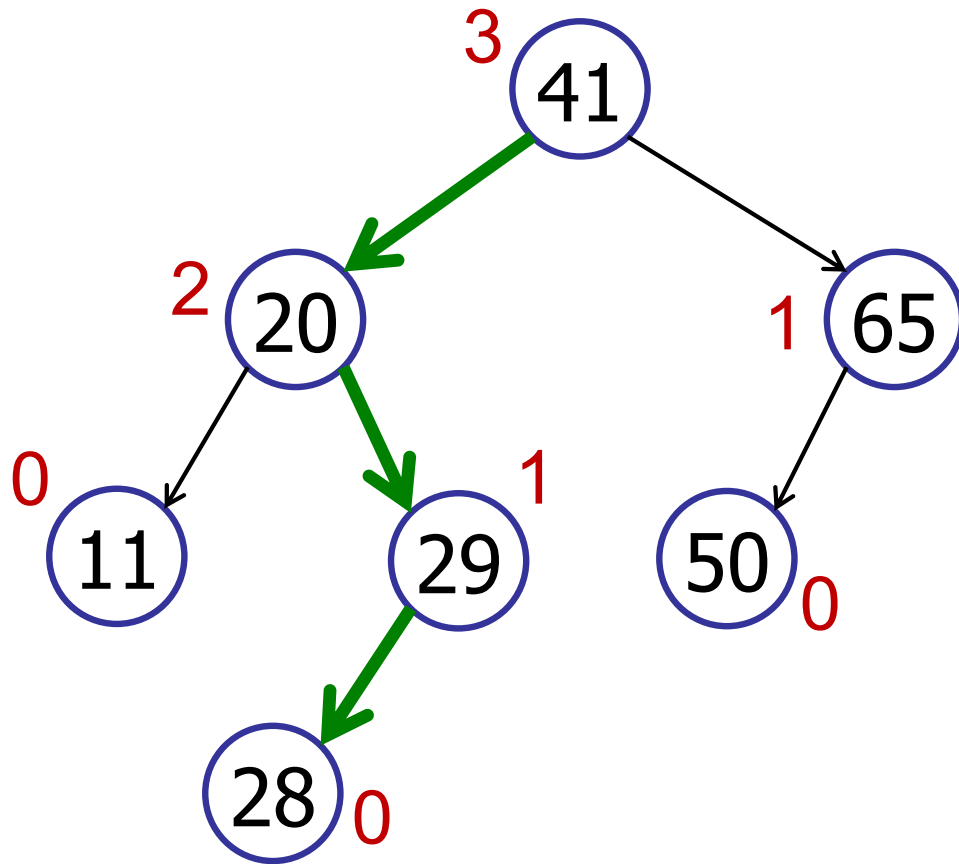
## Key observation:

- Only need to fix *lowest* out-of-balance node.
- Only need at most two rotations to fix.

# Example

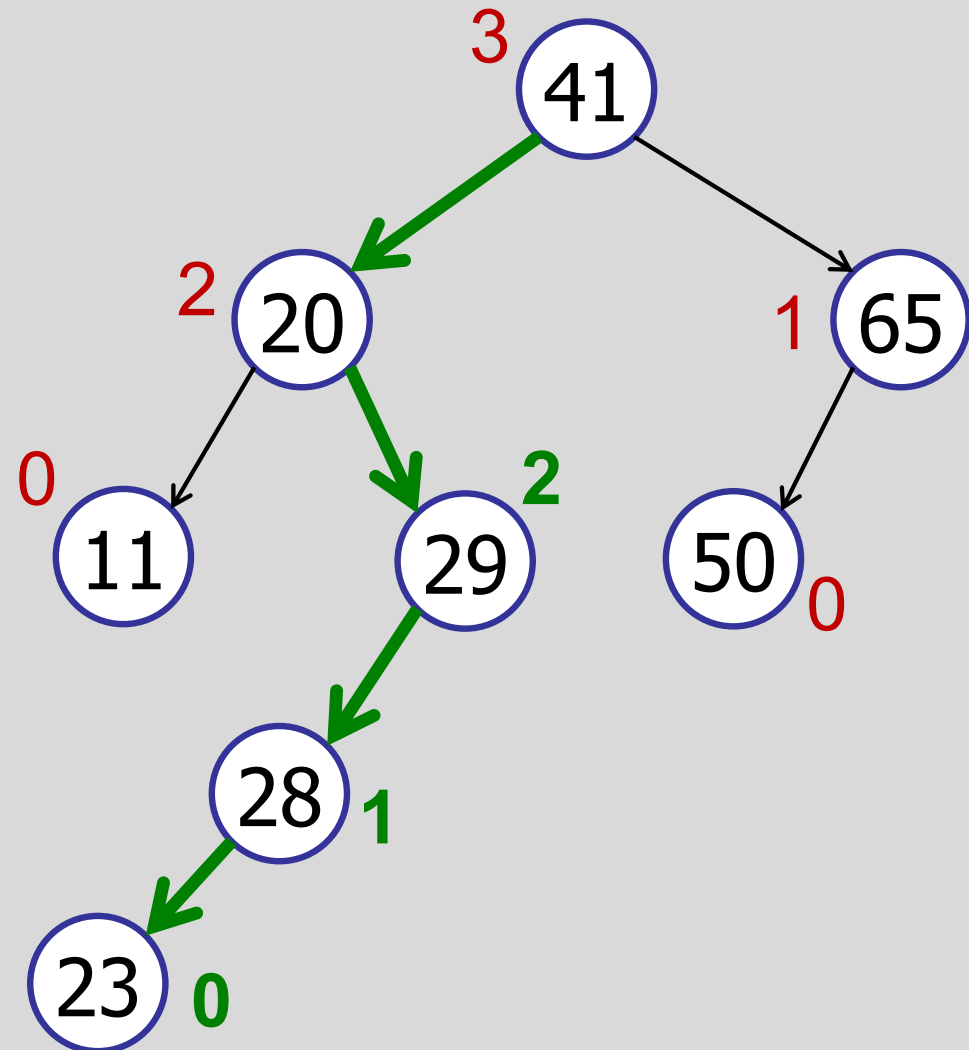
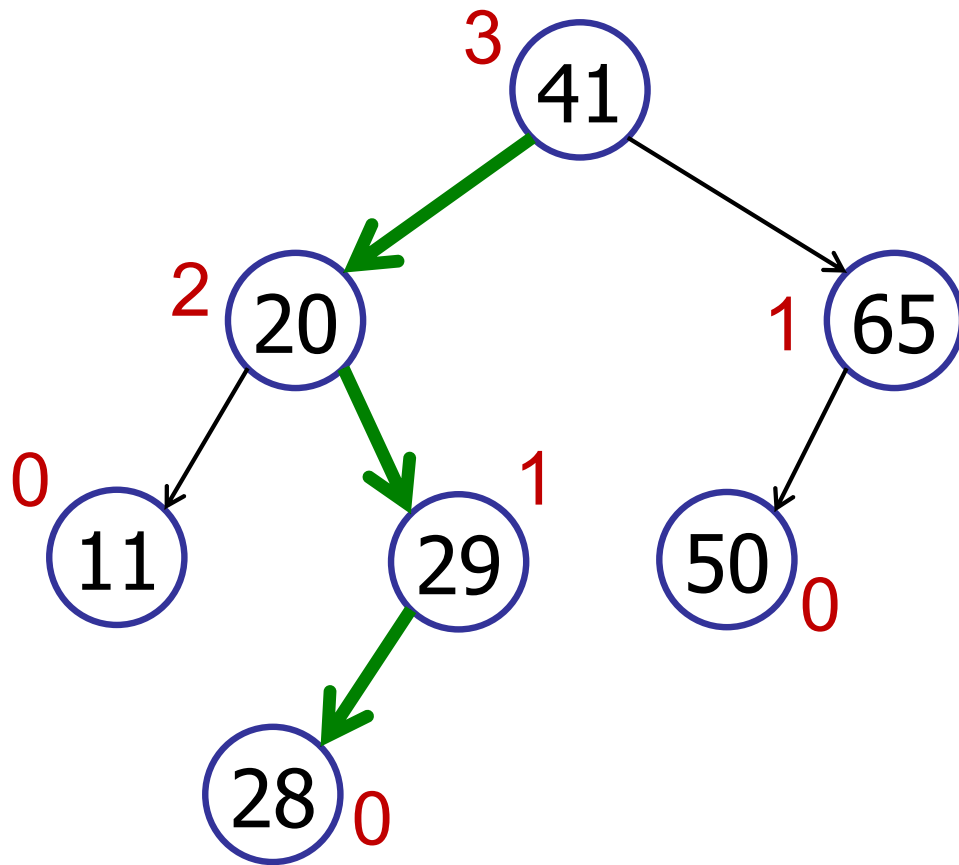
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insert(23)



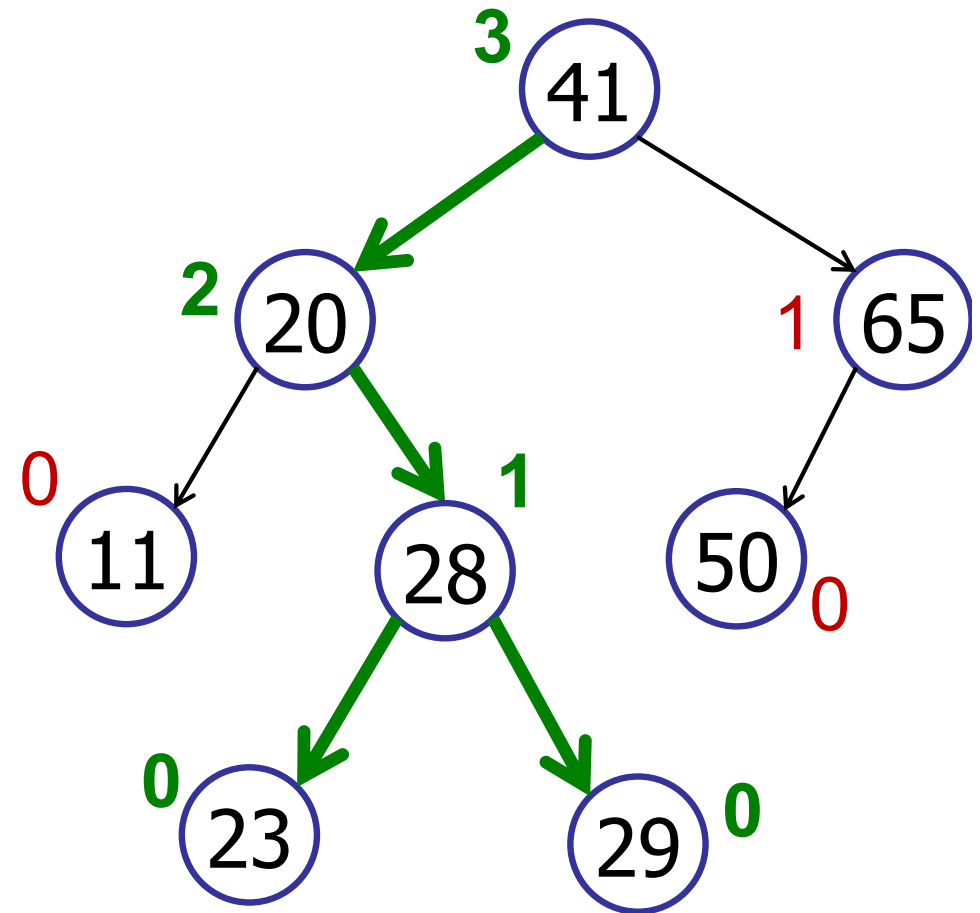
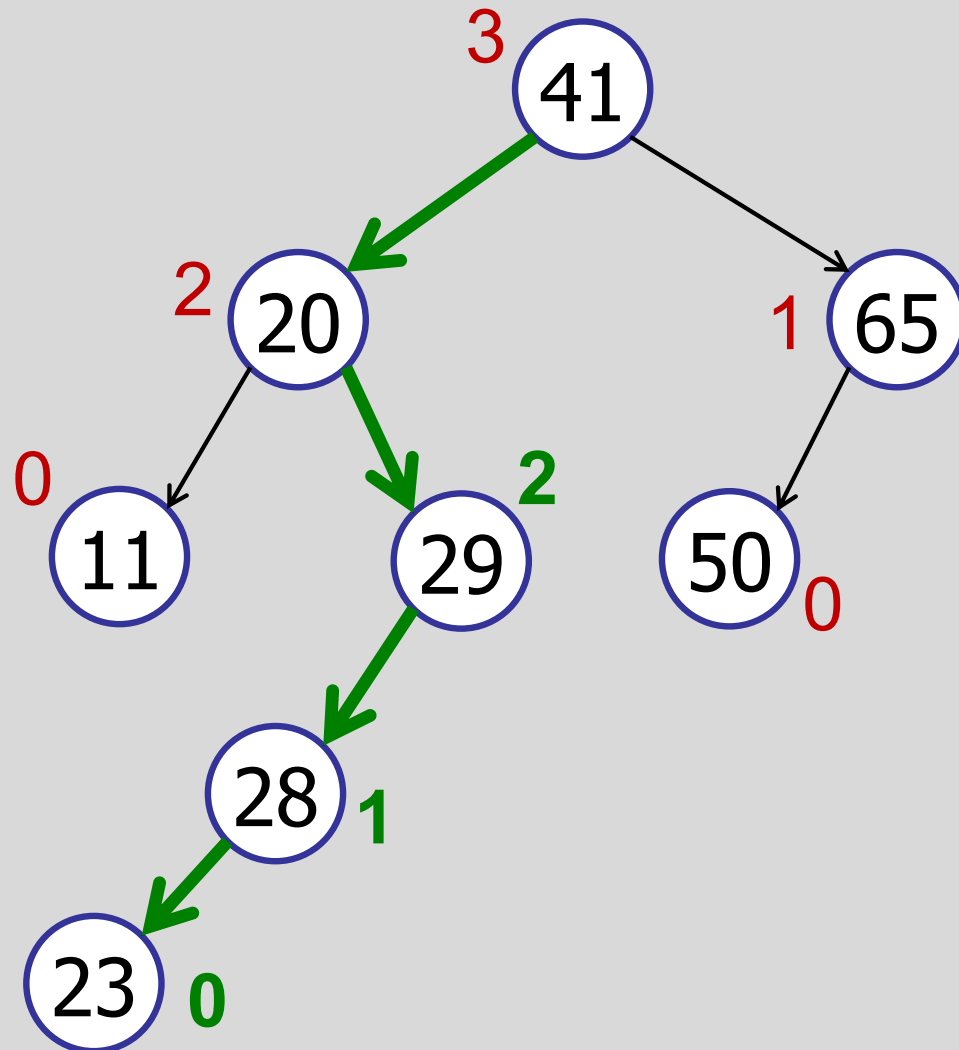
# Example

insert(23)



# Example

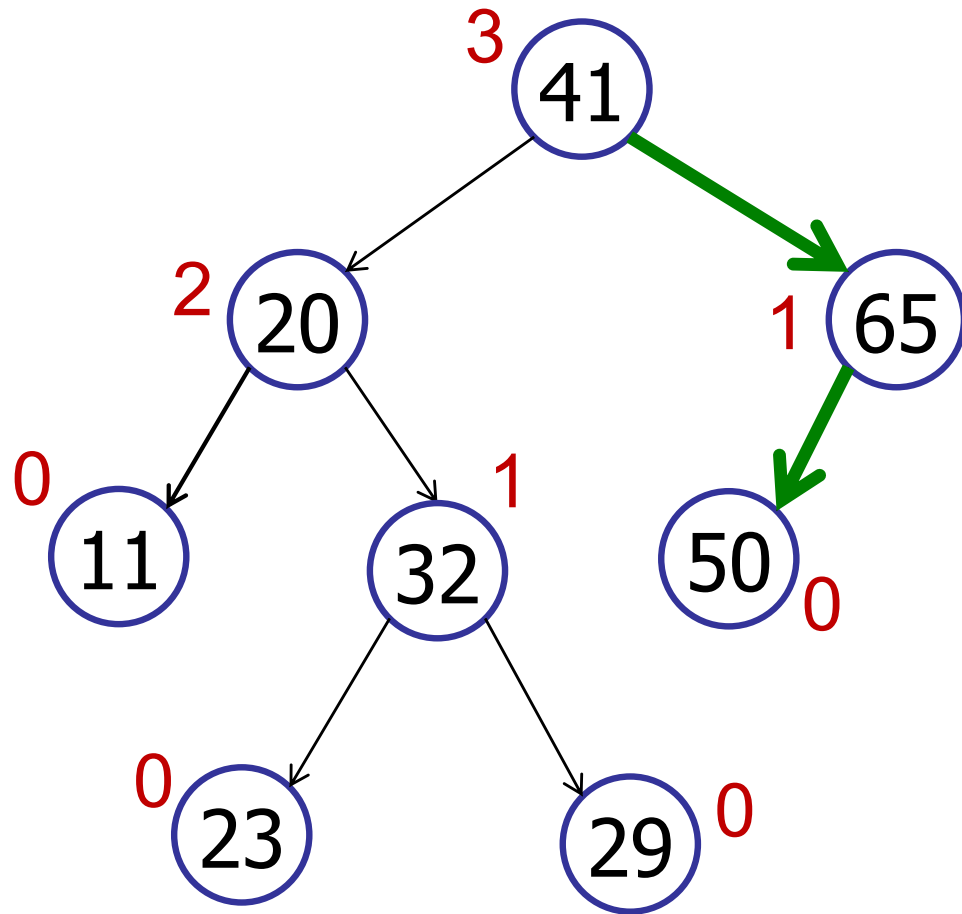
right-rotate(29)



# Example

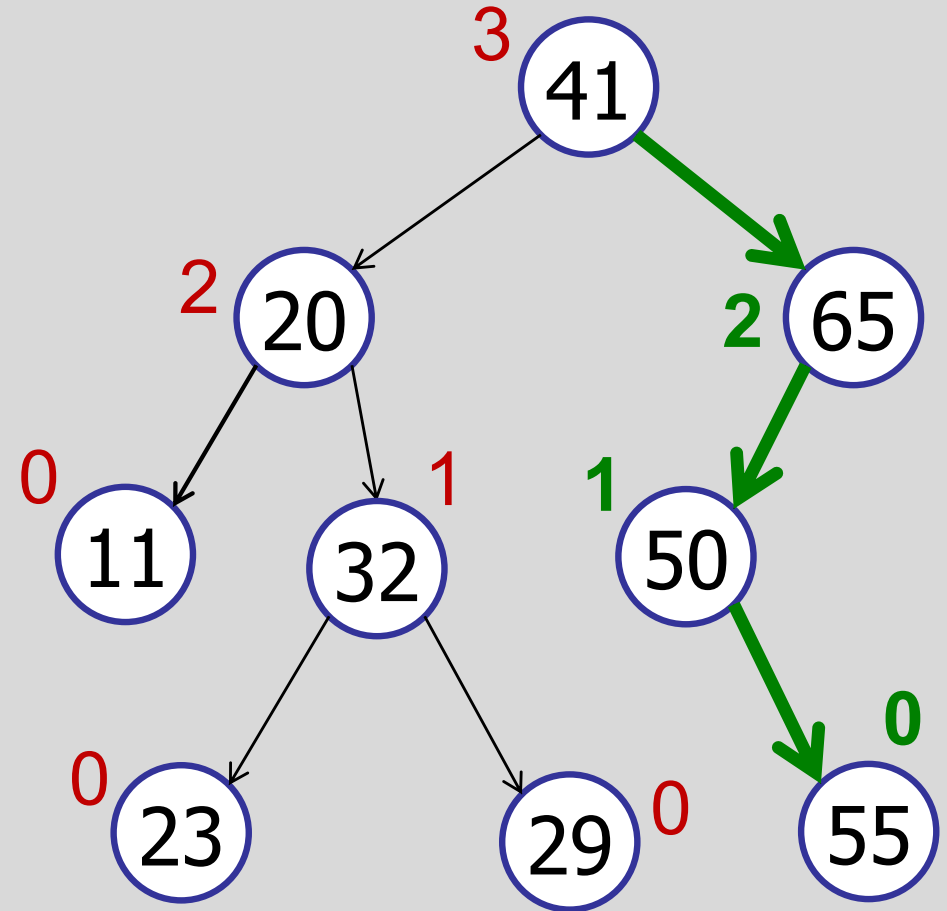
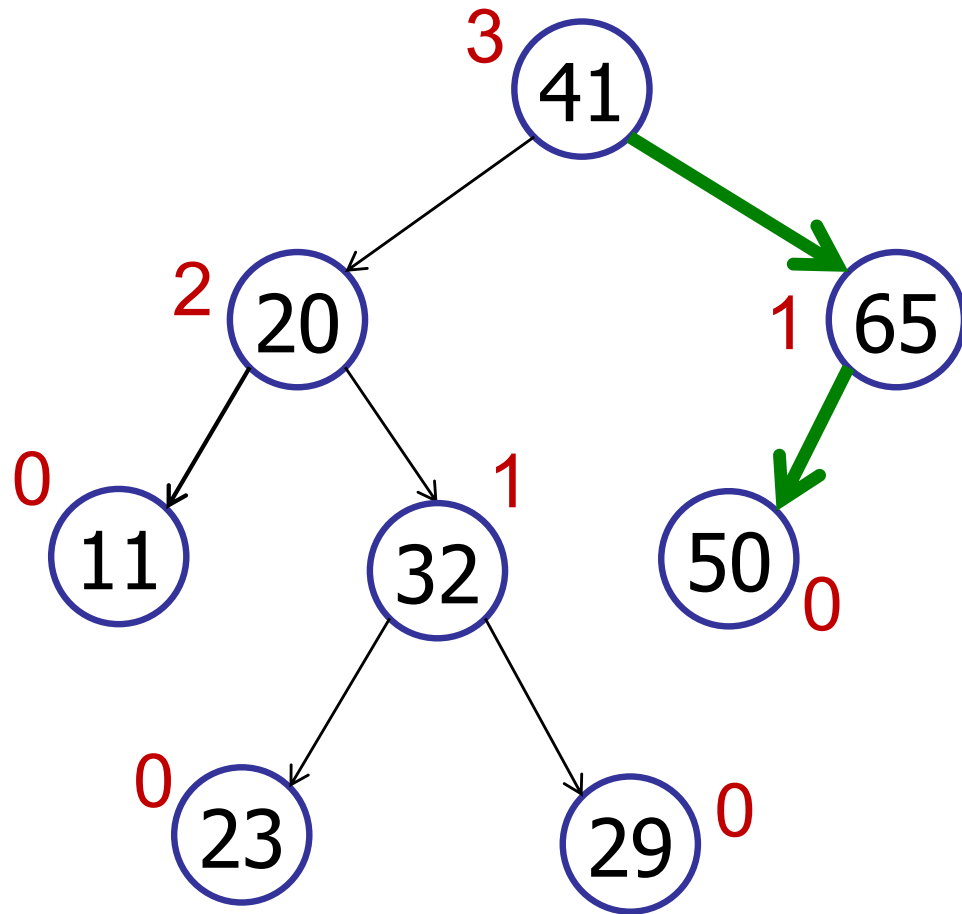
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insert(55)



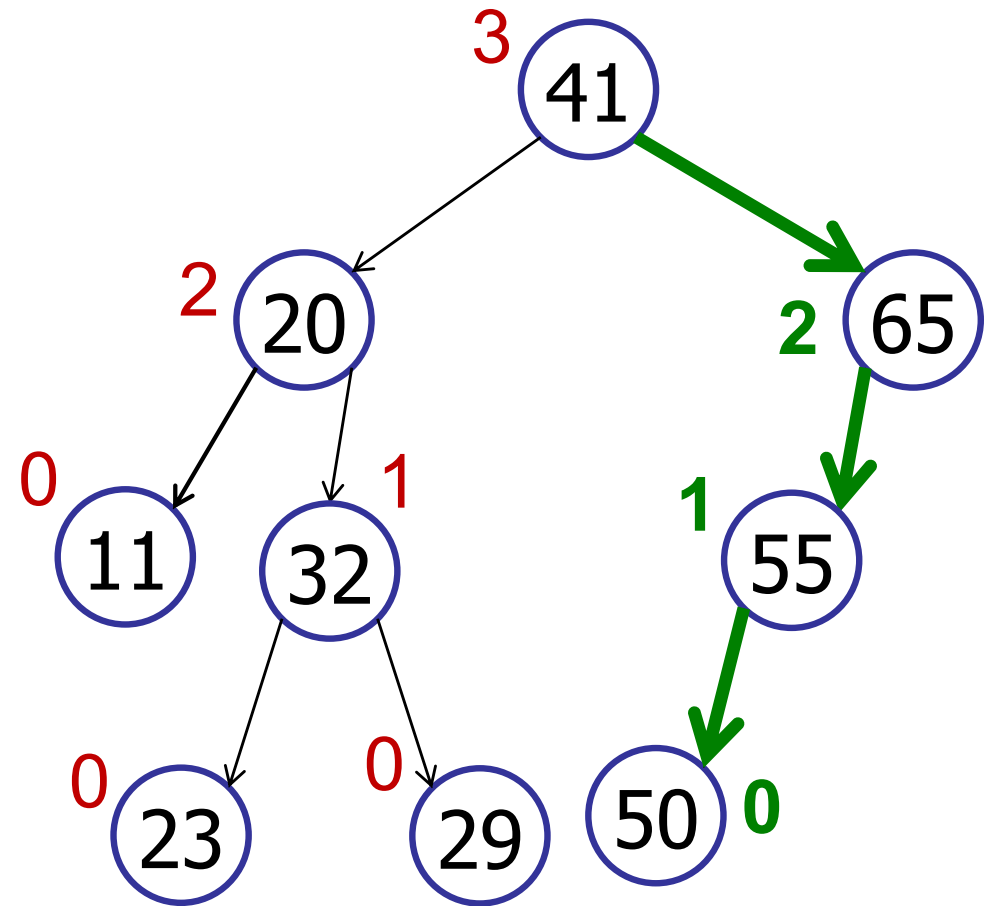
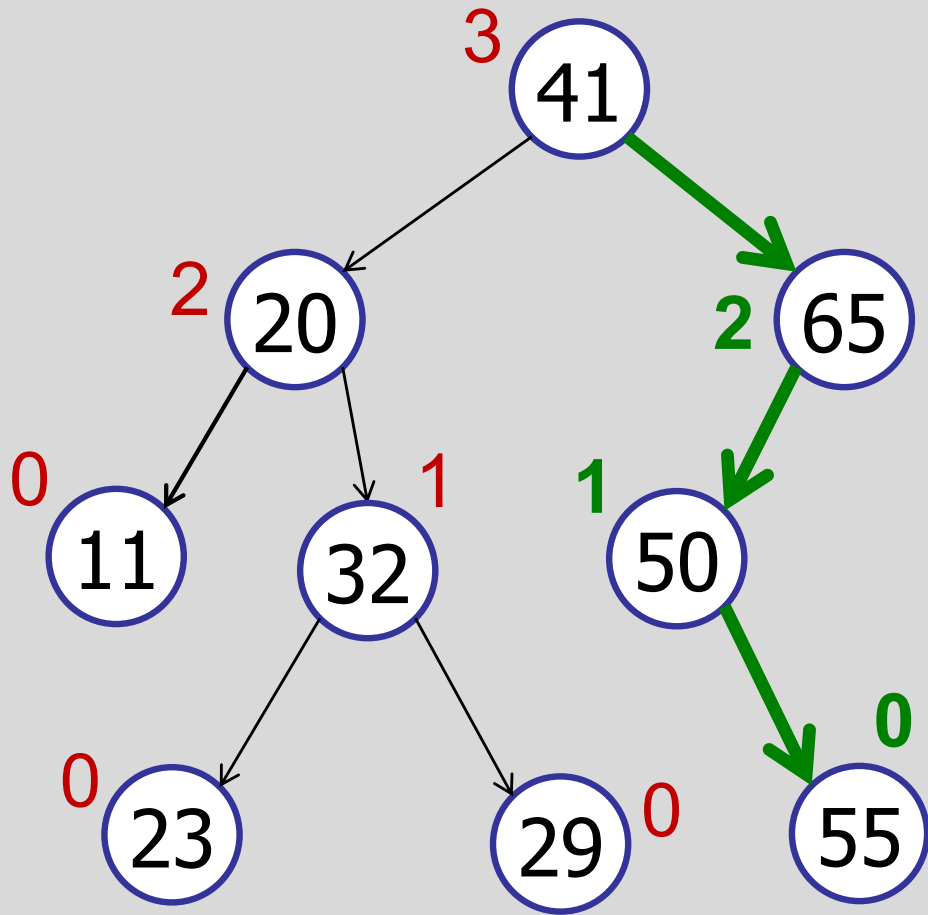
# Example

insert(55)



