

Data Structures and Algorithms – COMS21103

2014/2015

Minimum Spanning Trees via Disjoint Sets

Benjamin Sach

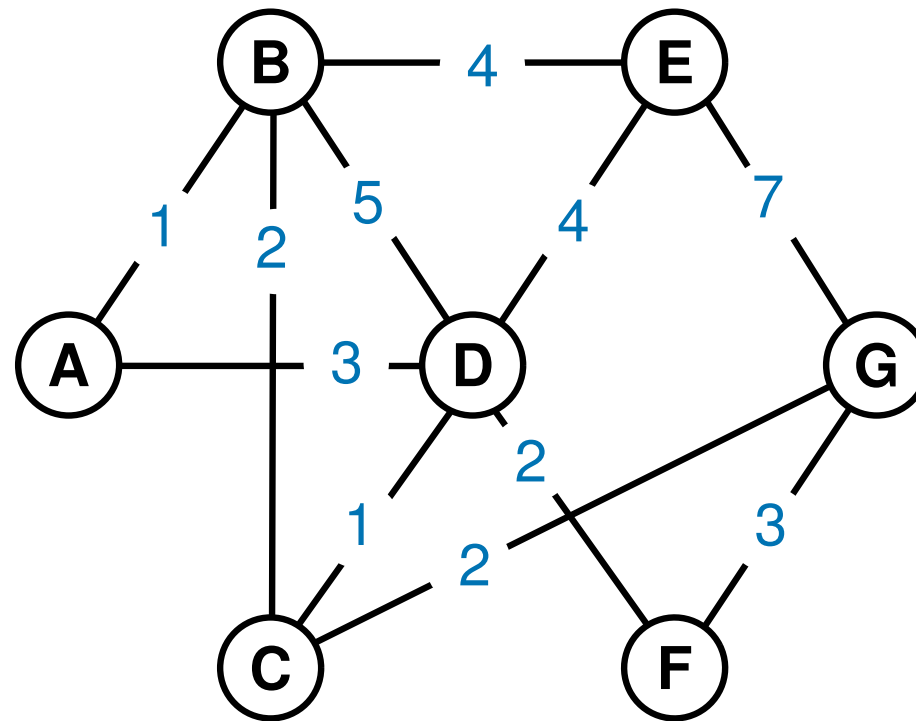
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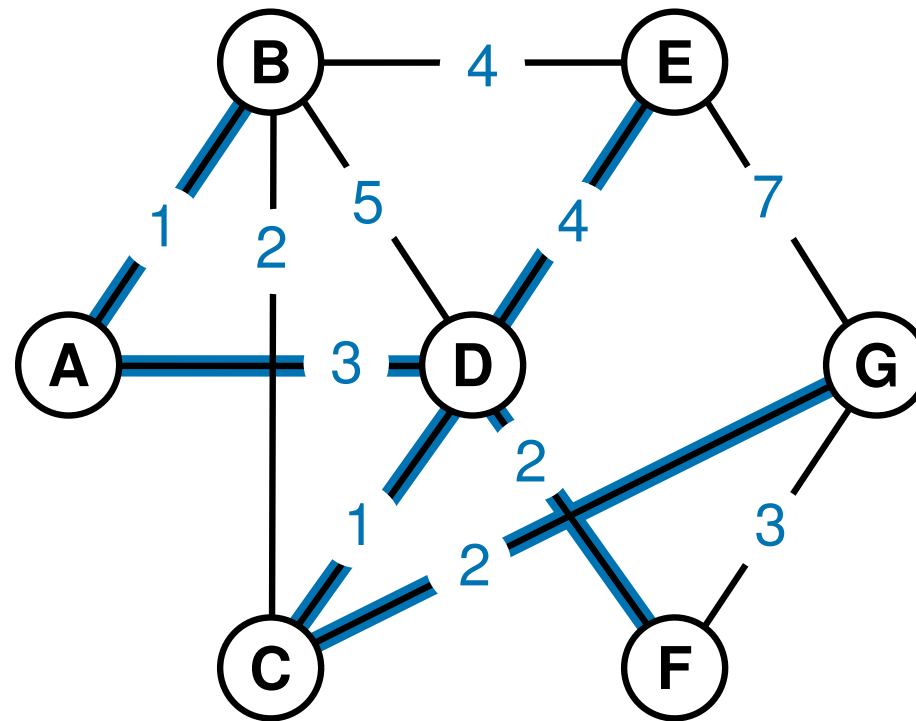
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In this lecture we will see an efficient data structure
for maintaining a collection of disjoint sets



We will then see how this data structure can be used to efficiently implement
Kruskal's algorithm which finds a minimum spanning tree in an undirected graph

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Disjoint set data structures

We will be interested in a **data structure** which
stores a collection of disjoint sets

The elements of the sets are numbers from $\{1, 2, \dots, n\}$

The following operations are supported:

MAKESET(x) - make a new set containing only x
 x cannot be a member of any existing set

UNION(x, y) - merge the sets containing x and y into a single set

FINDSET(x) - returns the *identity* of the set containing x
the identity of a set is any unique identifier of the set.

All we require from **FINDSET** is that **FINDSET**(x) = **FINDSET**(y)
if and only if x and y are **currently** in the same set

Example

MAKESET(x) - make a new set containing only x (which is not already in a set)

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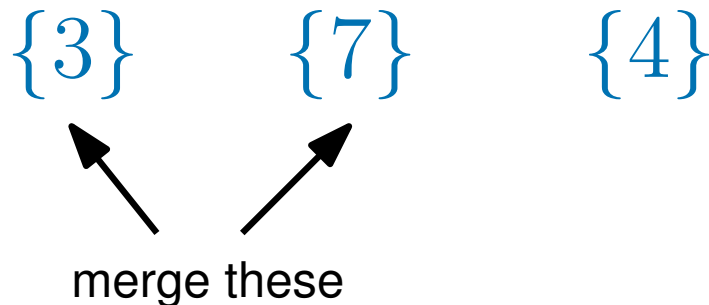
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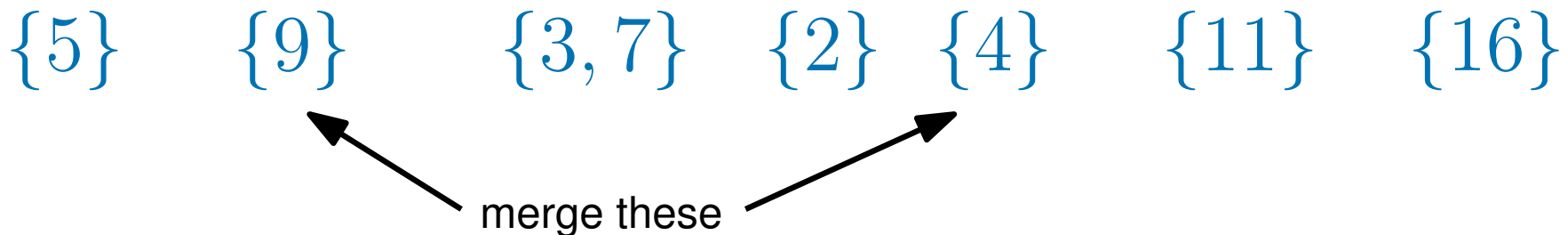
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UNION(2, 16)

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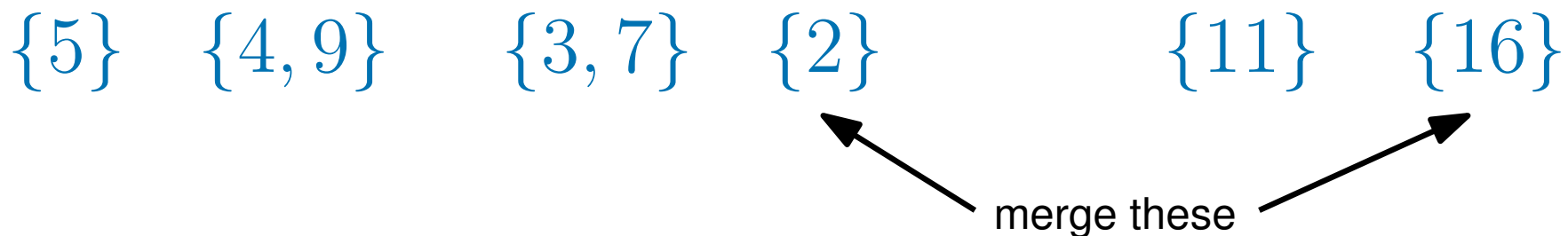
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UNION(7, 2)

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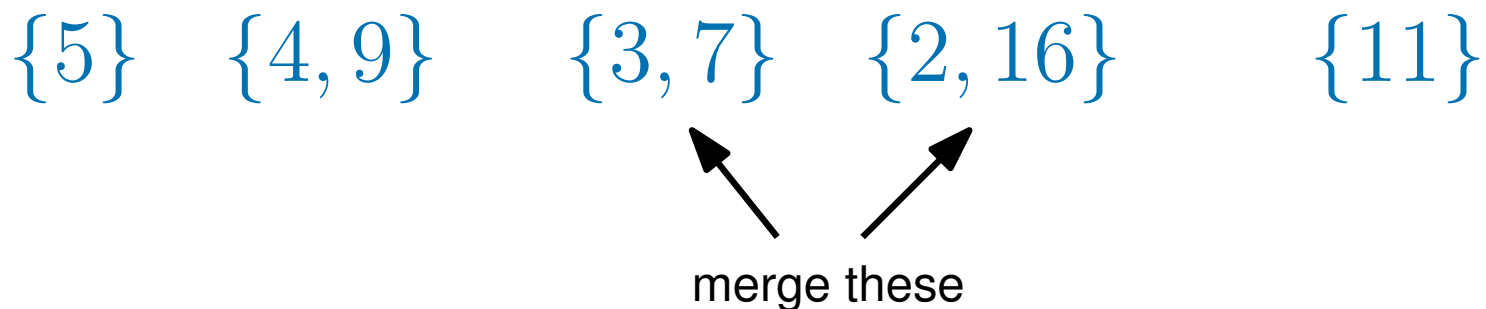
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UNION(3, 5)

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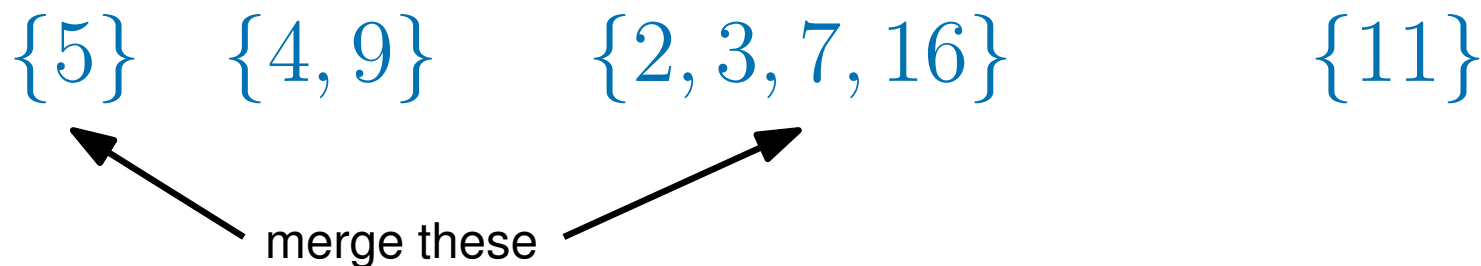
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FINDSET(2) returns 3

$\{4, 9\}$ $\{2, 3, 5, 7, 16\}$ $\{11\}$

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$\text{FINDSET}(2)$ returns 3

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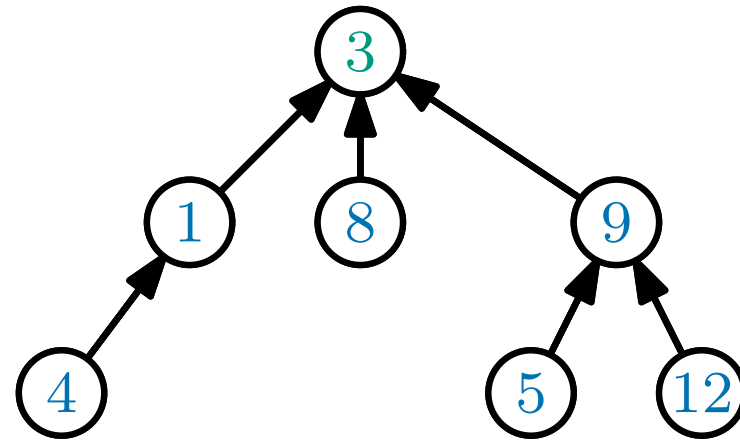
$\text{FINDSET}(4)$ returns 9

$\text{FINDSET}(9)$ returns 9

In our data structure,
the *identity* will be an
element of the set

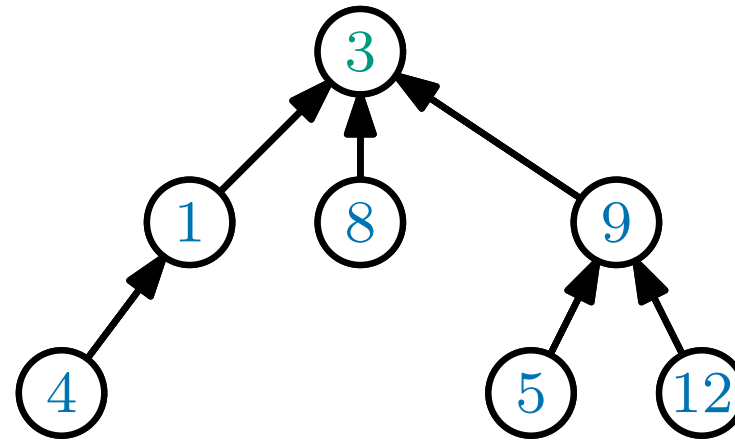
Reverse Trees

The data structure we will discuss stores each set as a reverse tree:



Reverse Trees

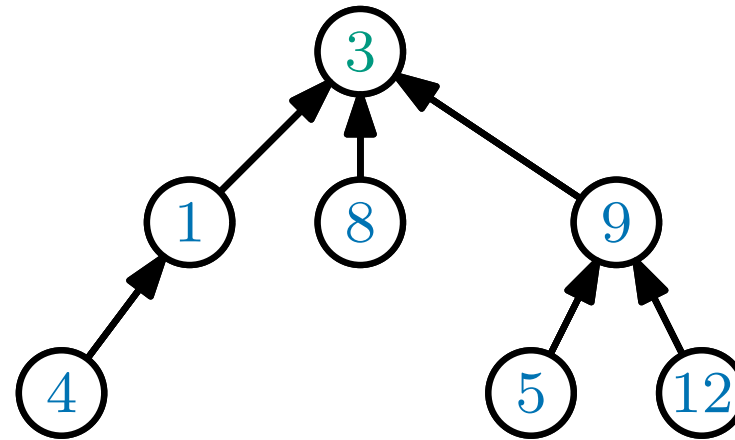
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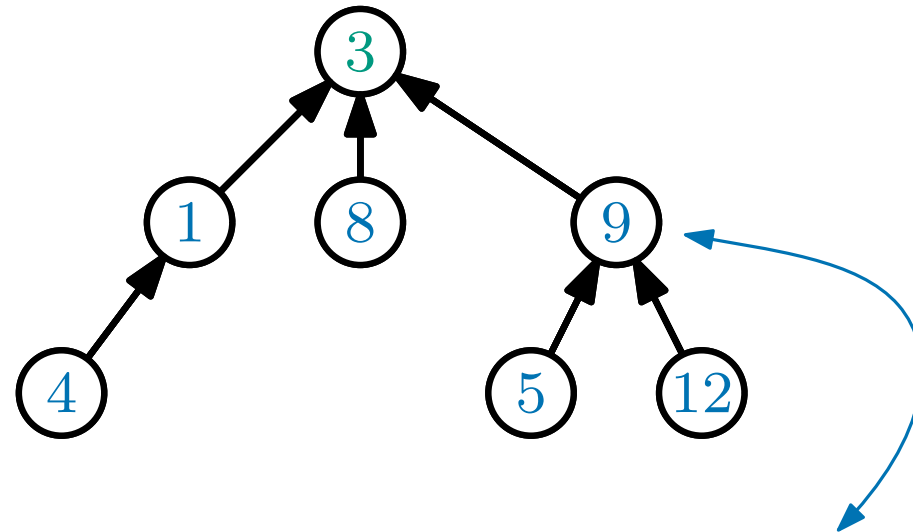


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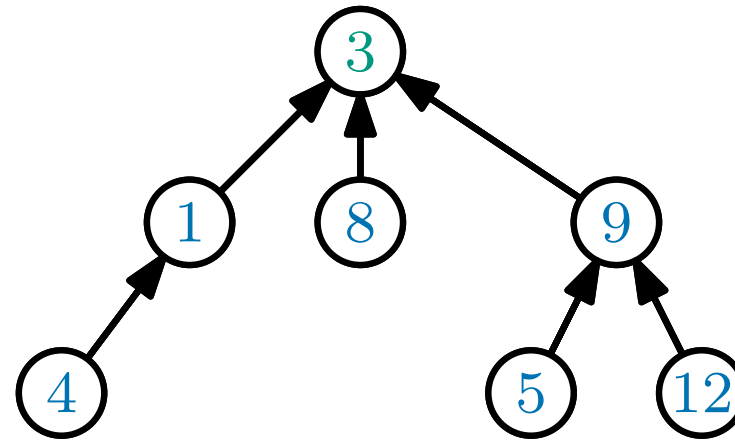


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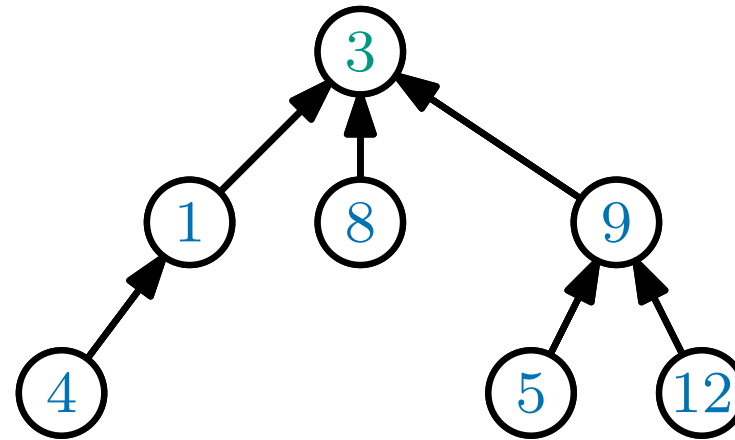


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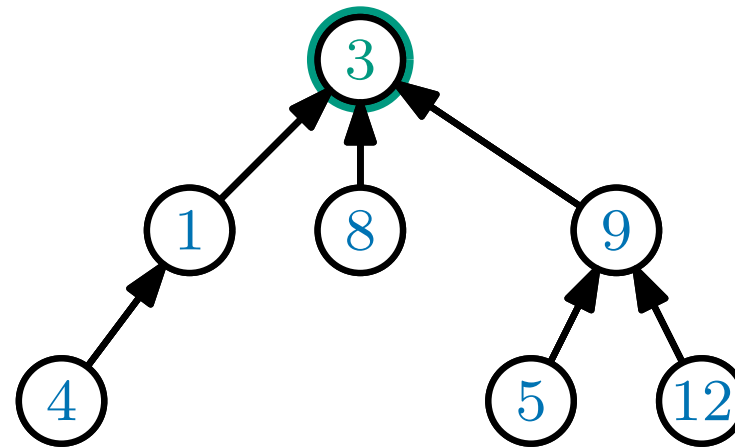
This reverse tree stores the set $\{1, 3, 4, 5, 8, 9, 12\}$

Each node stores an element from the set

The **identity** of a set is element at the root (here 3)

Reverse Trees

The data structure we will discuss stores each set as a reverse tree:



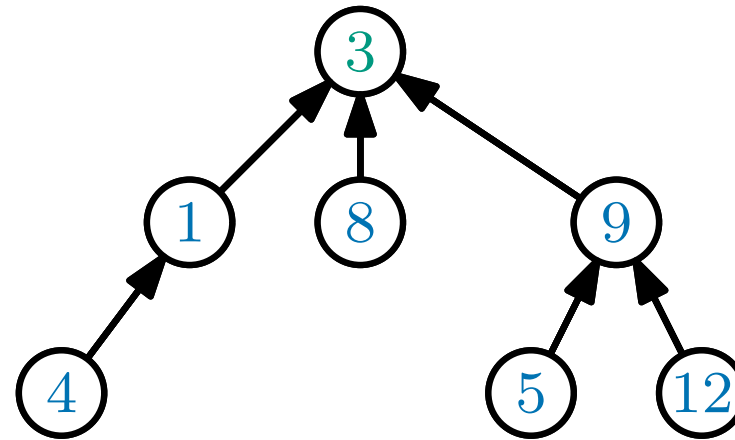
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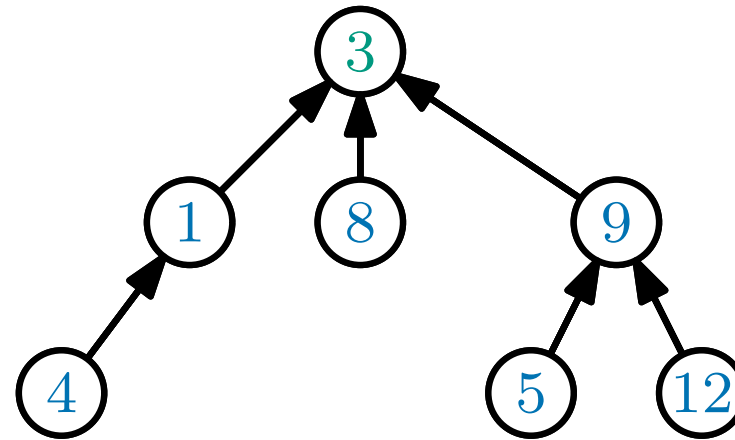
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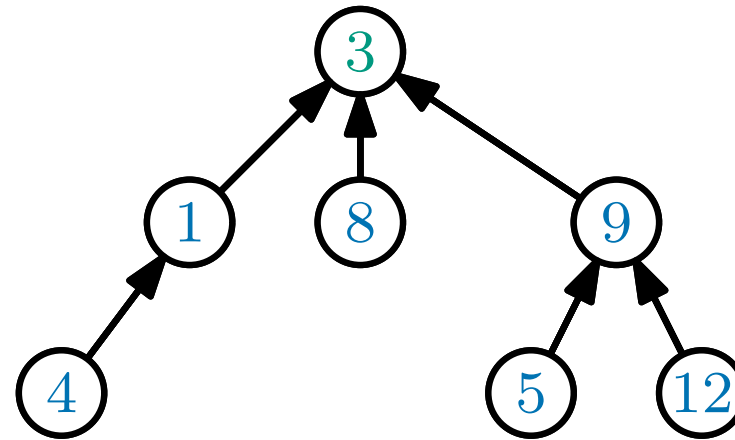
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The **identity** of a set is element at the root (here 3)

In a reverse tree, each element stores a pointer to its parent
but no pointers to its children

Reverse Trees

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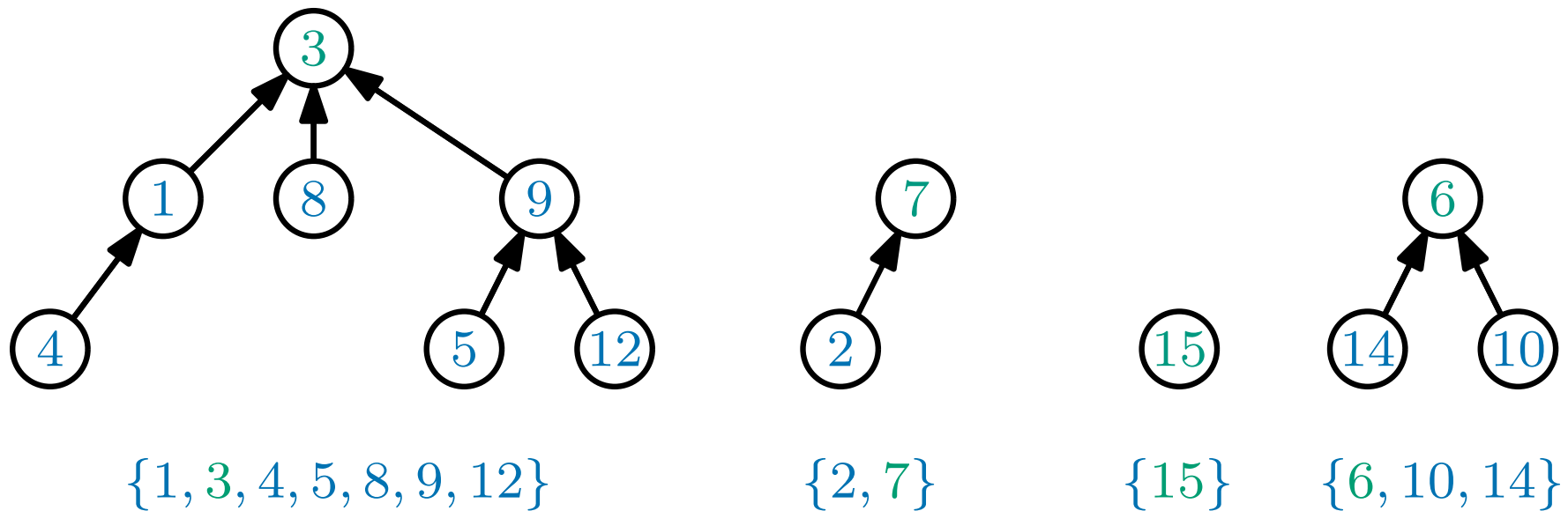
The **identity** of a set is element at the root (here 3)

In a reverse tree, each element stores a pointer to its parent
but no pointers to its children

- there will be no limit on the number of children each node can have

Reverse Forests

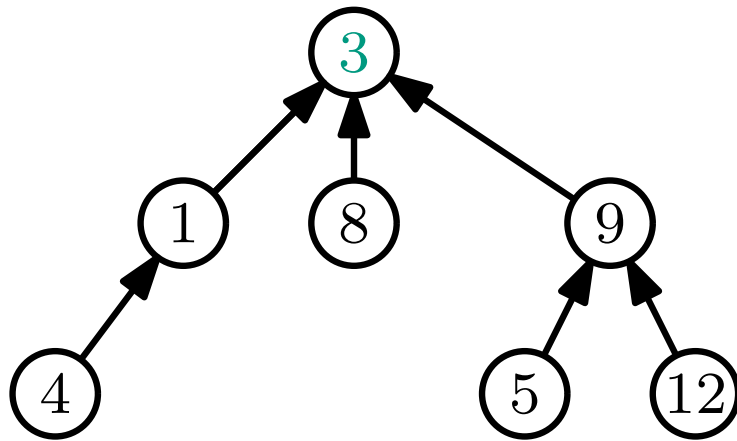
The data structure consists of a forest of reverse trees, one for each set



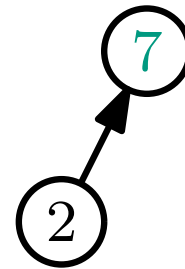
Each node stores an element from the set

The **identity** of a set is element at the root

How are these trees stored?



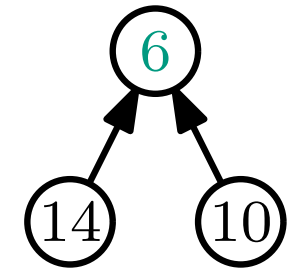
{1, 3, 4, 5, 8, 9, 12}



{2, 7}

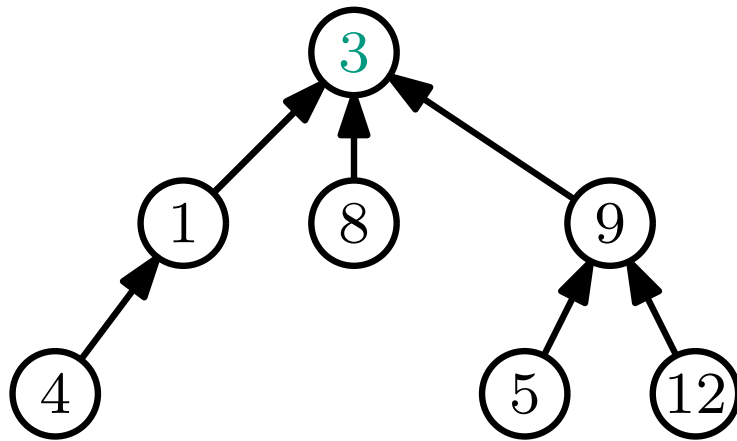


{15}

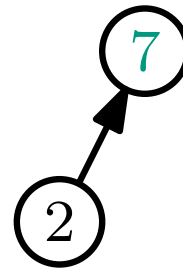


{6, 10, 14}

How are these trees stored?



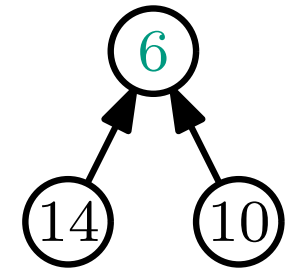
$\{1, 3, 4, 5, 8, 9, 12\}$



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$\{15\}$

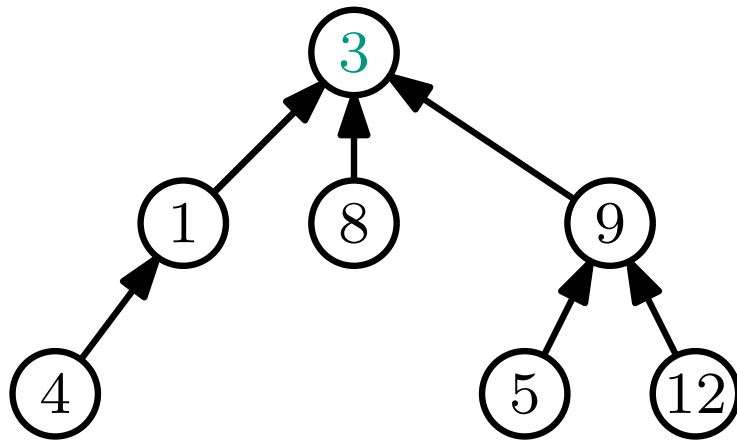


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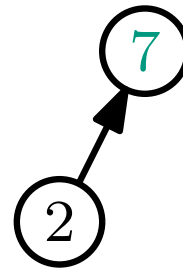
The elements are stored in an array of length n :



How are these trees stored?



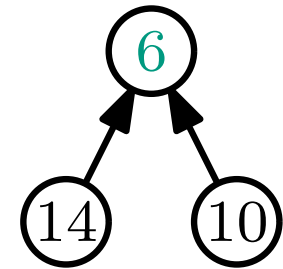
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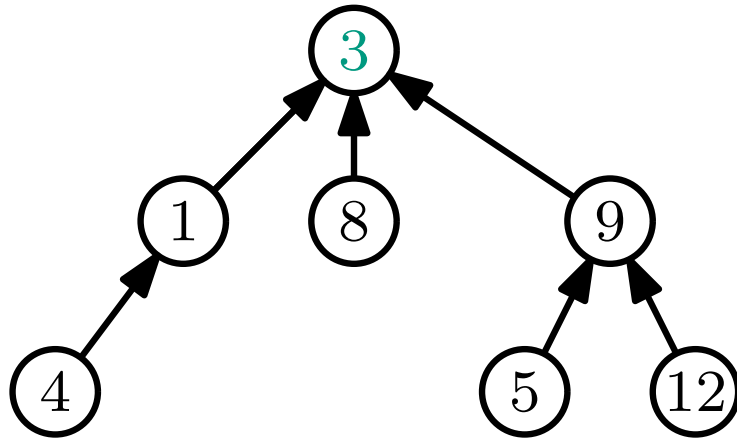
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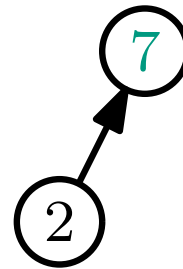


recall that the elements of the sets are numbers from $\{1, 2, \dots, n\}$

How are these trees stored?



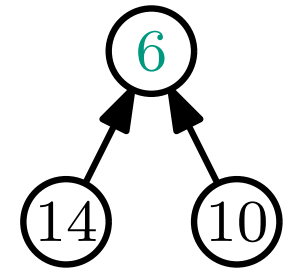
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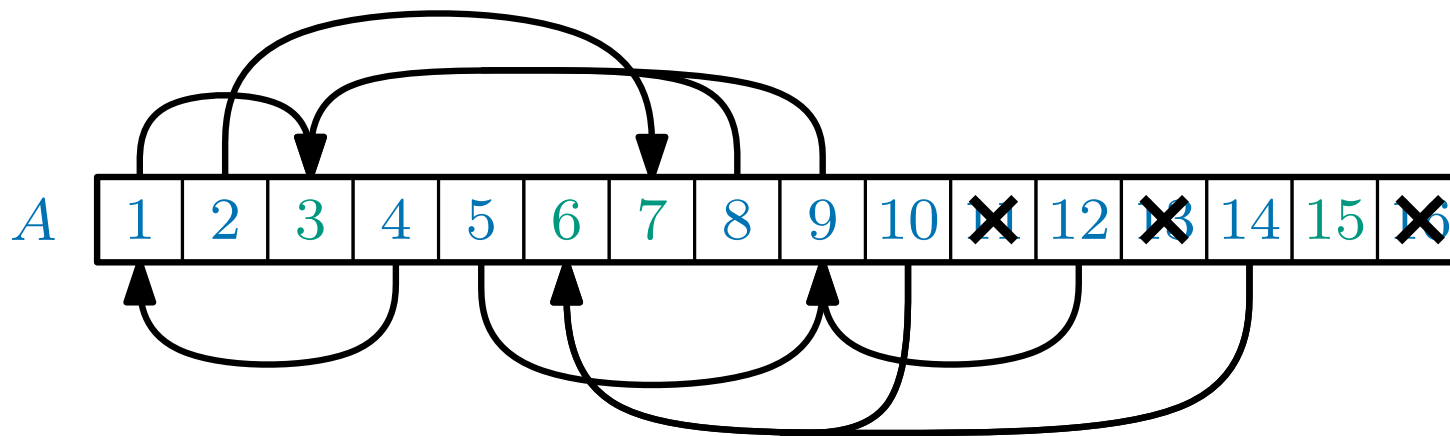


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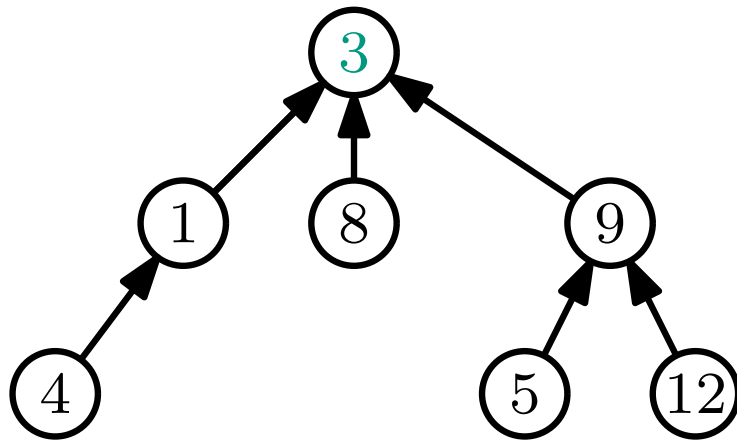


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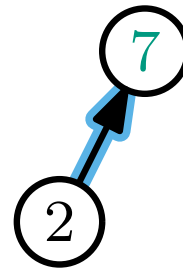
The elements are stored in an array of length n :



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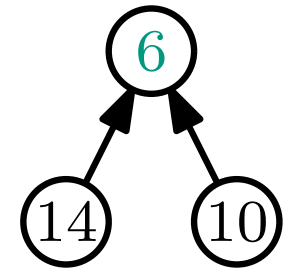
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$\{2, 7\}$

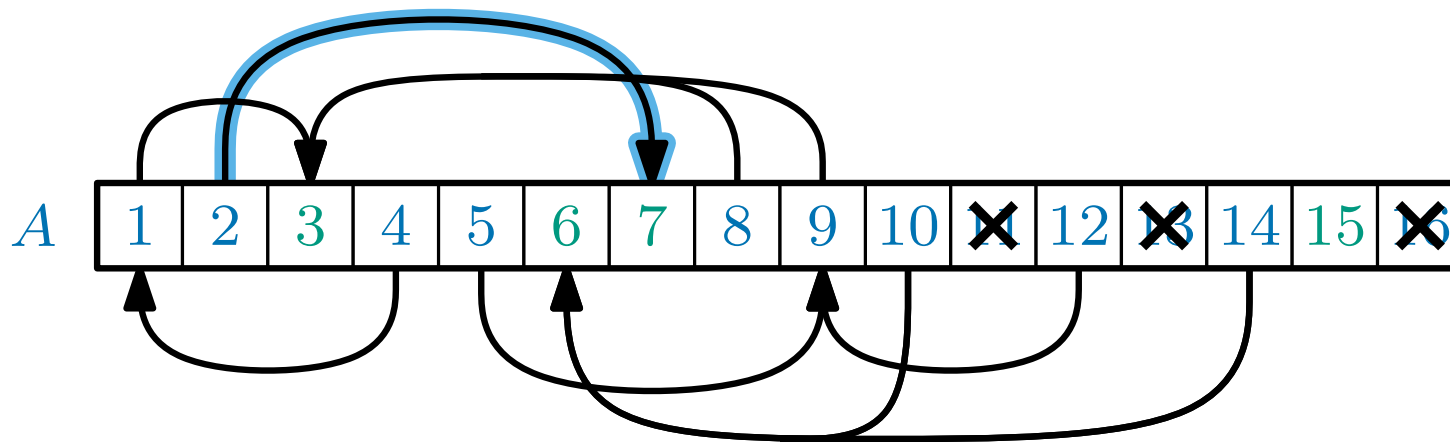


$\{15\}$

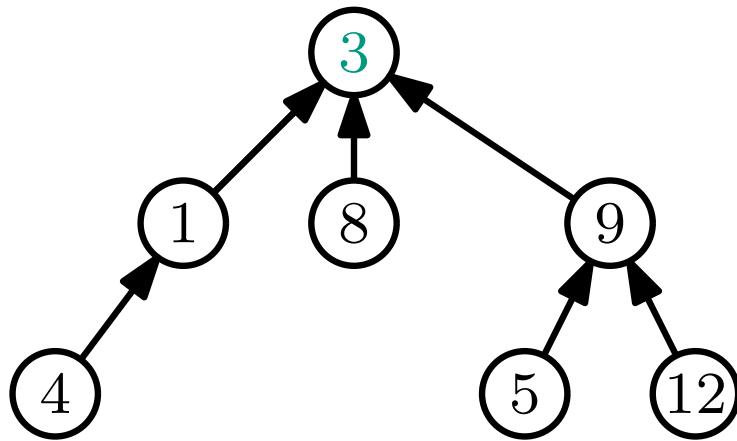


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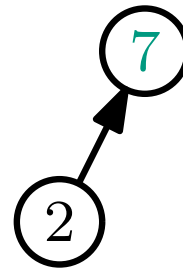
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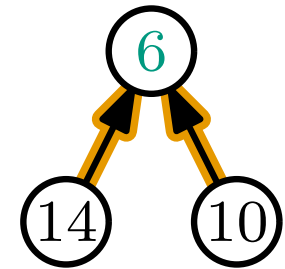
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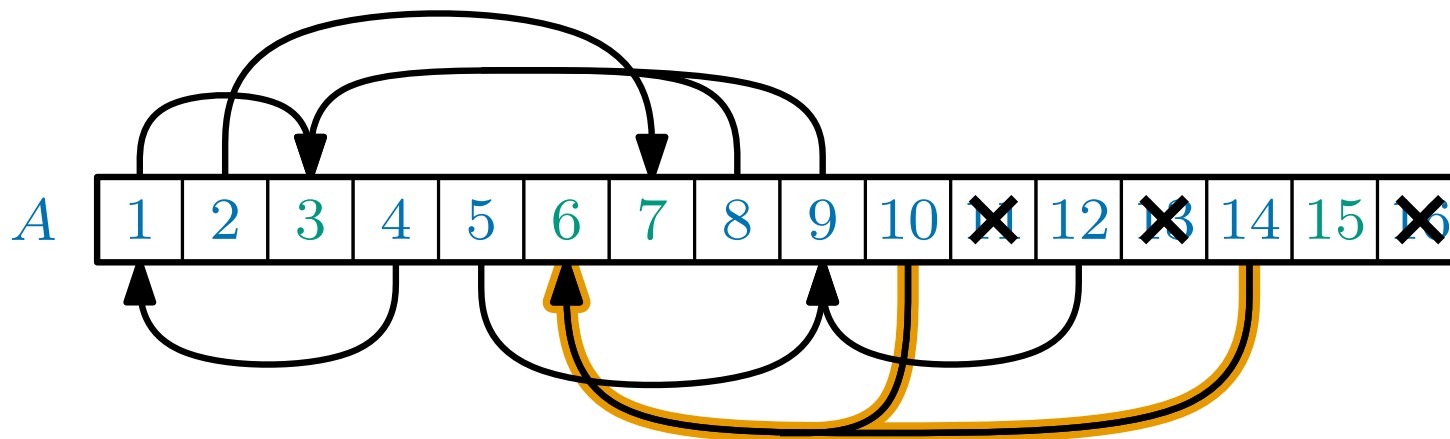