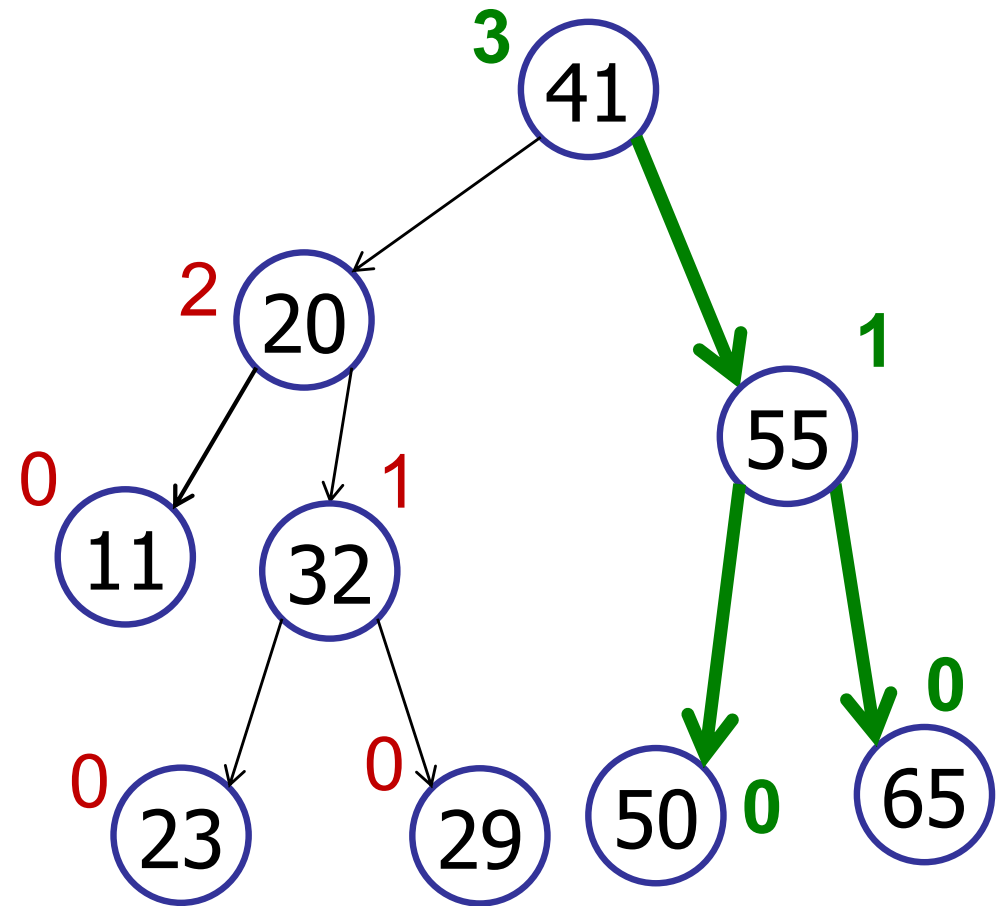
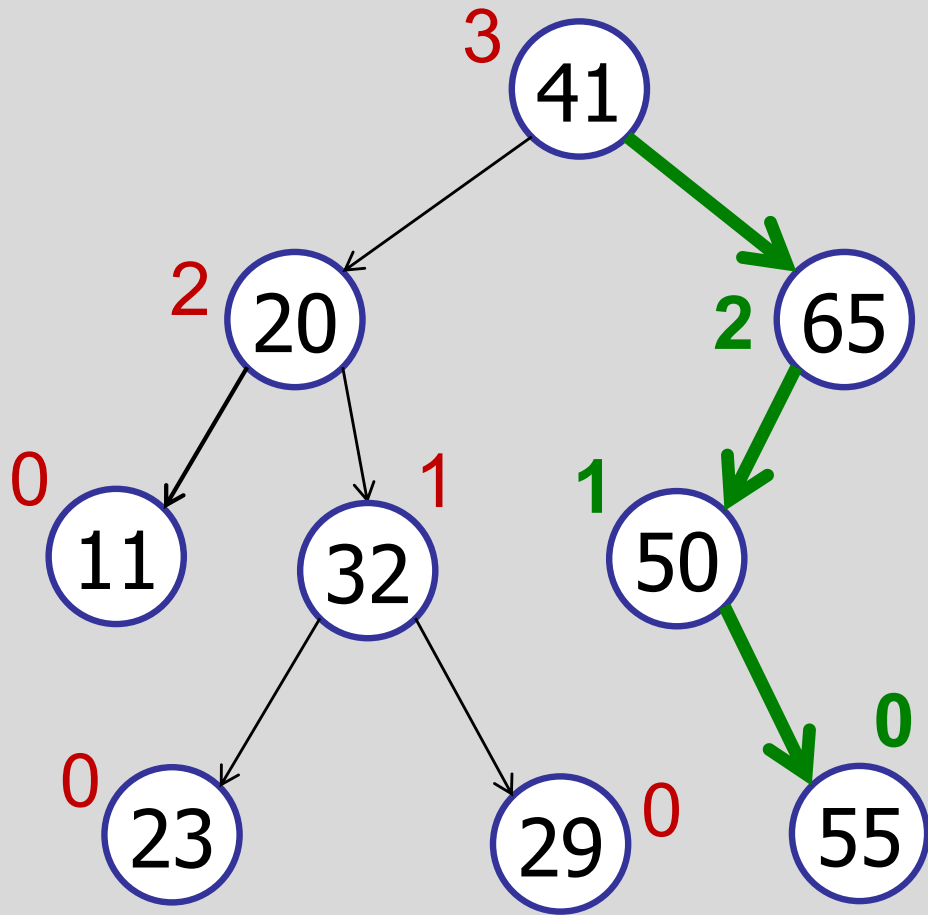
# Example

left-rotate(50)



# Example

right-rotate(65)





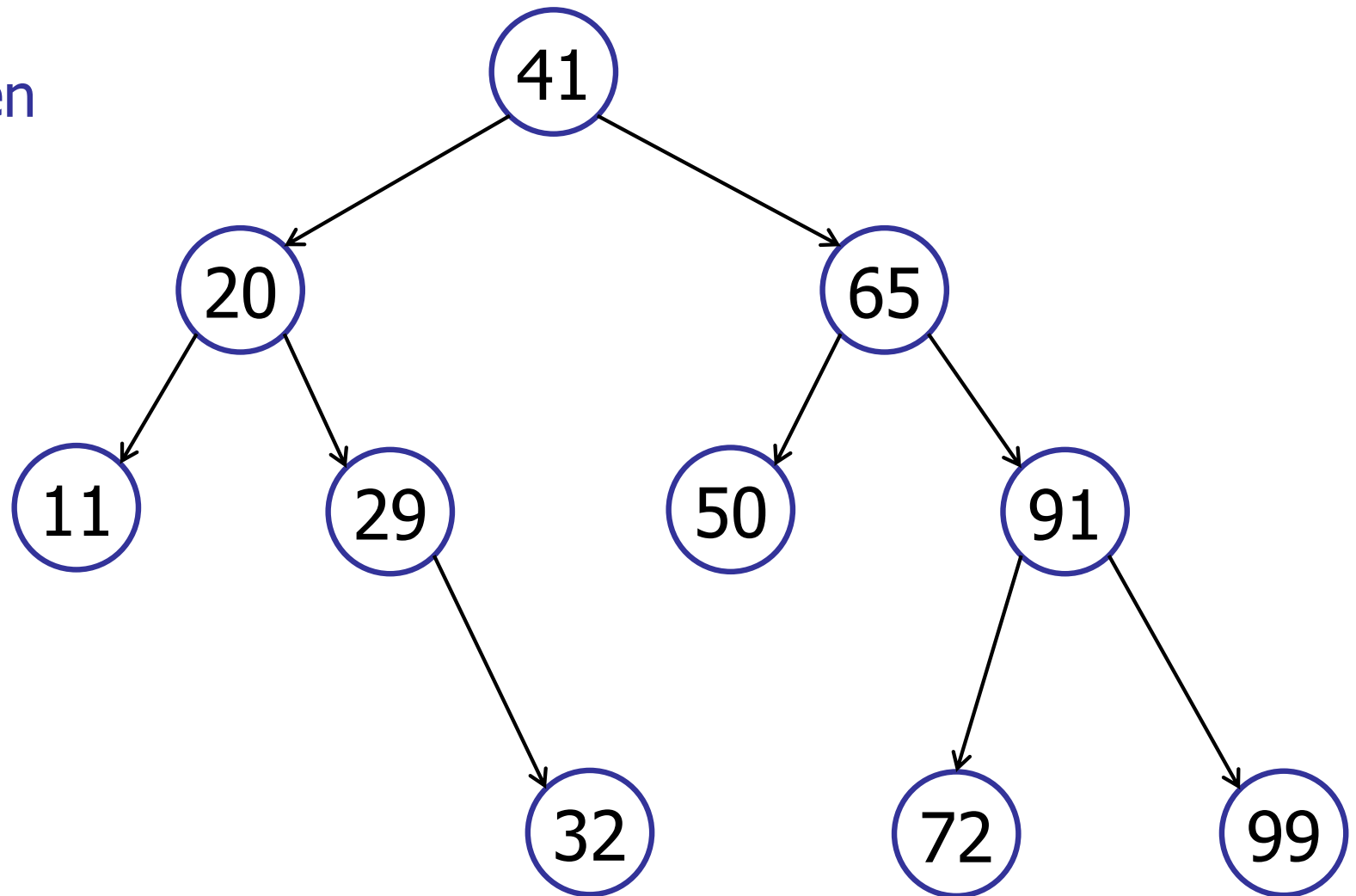
# Binary Search Tree

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delete(v)

Three cases:

1. No children
2. 1 child
3. 2 children



# Binary Search Tree

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## delete(v)

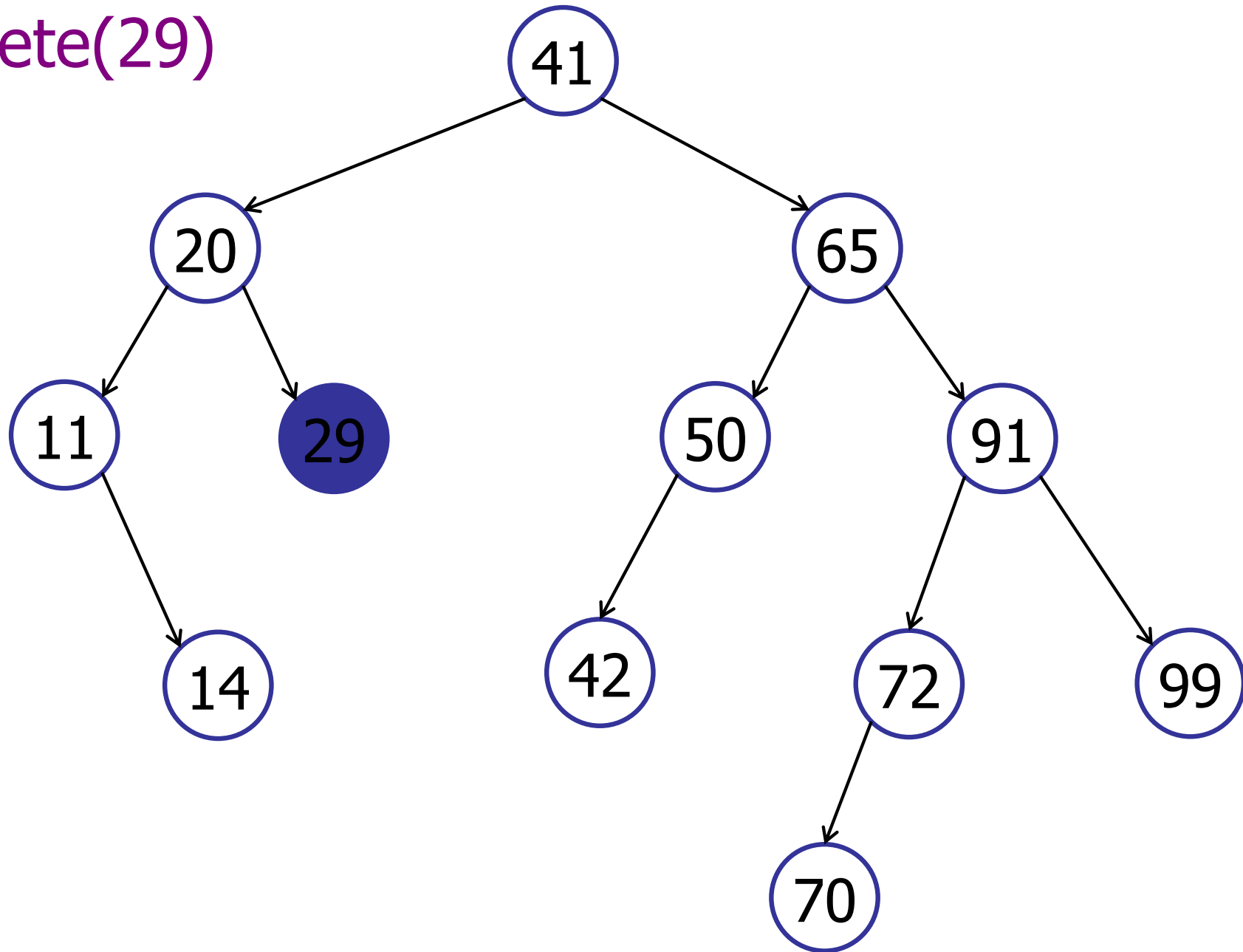
1. If **v** has two children, swap it with its successor.
2. Delete node v from binary tree (and reconnect children).
3. For every ancestor of the deleted node:
  - Check if it is height-balanced.
  - If not, perform a rotation.
  - Continue to the root.

Deletion may take up to  $O(\log(n))$  rotations.

# Binary Search Tree

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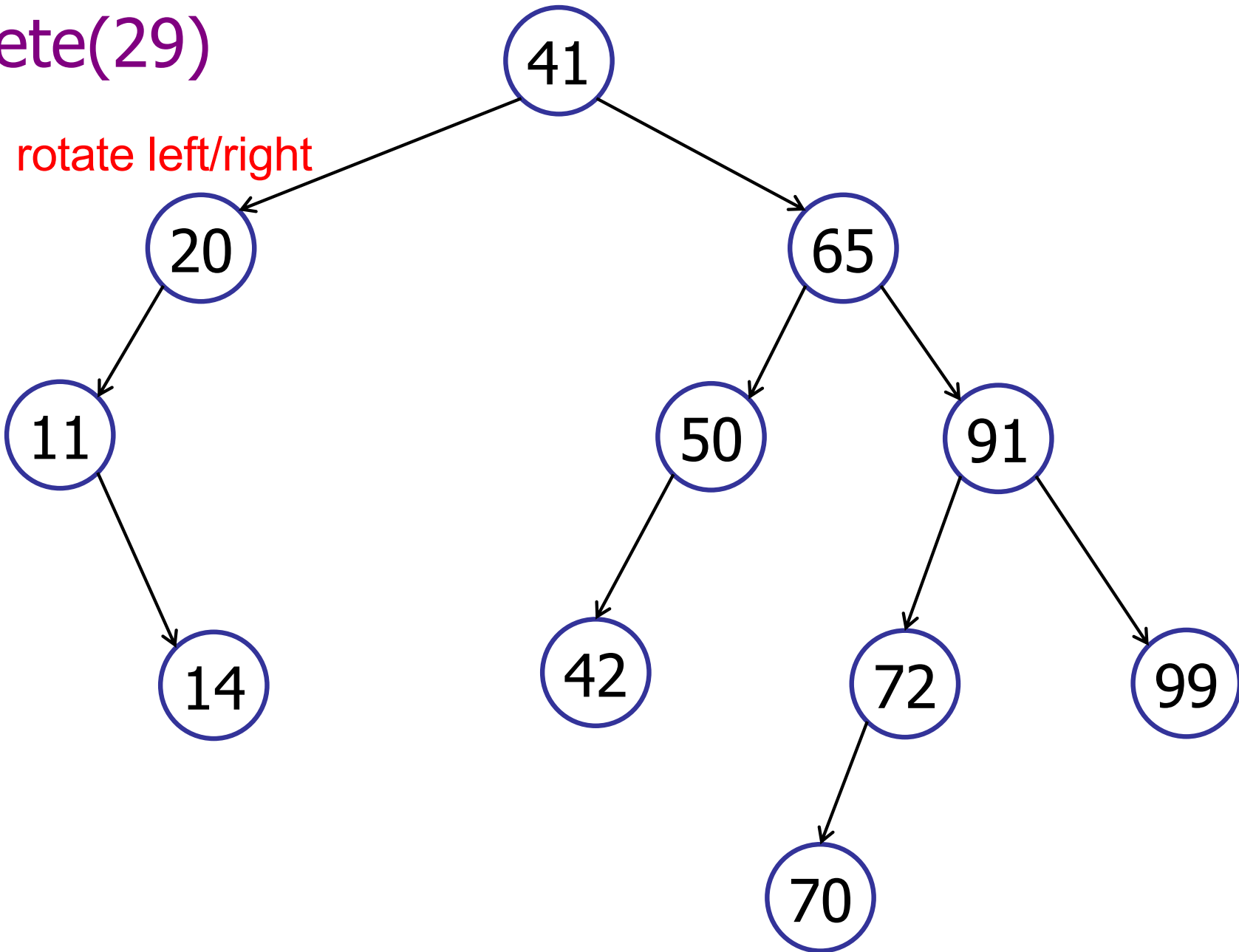
delete(29)



# Binary Search Tree

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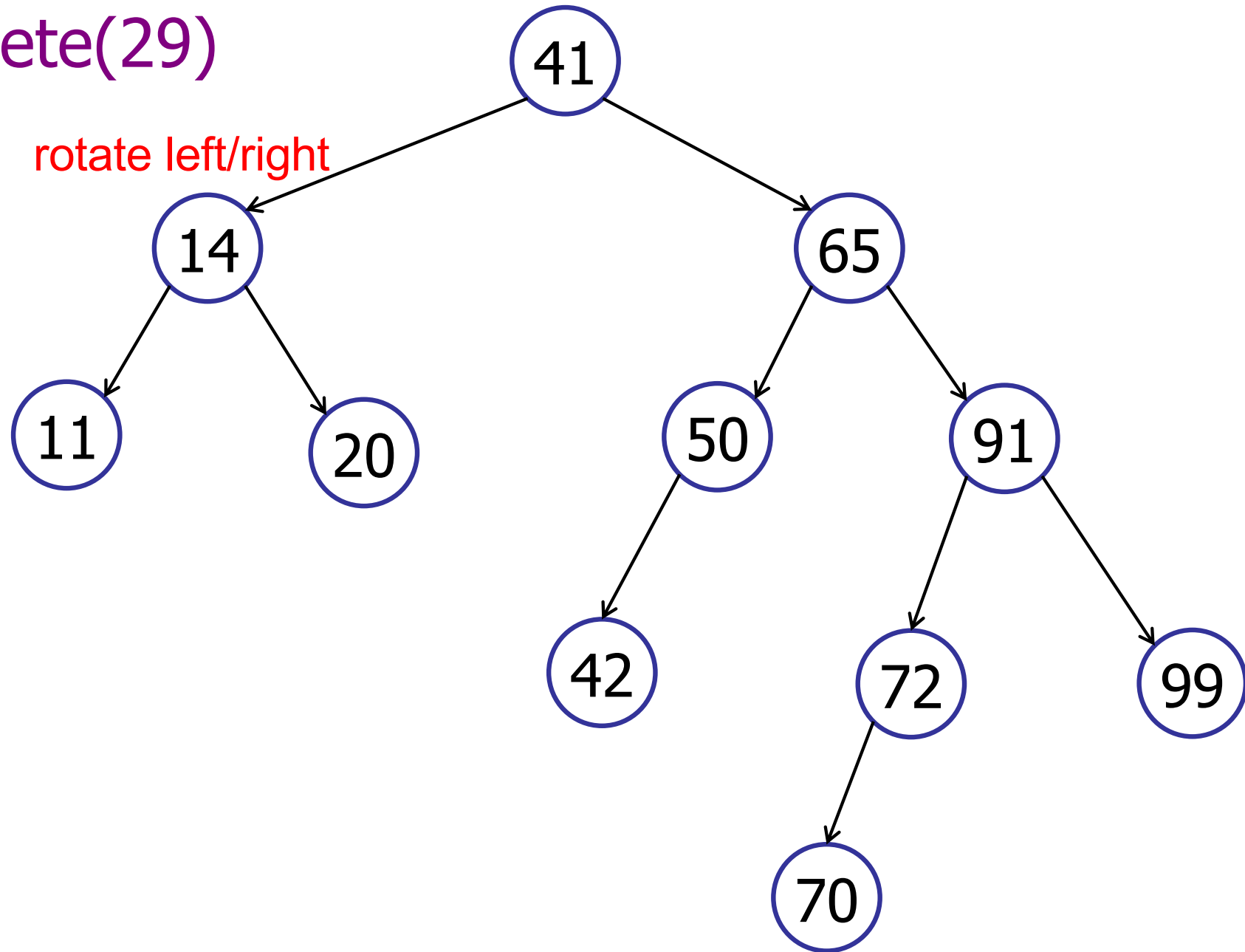
delete(29)



# Binary Search Tree

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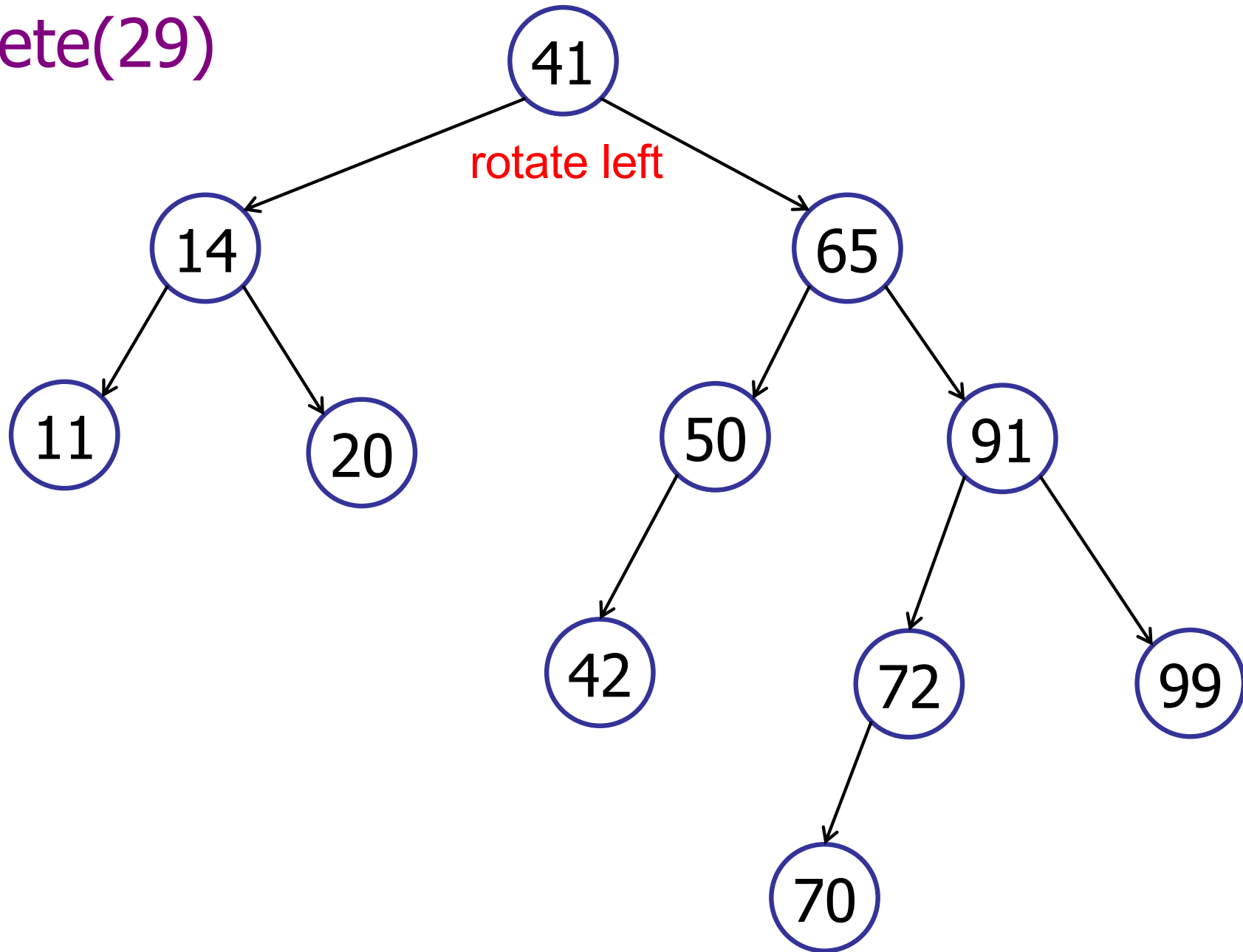
delete(29)



# Binary Search Tree

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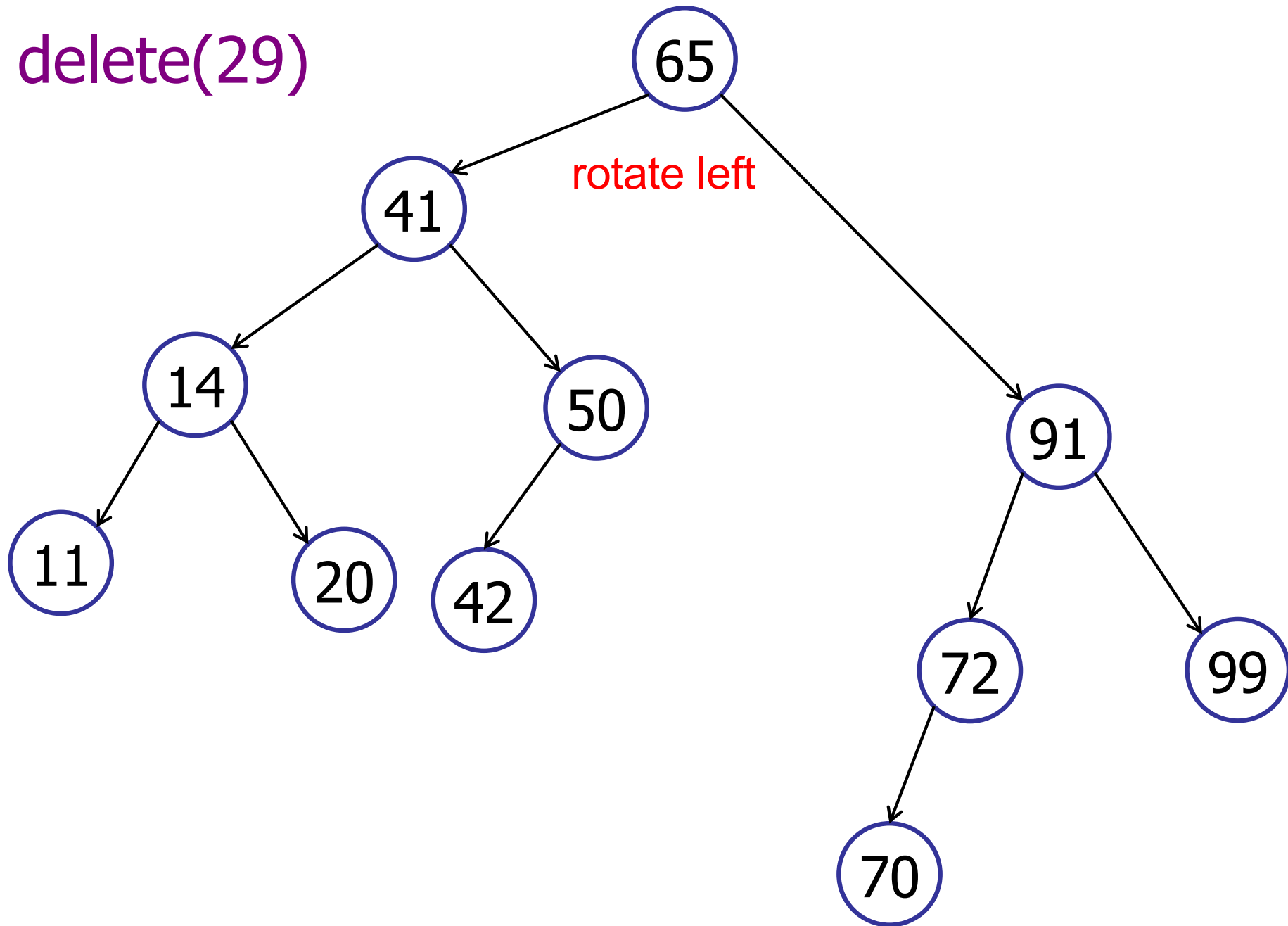
delete(29)



# Binary Search Tree

---

delete(29)



# How many rebalances?

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Why are two rotations not enough?