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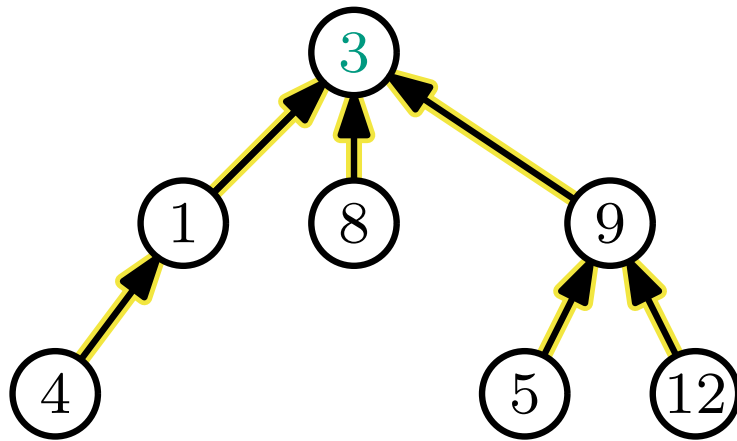


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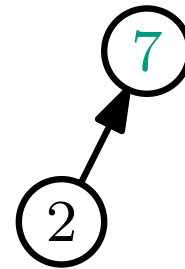
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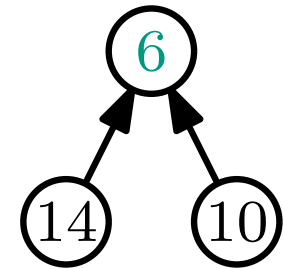
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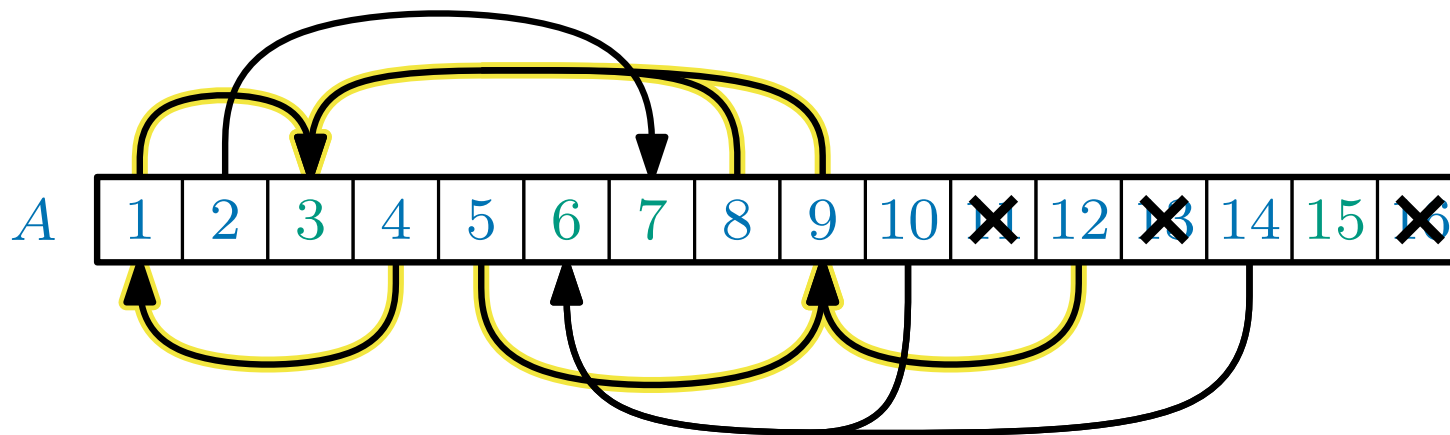


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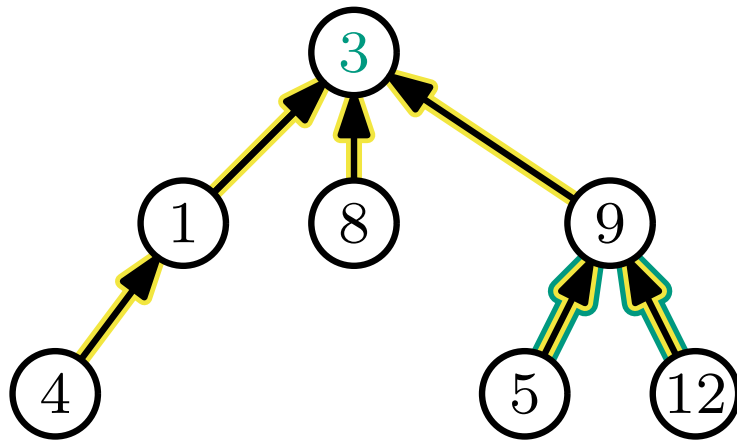


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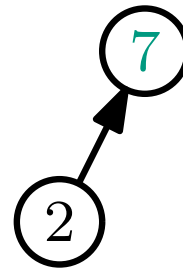
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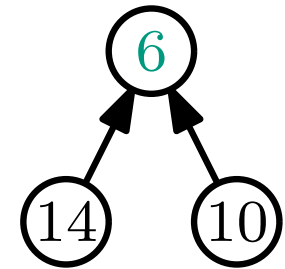
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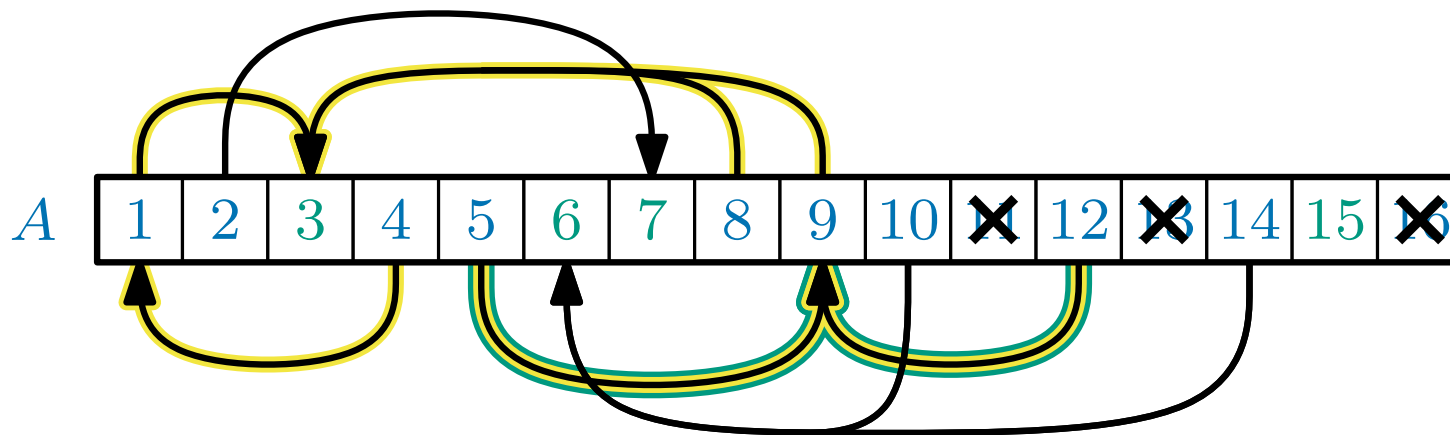


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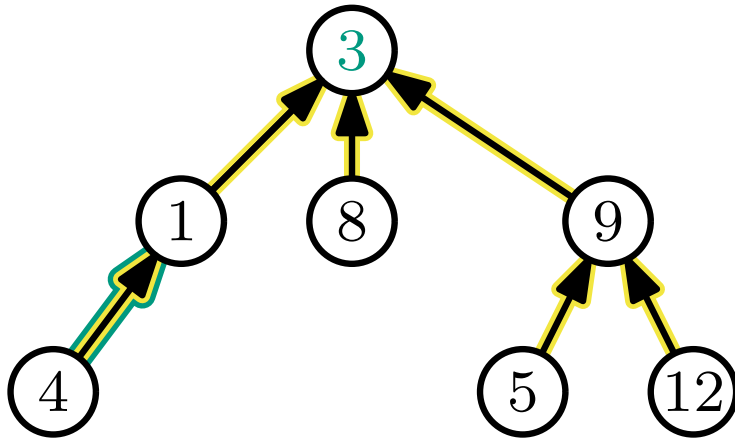


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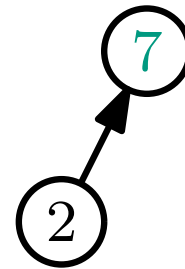
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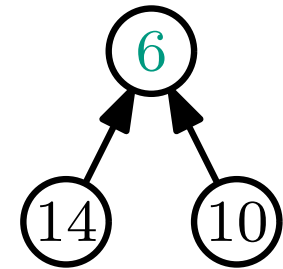
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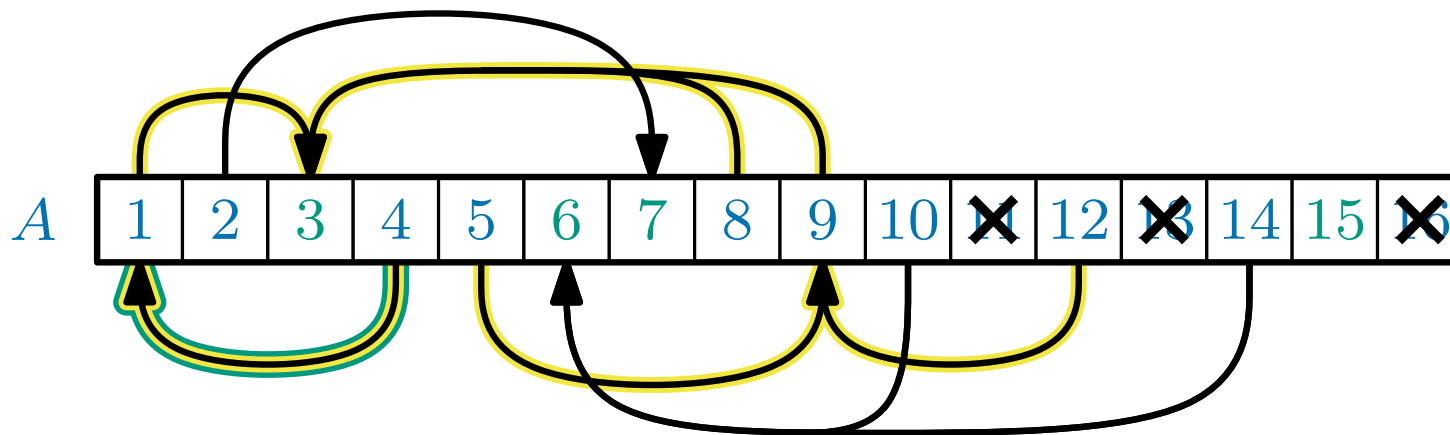


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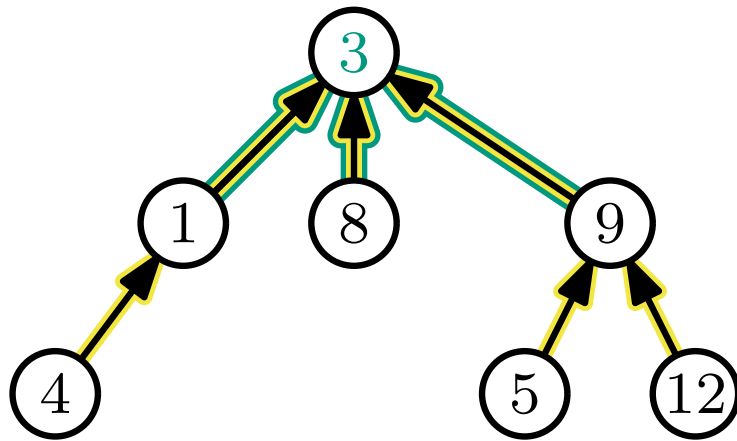


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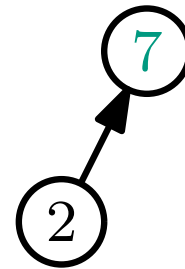
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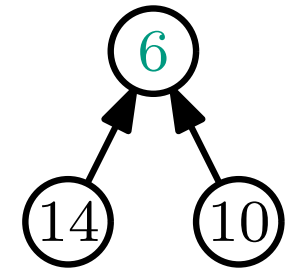
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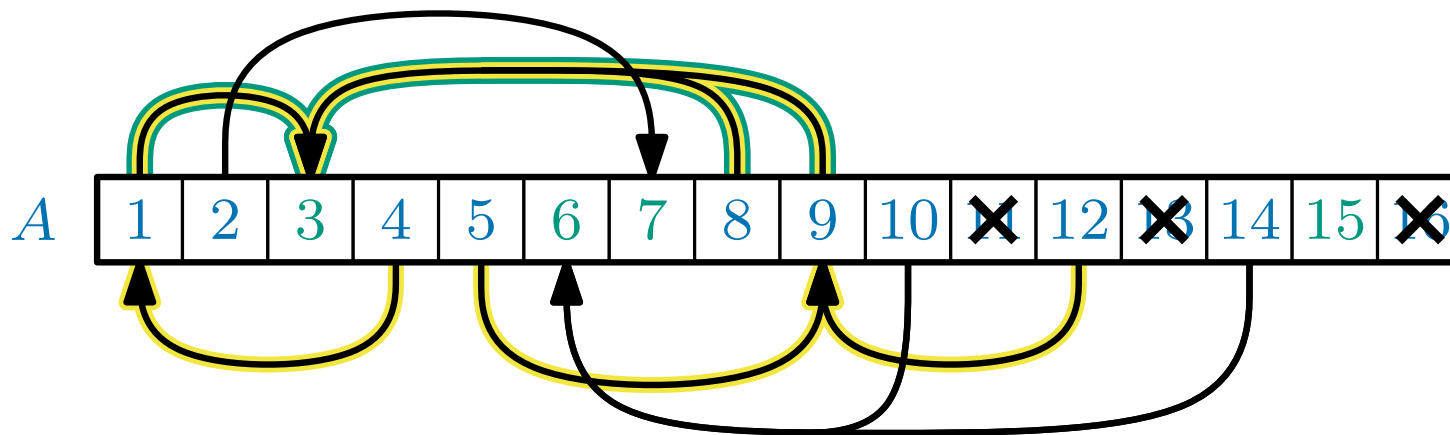


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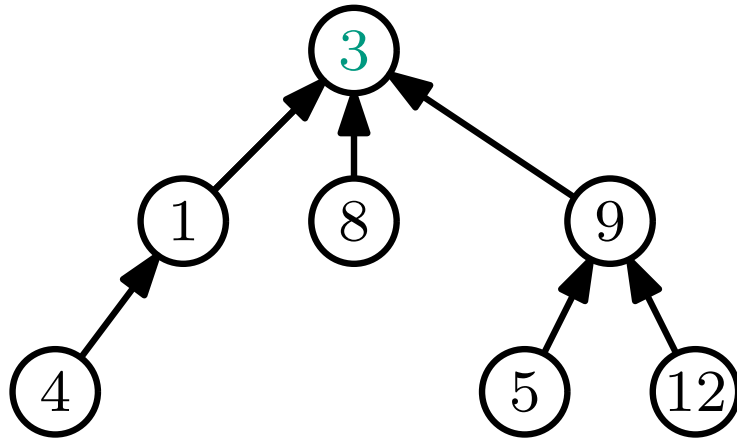


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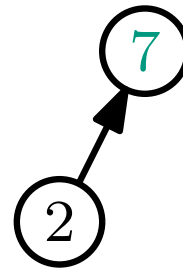
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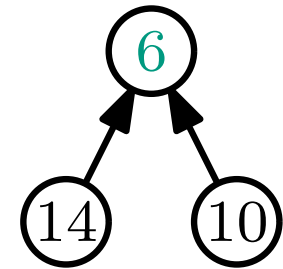
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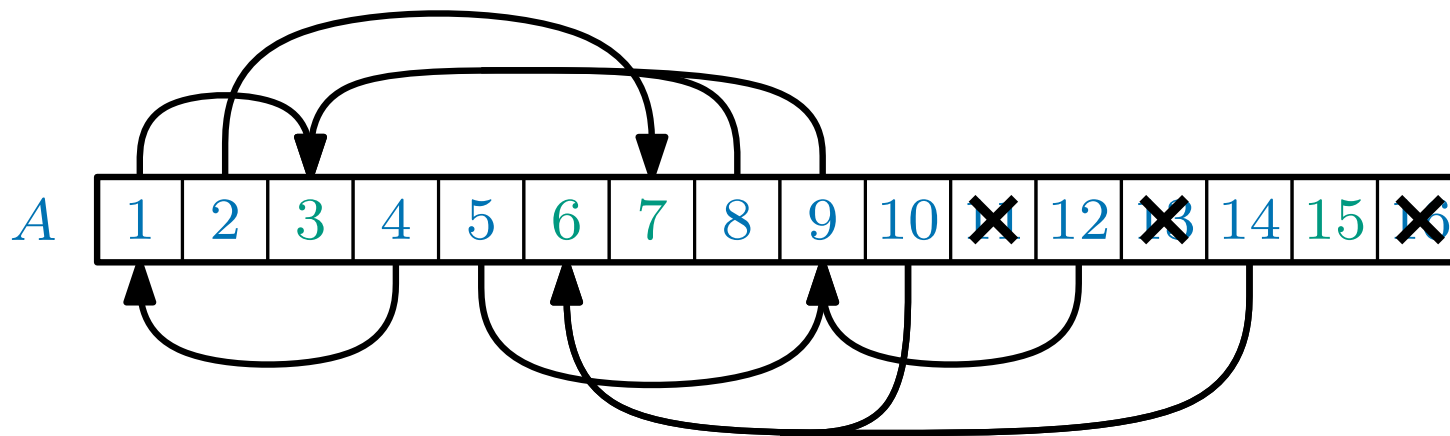


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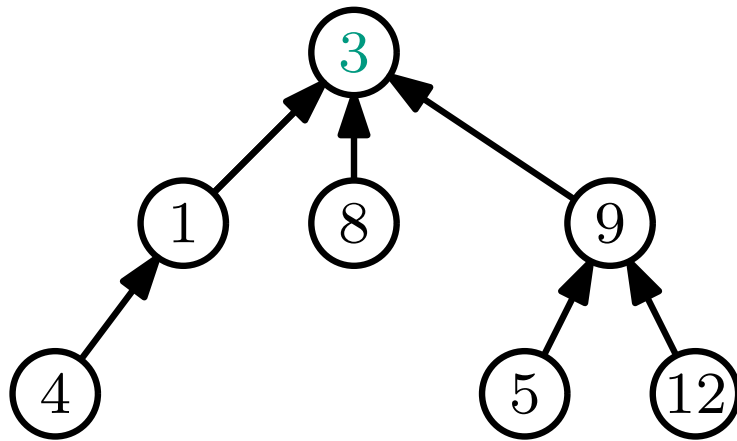


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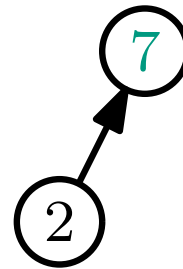
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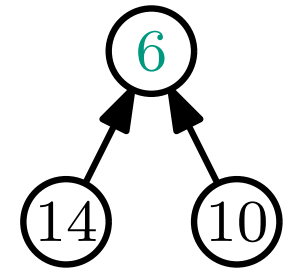
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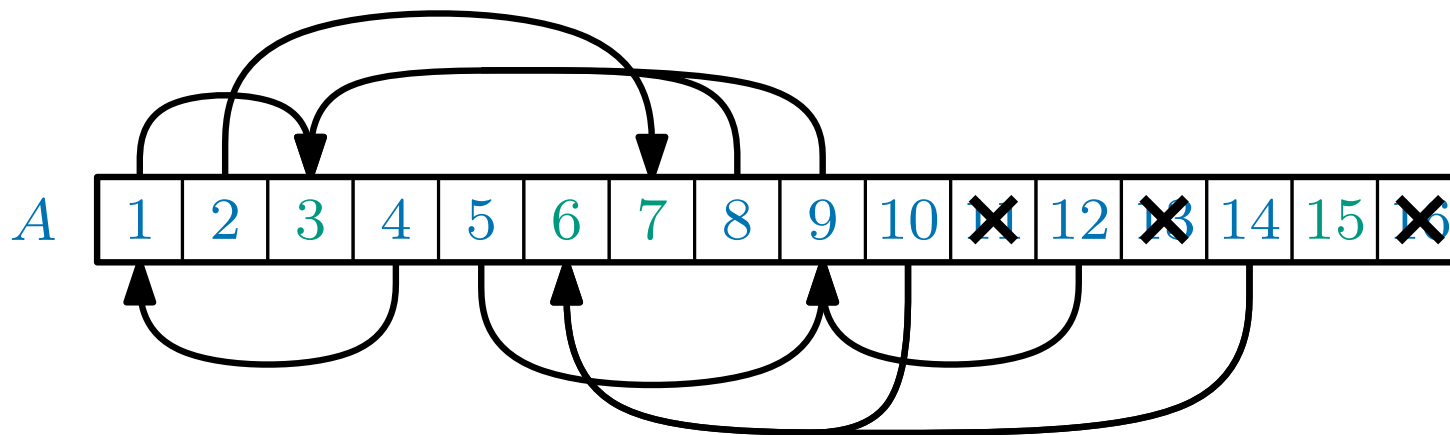


$\{15\}$



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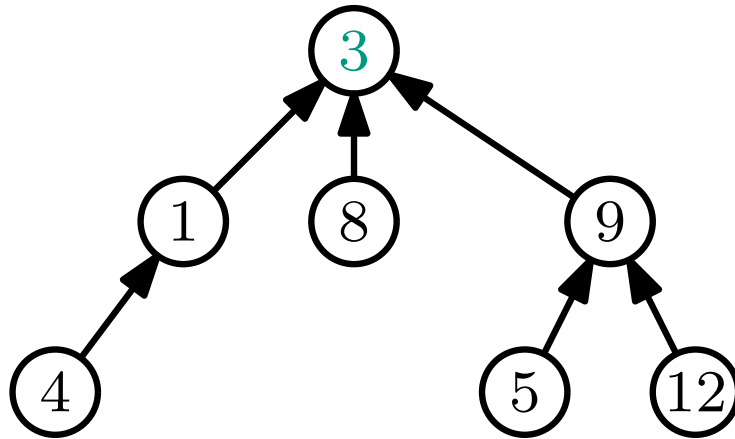
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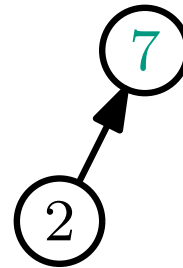
This allows us to find any element x in $O(1)$ time (x is stored in $A[x]$)

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



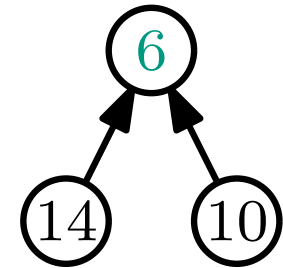
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$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

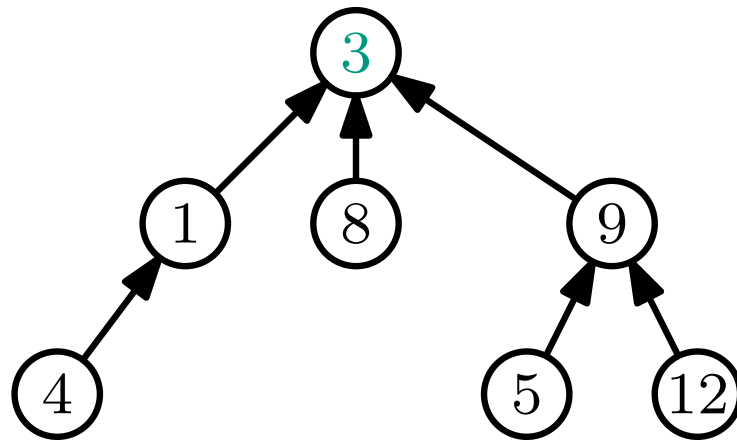
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

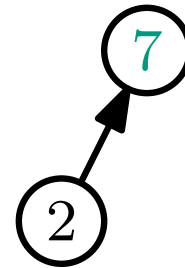
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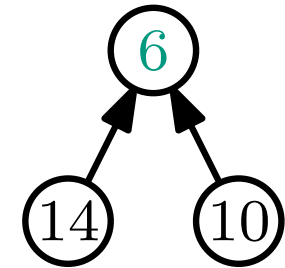
$\text{FINDSET}(5)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

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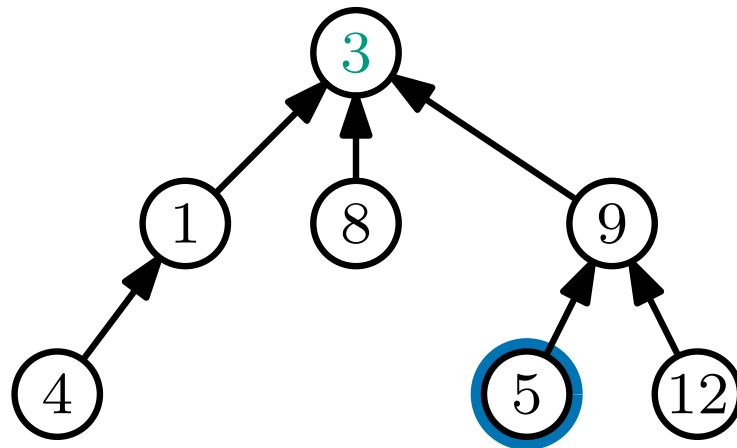
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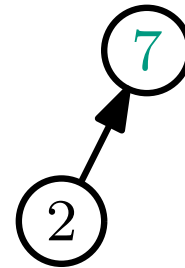
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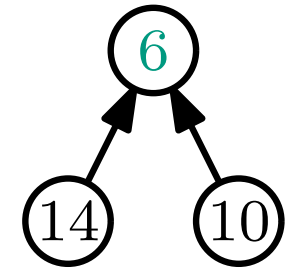
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$\{15\}$



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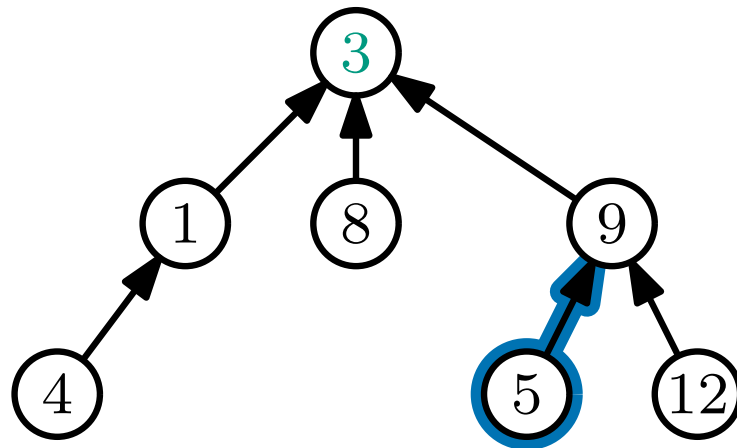
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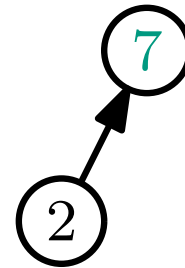
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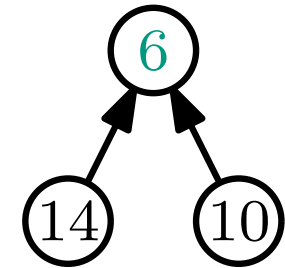
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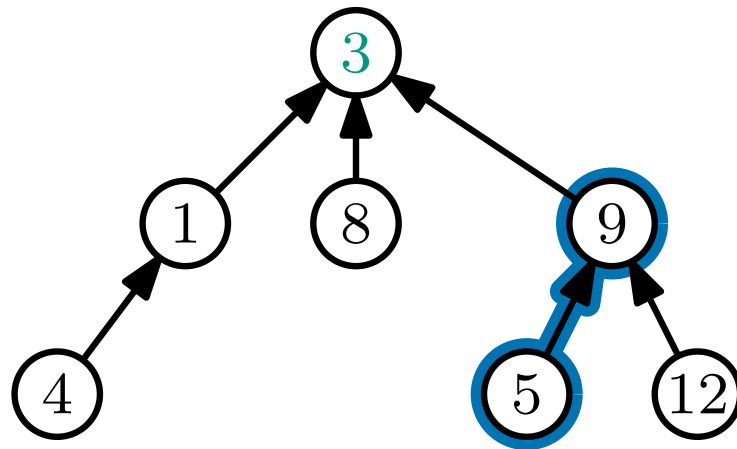
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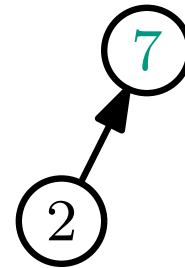
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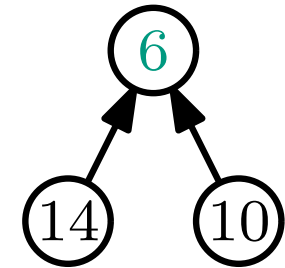
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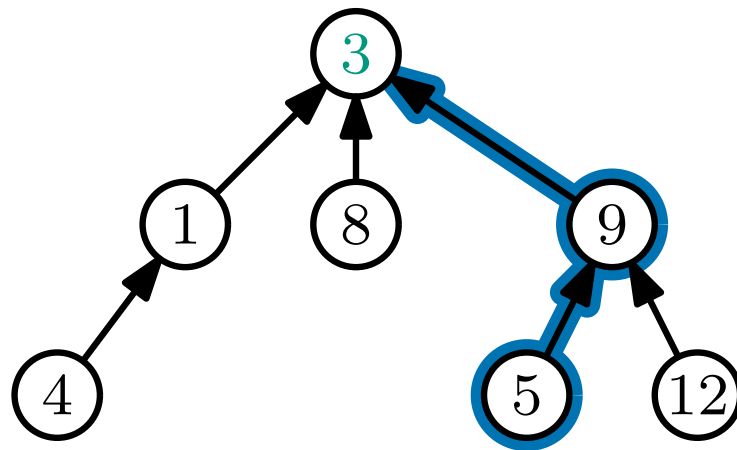
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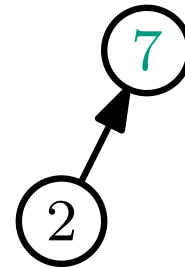
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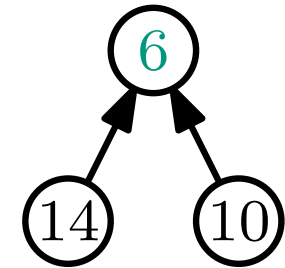
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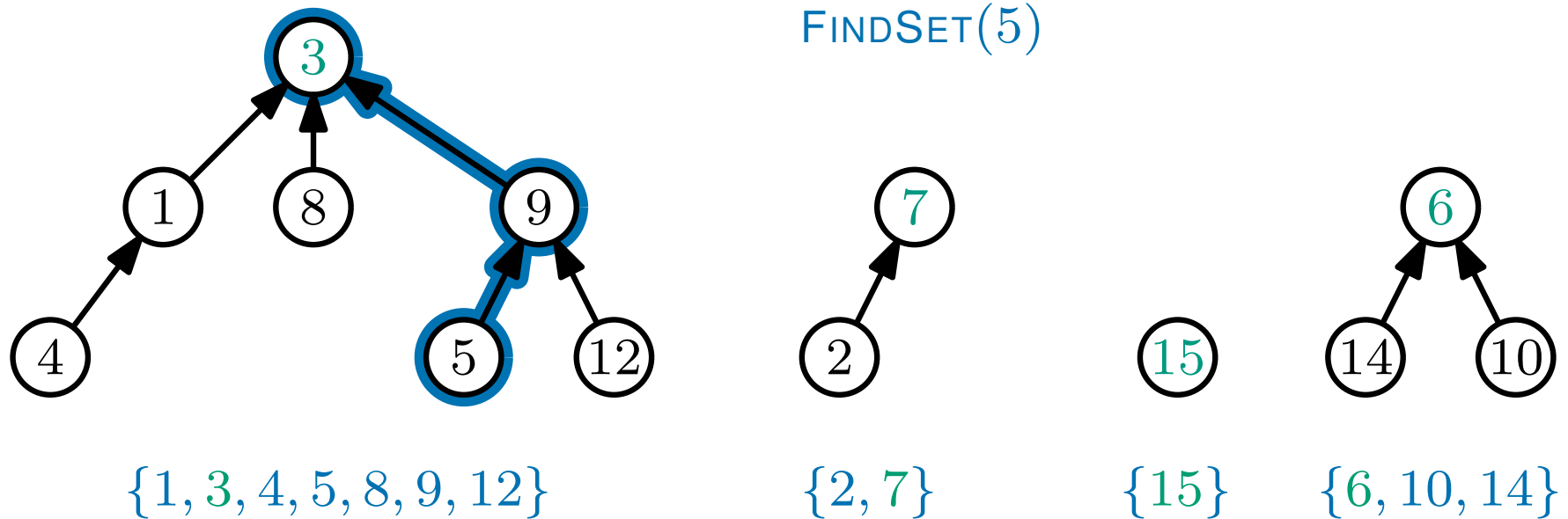
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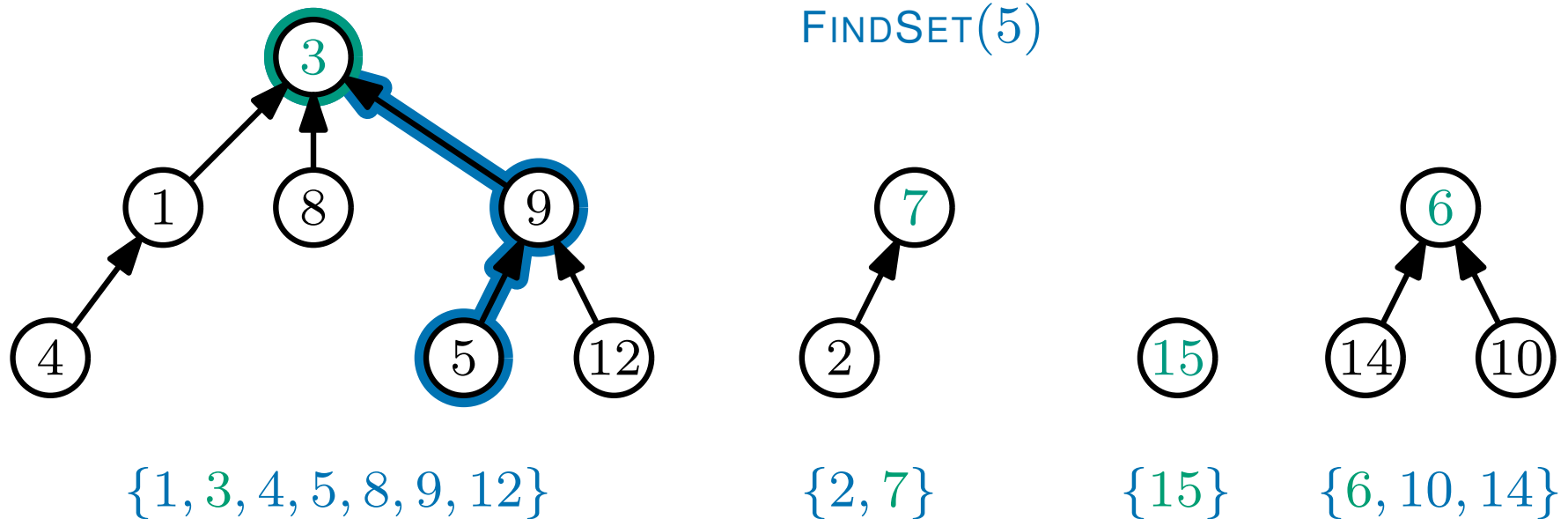
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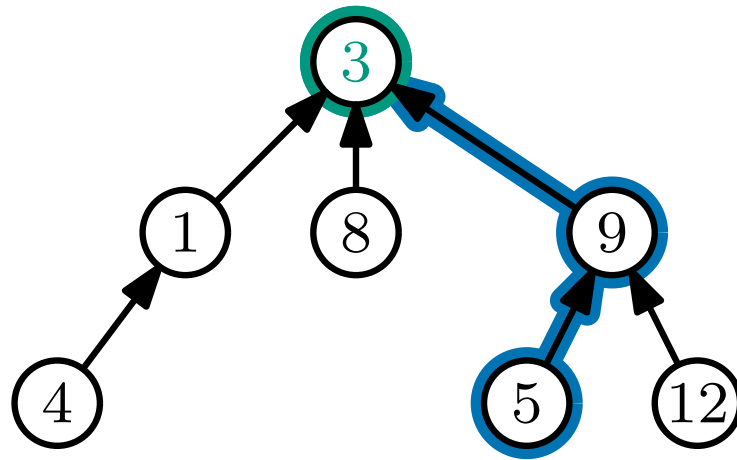
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follow the pointer to the parent of the current node

Step 3: Output the element at the root

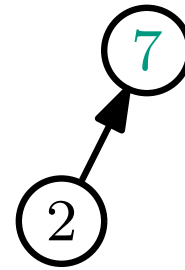
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

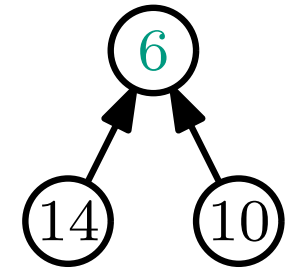
$\text{FINDSET}(5)$ returns 3



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

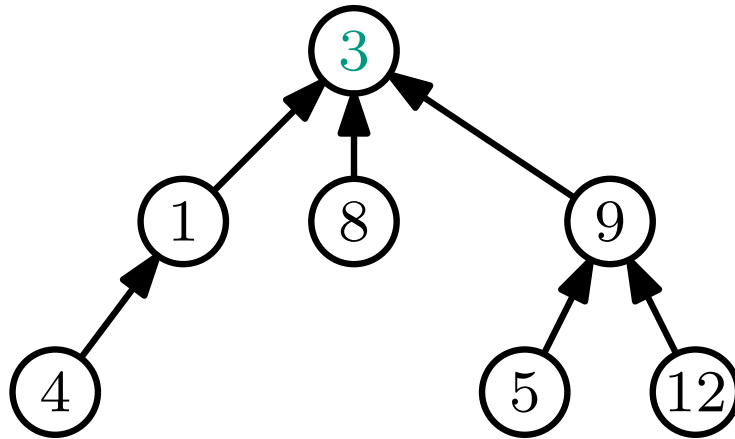
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

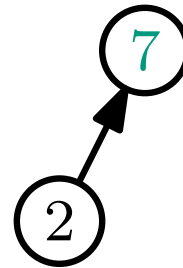
Step 3: Output the element at the root

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



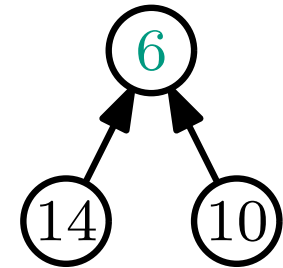
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

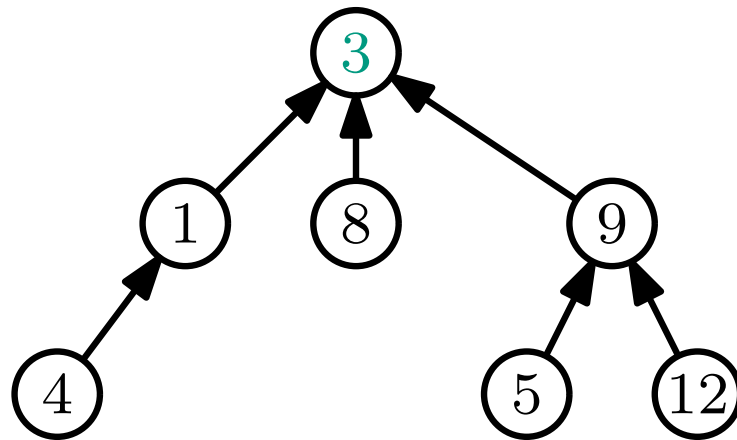
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

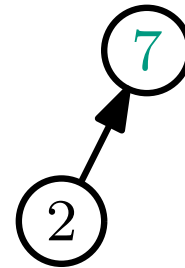
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

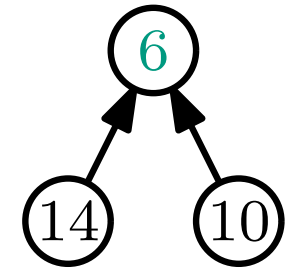
$\text{FINDSET}(1)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

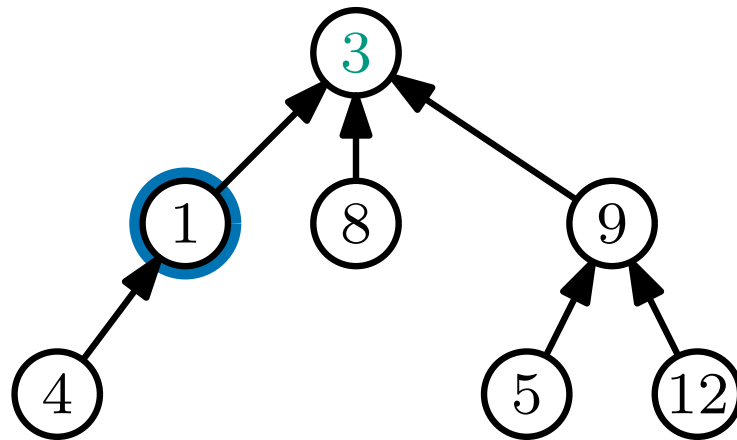
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

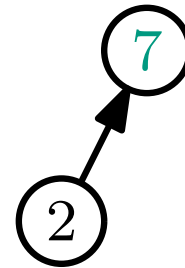
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

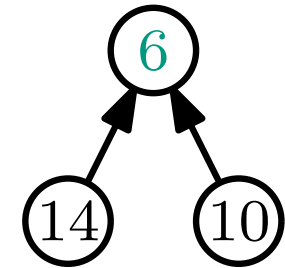
$\text{FINDSET}(1)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