- Delete reduced height.
- Rotations (to rebalance) reduce height!

Key observation:

- Rebalancing does not “undo” the change in height caused by insertion.



# Delete in AVL Tree

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## Summary:

- Delete key from BST.
- Walk up tree:
  - At every step, check for balance.
  - If out-of-balance, use rotations to rebalance.
  - Continue to root.

## Key observation:

- It is *not* sufficient to only fix lowest out-of-balance node in tree.

Every insertion requires 1 or 2 rotations?

1. Yes
- ✓ 2. No
3. I don't know



A tree is **balanced** if every node's children differ in height be at most 1?

- ✓ 1. Yes
- 2. No
- 3. I don't know



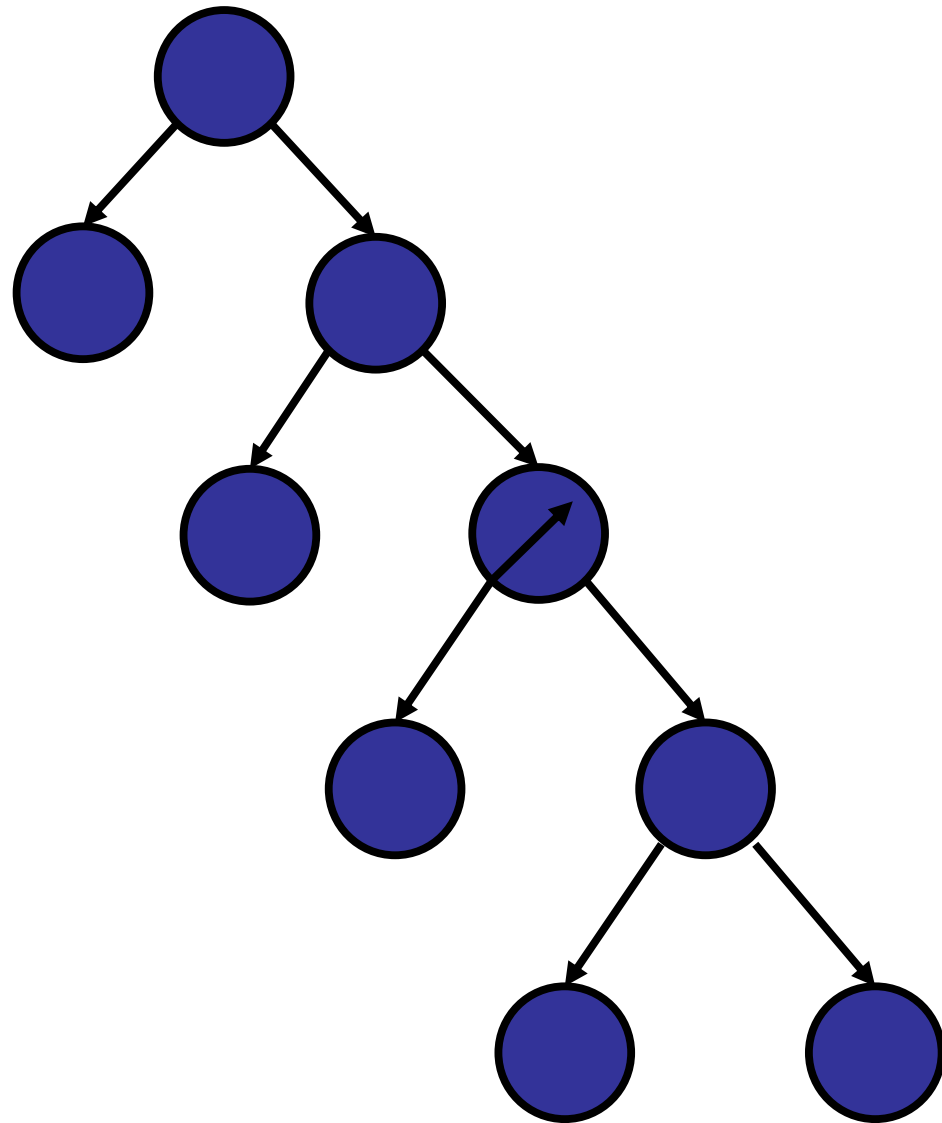
A tree is **balanced** if every node either has two children or zero children?

1. Yes
- ✓ 2. No
3. I don't know



A tree is balanced if every node either has two children or zero children?

1. Yes
- ✓ 2. No
3. I don't know



Using rotations, you can create every possible “tree shape.”

- ✓ 1. True
- 2. False
- 3. I don't know



# AVL Trees

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What if you do not remove deleted nodes?

- Mark a node “deleted” and leave it in the tree.

Logical deletes:

- Performance degrades over time.
- Clean up later? (Amortized performance...)

# AVL Trees

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What if you do not want to store the height in every node?

- Only store difference in height from parent.



# Today's Plan

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## **On the importance of being balanced**

- Height-balanced binary search trees
- AVL trees
- Rotations

## **Tries**

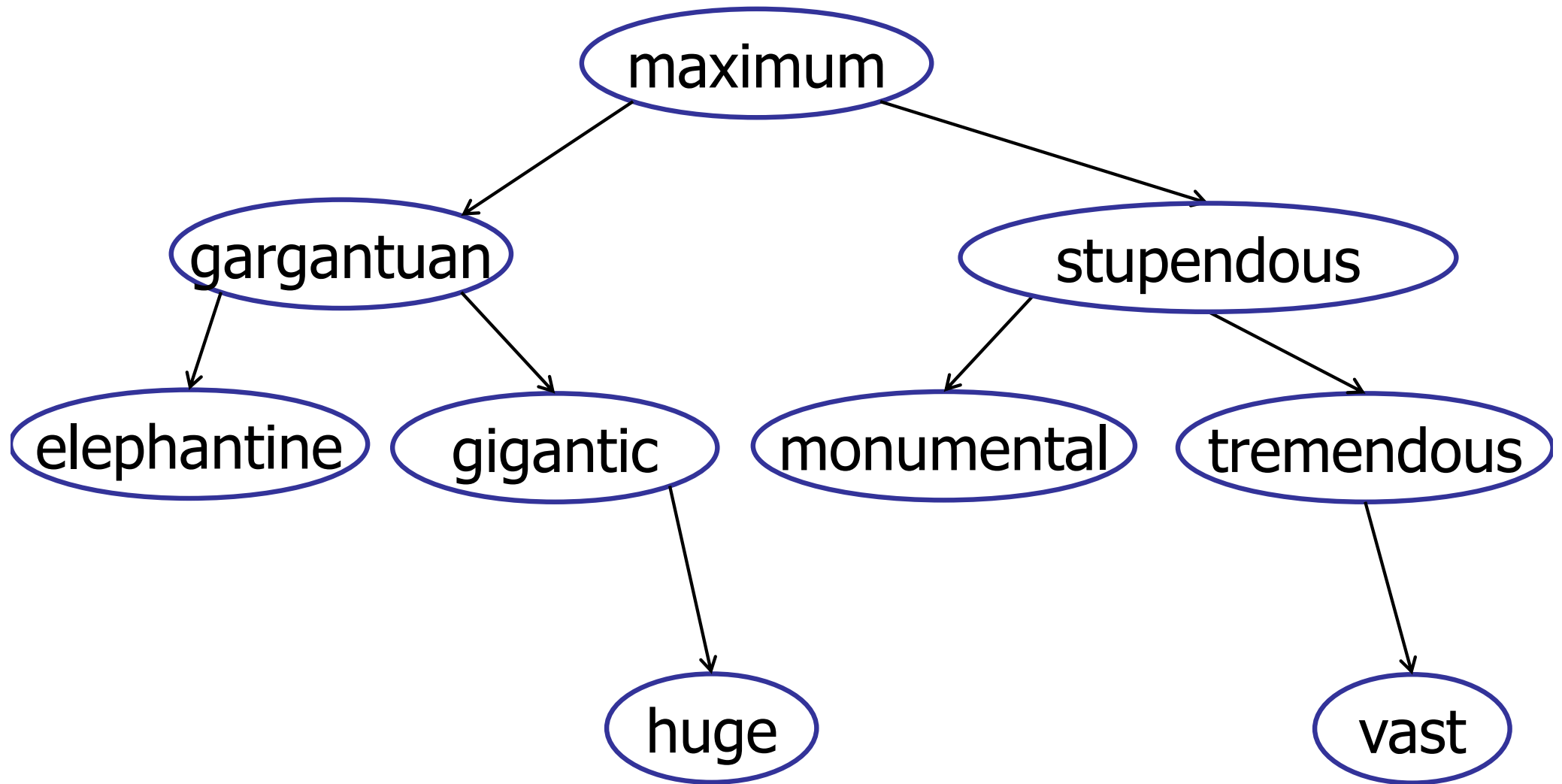
- How to handle text?

## **Data structure design**

- How to build new structures on existing ideas?

# What about text strings?

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Implement a searchable dictionary!

# What about text strings?

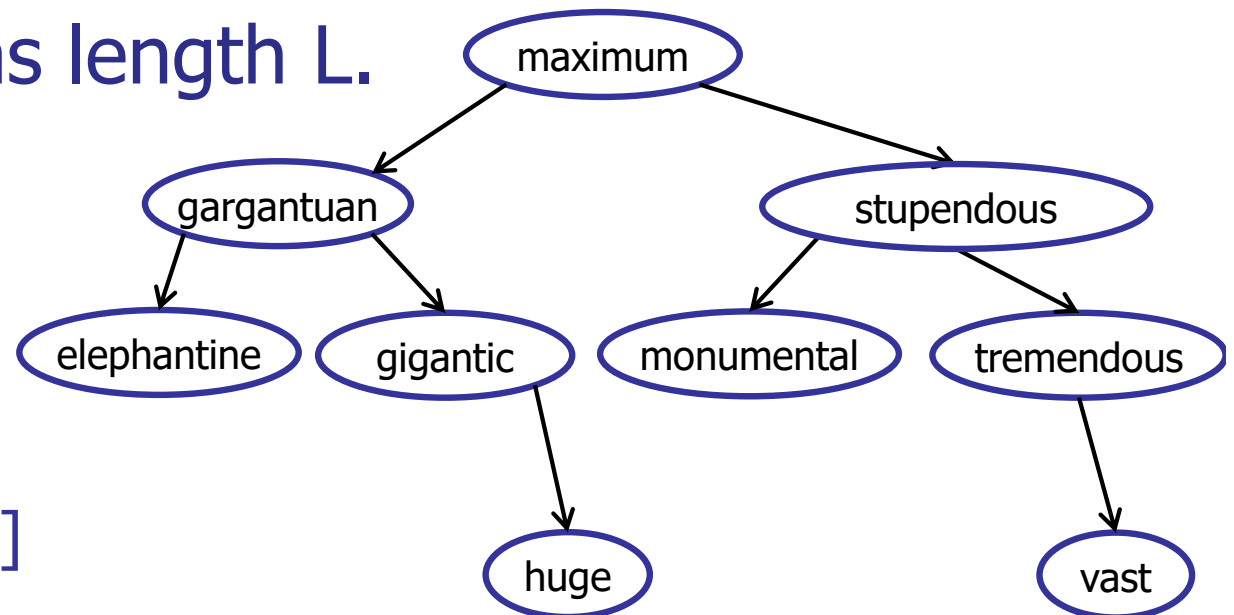
---

Cost of comparing two strings:

- $\text{Cost}[A \neq B] = \min(A.\text{length}, B.\text{length})$
- Compare strings letter by letter

Cost of tree operation:

- Assume string has length  $L$ .
- Cost:  $O(hL)$



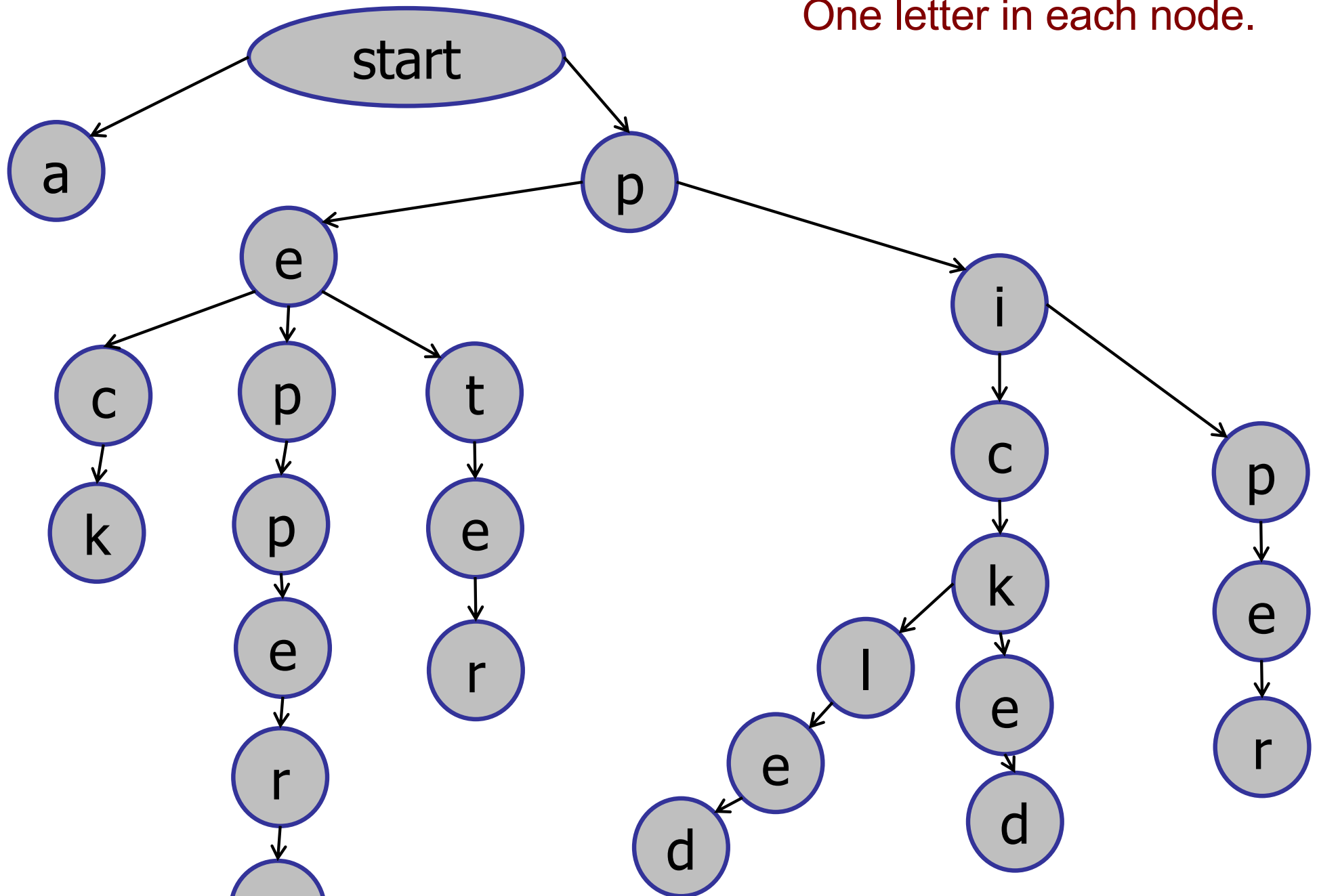
[In the worst case.]

[Optimizations are possible.]

# Trie [pronounced: try]

---

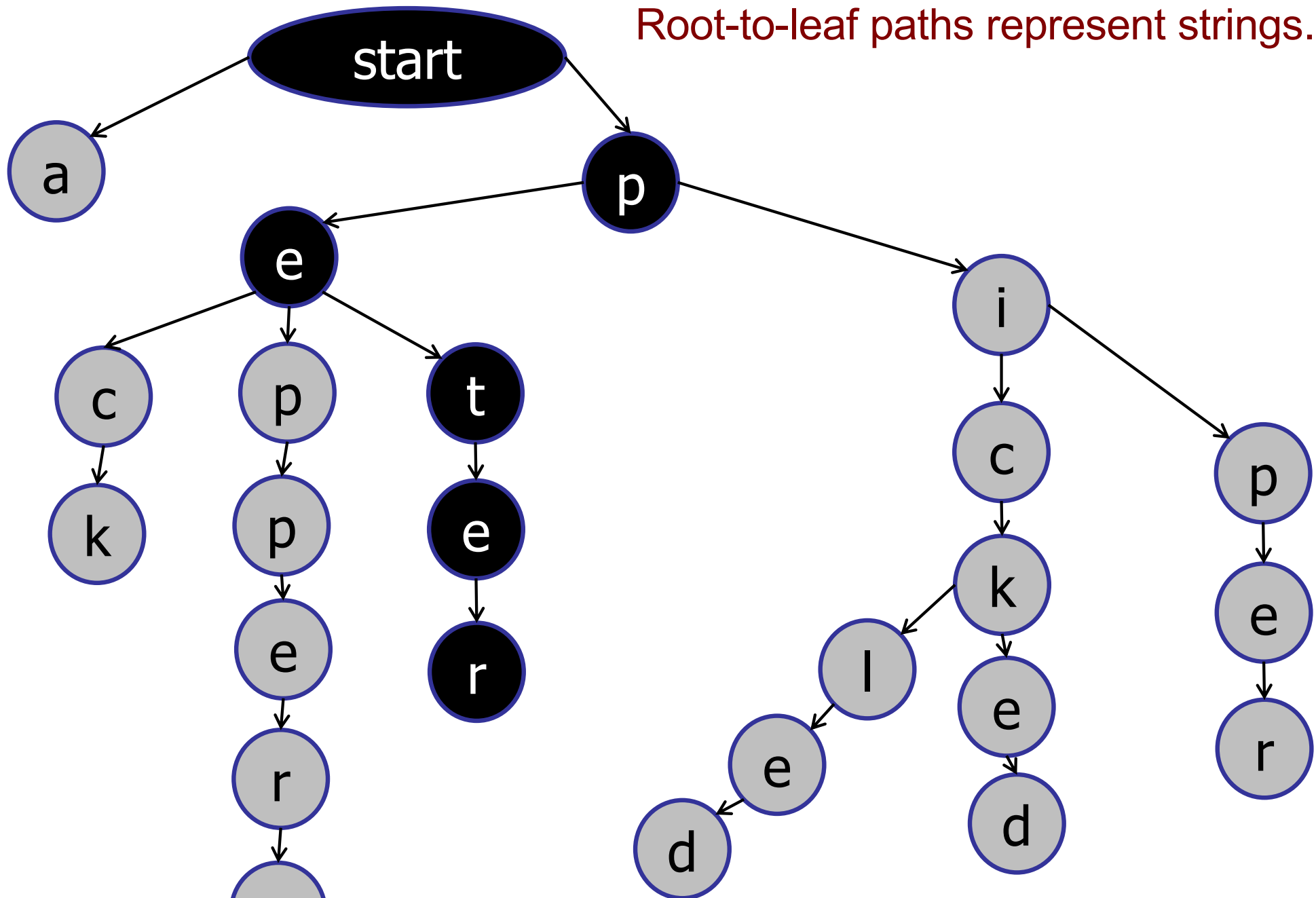
One letter in each node.



# Trie [pronounced: try]

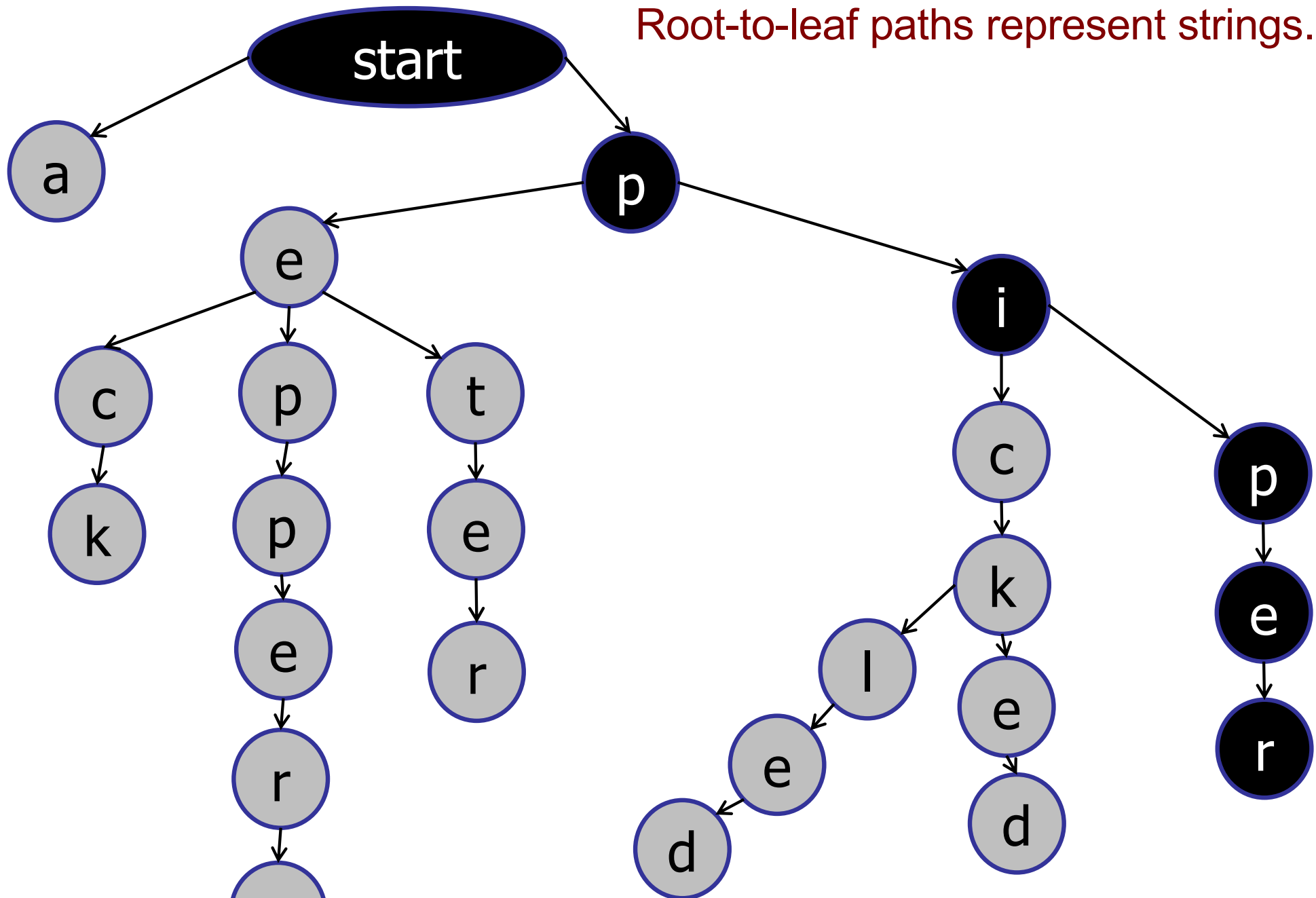
---

Root-to-leaf paths represent strings.



# Trie [pronounced: try]

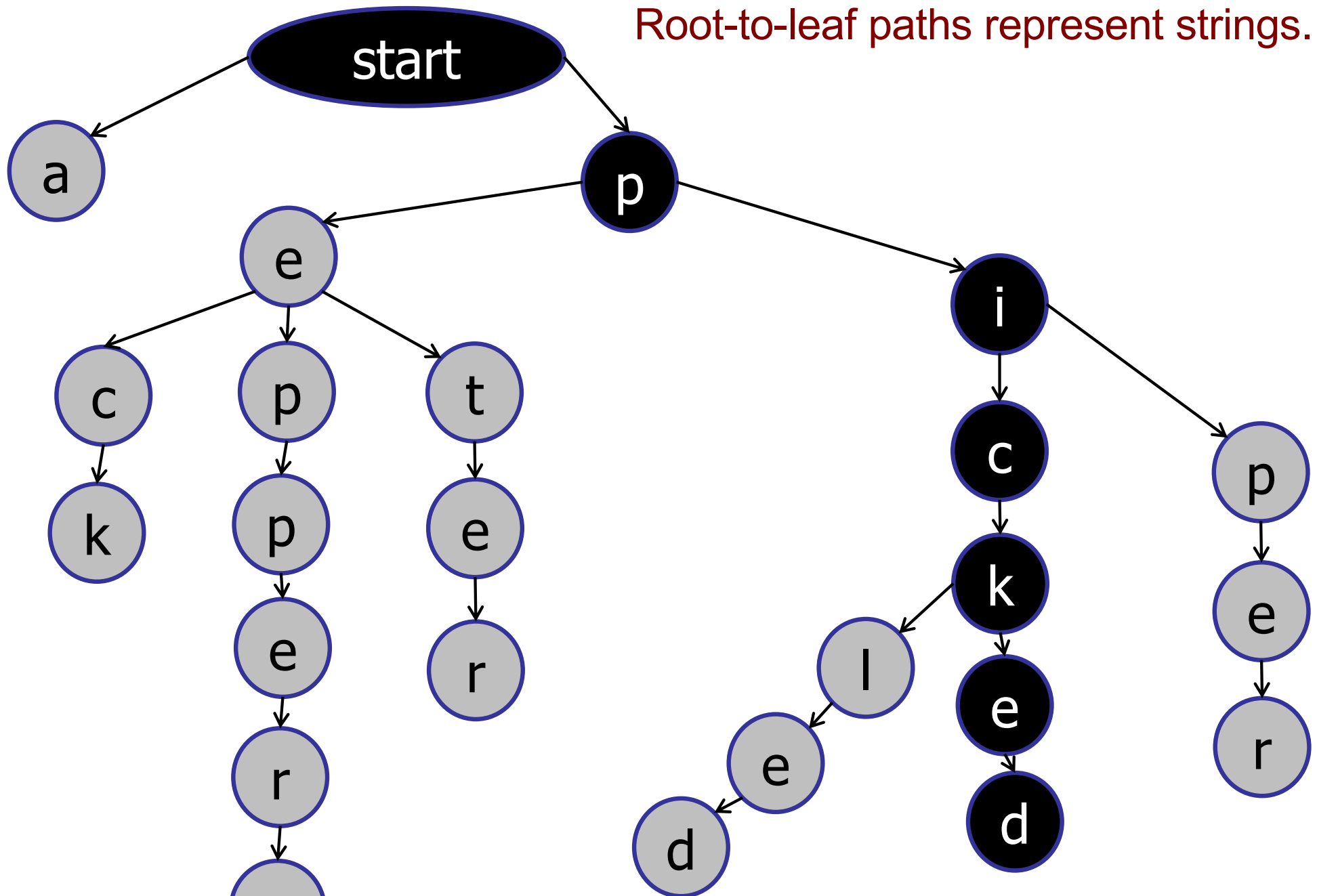
Root-to-leaf paths represent strings.