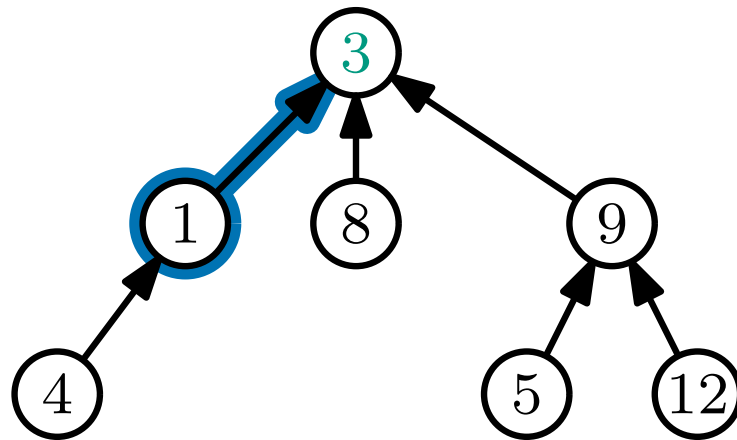
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

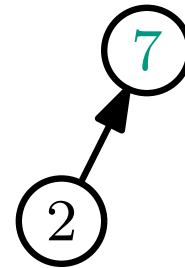
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

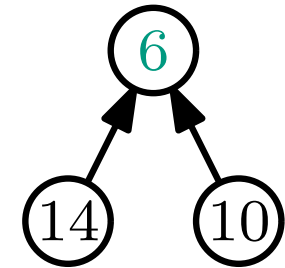
$\text{FINDSET}(1)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

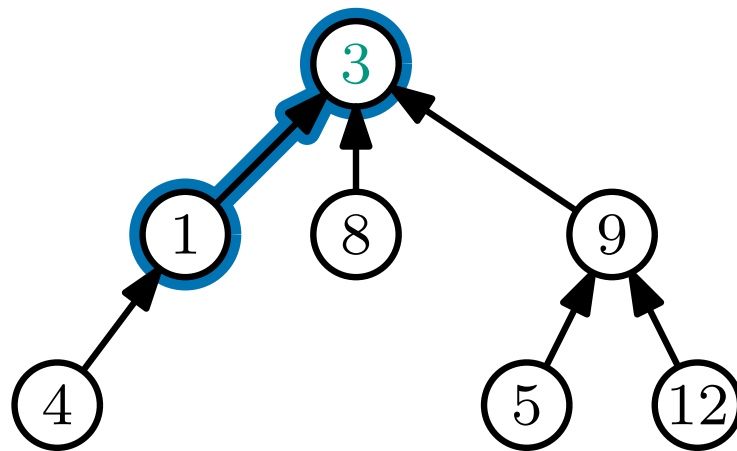
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

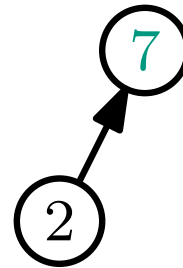
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

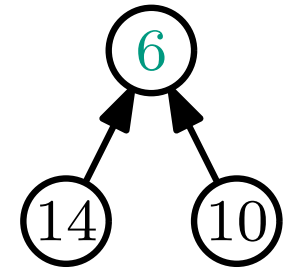
$\text{FINDSET}(1)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

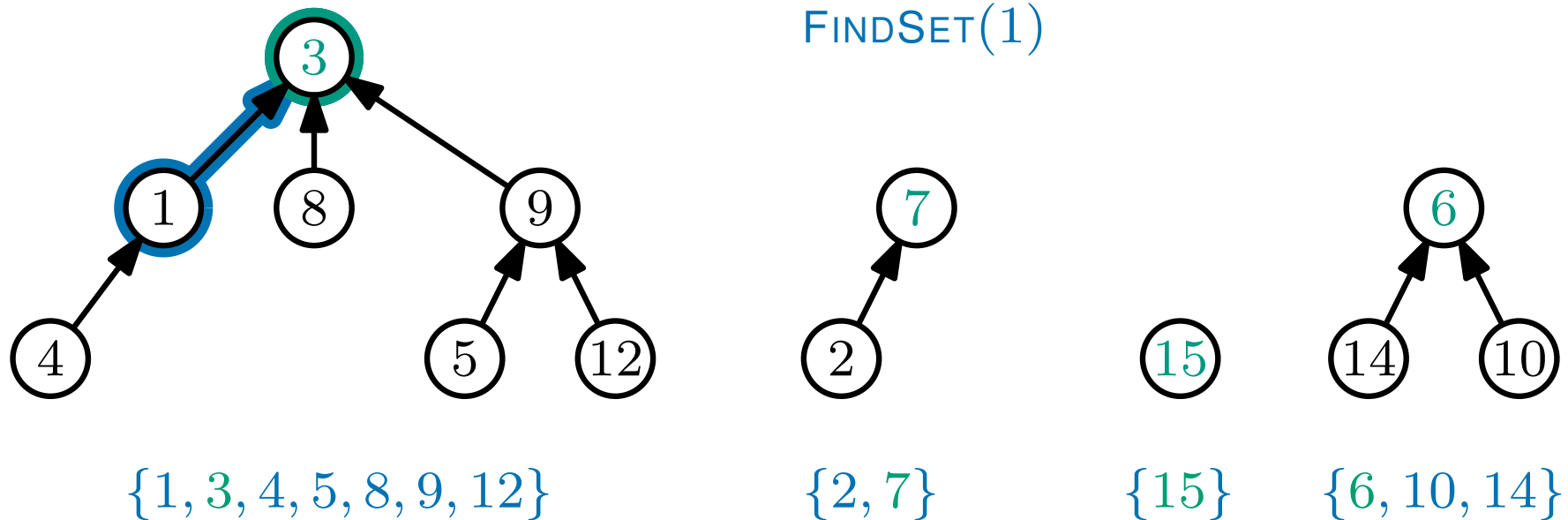
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



Step 1: Find the node storing element x

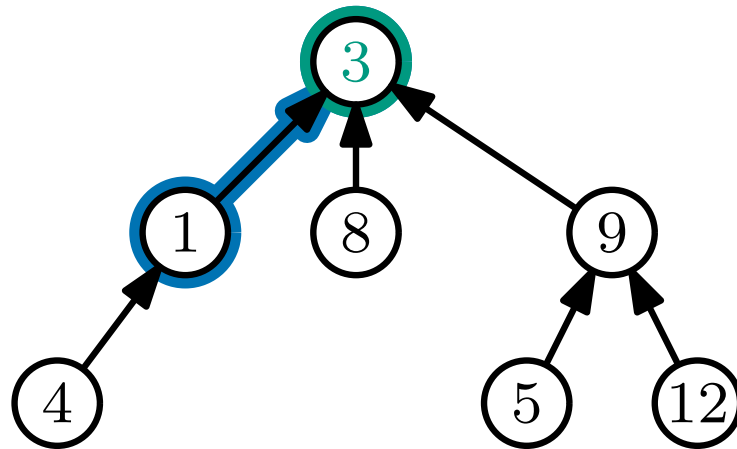
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

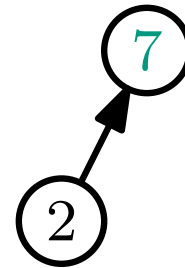
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

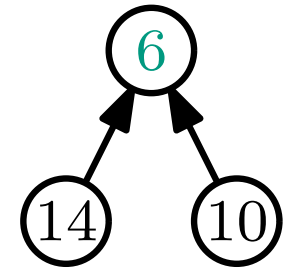
$\text{FINDSET}(1)$ also returns 3



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

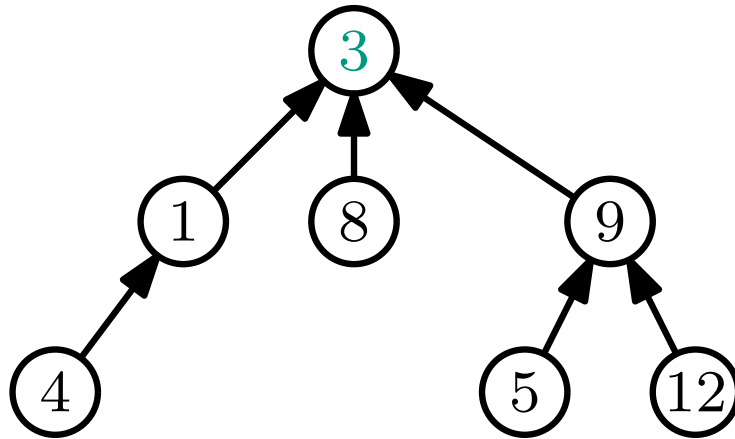
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

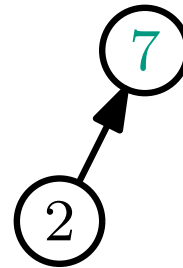
Step 3: Output the element at the root

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



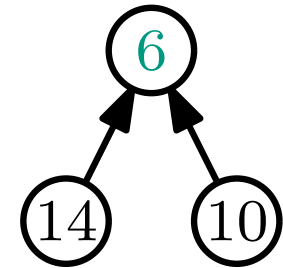
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

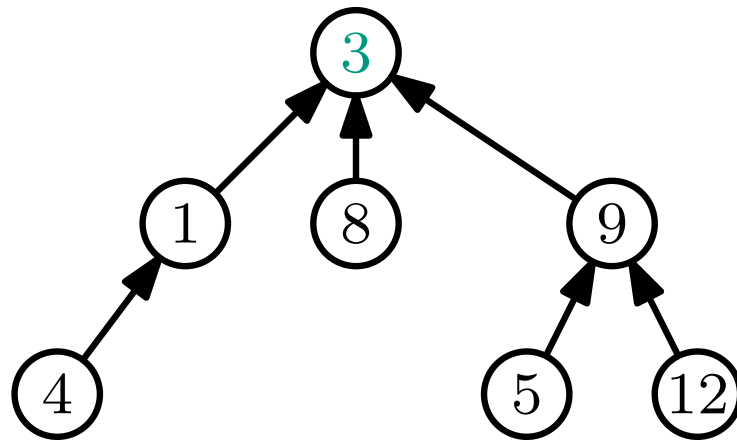
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

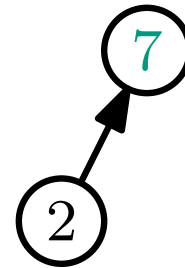
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

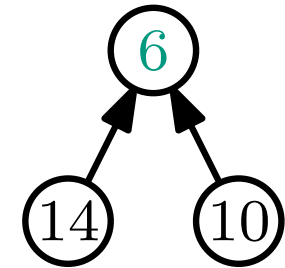
$\text{FINDSET}(15)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

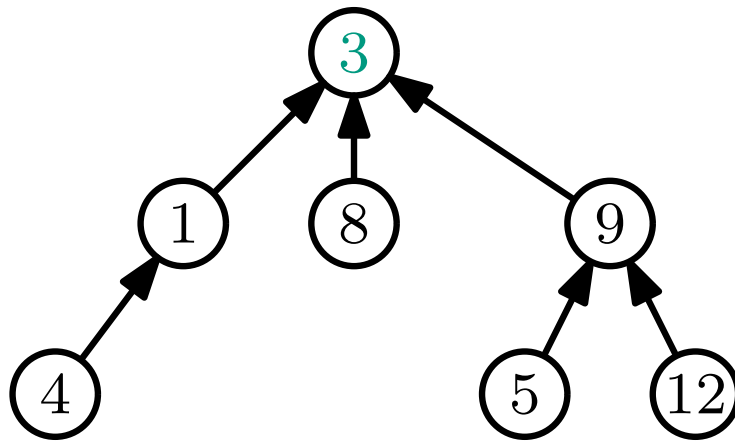
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

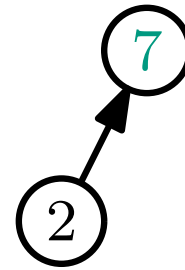
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

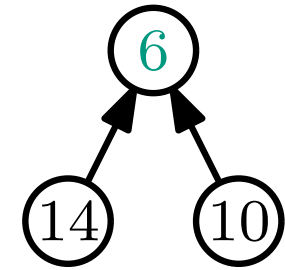
$\text{FINDSET}(15)$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

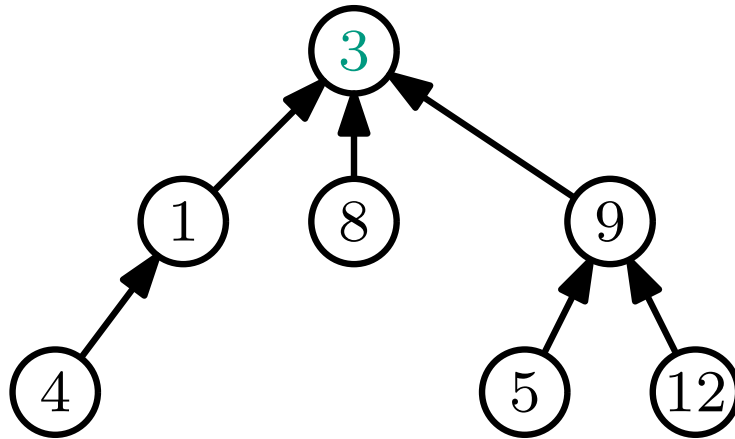
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

Step 3: Output the element at the root

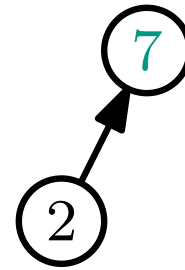
The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



$\{1, 3, 4, 5, 8, 9, 12\}$

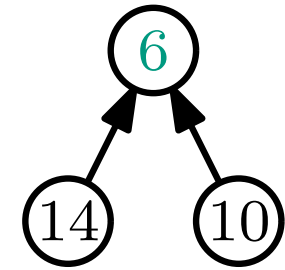
$\text{FINDSET}(15)$ returns 15



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

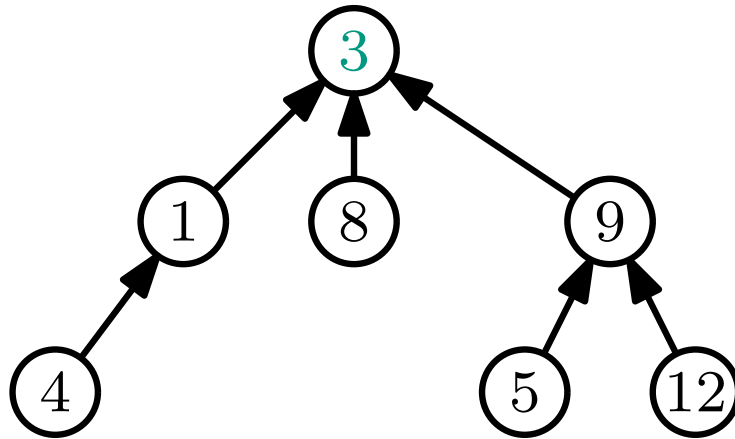
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

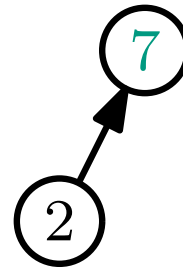
Step 3: Output the element at the root

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



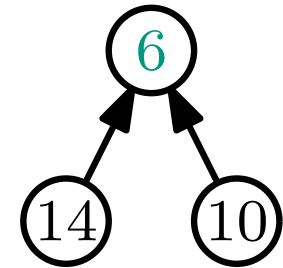
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

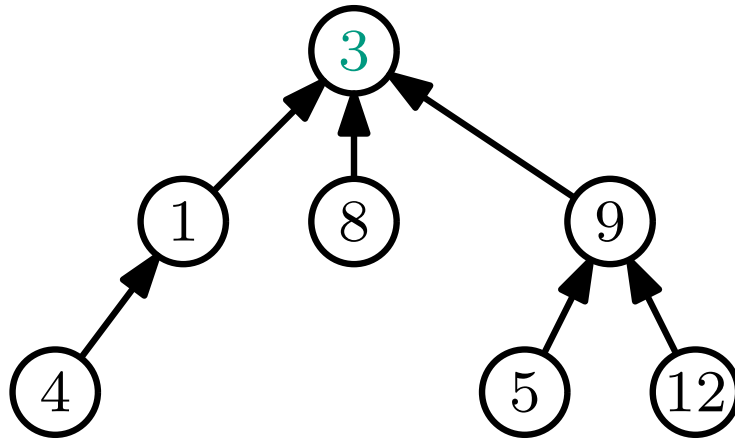
Step 2: Until you are at the root,

follow the pointer to the parent of the current node

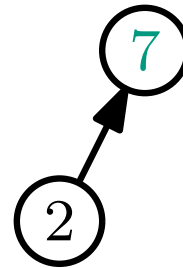
Step 3: Output the element at the root

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



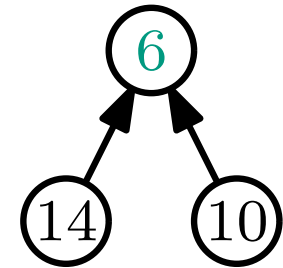
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

Step 1: Find the node storing element x

Step 2: Until you are at the root,

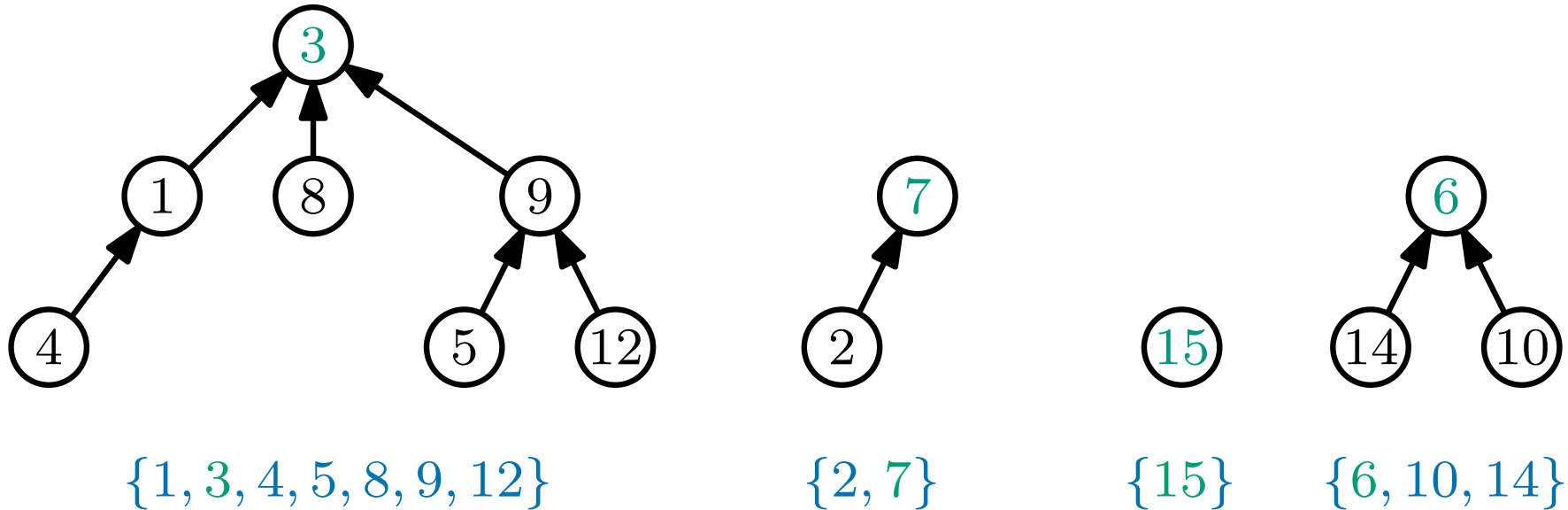
follow the pointer to the parent of the current node

Step 3: Output the element at the root

What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



Step 1: Find the node storing element x

Step 2: Until you are at the root,

follow the pointer to the parent of the current node

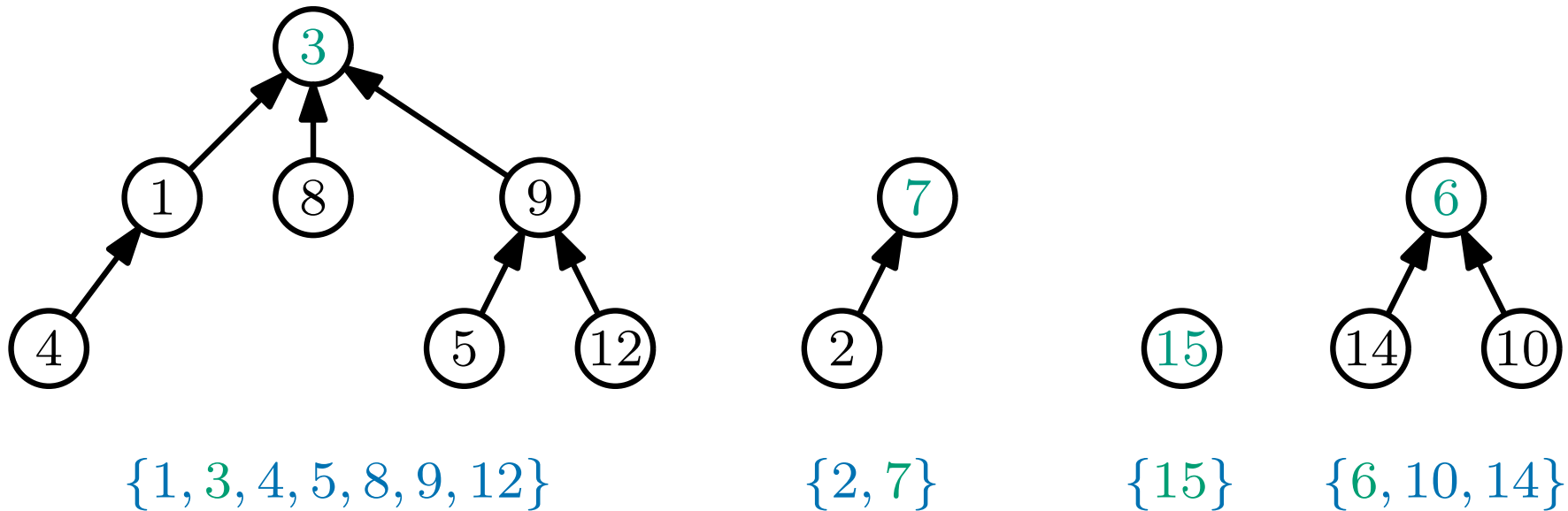
Step 3: Output the element at the root

$O(1)$ time

What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



Step 1: Find the node storing element x

$O(1)$ time

because x is stored in $A[x]$

Step 2: Until you are at the root,

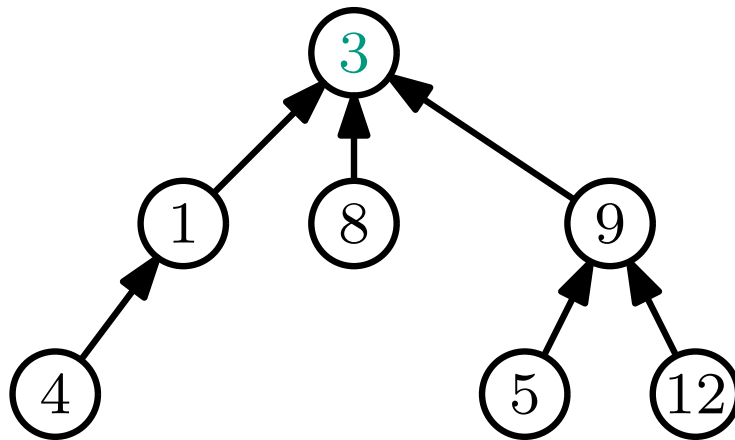
follow the pointer to the parent of the current node

Step 3: Output the element at the root

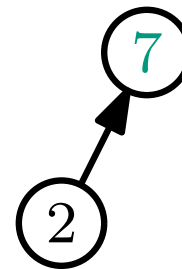
What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x



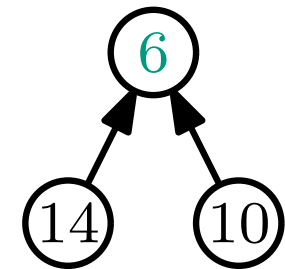
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

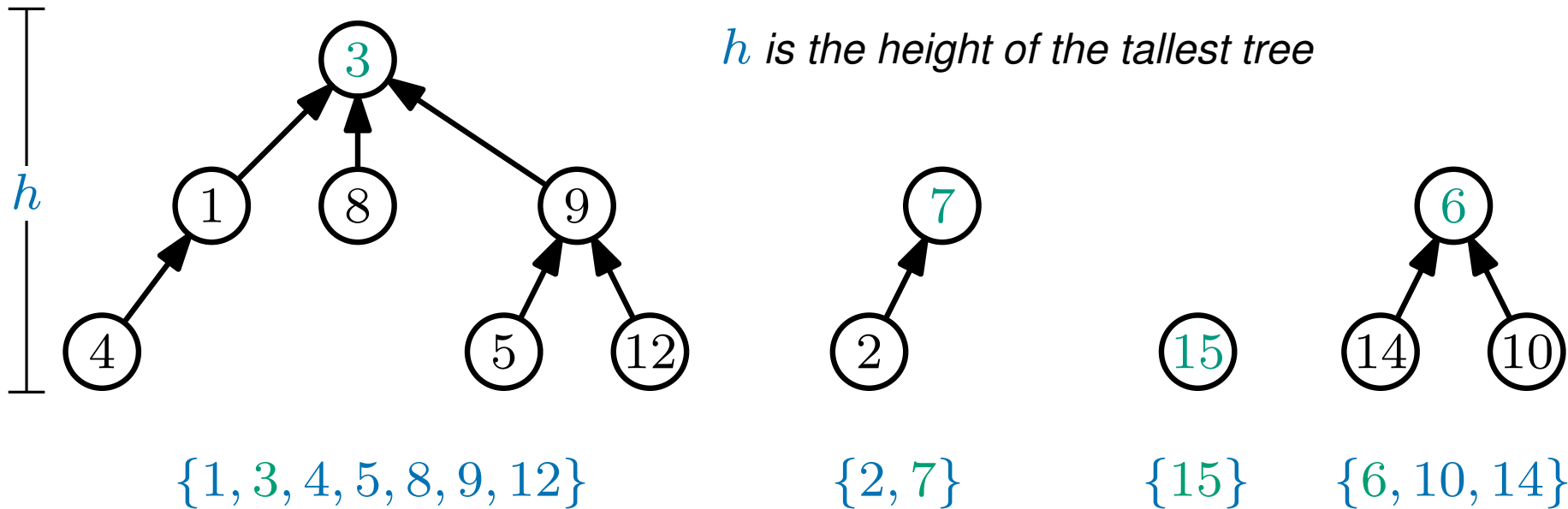
- $O(h)$ time
- Step 1:** Find the node storing element x
 - Step 2:** Until you are at the root,
follow the pointer to the parent of the current node
 - Step 3:** Output the element at the root
- $O(1)$ time
because x is stored in $A[x]$

What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x

h is the height of the tallest tree



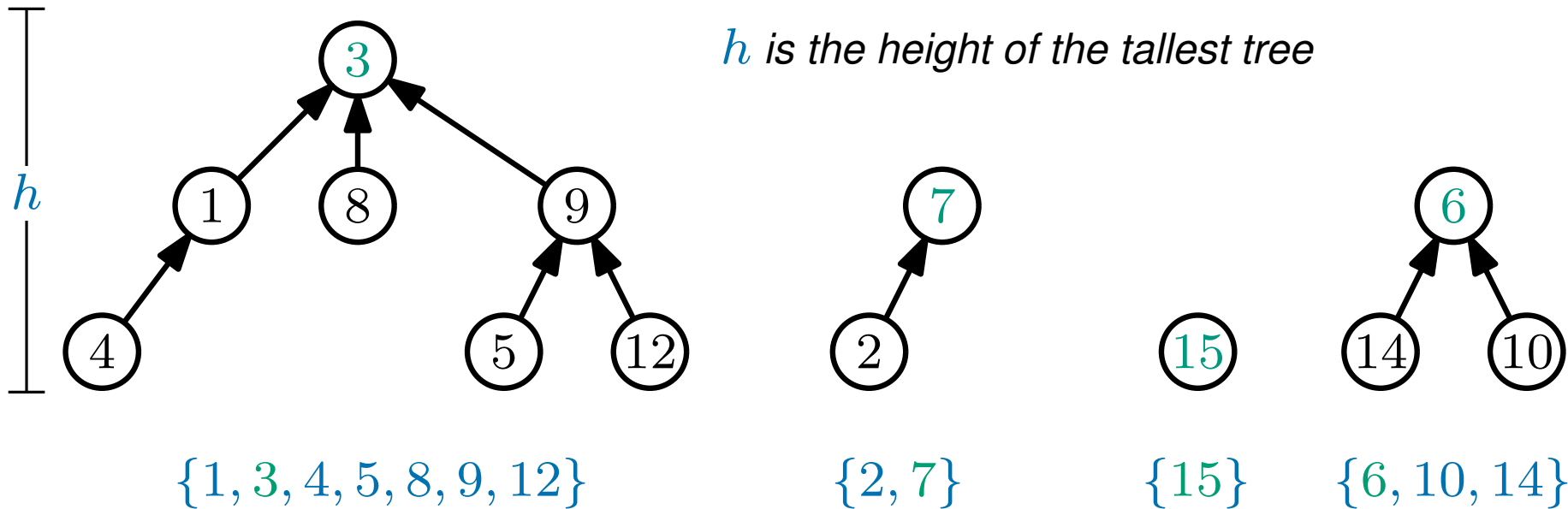
- $O(h)$ time
- Step 1:** Find the node storing element x \swarrow $O(1)$ time
because x is stored in $A[x]$
- Step 2:** Until you are at the root,
follow the pointer to the parent of the current node
- Step 3:** Output the element at the root

What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x

h is the height of the tallest tree



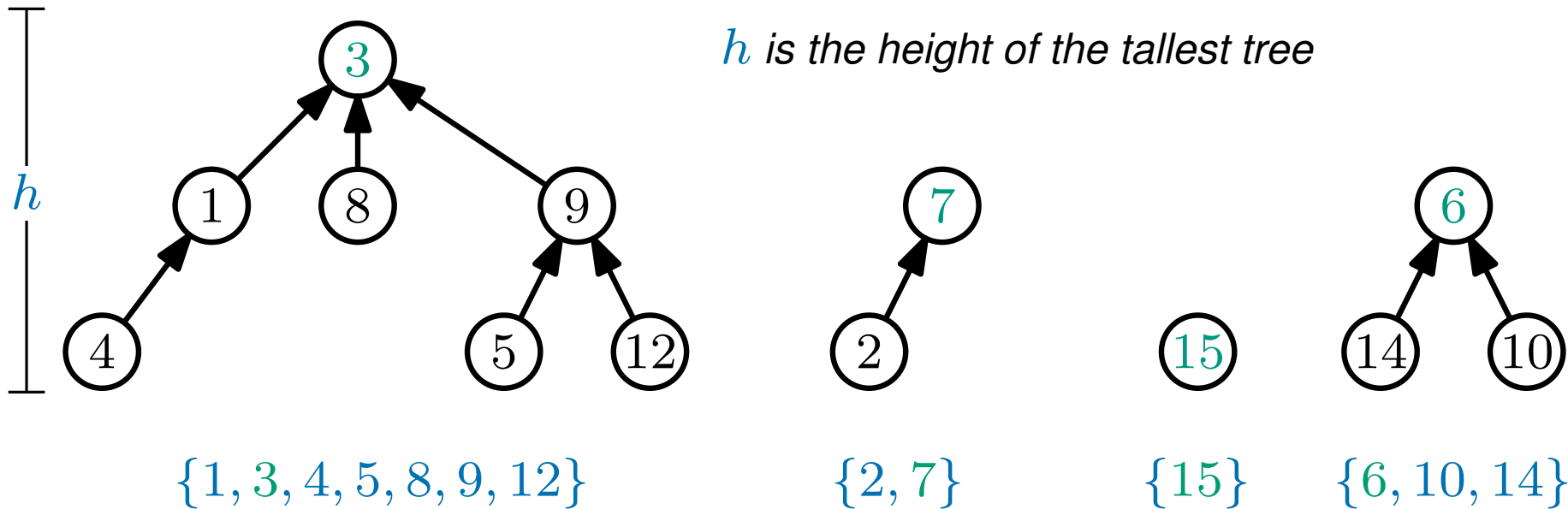
- $O(h)$ time
- Step 1:** Find the node storing element x $\leftarrow O(1)$ time
because x is stored in $A[x]$
 - Step 2:** Until you are at the root,
follow the pointer to the parent of the current node
 - Step 3:** Output the element at the root $\leftarrow O(1)$ time

What is the worst-case time complexity of this operation?

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x

h is the height of the tallest tree

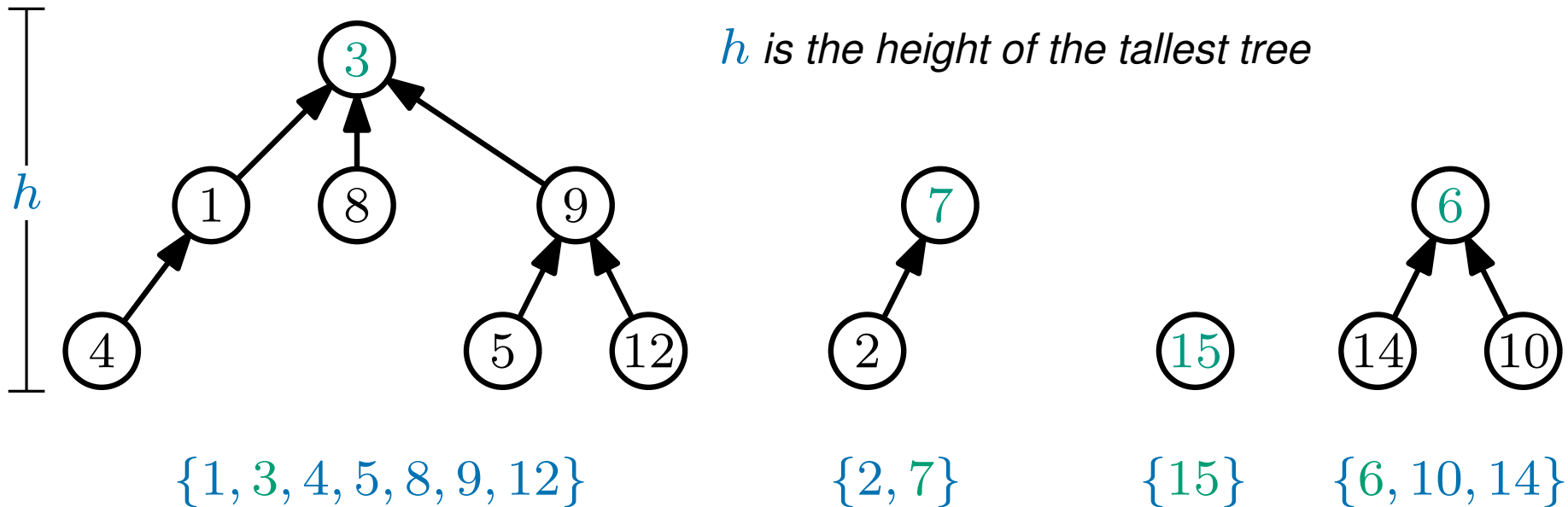


- Step 1:** Find the node storing element x $\leftarrow O(1)$ time
because x is stored in $A[x]$
- $O(h)$ time $\left[\begin{array}{l} \textbf{Step 2:} \text{ Until you are at the root,} \\ \text{follow the pointer to the parent of the current node} \end{array} \right.$
- Step 3:** Output the element at the root $\leftarrow O(1)$ time
- The overall worst-case time complexity is $O(h)$

The FINDSET operation

$\text{FINDSET}(x)$ - returns the *identity* of the set containing x

h is the height of the tallest tree



Step 1: Find the node storing element x

Step 2: Until you are at the root,

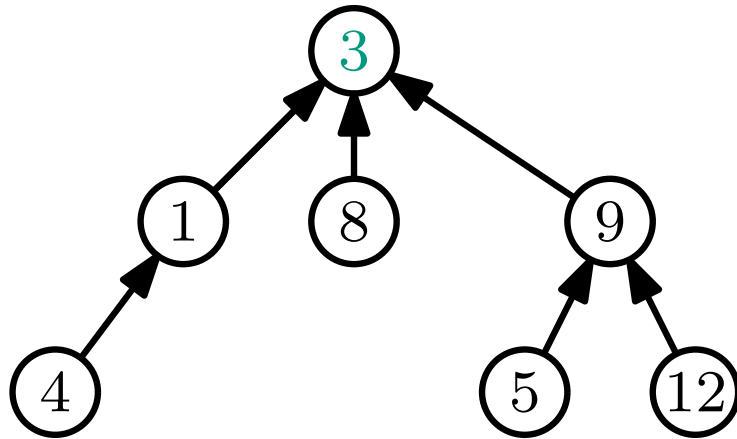
follow the pointer to the parent of the current node

Step 3: Output the element at the root

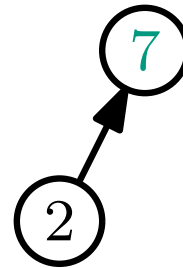
The overall worst-case time complexity is $O(h)$

The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



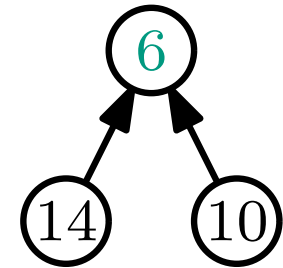
{1, 3, 4, 5, 8, 9, 12}



{2, 7}



{15}

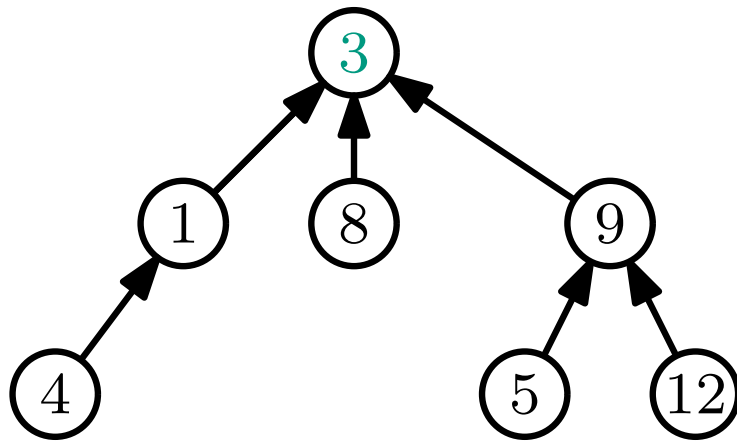


{6, 10, 14}

Step 1: Make a new tree containing x as the root

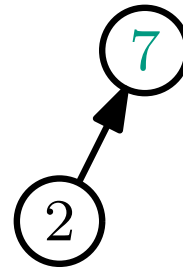
The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



{1, 3, 4, 5, 8, 9, 12}

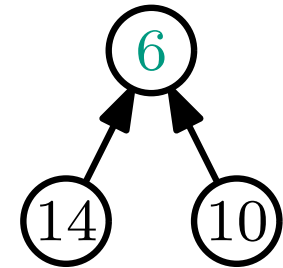
MAKESET(11)



{2, 7}



{15}

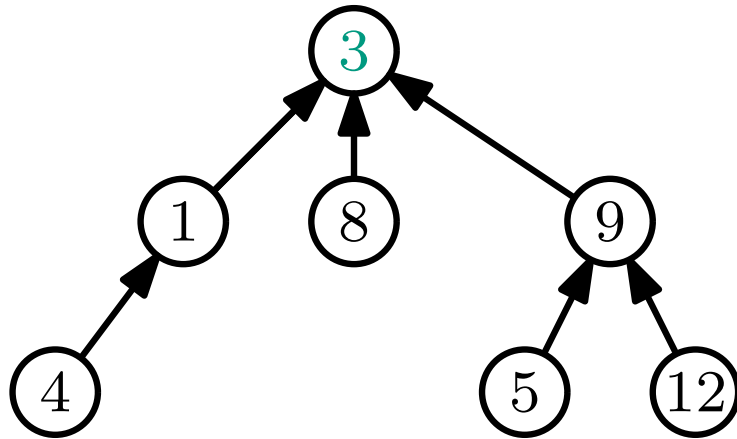


{6, 10, 14}

Step 1: Make a new tree containing x as the root

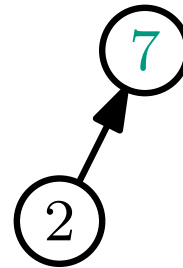
The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



{1, 3, 4, 5, 8, 9, 12}

MAKESET(11)



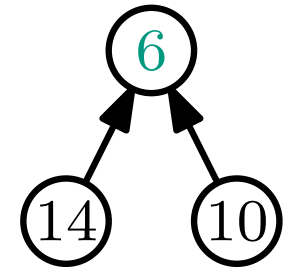
{2, 7}



{11}



{15}

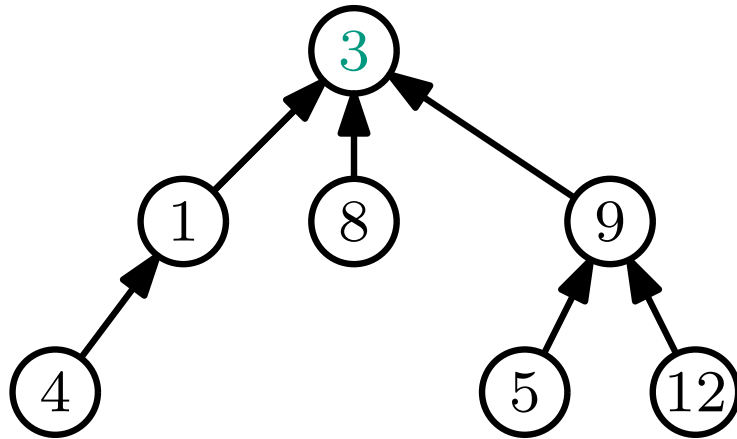


{6, 10, 14}

Step 1: Make a new tree containing x as the root

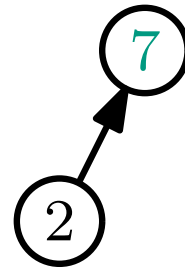
The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