# Trie [pronounced: try]

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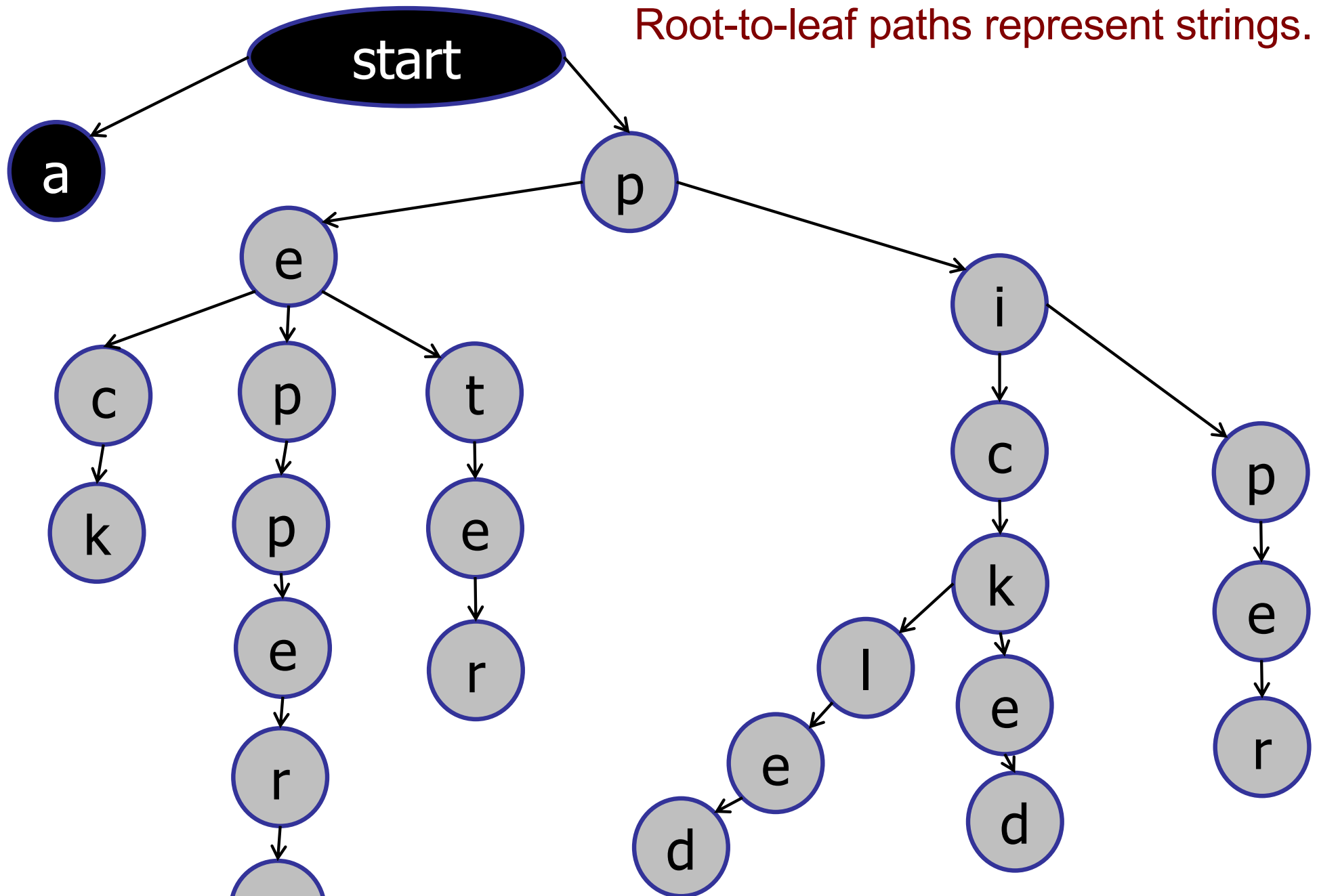
Root-to-leaf paths represent strings.



# Trie [pronounced: try]

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Root-to-leaf paths represent strings.

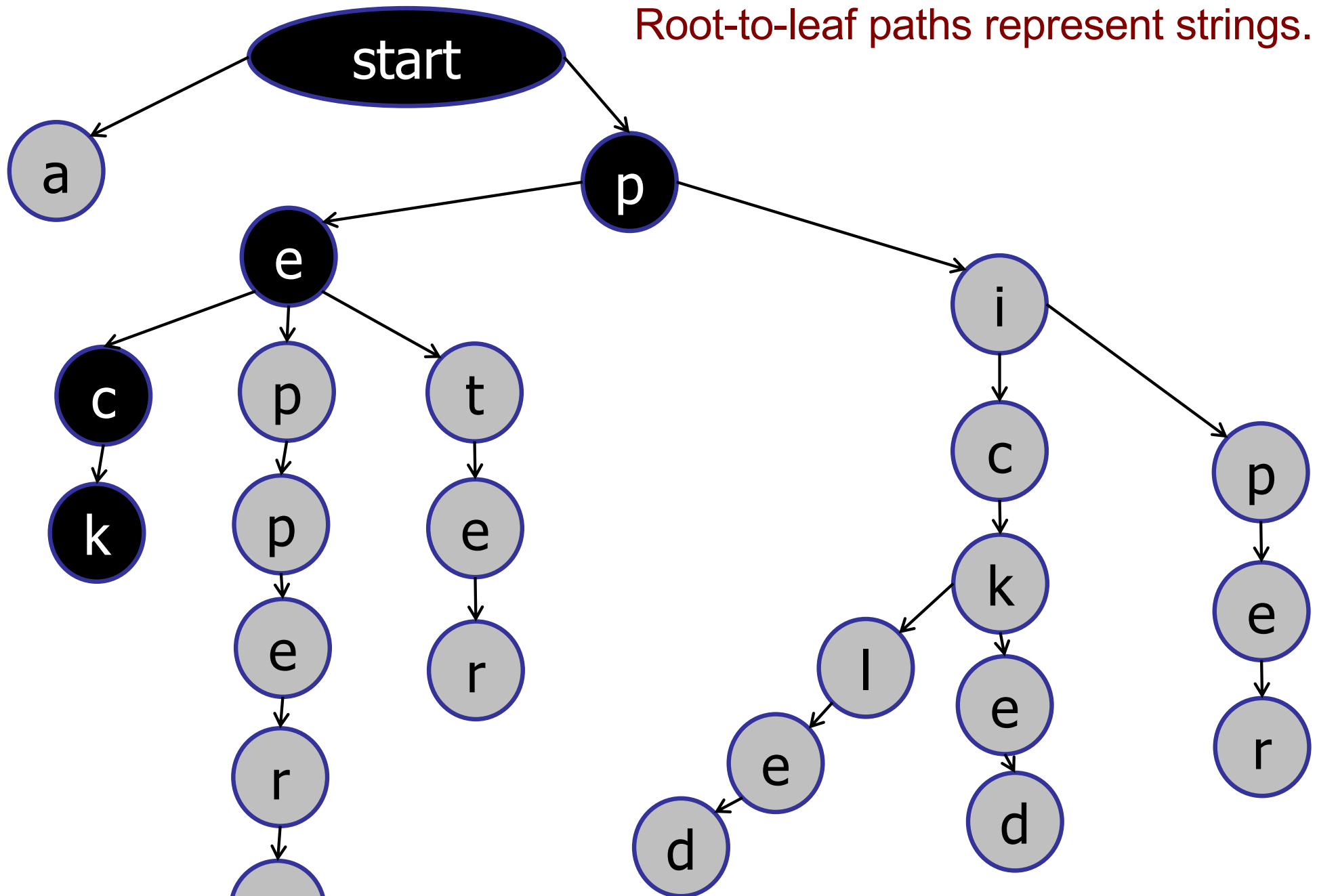




# Trie [pronounced: try]

---

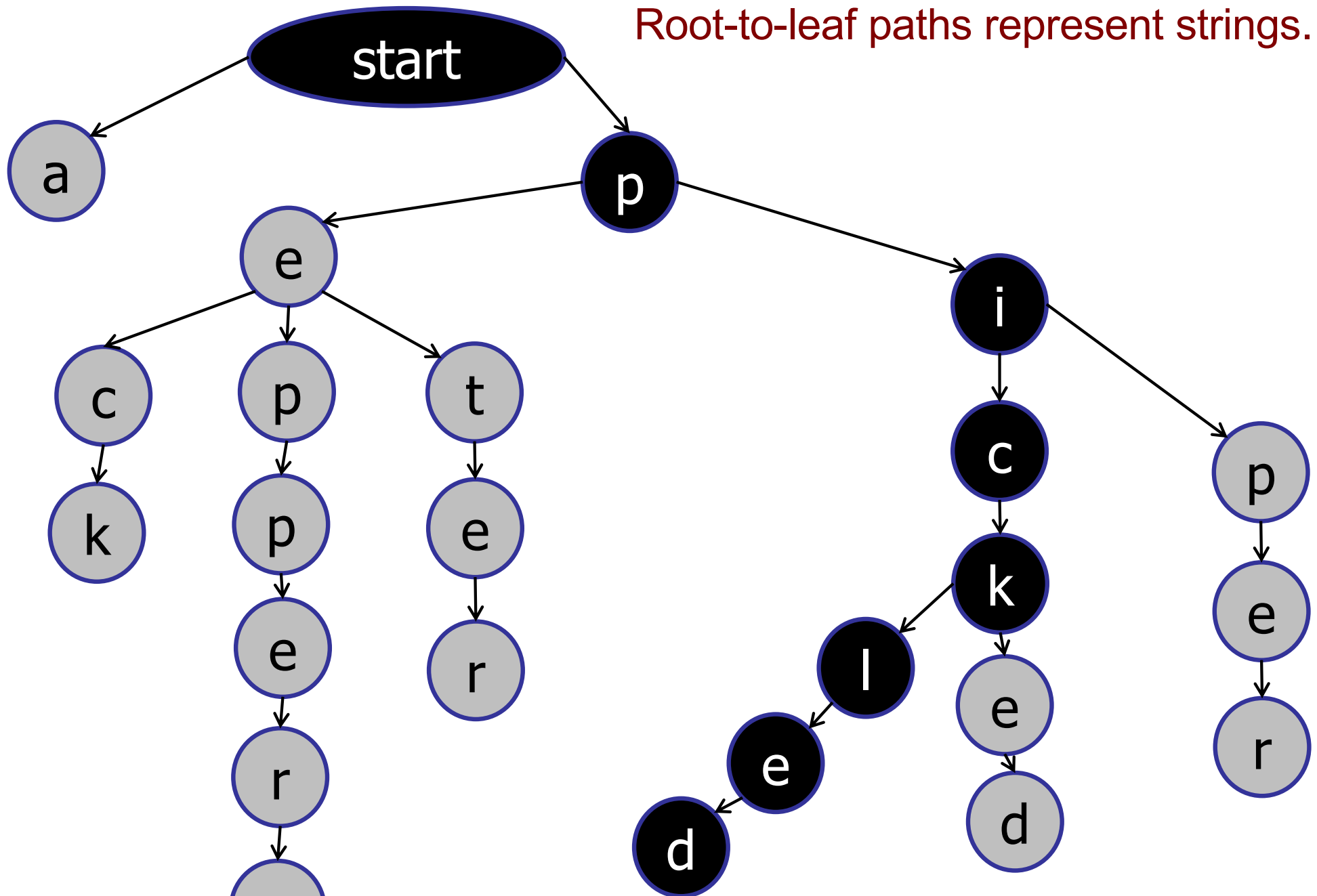
Root-to-leaf paths represent strings.



# Trie [pronounced: try]

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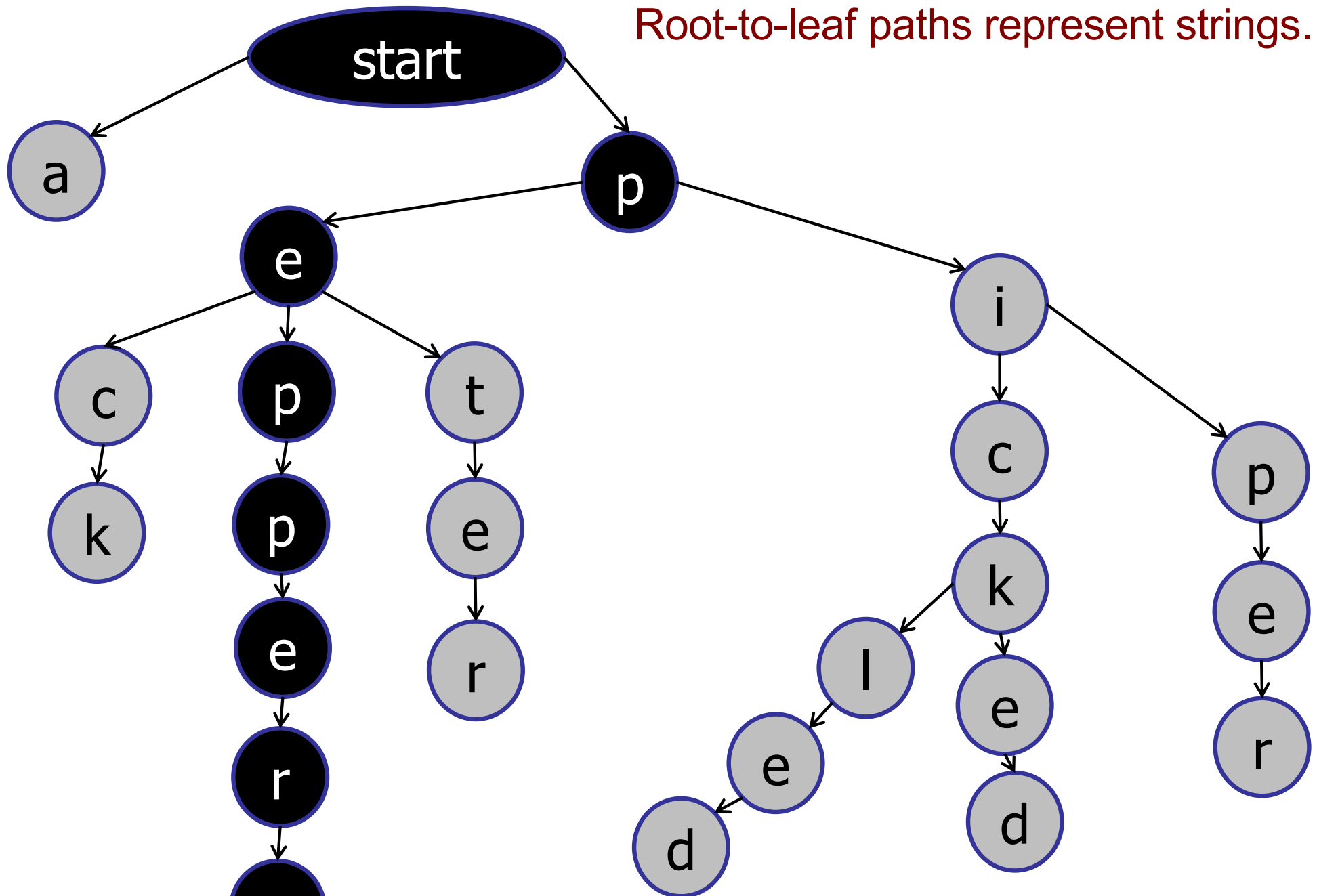
Root-to-leaf paths represent strings.



# Trie [pronounced: try]

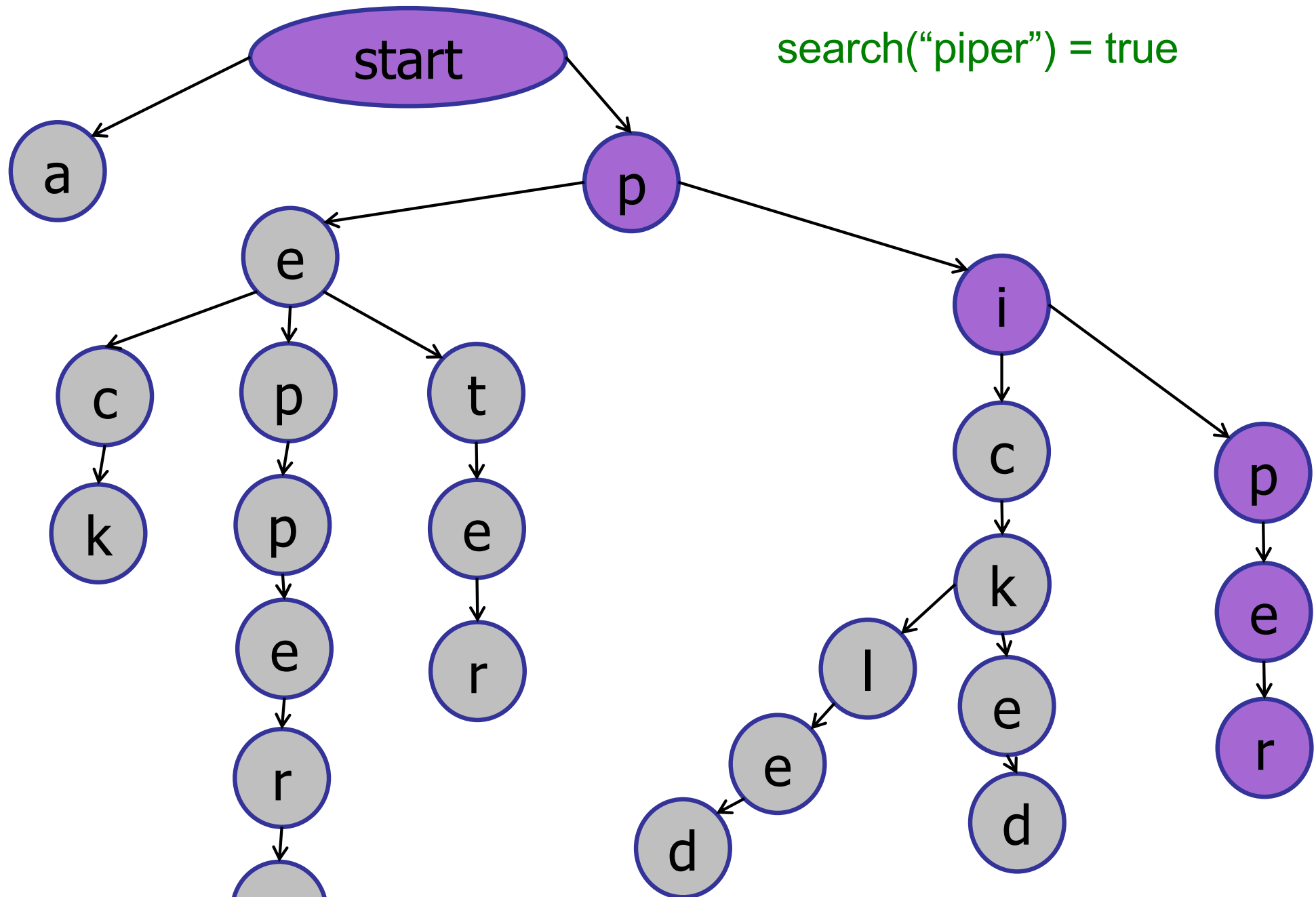
---

Root-to-leaf paths represent strings.

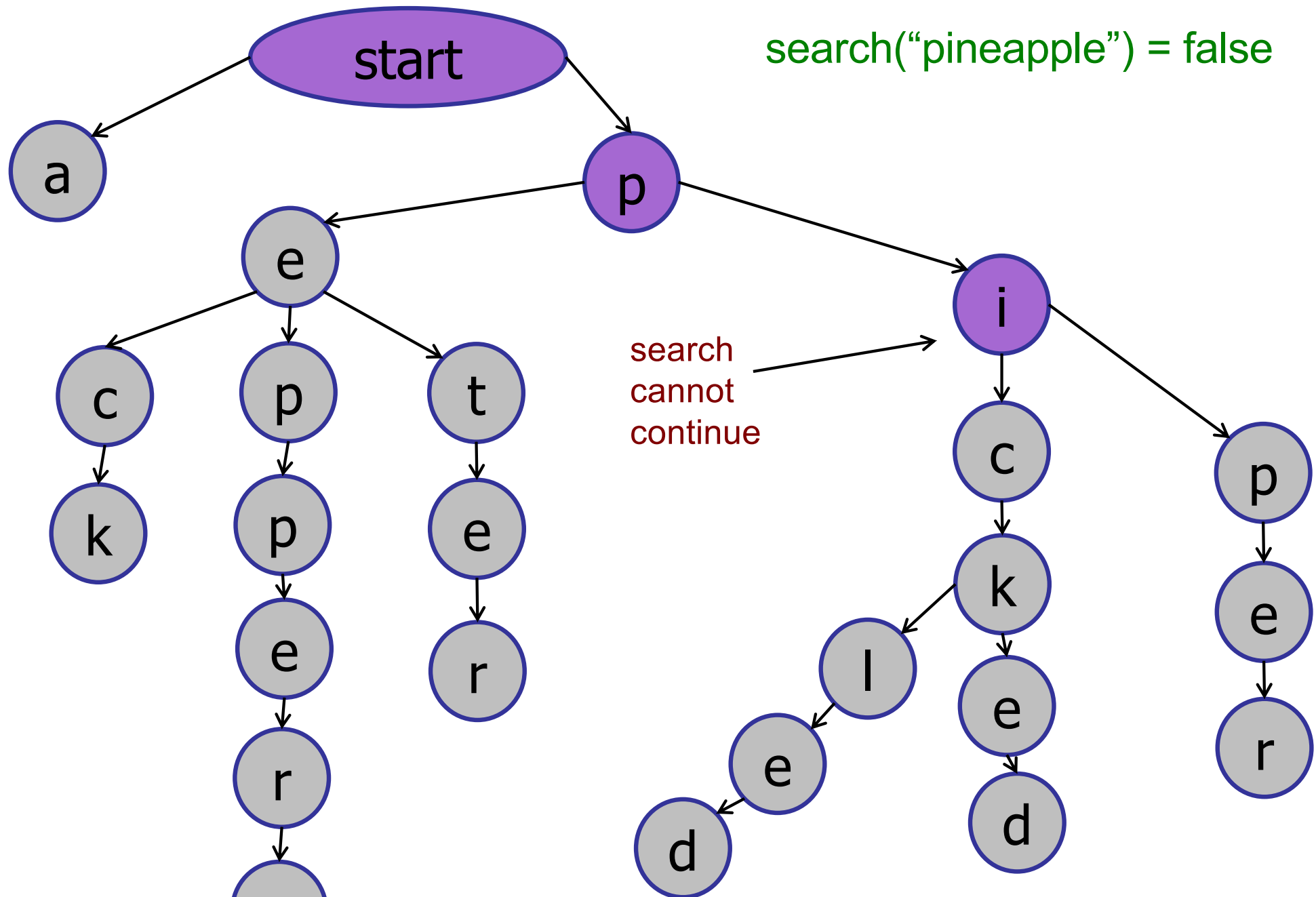


# Searching a Trie

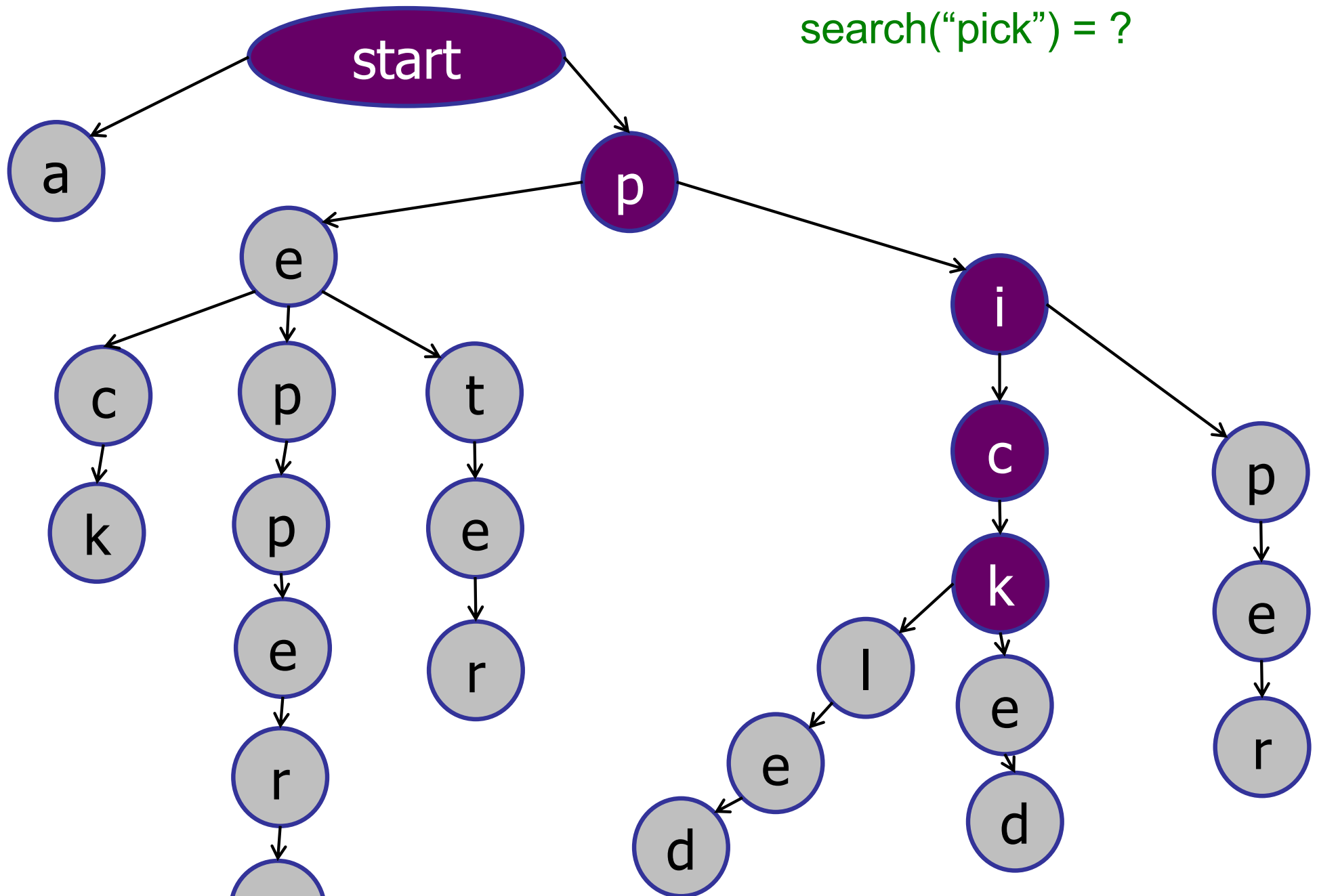
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# Searching a Trie

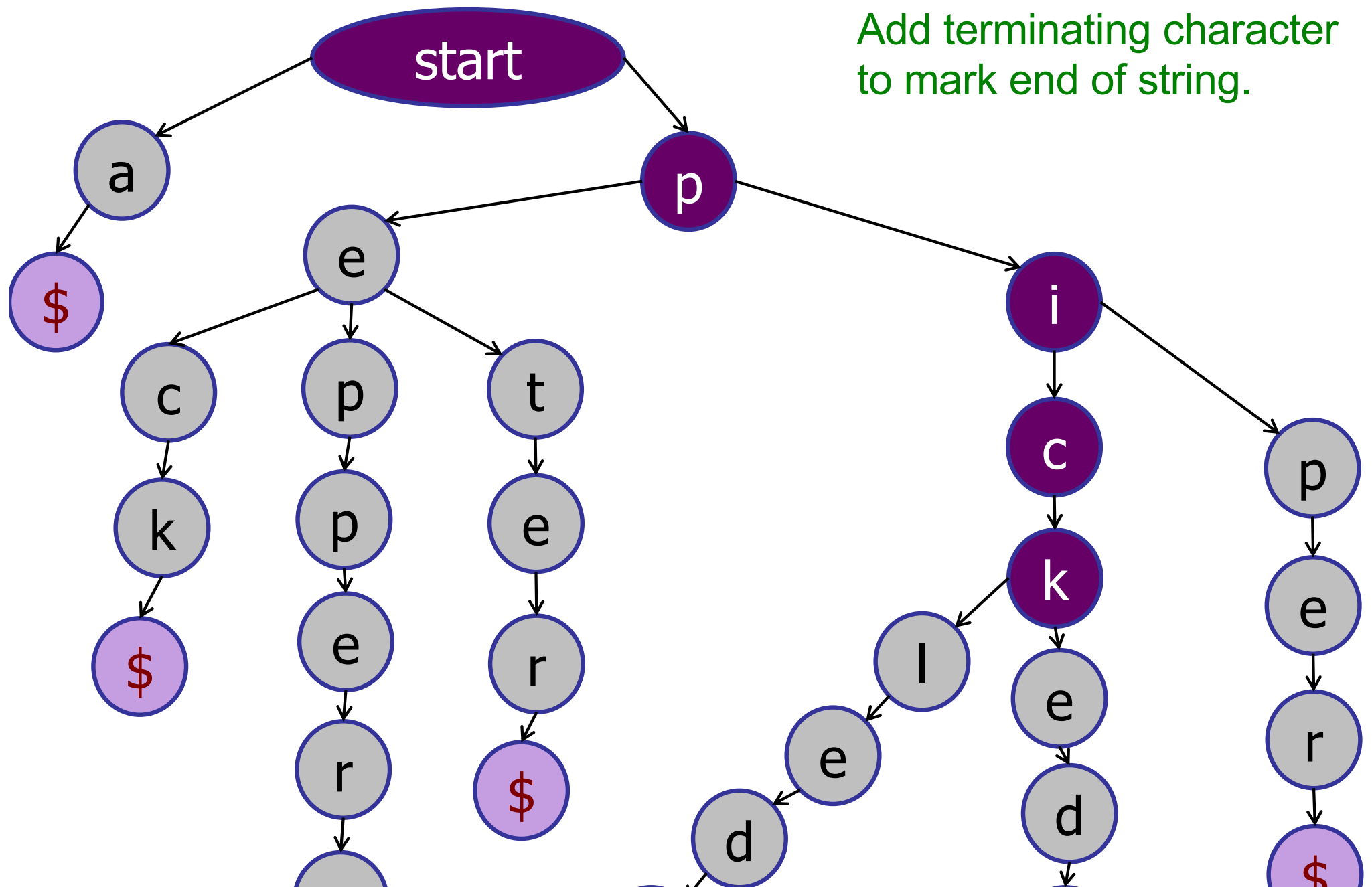


search("pick") = ?