{1, 3, 4, 5, 8, 9, 12}

MAKESET(11)



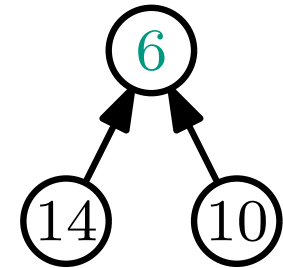
{2, 7}



{11}



{15}

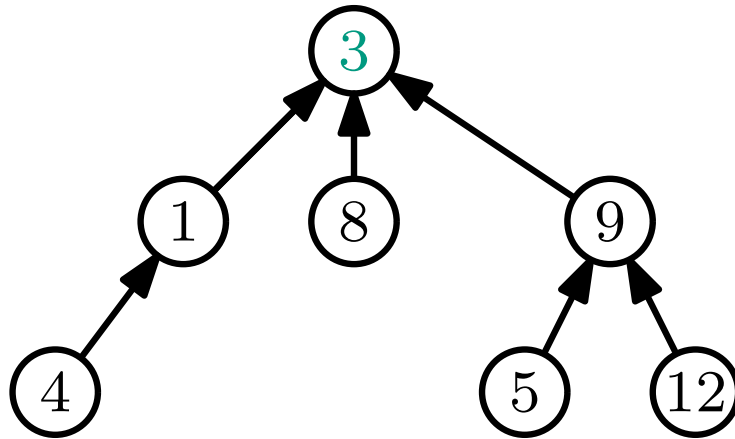


{6, 10, 14}

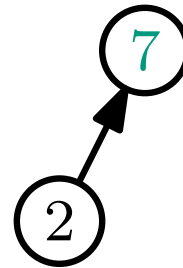
Step 1: Make a new tree containing x as the root
(that's it)

The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



{1, 3, 4, 5, 8, 9, 12}



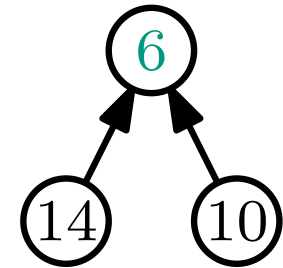
{2, 7}



{11}



{15}



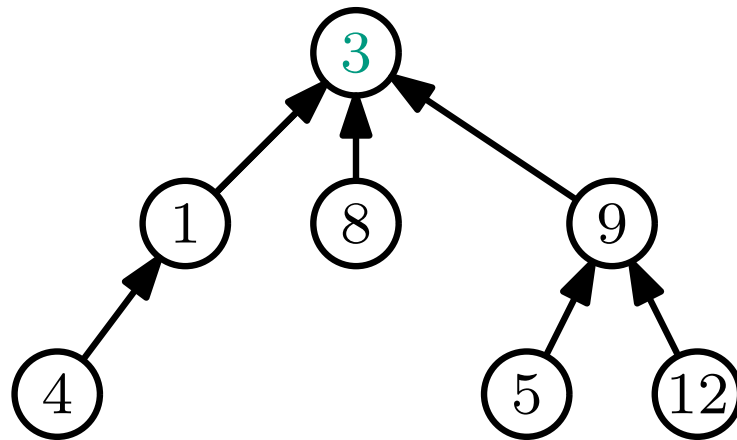
{6, 10, 14}

Step 1: Make a new tree containing x as the root
(that's it)

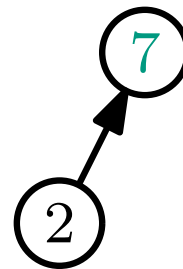
What is the worst-case time complexity of this operation?

The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



$\{1, 3, 4, 5, 8, 9, 12\}$



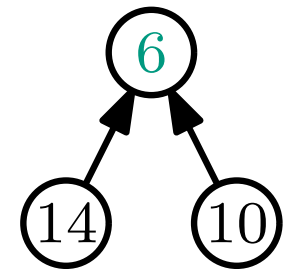
$\{2, 7\}$



$\{11\}$



$\{15\}$



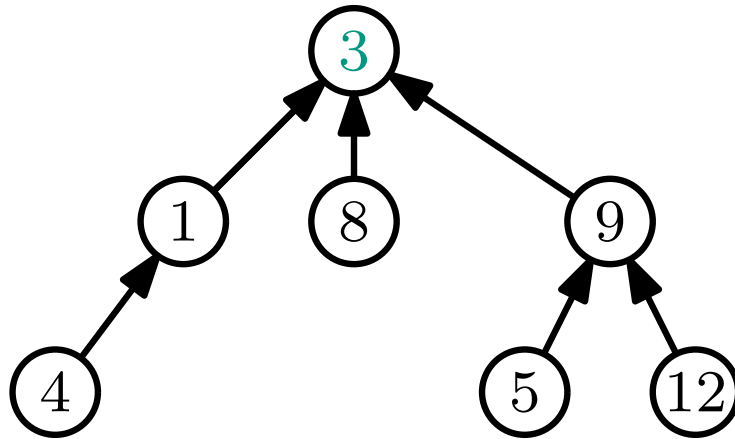
$\{6, 10, 14\}$

$O(1)$ time \rightarrow **Step 1:** Make a new tree containing x as the root
(that's it)
because x should be stored in $A[x]$

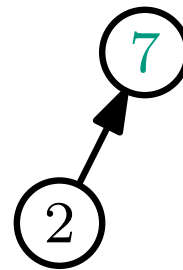
What is the worst-case time complexity of this operation?

The MAKESET operation

MAKESET(x) - make a new set containing only x (which is not already in a set)



$\{1, 3, 4, 5, 8, 9, 12\}$



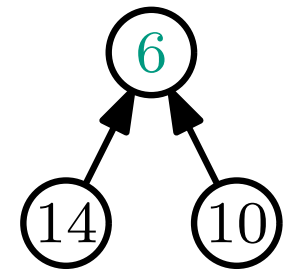
$\{2, 7\}$



$\{11\}$



$\{15\}$



$\{6, 10, 14\}$

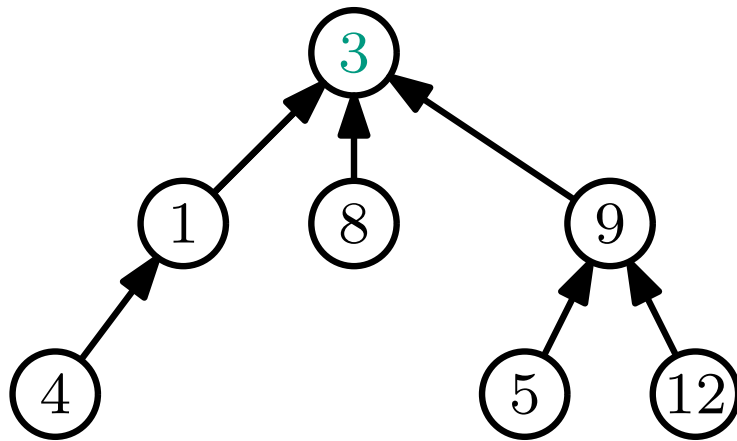
$O(1)$ time \rightarrow **Step 1:** Make a new tree containing x as the root
(that's it)
because x should be stored in $A[x]$

What is the worst-case time complexity of this operation?

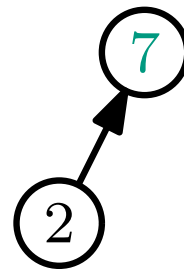
The overall worst-case time complexity is $O(1)$

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



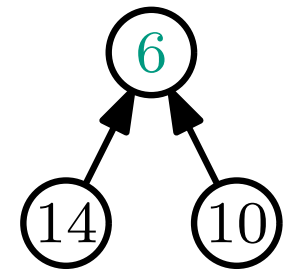
$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7\}$



$\{15\}$



$\{6, 10, 14\}$

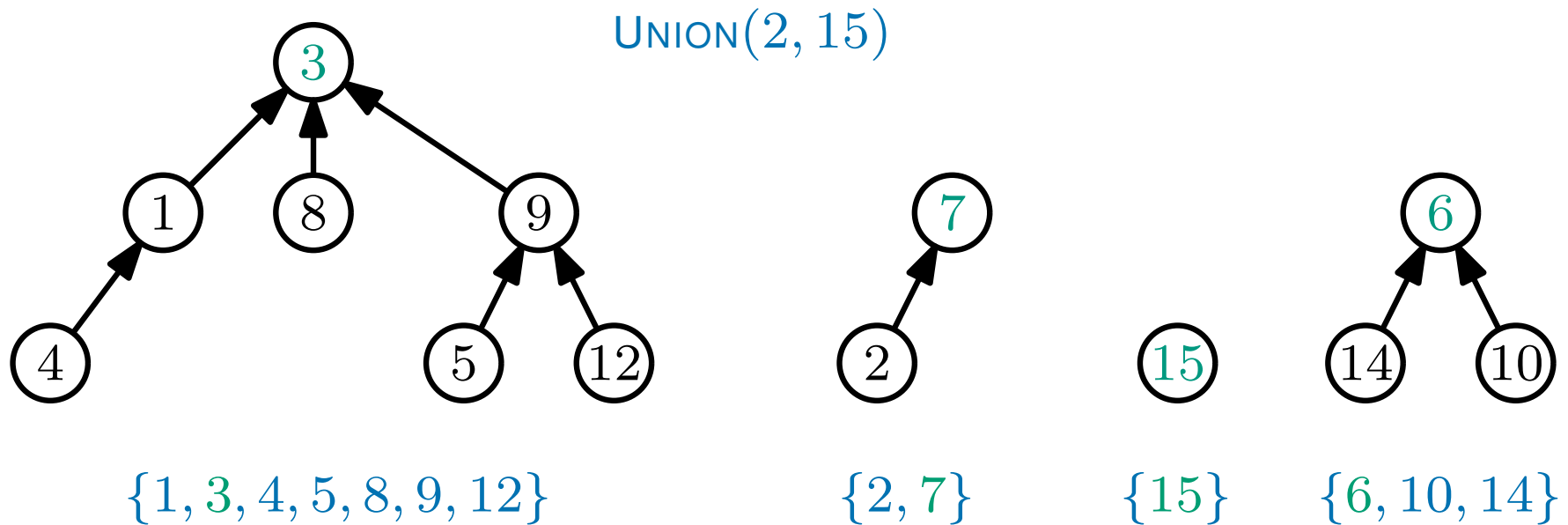
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



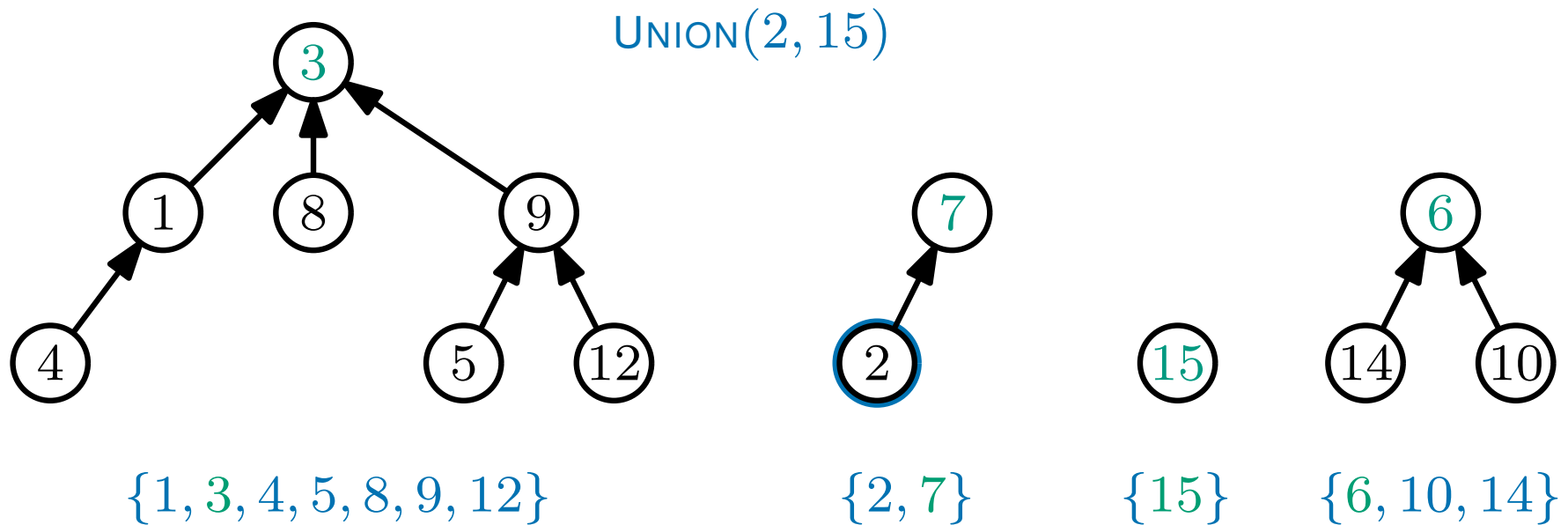
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



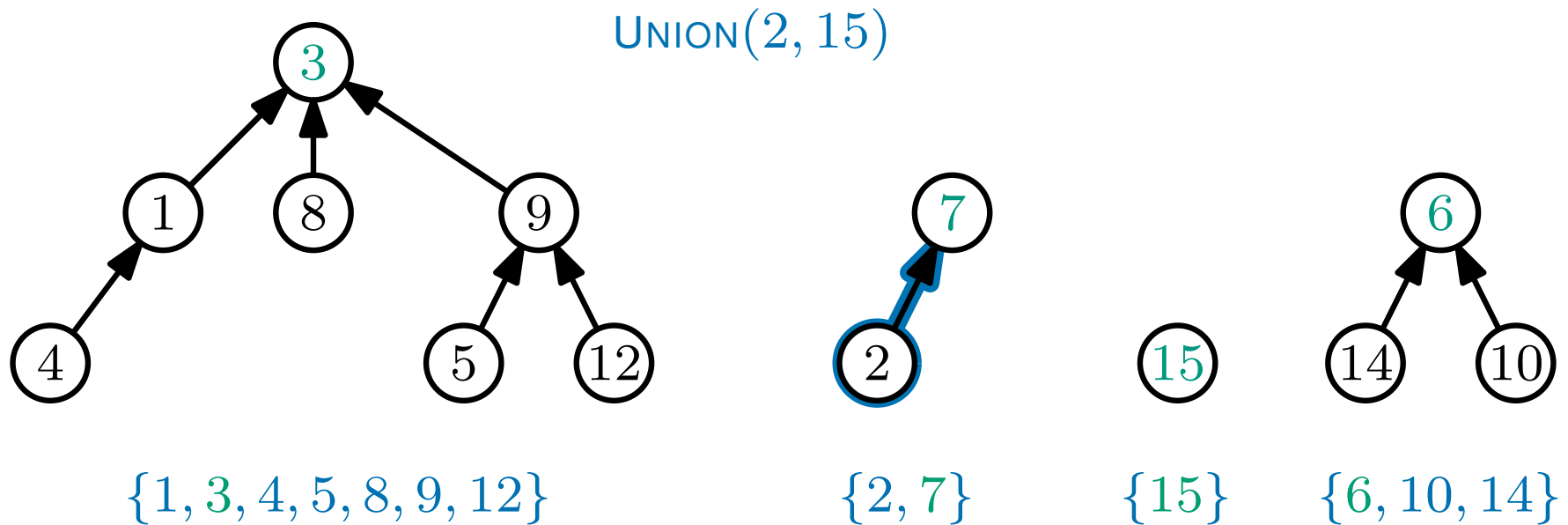
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



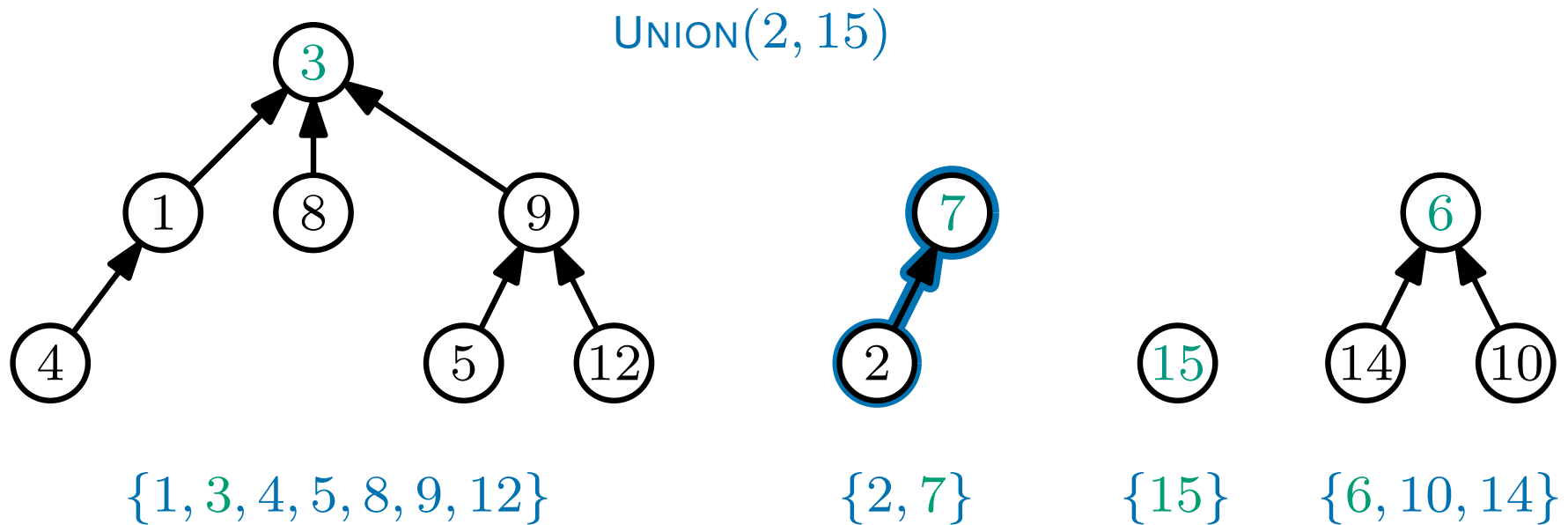
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The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



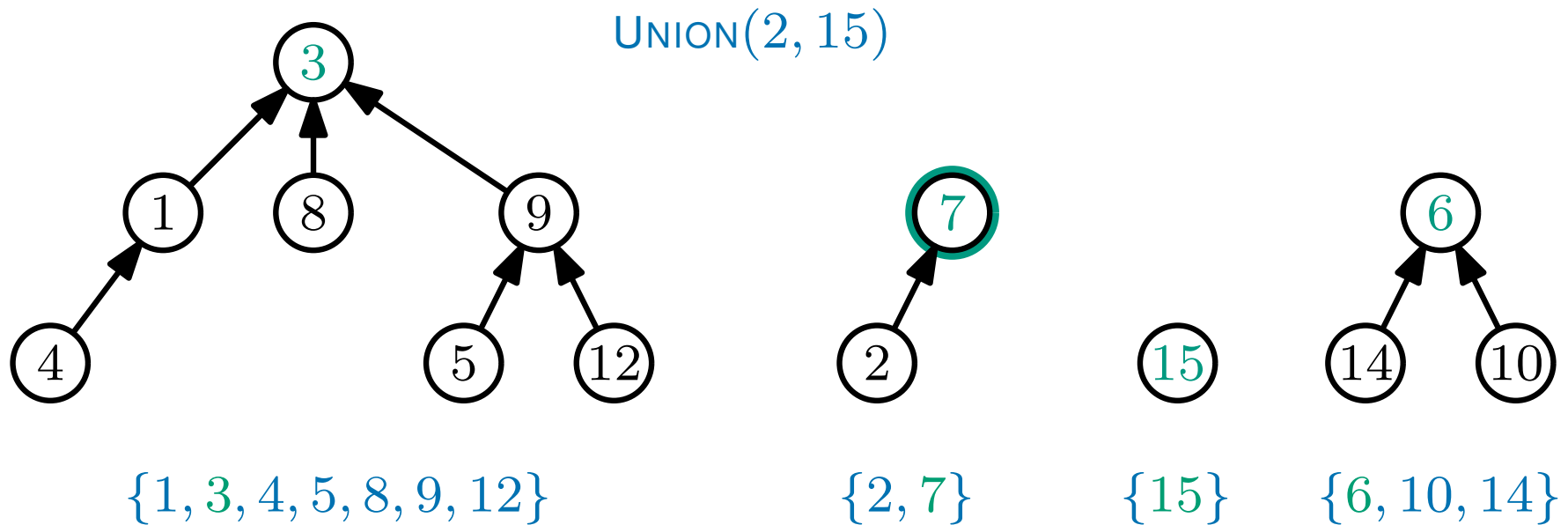
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