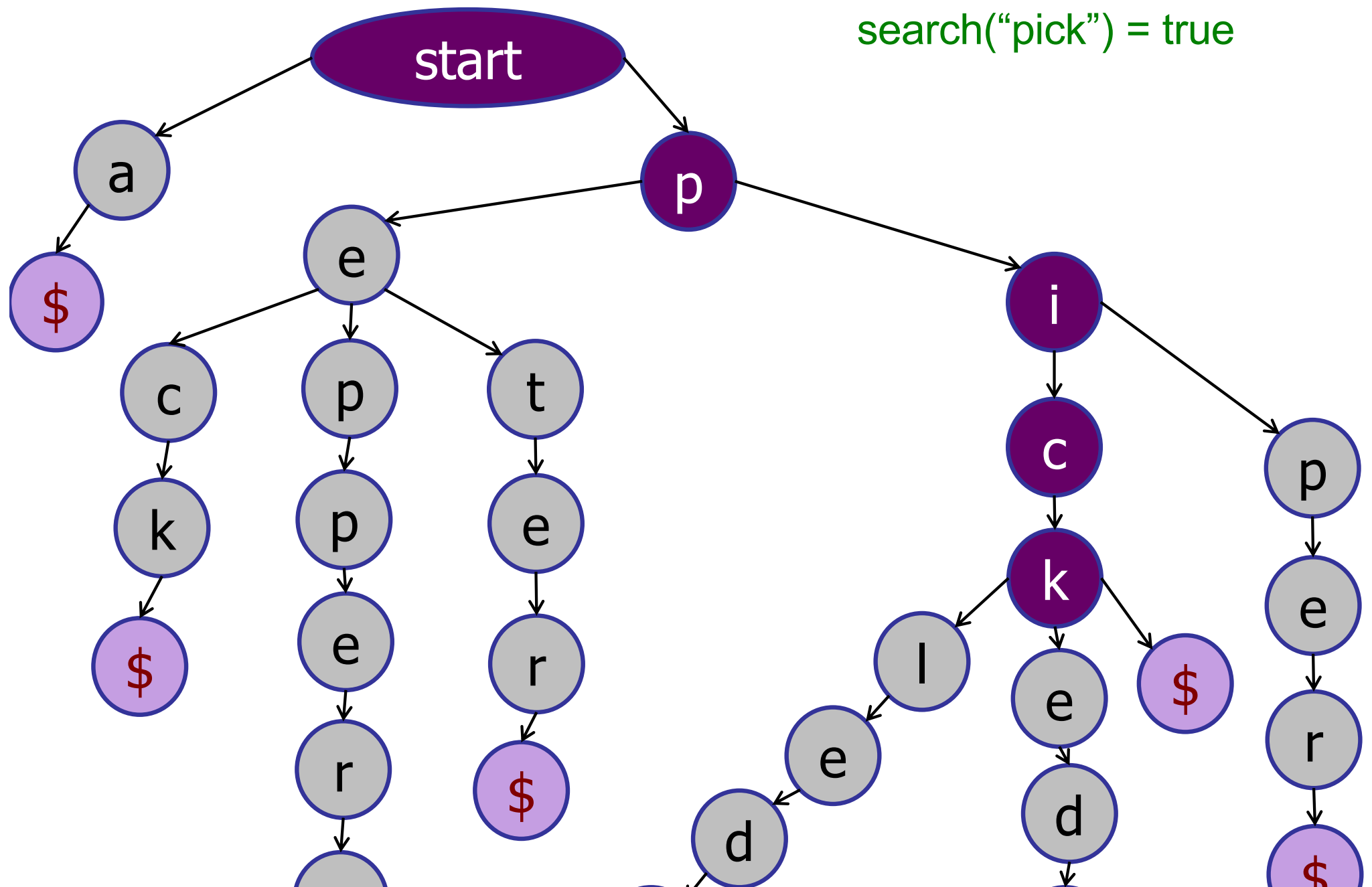


# Trie Details

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# Trie Details

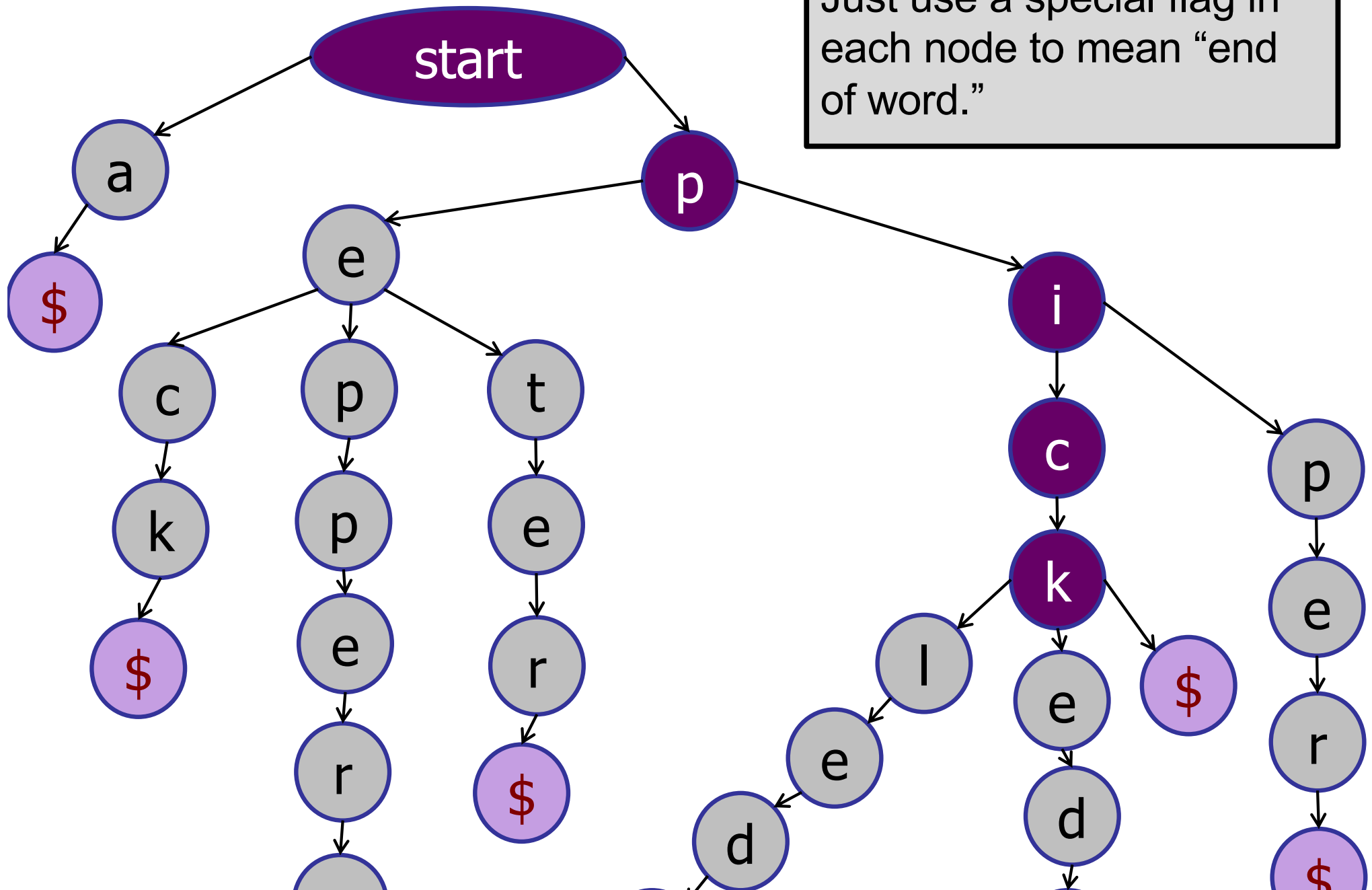




# Trie Details

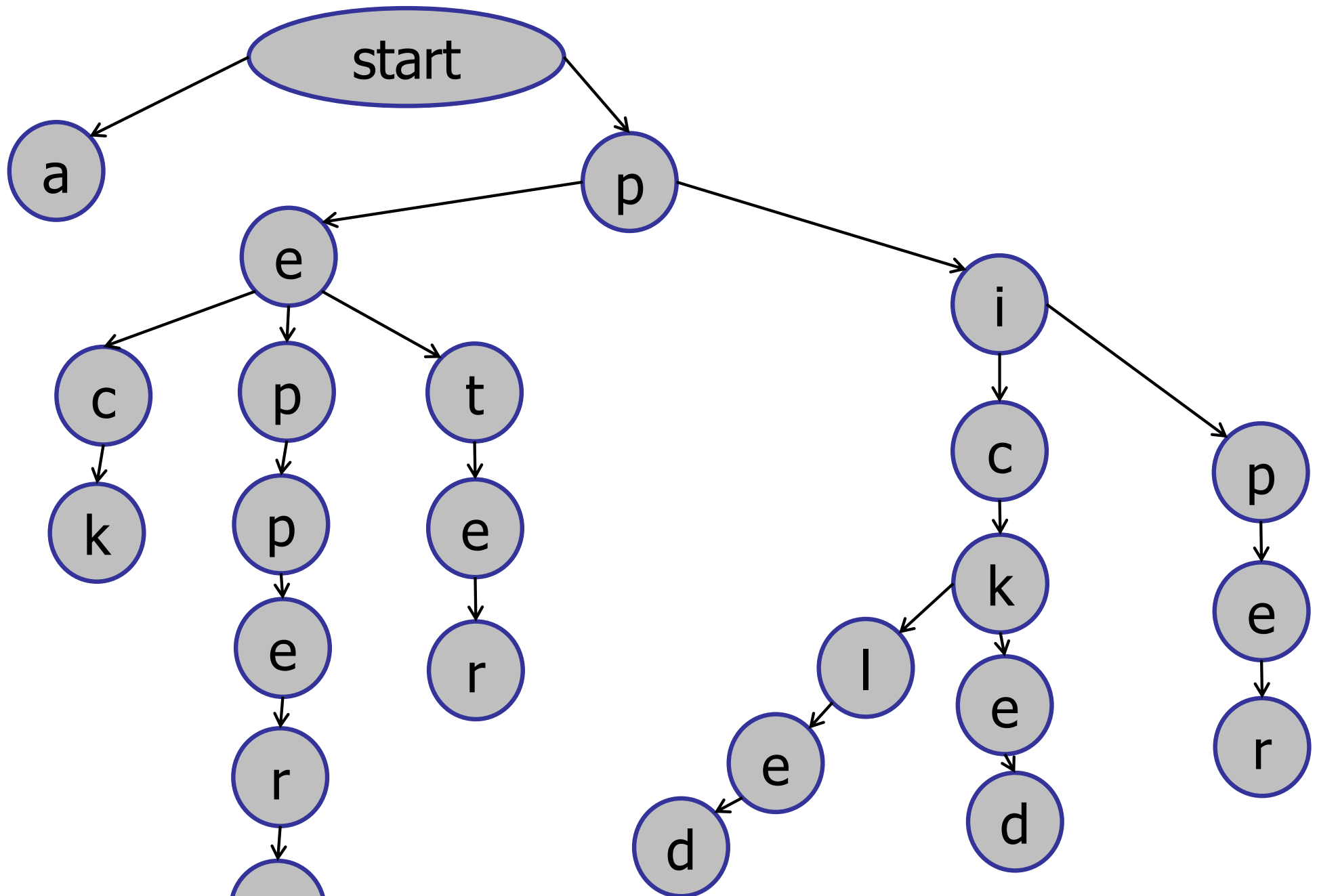
Or:

Just use a special flag in each node to mean “end of word.”



# Trie

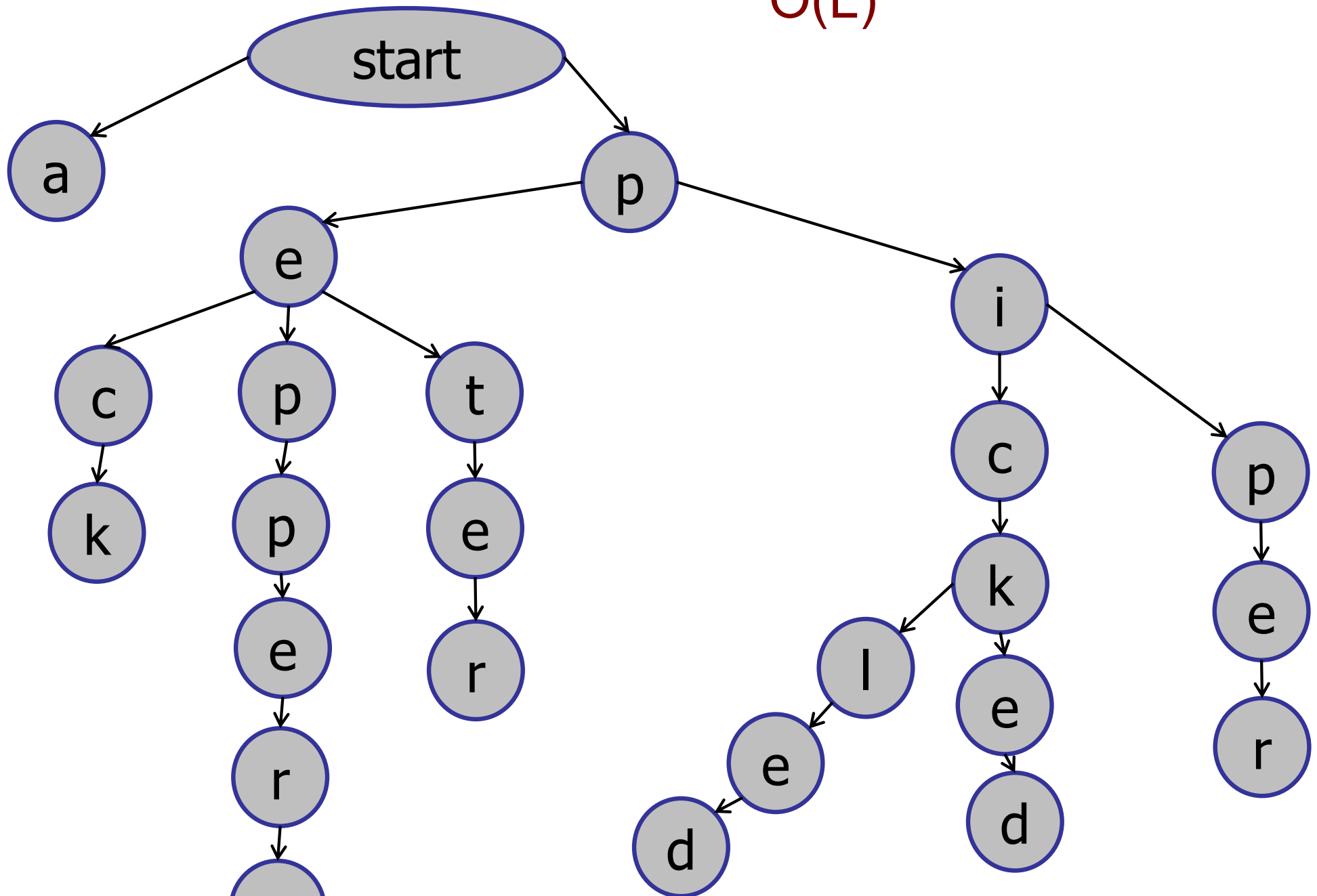
Cost to search for a string of length L?



# Trie

Cost to search for a string of length L?

$O(L)$



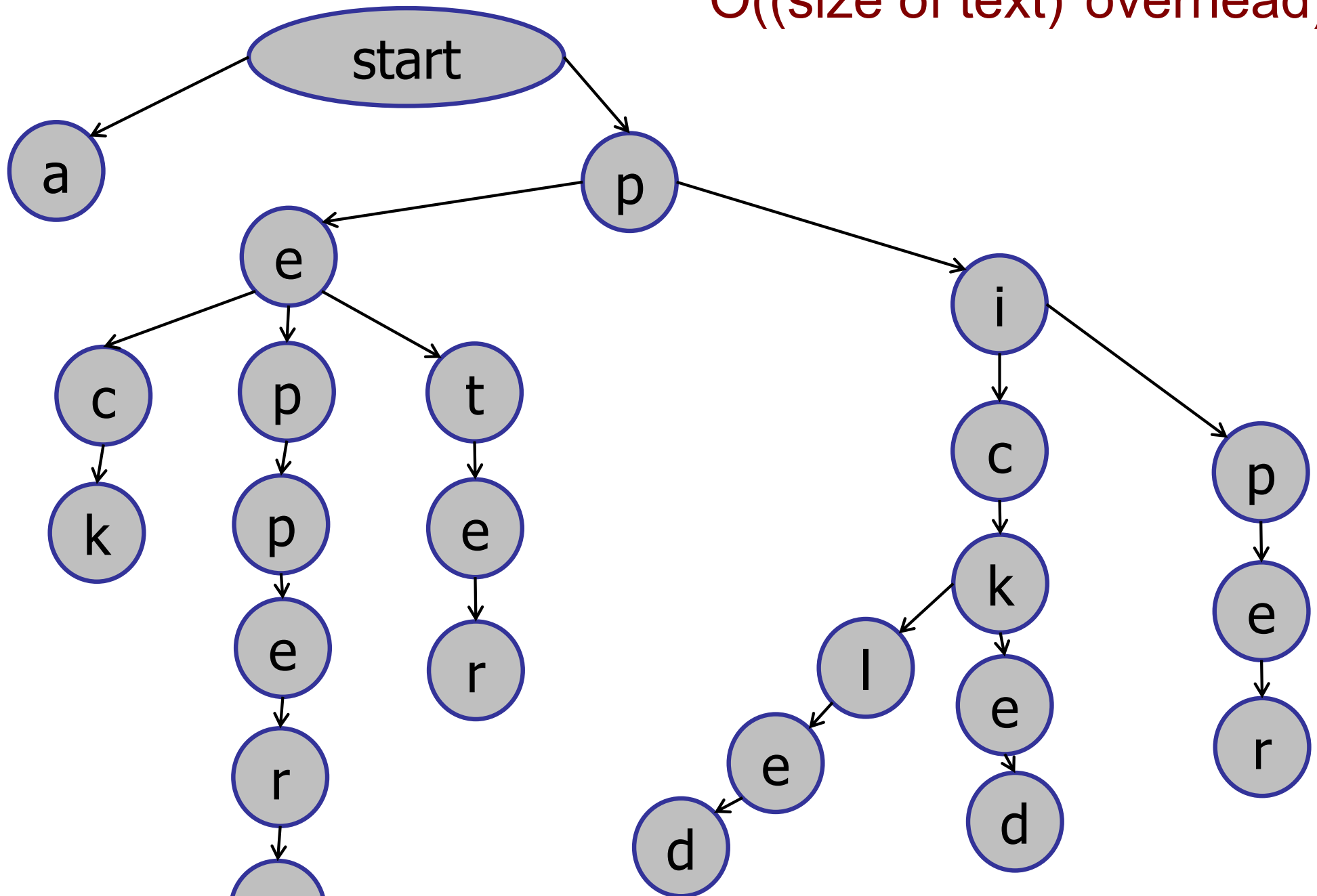
Space for storing a try?  
O(size of text)



# Trie

Space for storing a try?

$O((\text{size of text}) * \text{overhead})$



# Trie Tradeoffs

---

Time:

- Trie tends to be faster:  $O(L)$  vs.  $O(Lh)$ .
- Does not depend on number of strings.

Even faster if string is not in trie!

# Trie Tradeoffs

---

## Time:

- Trie tends to be faster:  $O(L)$ .
- Does not depend on size of total text.
- Does not depend on number of strings.

## Space:

- Trie tends to use more space.
- BST and Trie use  $O(\text{text size})$  space.
- But Trie has more nodes and more overhead.

# Trie Space

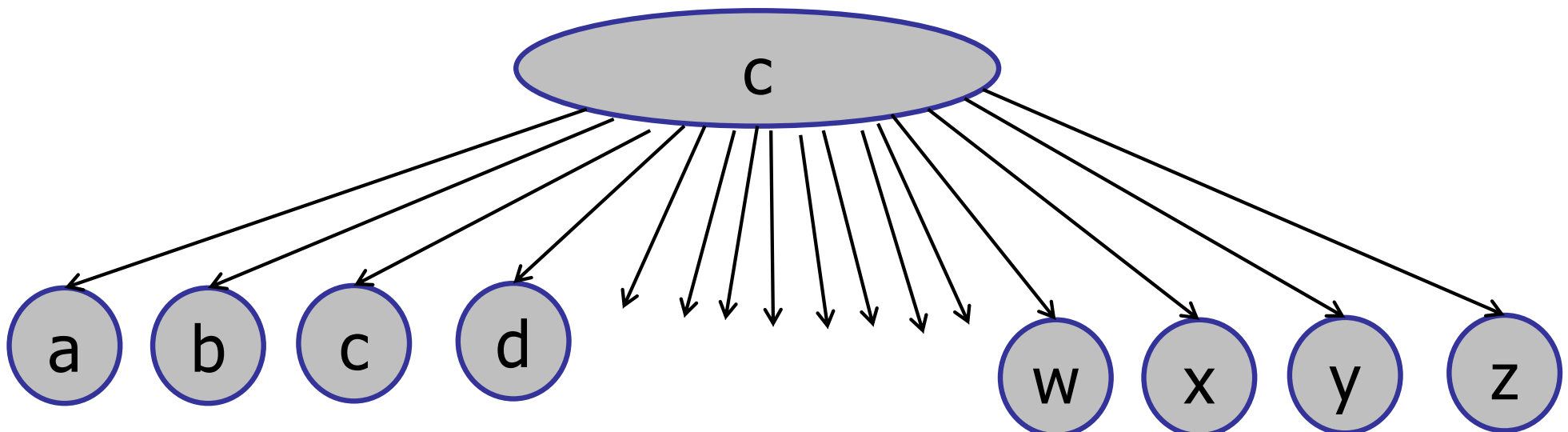
---

## Trie node:

- Has many children.
- For strings: fixed degree.
- Ascii character set: 256

wasted space?

```
TrieNode children[] = new TrieNode[256];
```





# Trie Applications

---

## String dictionaries

- Searching
- Sorting / enumerating strings

## Partial string operations:

- Prefix queries: find all the strings that start with pi.
- Long prefix: what is the longest prefix of “pickling” in the trie?
- Wildcards: find a string of the form “pi??le” in the trie.

# Today's Plan

---

## **On the importance of being balanced**

- Height-balanced binary search trees
- AVL trees
- Rotations

## **Tries**

- How to handle text?

## **Data structure design**

- How to build new structures on existing ideas?

# Dynamic Data Structures

---

1. Maintain a set of items
2. Modify the set of items
3. Answer queries.

Big picture idea:

Trees are a good way to store, summarize, and search dynamic data.

# Dynamic Data Structures

---

- Operations that create a data structure
  - build (preprocess)
- Operations that modify the structure
  - insert
  - delete
- Query operations
  - search, select, etc.

“Why do we need to learn how an AVL tree works?”

Just use a Java TreeMap, right?

“Why do we need to learn how an AVL tree works?”

1. Learn how to think like a computer scientist.

“Why do we need to learn how an AVL tree works?”

1. Learn how to think like a computer scientist.
2. Learn to modify existing data structures to solve new problems.

# Augmented Data Structures

---

Many problems require storing additional data in a standard data structure.

Augment more frequently than invent...



# Plan

---

Three examples of augmenting balanced BSTs

1. Order Statistics
2. Interval Queries
3. Orthogonal Range Searching

# Augmenting data structures

---

## Basic methodology:

1. Choose underlying data structure  
(tree, hash table, linked list, stack, etc.)

# Augmenting data structures

---

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---

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1. Choose underlying data structure  
(tree, hash table, linked list, stack, etc.)
2. Determine additional info needed.
3. Modify data structure to *maintain* additional info when the structure changes.  
(subject to insert/delete/etc.)

# Augmenting data structures

---

## Basic methodology:

1. Choose underlying data structure  
(tree, hash table, linked list, stack, etc.)
2. Determine additional info needed.
3. Modify data structure to *maintain* additional info when the structure changes.  
(subject to insert/delete/etc.)
4. Develop new operations.

# Plan

---

Three examples of augmenting balanced BSTs

1. Order Statistics
2. Interval Queries
3. Orthogonal Range Searching

# Order Statistics

---

Input

A set of integers.

Output: `select(k)`

The  $k^{\text{th}}$  item in the set.

52	7	13	43	22	92	18	9	65	67	87	25
----	---	----	----	----	----	----	---	----	----	----	----



`select(4)`

select(2) returns:

52	7	13	43	22	92	18	9	65	67	87	25
----	---	----	----	----	----	----	---	----	----	----	----

1. 52
- ✓ 2. 9
3. 13
4. 43
5. 25





# Order Statistics

---

Input

A set of integers.

Output: `select(k)`

The  $k^{\text{th}}$  item in the set.

52	7	13	43	22	92	18	9	65	67	87	25
----	---	----	----	----	----	----	---	----	----	----	----



`select(4)`

# Order Statistics

---

Input

A set of integers.

Output: `select(k)`

The  $k^{\text{th}}$  item in the set.

7	9	13	18	22	25	43	52	65	67	87	92
---	---	----	----	----	----	----	----	----	----	----	----



`select(4)`

# Order Statistics

---

Input

A set of integers.

Output:  $\text{select}(k)$   $\longrightarrow$  Sort:  $O(n \log n)$

The  $k^{\text{th}}$  item in the set.

7	9	13	18	22	25	43	52	65	67	87	92
---	---	----	----	----	----	----	----	----	----	----	----

$\uparrow$   
 $\text{select}(4)$

# Order Statistics

---

Input

A set of integers.

Output:  $\text{select}(k)$   $\longrightarrow$  QuickSelect:  $O(n)$

The  $k^{\text{th}}$  item in the set.

7	9	13	18	22	25	43	52	65	67	87	92
---	---	----	----	----	----	----	----	----	----	----	----

$\uparrow$   
 $\text{select}(4)$

# Order Statistics

---

Solution 1:

Sort:  $O(n \log n)$

Solution 2:

QuickSelect:  $O(n)$

<b>7</b>	<b>9</b>	<b>13</b>	<b>18</b>	<b>22</b>	<b>25</b>	<b>43</b>	<b>52</b>	<b>65</b>	<b>67</b>	<b>87</b>	<b>92</b>
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**select(4)**

# Order Statistics

---

## Solution 1:

Preprocess: sort ---  $O(n \log n)$

Select:  $O(1)$

## Solution 2:

Preprocess: nothing ---  $O(1)$

QuickSelect:  $O(n)$

# Order Statistics

---

## Solution 1:

Preprocess: sort ---  $O(n \log n)$

Select:  $O(1)$

## Solution 2:

Preprocess: nothing ---  $O(1)$

QuickSelect:  $O(n)$

Trade-off: how many items to select?

# Dynamic Order Statistics

---

Implement a data structure that supports:

- insert(int key)
- delete(int key)

and also:

- select(int k)

7	9	13	18	22	25	43	52	65	67	87	92
---	---	----	----	----	----	----	----	----	----	----	----



select(4)



# Dynamic Order Statistics

---

Solution 1:

**Basic structure:** sorted array A.

**insert(int item):** add item to sorted array A.

**select(int k):** return A[k]

<b>7</b>	<b>9</b>	<b>13</b>	<b>18</b>	<b>22</b>	<b>25</b>	<b>43</b>	<b>52</b>	<b>65</b>	<b>67</b>	<b>87</b>	<b>92</b>
----------	----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------

# Dynamic Order Statistics

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Solution 2:

**Basic structure:** unsorted array A.

**insert(int item):** add item to end of array A.

**select(int k):** run QuickSelect(k)

<b>7</b>	<b>9</b>	<b>13</b>	<b>18</b>	<b>22</b>	<b>25</b>	<b>43</b>	<b>52</b>	<b>65</b>	<b>67</b>	<b>87</b>	<b>92</b>
----------	----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------

When is it more efficient to maintain a sorted array (Solution 1)?

- A. Always
- B. When there are more inserts than selects.
- ✓ C. When there are more selects than inserts.
- D. Never
- E. I'm confused.

ARCHIPELAGO

is open

# Dynamic Order Statistics

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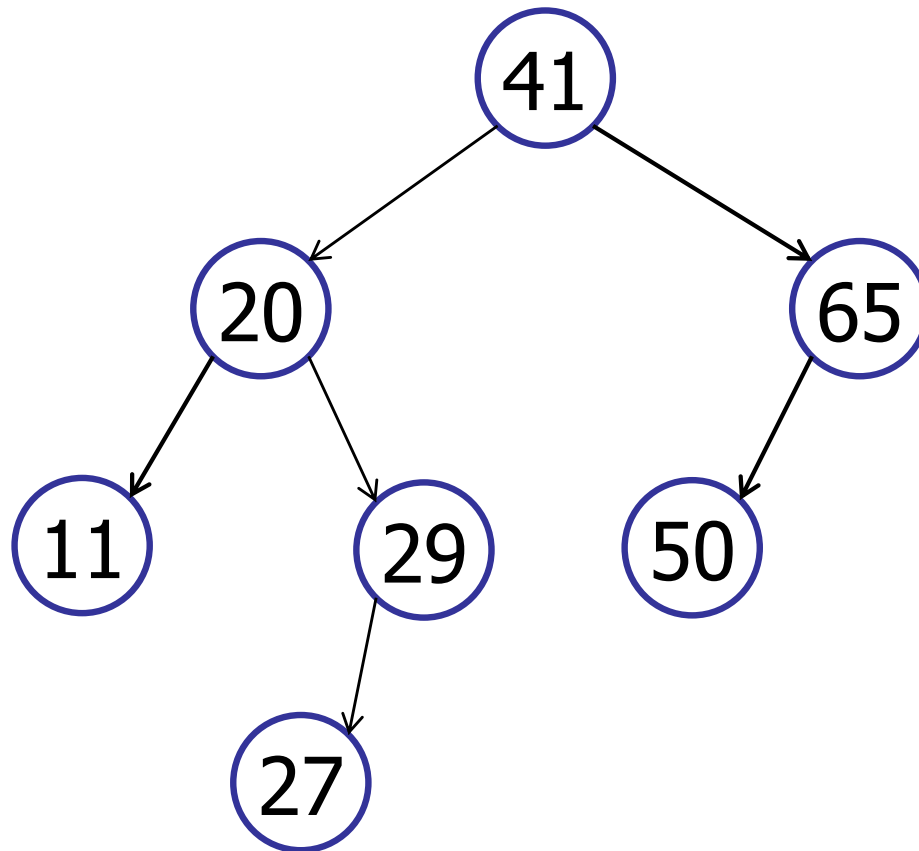
	Insert	Select
Solution 1: Sorted Array	$O(n)$	$O(1)$
Solution 2: Unsorted Array	$O(1)$	$O(n)$

7	9	13	18	22	25	43	52	65	67	87	92
---	---	----	----	----	----	----	----	----	----	----	----

# Dynamic Order Statistics

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**Today:** use a (balanced) tree

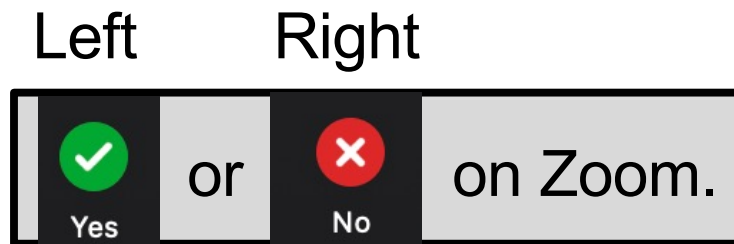
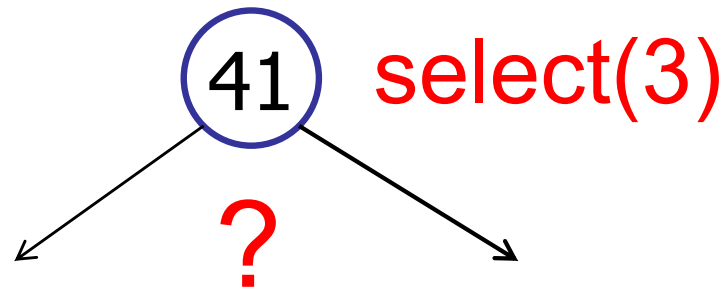


<b>11</b>	<b>20</b>	<b>27</b>	<b>29</b>	<b>41</b>	<b>50</b>	<b>65</b>
-----------	-----------	-----------	-----------	-----------	-----------	-----------

# Dynamic Order Statistics

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How to find the right item?

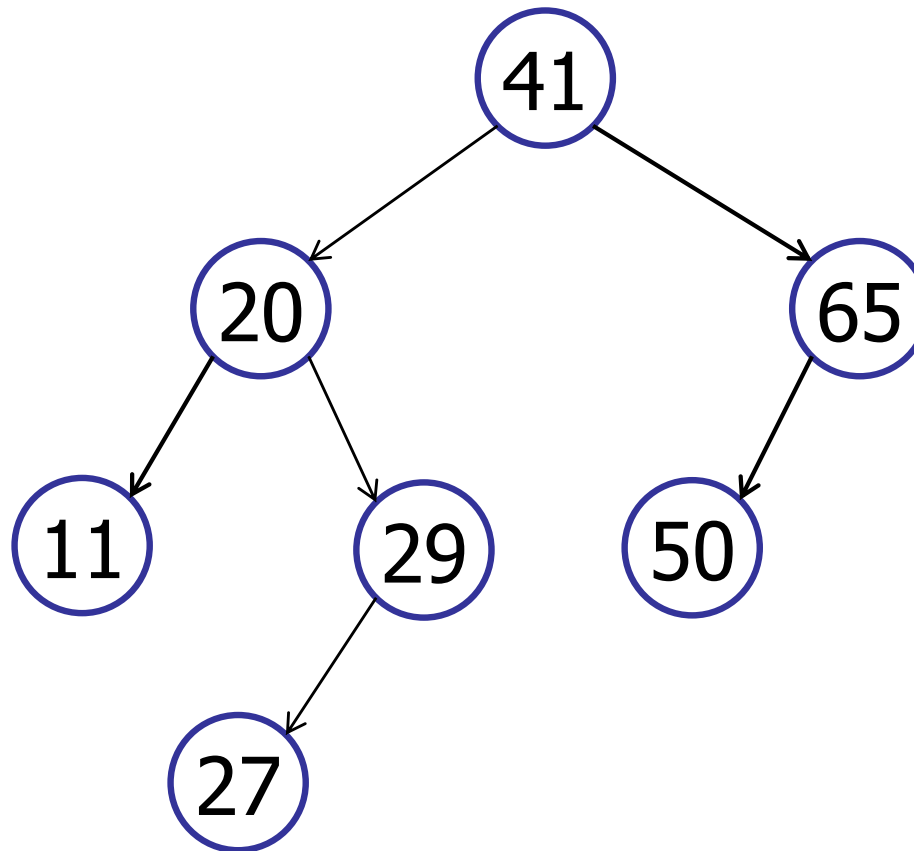


# Dynamic Order Statistics

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Simple solution: traversal

**select(k):**  $O(k)$   
in-order  
traversal



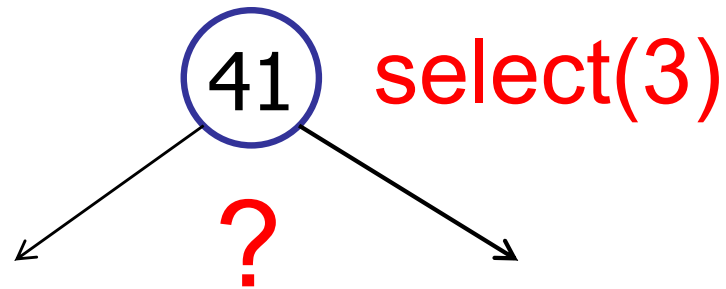
<b>11</b>	<b>20</b>	<b>27</b>	<b>29</b>	<b>41</b>	<b>50</b>	<b>65</b>
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# Dynamic Order Statistics

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Augment!

What extra information would help?

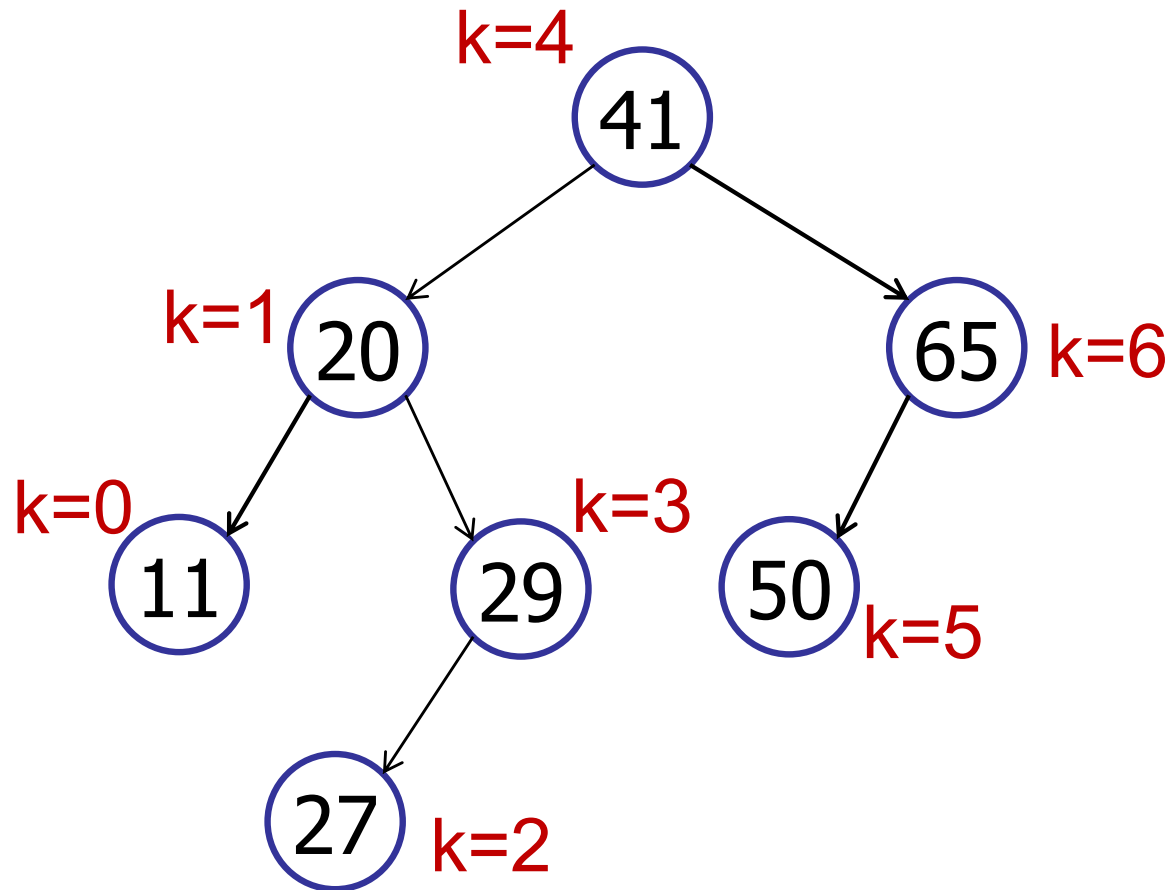




# Dynamic Order Statistics

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**Idea:** store rank in every node

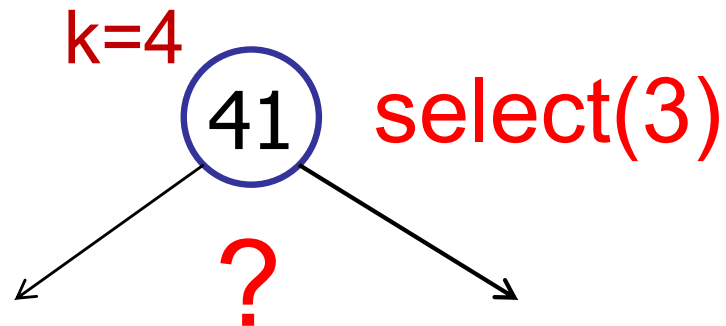


11	20	27	29	41	50	65
----	----	----	----	----	----	----

# Dynamic Order Statistics

---

**Idea:** store rank in every node

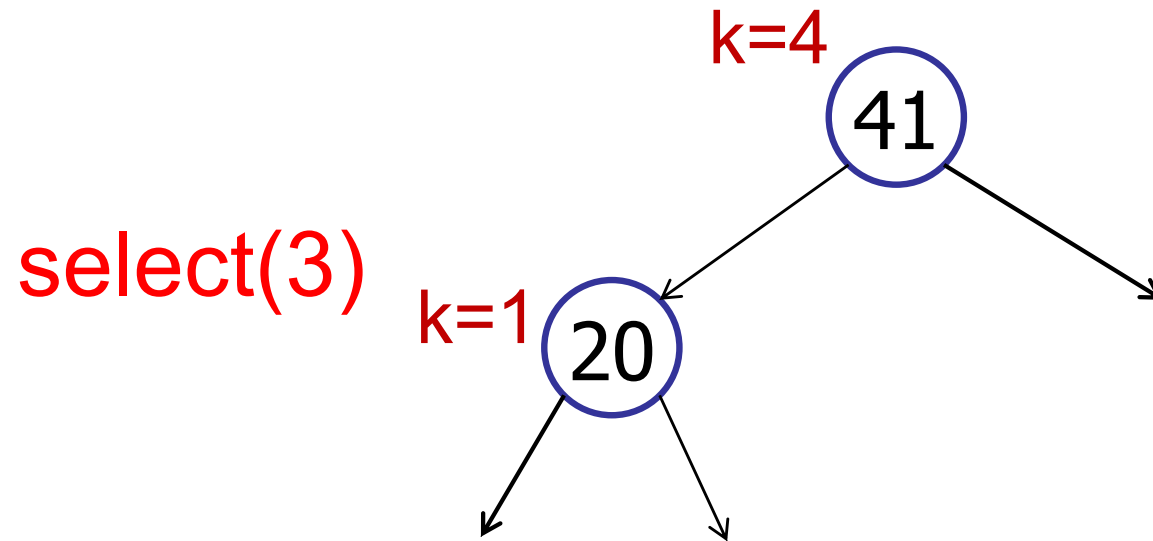


11	20	27	29	41	50	65
----	----	----	----	----	----	----

# Dynamic Order Statistics

---

**Idea:** store rank in every node

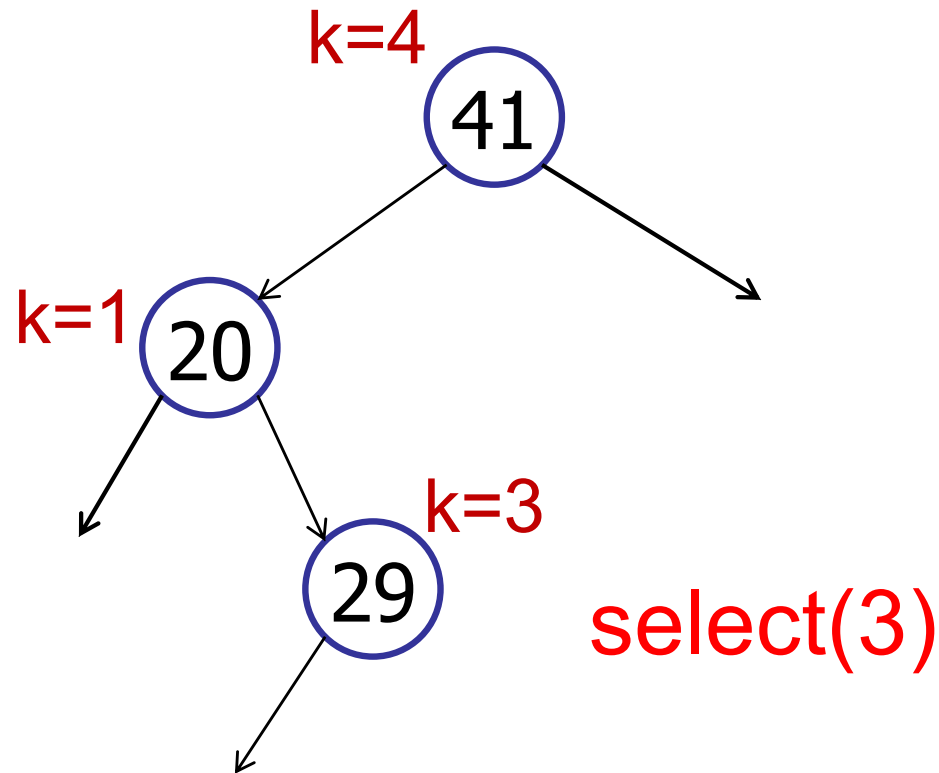


<b>11</b>	<b>20</b>	<b>27</b>	<b>29</b>	<b>41</b>	<b>50</b>	<b>65</b>
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# Dynamic Order Statistics

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**Idea:** store rank in every node

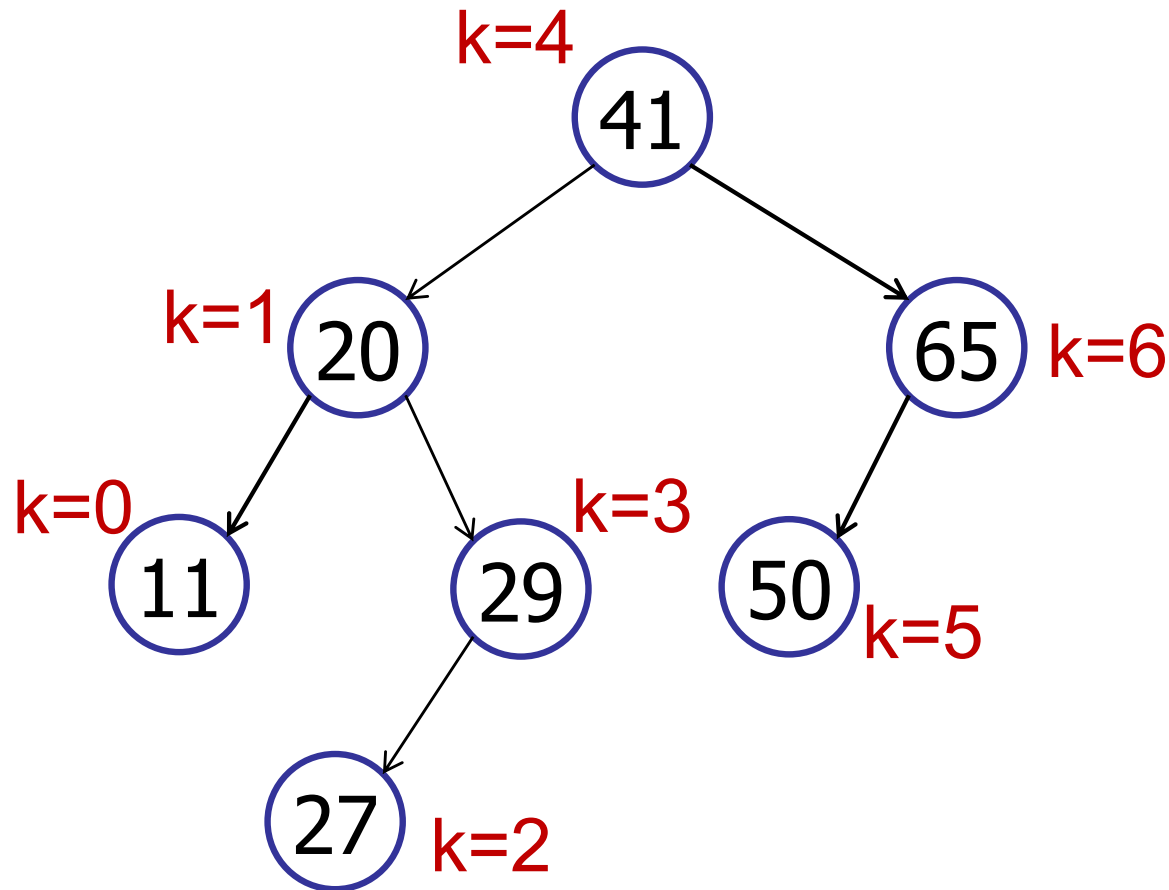


<b>11</b>	<b>20</b>	<b>27</b>	<b>29</b>	<b>41</b>	<b>50</b>	<b>65</b>
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# Dynamic Order Statistics

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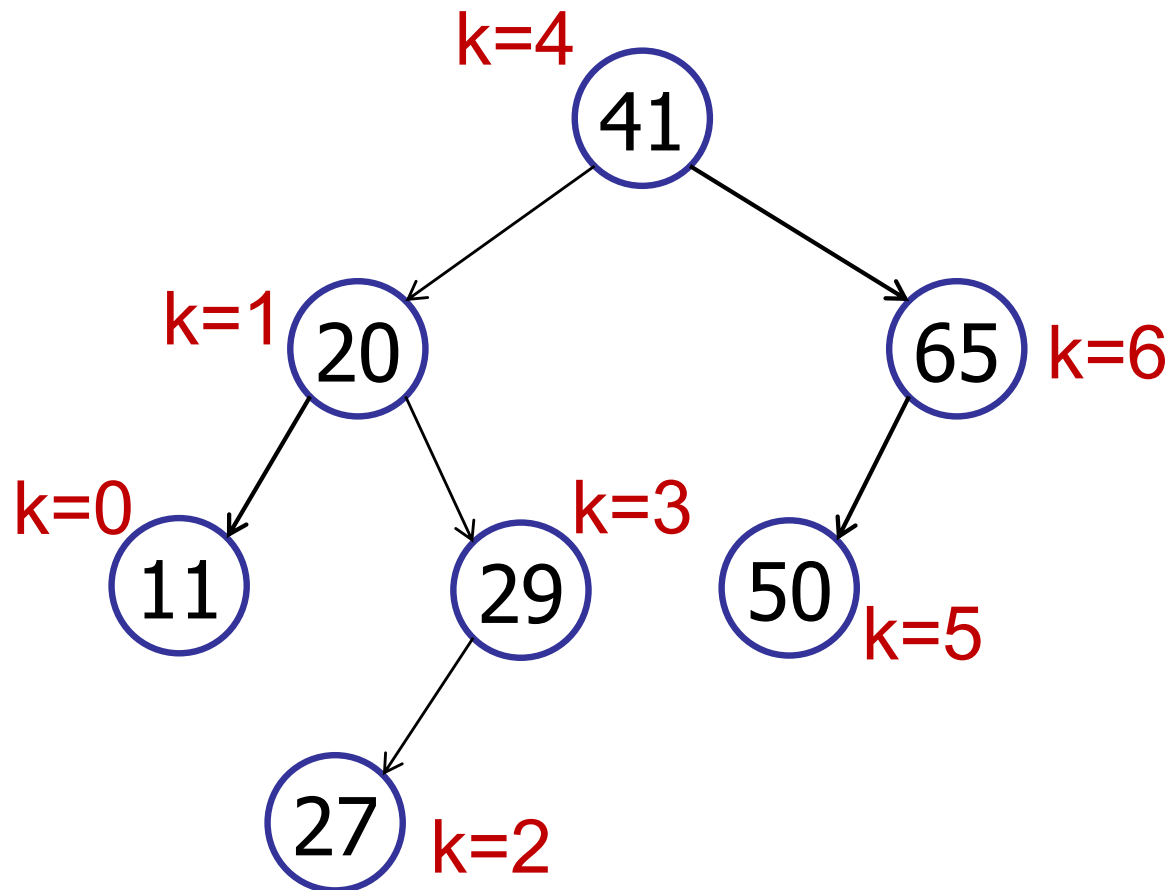
**Idea:** store rank in every node



11	20	27	29	41	50	65
----	----	----	----	----	----	----

# Dynamic Order Statistics

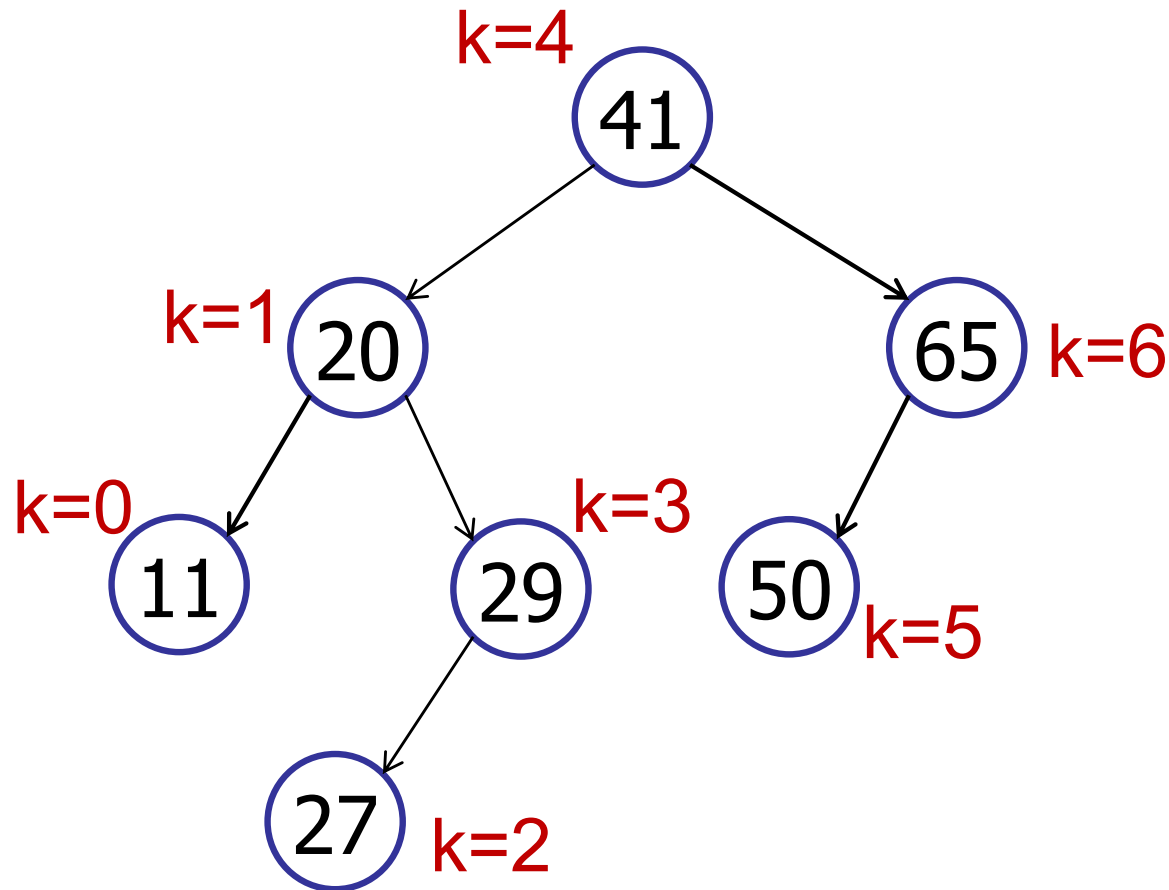
**Question:** What goes wrong if you store ranks on every node??



# Dynamic Order Statistics

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**Idea:** store rank in every node

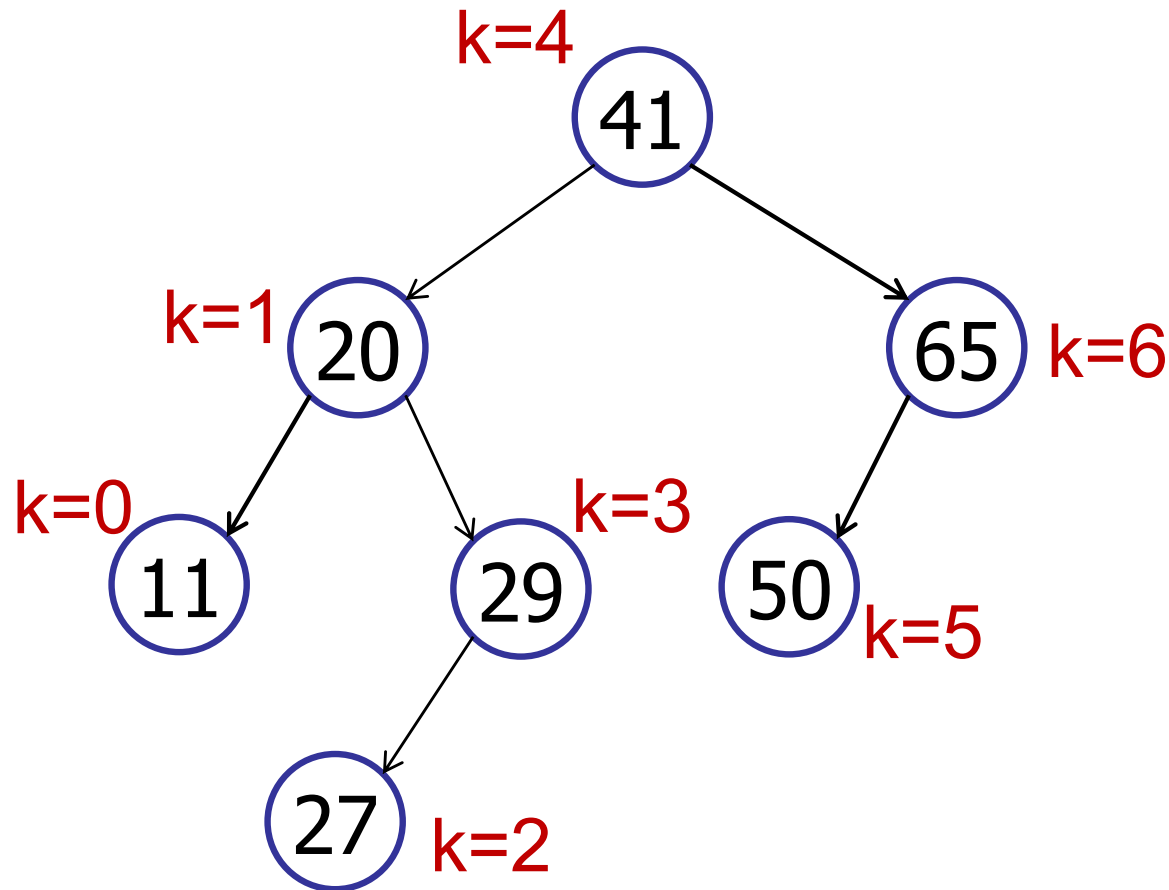


**Problem:** insert(5)

# Dynamic Order Statistics

---

**Idea:** store rank in every node



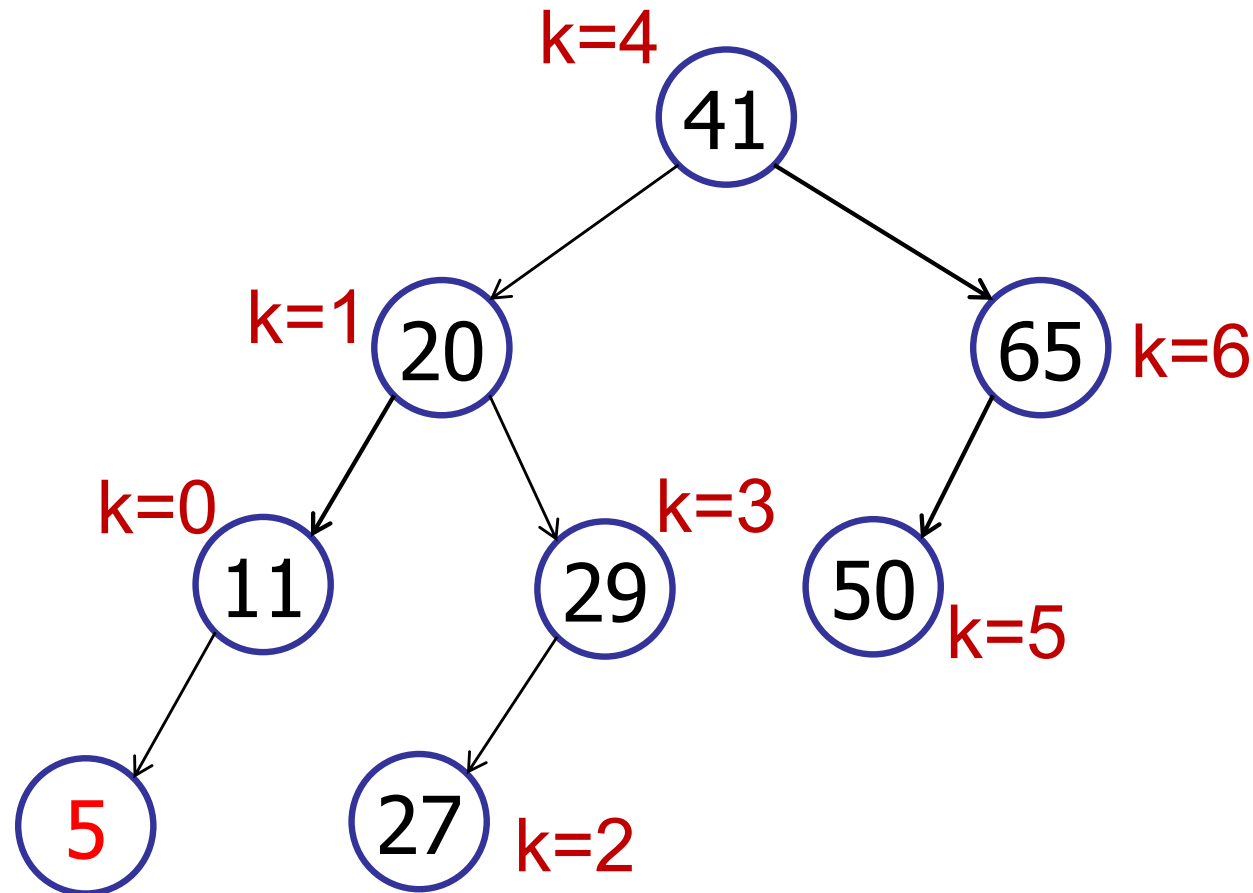
**Problem:** insert(5) requires updating *all* the ranks!



# Dynamic Order Statistics

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**Idea:** store rank in every node

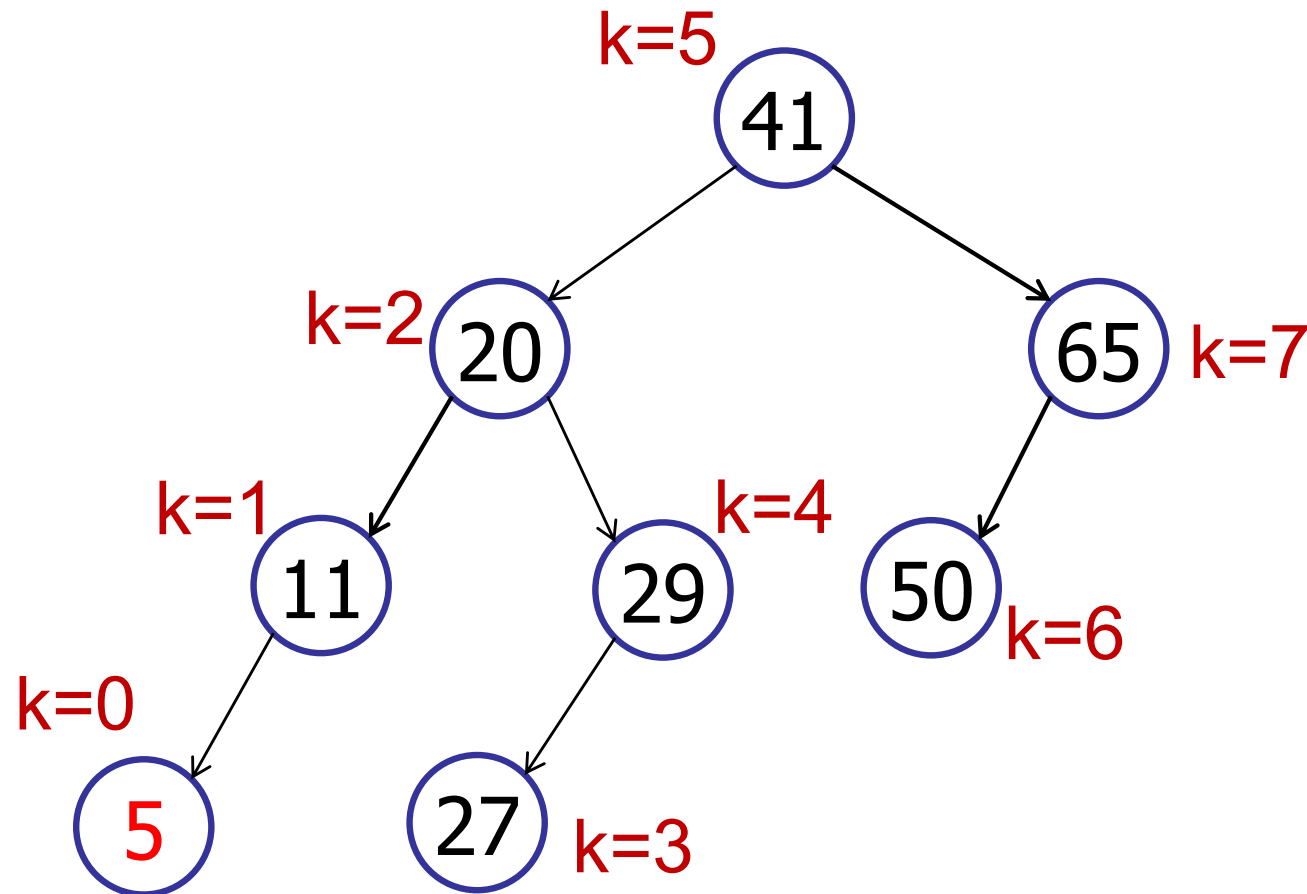


<b>5</b>	<b>11</b>	<b>20</b>	<b>27</b>	<b>29</b>	<b>41</b>	<b>50</b>	<b>65</b>
----------	-----------	-----------	-----------	-----------	-----------	-----------	-----------

# Dynamic Order Statistics

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**Conclusion:** too expensive to store rank in every node!

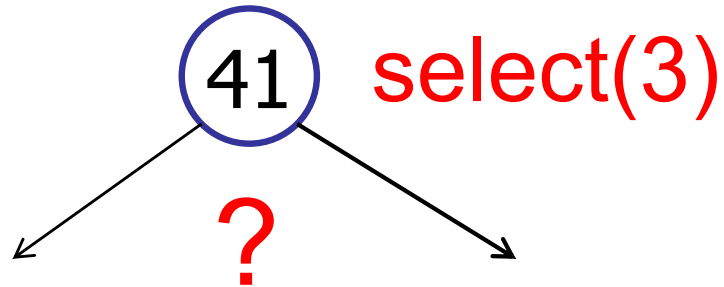


5	11	20	27	29	41	50	65
---	----	----	----	----	----	----	----

# Dynamic Order Statistics

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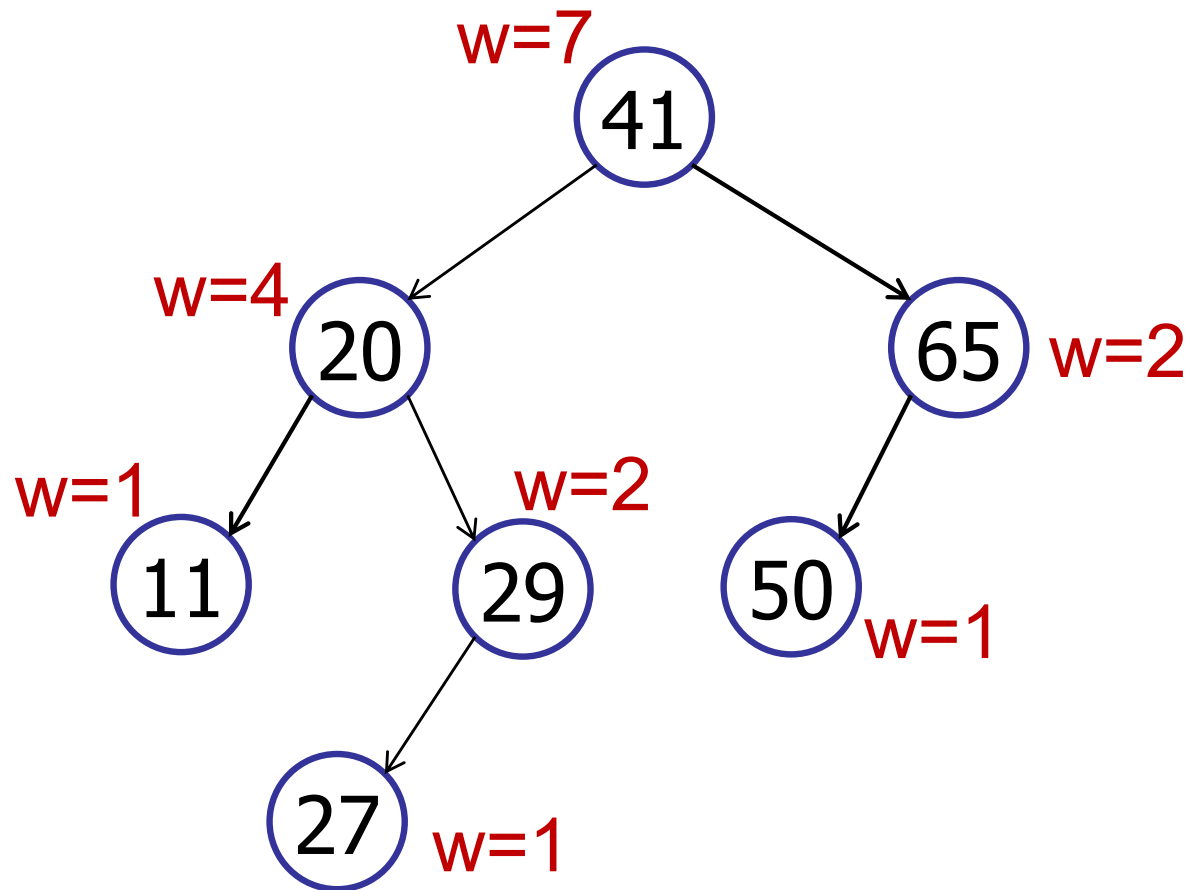
What should we store in each node?



# Dynamic Order Statistics

---

**Idea:** store *size* of sub-tree in every node



# Dynamic Order Statistics

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**Idea:** store size of sub-tree in every node

The weight of a node is the size of the tree rooted at that node.

Define weight:

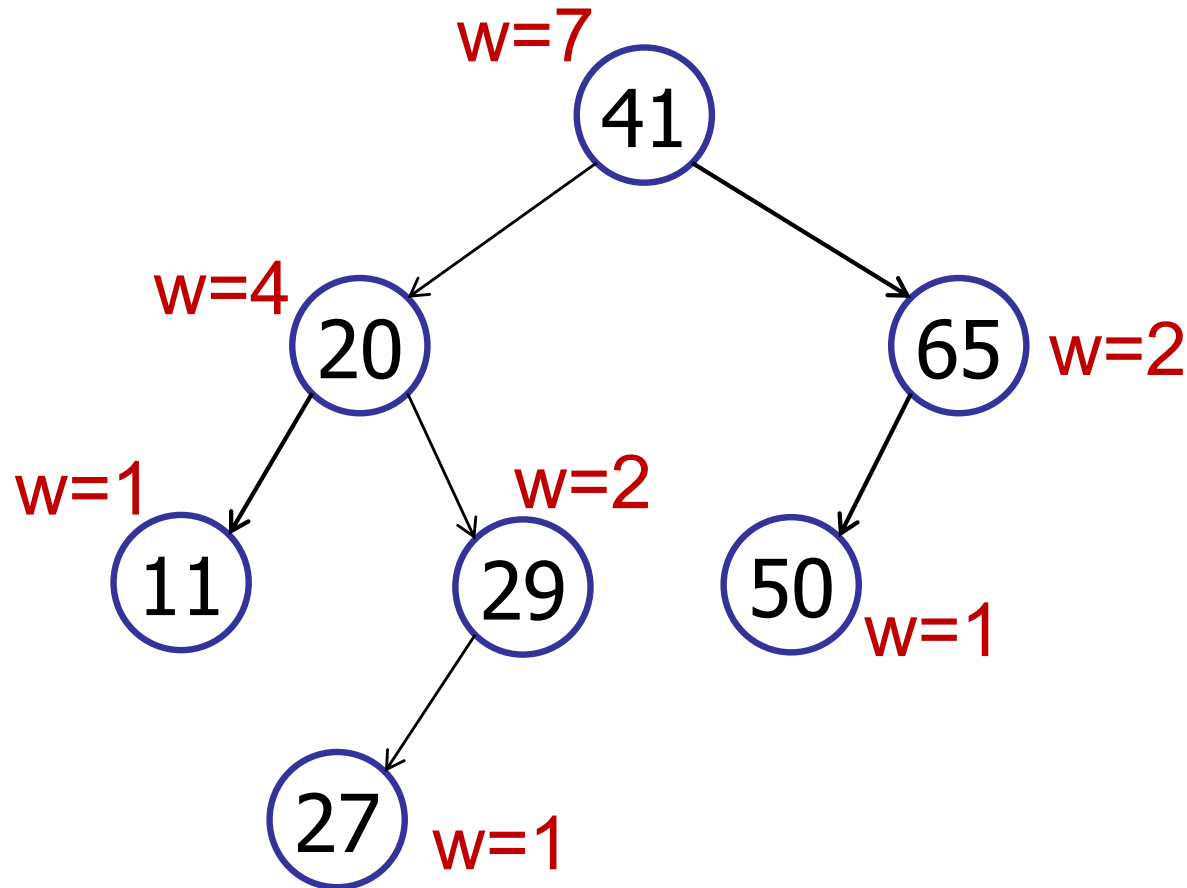
$$w(\text{leaf}) = 1$$

$$w(v) = w(v.\text{left}) + w(v.\text{right}) + 1$$

# Dynamic Order Statistics

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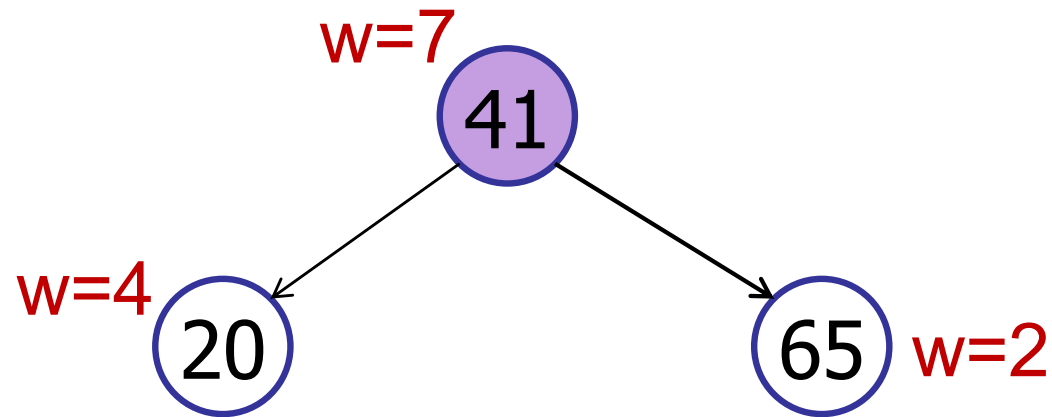
**Idea:** store *size* of sub-tree in every node



# Dynamic Order Statistics

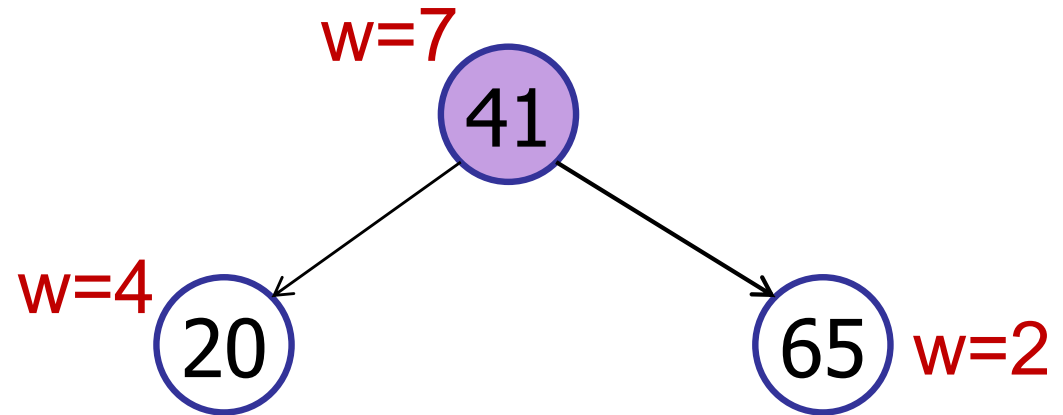
---

Example: `select(3)`



What is the rank of 41?

- 1. 1
- 2. 3
- ✓ 3. 5
- 4. 7
- 5. 9
- 6. Can't tell.



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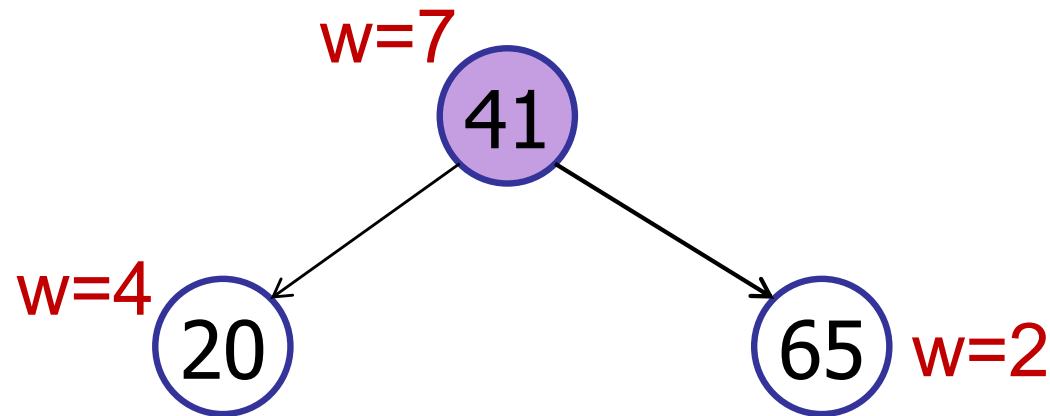
is open



# Dynamic Order Statistics

---

Example: `select(3)`



"rank in subtree" = left.weight + 1

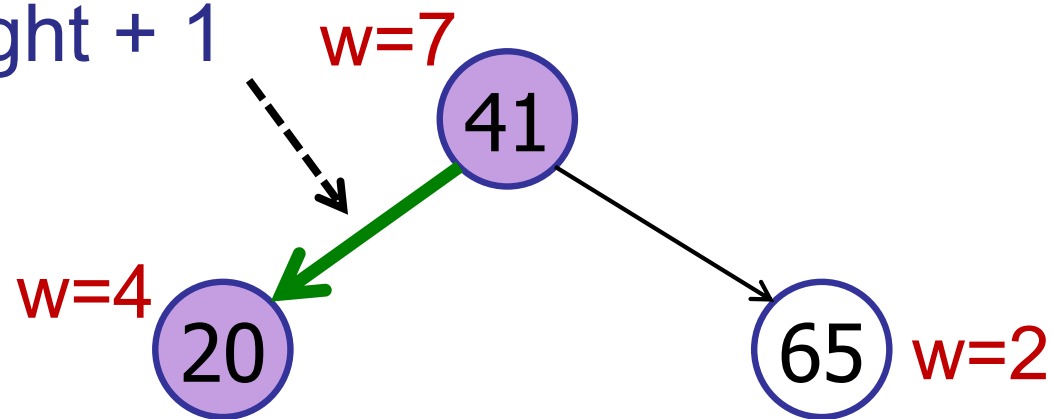
# Dynamic Order Statistics

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Example: `select(3)`

$3 < \text{left.weight} + 1$

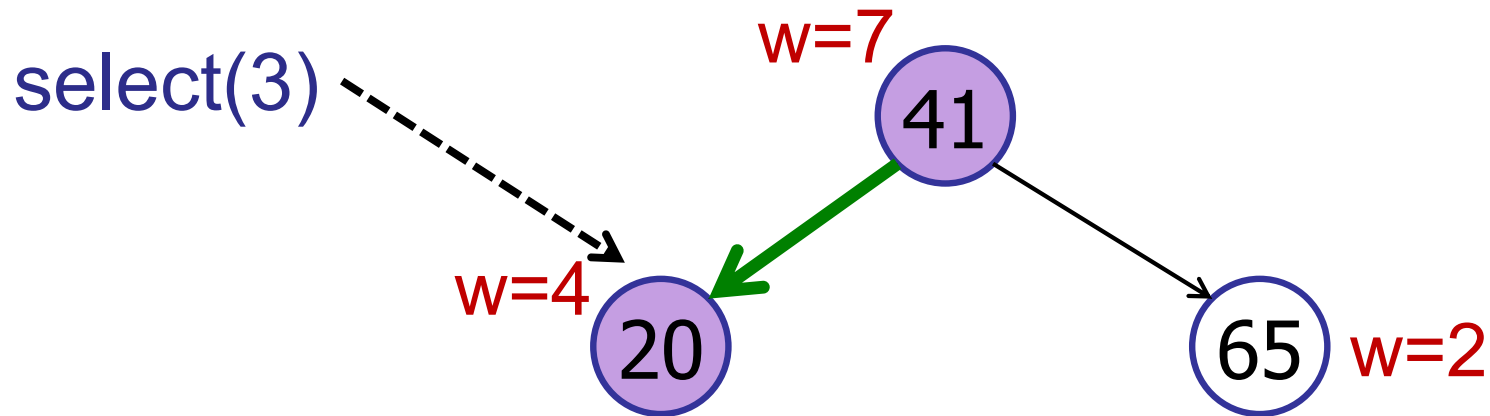
Go left!



# Dynamic Order Statistics

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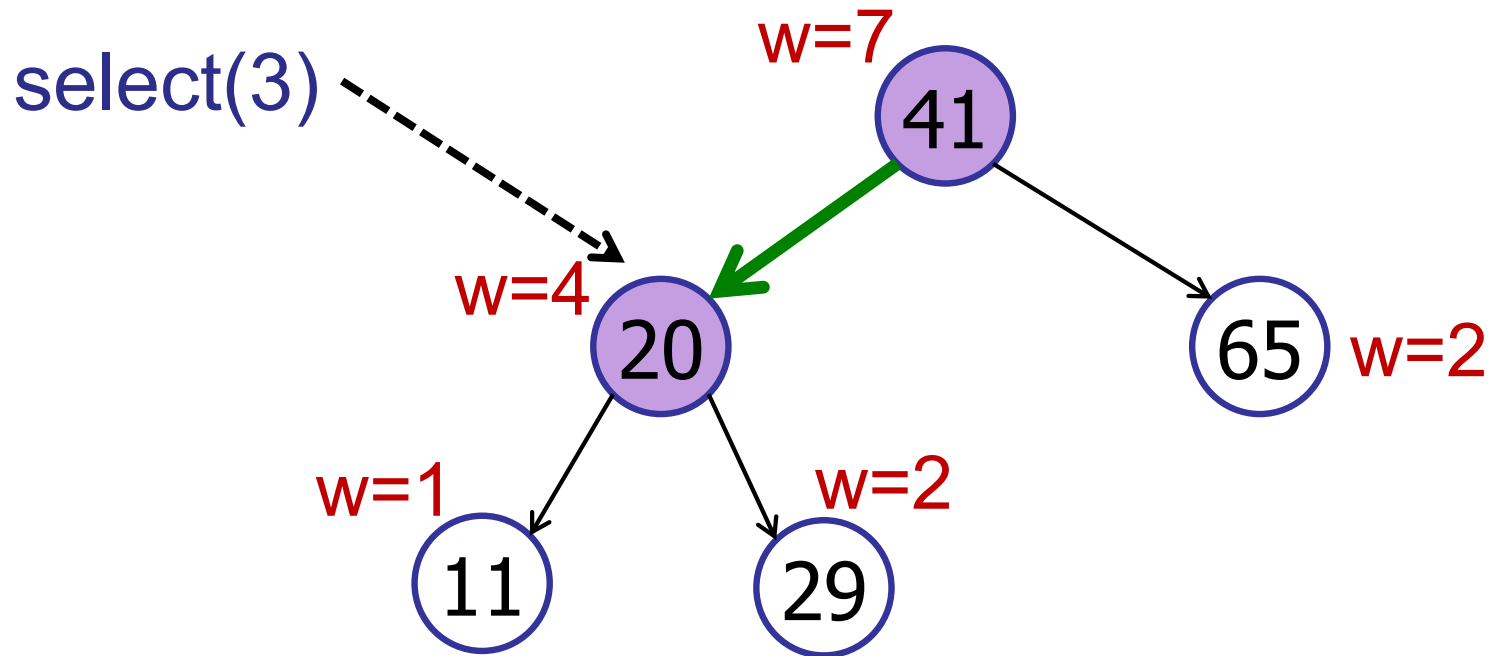
Example: `select(3)`



# Dynamic Order Statistics

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Example: `select(3)`

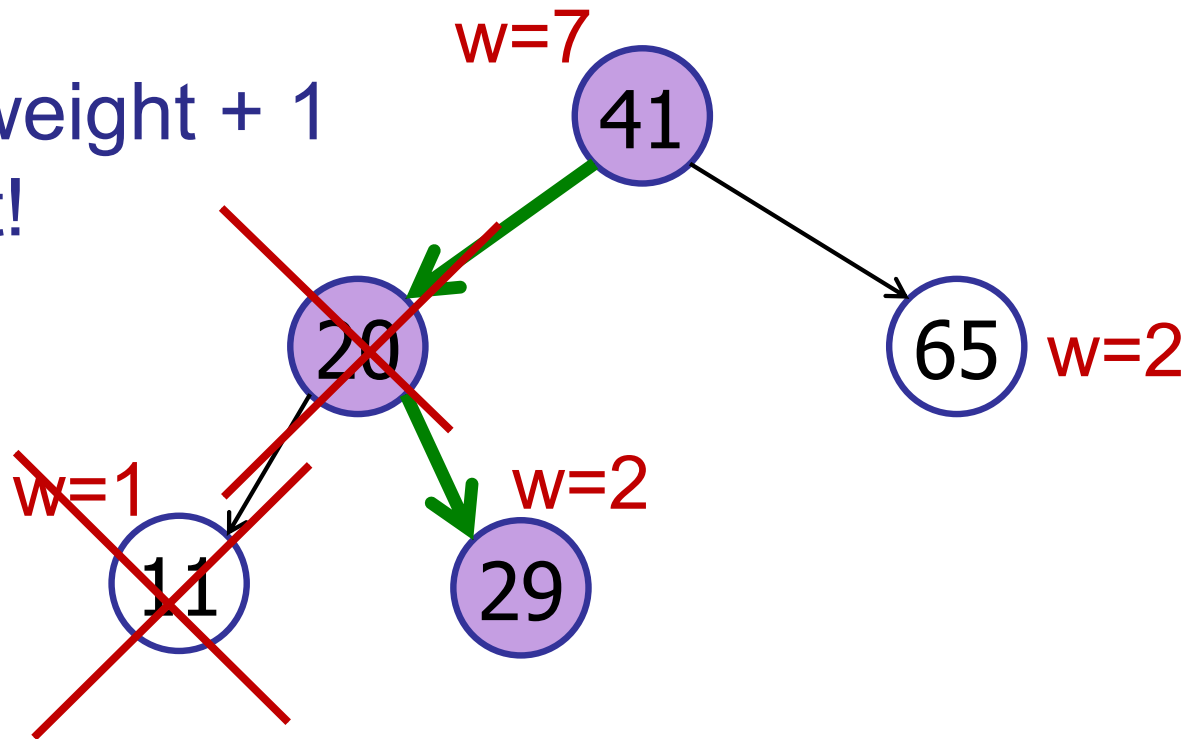


# Dynamic Order Statistics

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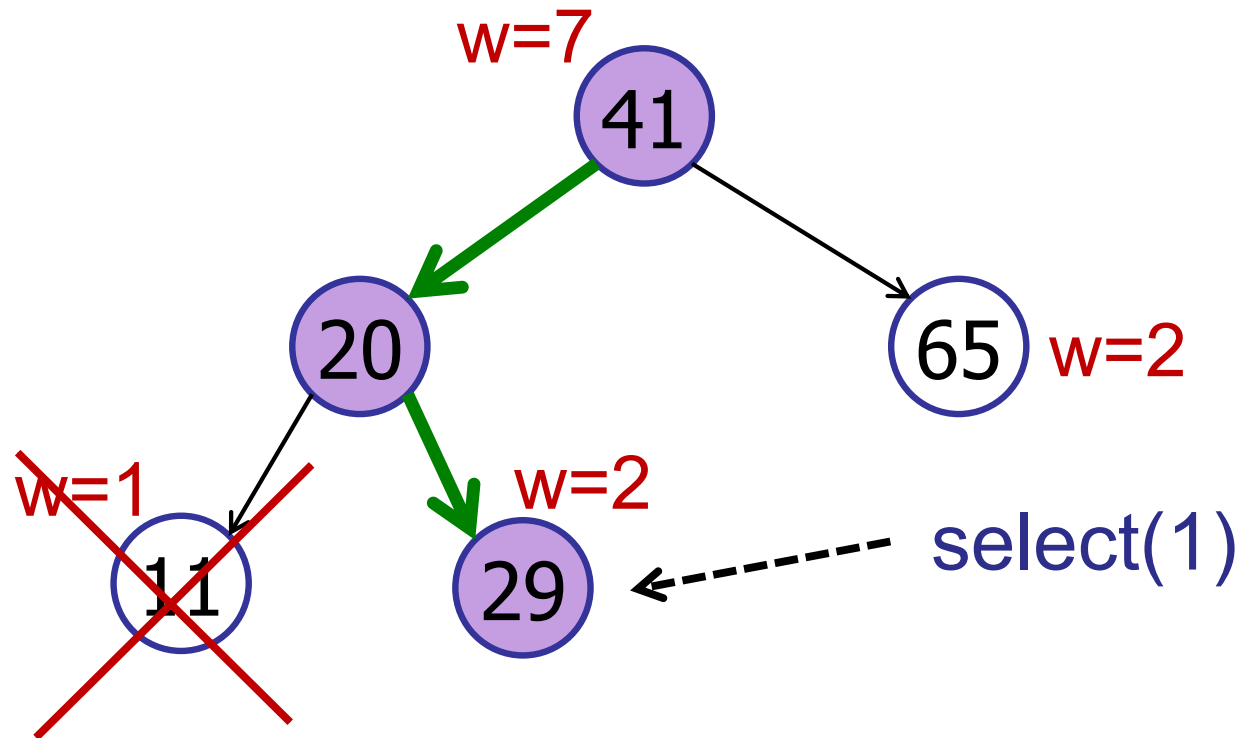
Example: `select(3)`

$3 > \text{left.weight} + 1$   
Go right!



# Dynamic Order Statistics

Example: `select(3)`



Item to select:

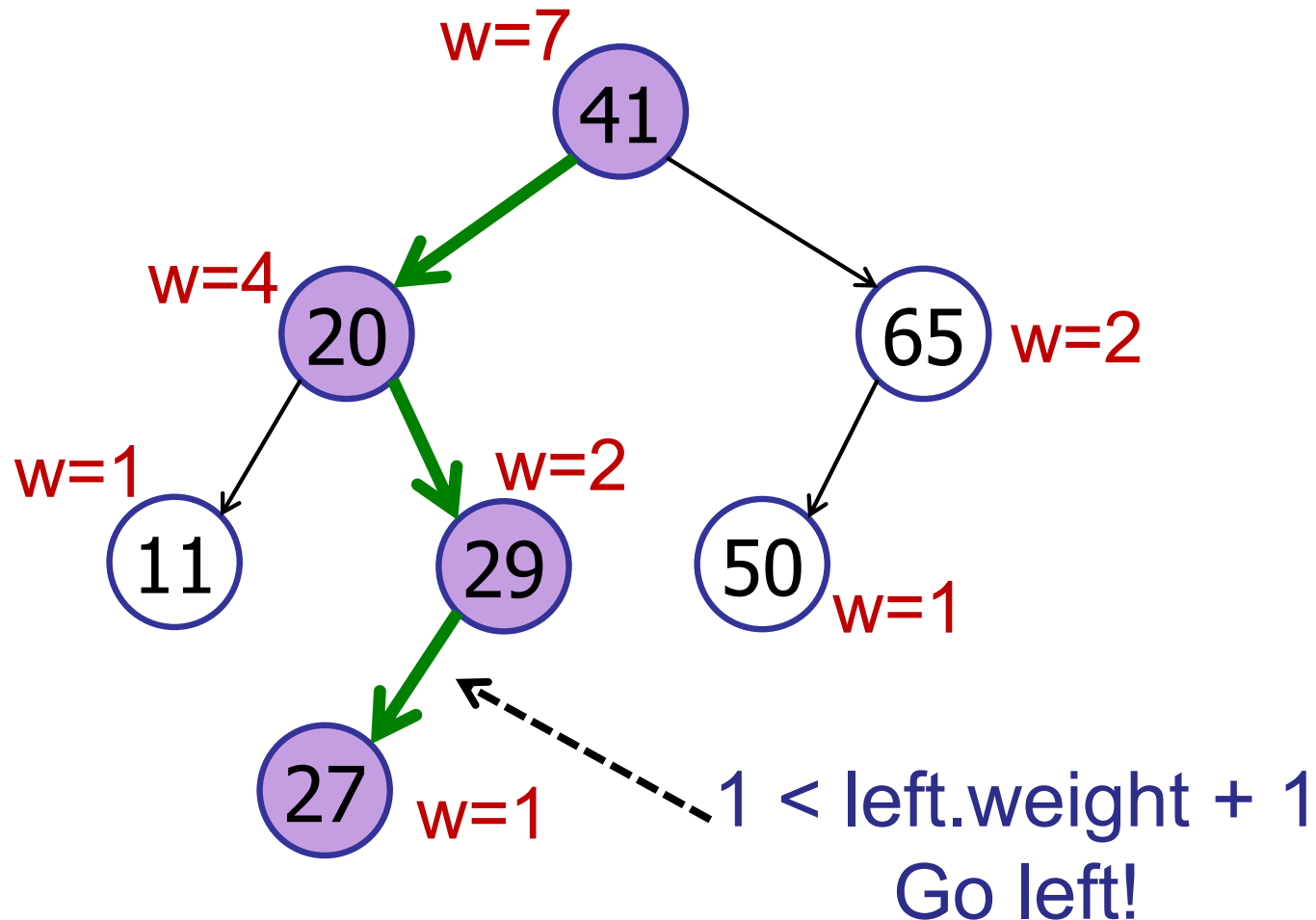
$$3 - (\text{left.weight} + 1) =$$

$$3 - (1 + 1) = 1$$

# Dynamic Order Statistics

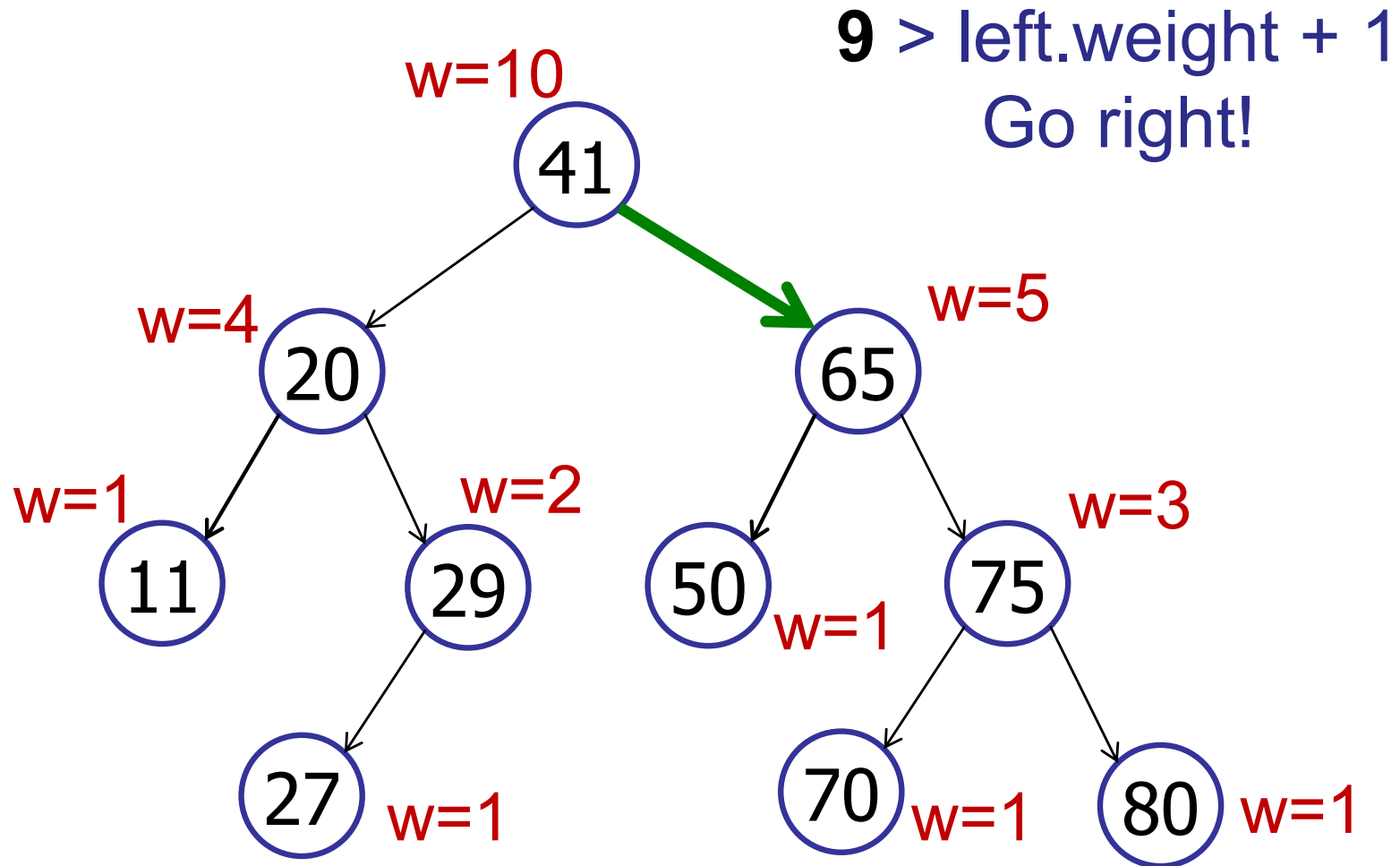
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Example: `select(3)`



# Dynamic Order Statistics

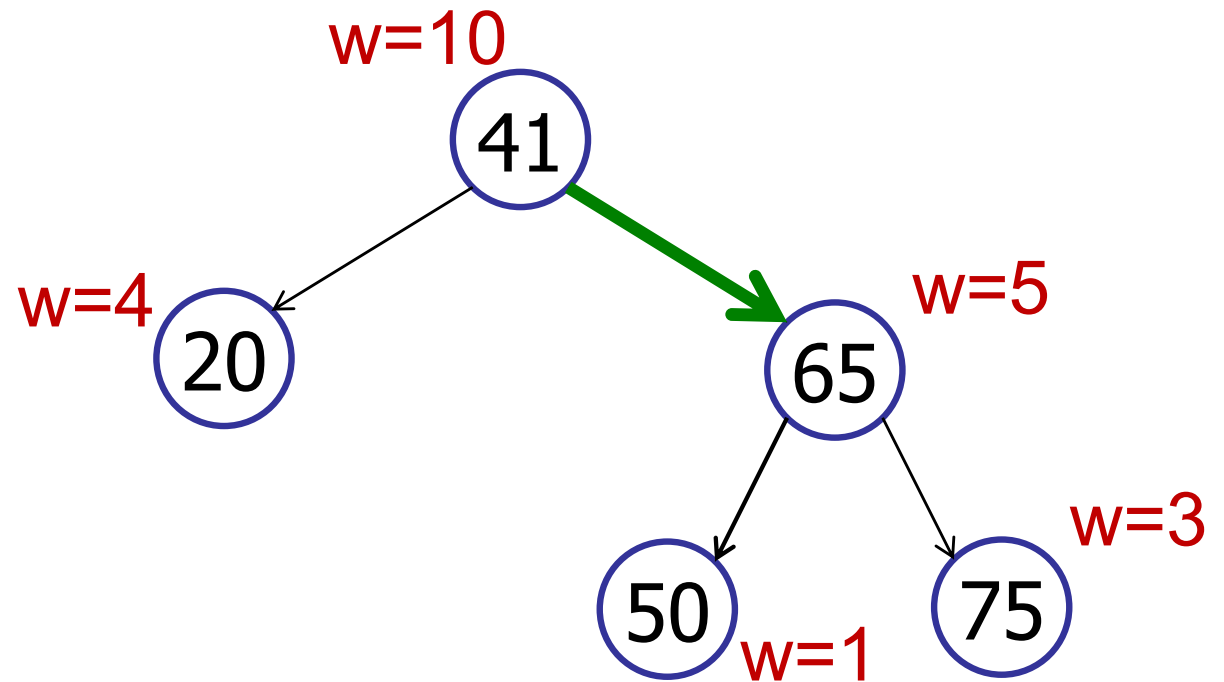
Example: `select(9)`





select(9)

1. Go left at 65
- ✓ 2. Go right at 65
3. Stop at 65
4. I'm confused

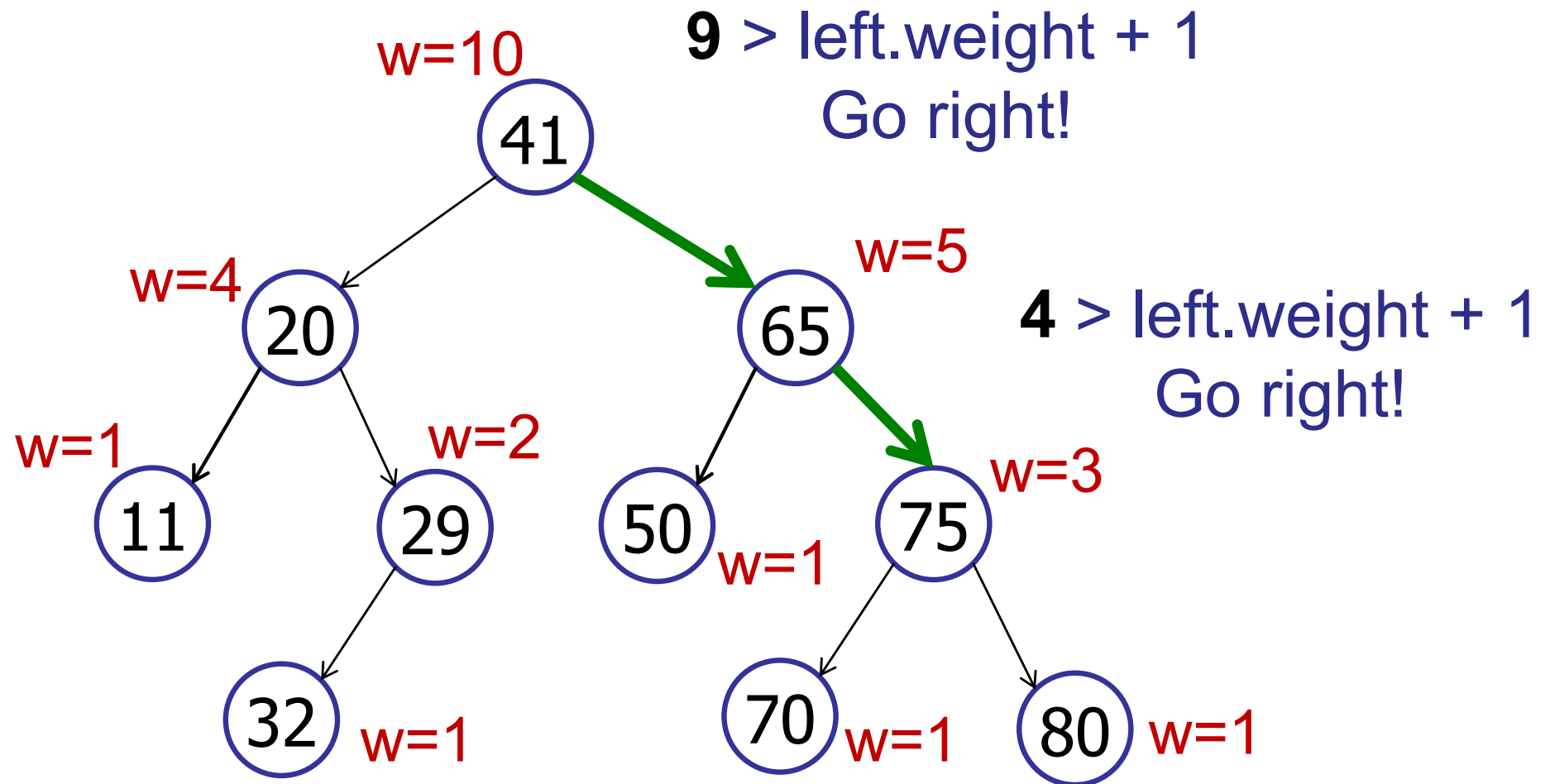


ARCHIPELAGO

is open

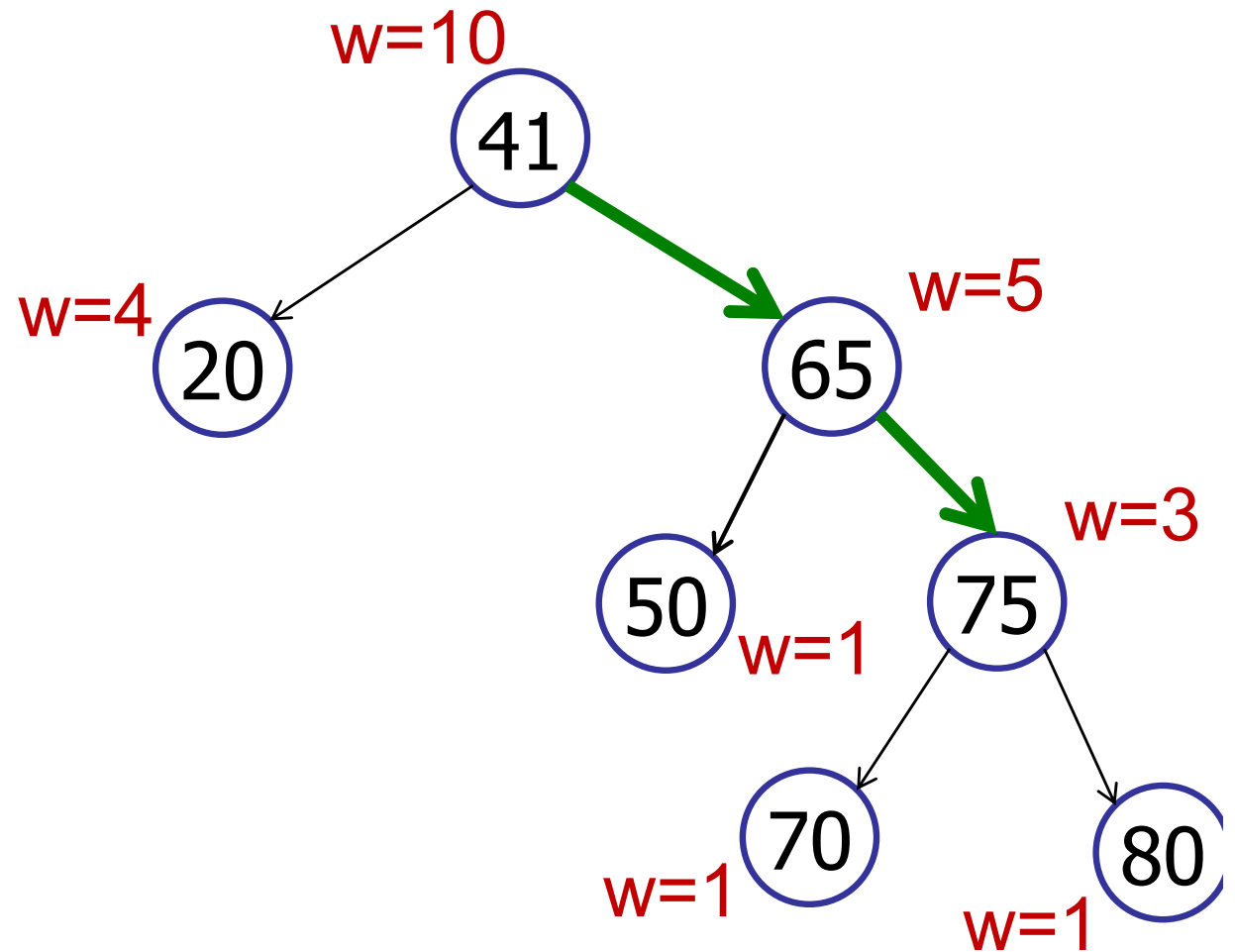
# Dynamic Order Statistics

select(9)



select(9)

1. Go left at 75
2. Go right at 75
- ✓ 3. Stop at 75
4. I'm confused

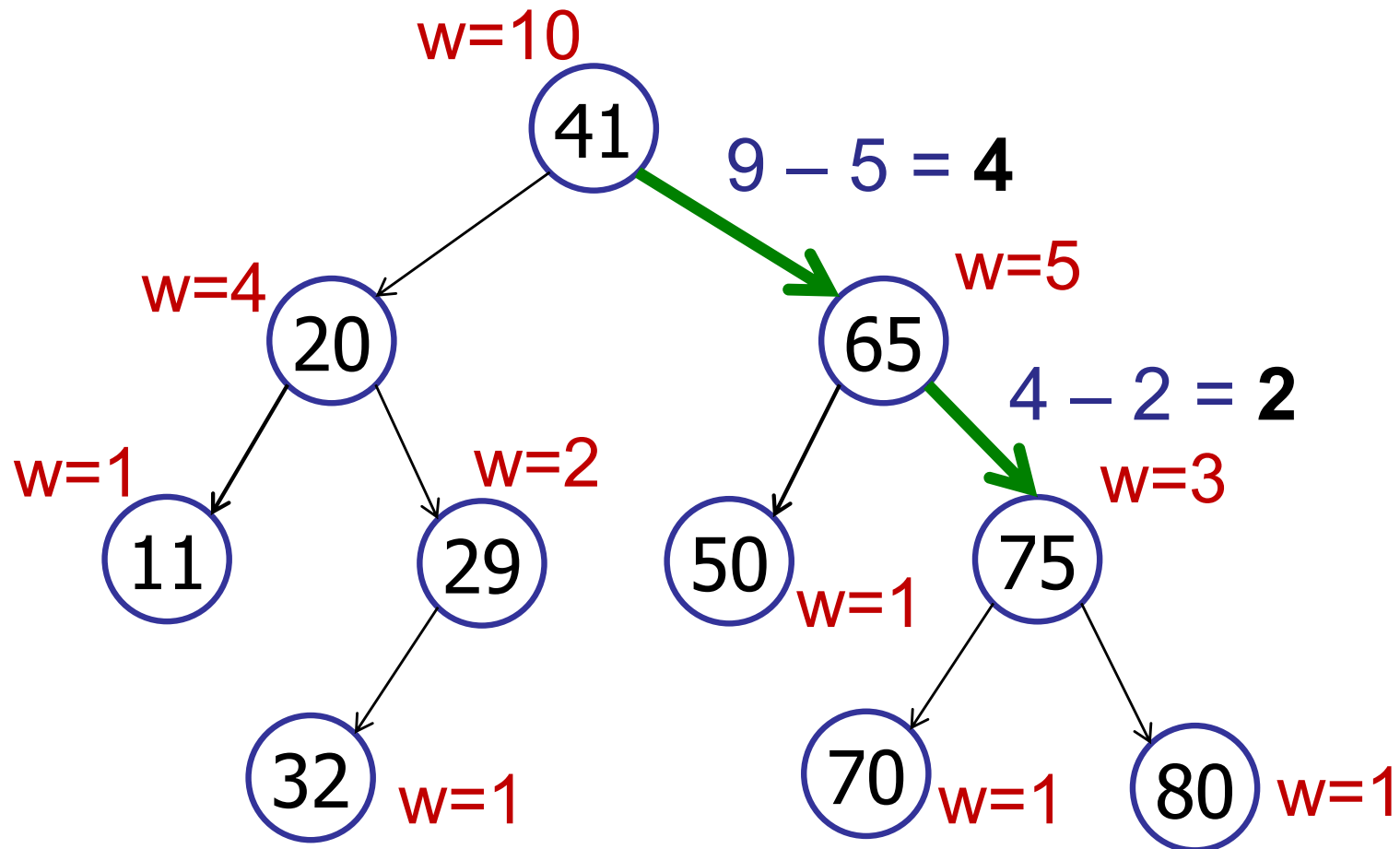


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is open

# Dynamic Order Statistics

select(9)



# Dynamic Order Statistics

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select(k)

rank = m\_left.weight + 1;

if (k == rank) then

return v;

else if (k < rank) then

return m\_left.select(k);

else if (k > rank) then

return m\_right.select(k-rank);

# Dynamic Order Statistics

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`select(k)` : finds the node with rank  $k$

Example: find the 10th tallest student in the class.

# Dynamic Order Statistics

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`select(k)` : finds the node with rank  $k$

Example: find the 10th tallest student in the class.

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`rank(v)` : computes the rank of a node  $v$

Example: determine the percentile of Johnny's height.  
Is Johnny in the 10<sup>th</sup> percentile or the 90<sup>th</sup> percentile?