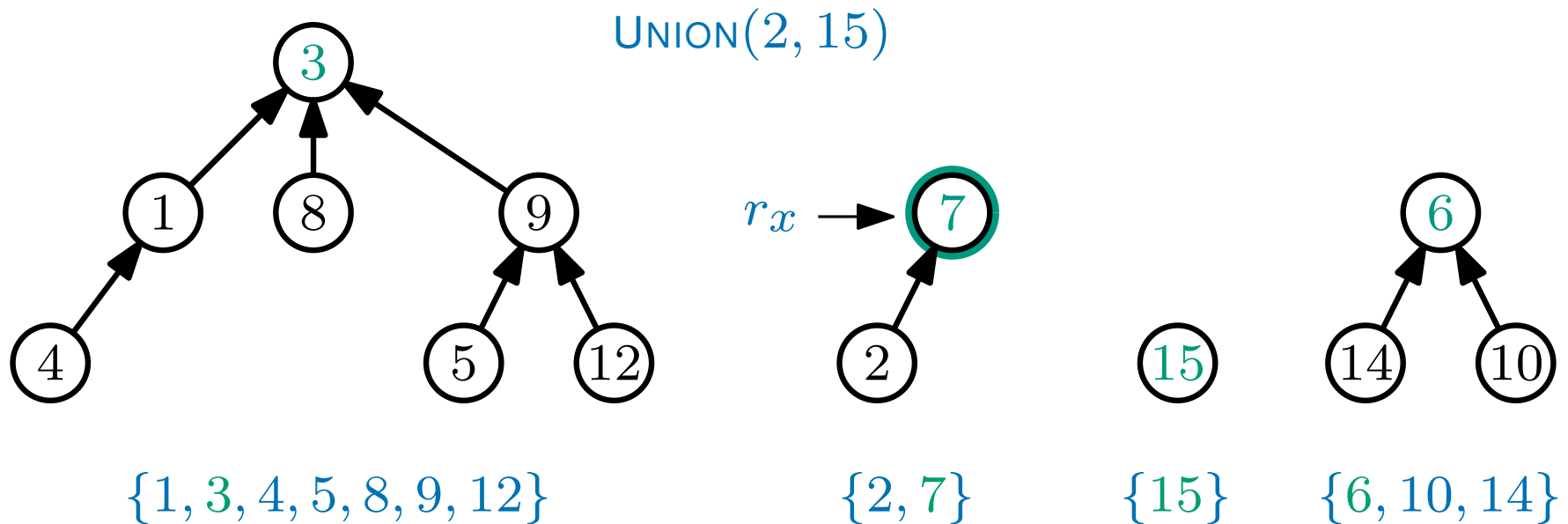
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



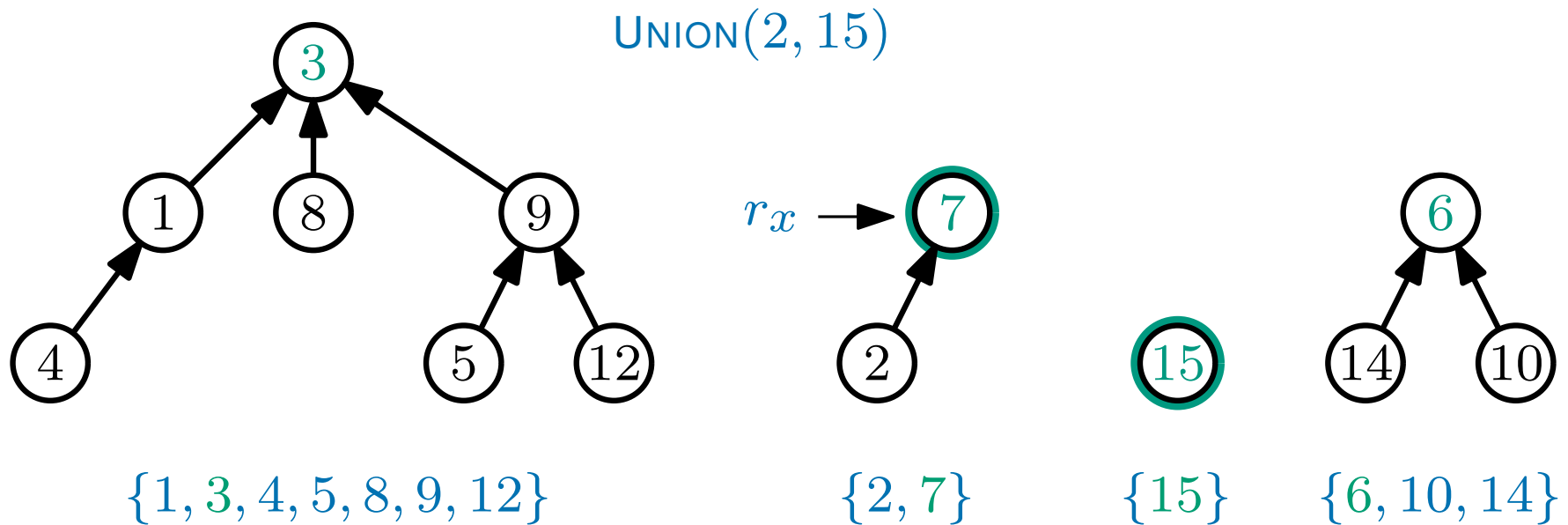
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



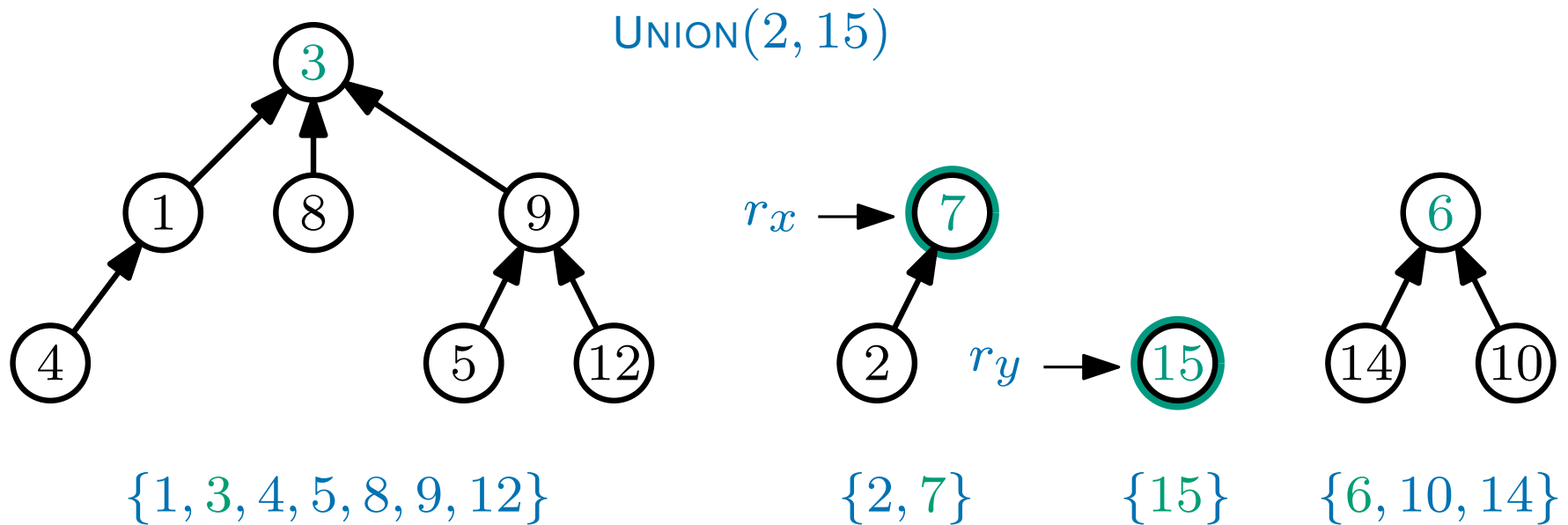
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



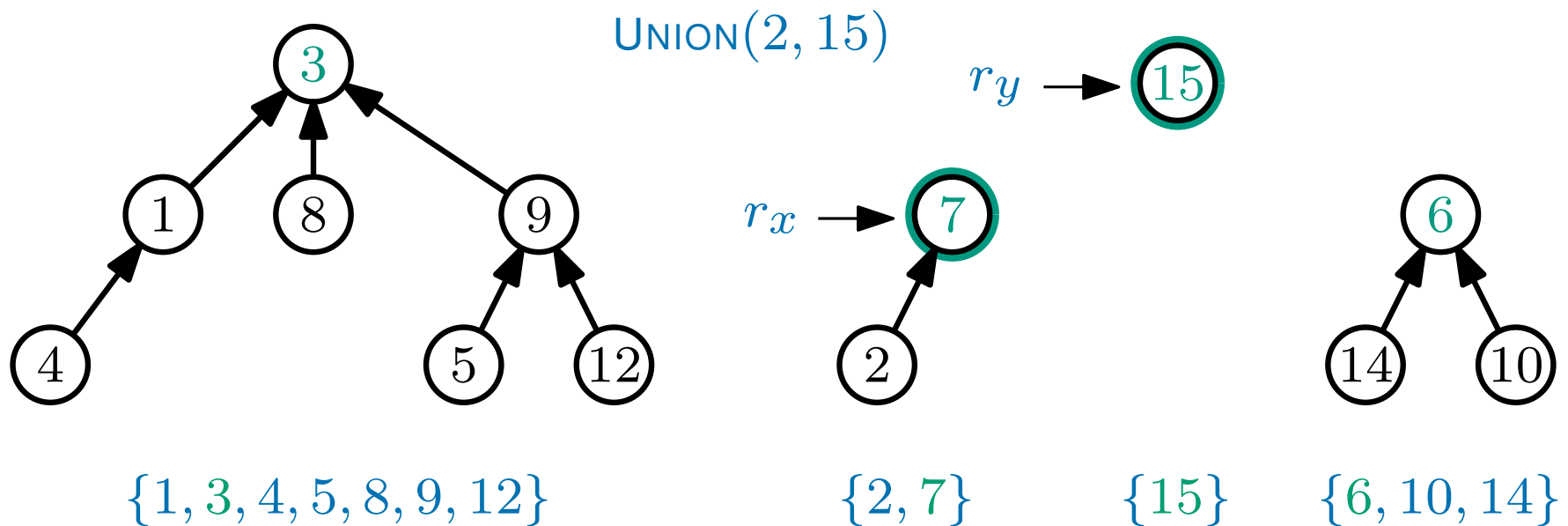
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

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The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



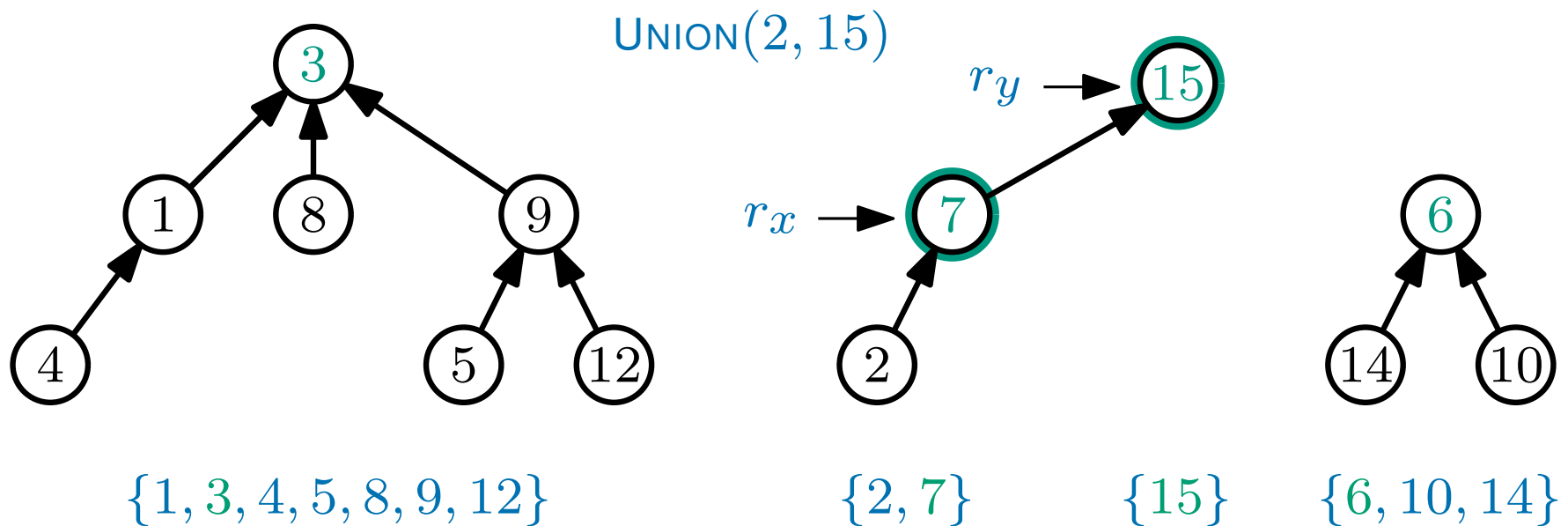
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



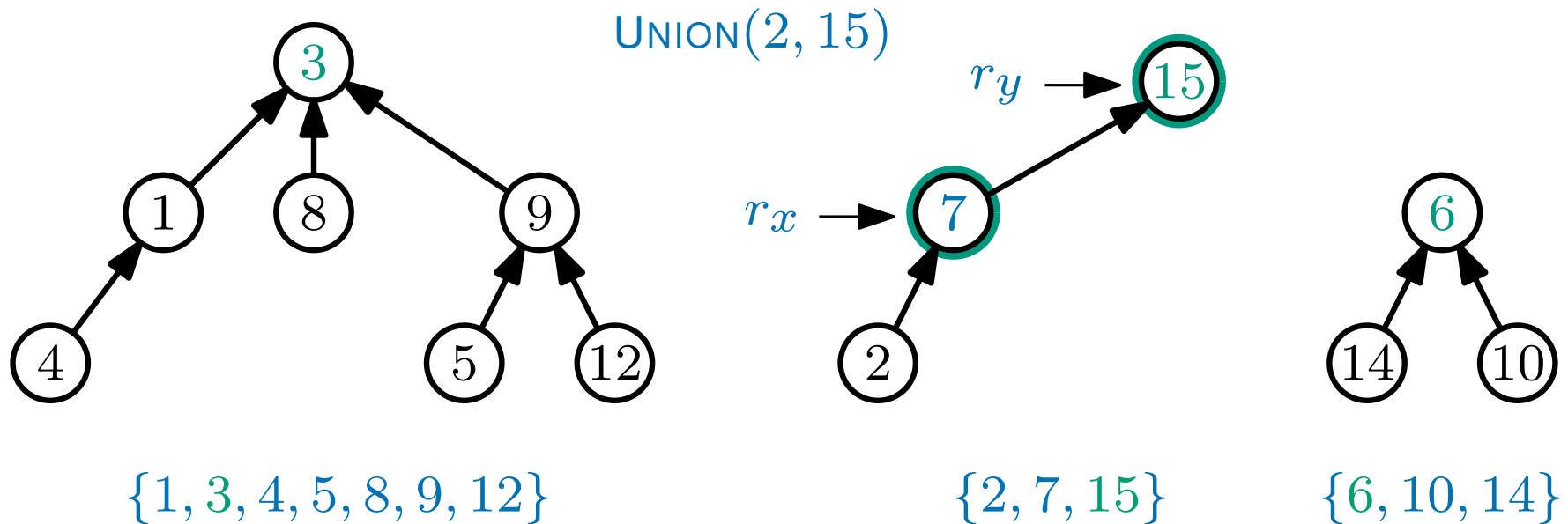
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



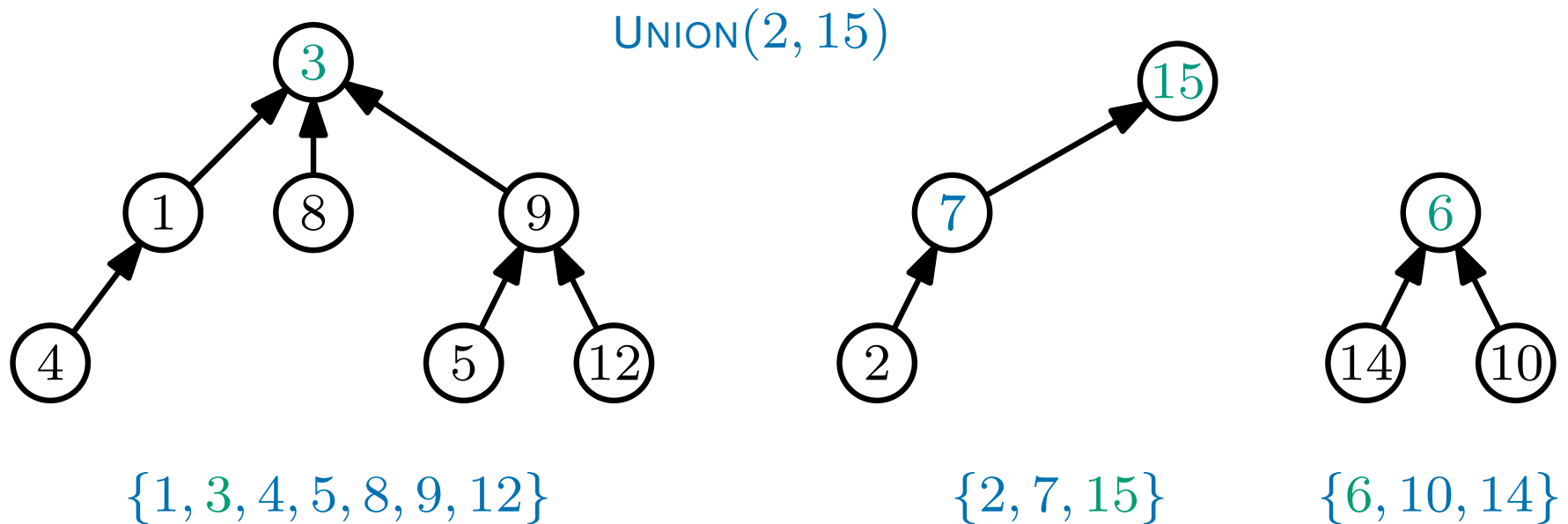
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

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Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



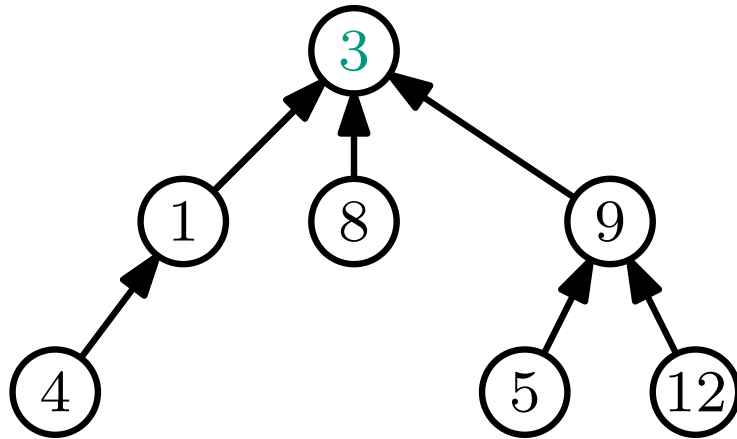
Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

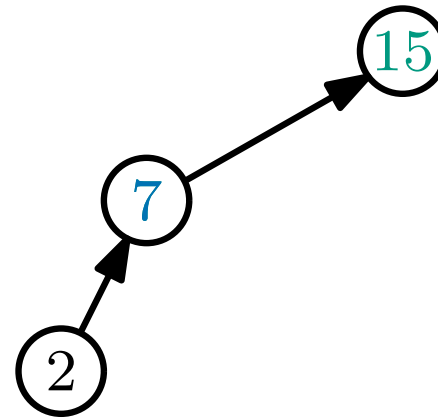
Step 3: Make r_x a child of r_y (which merges the two trees)

The UNION operation

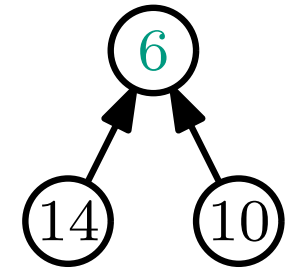
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7, 15\}$



$\{6, 10, 14\}$

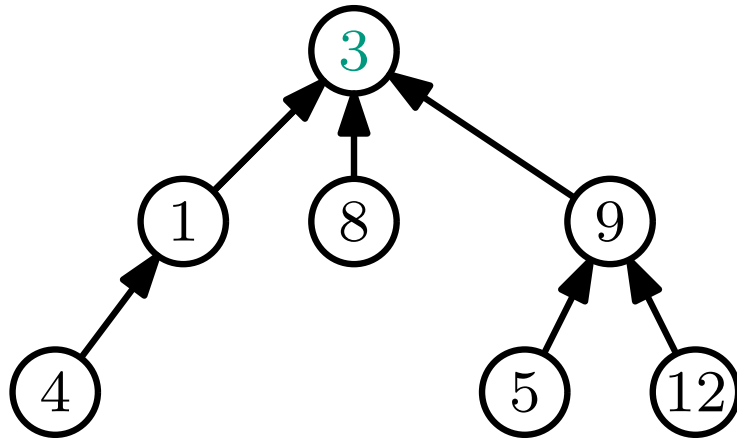
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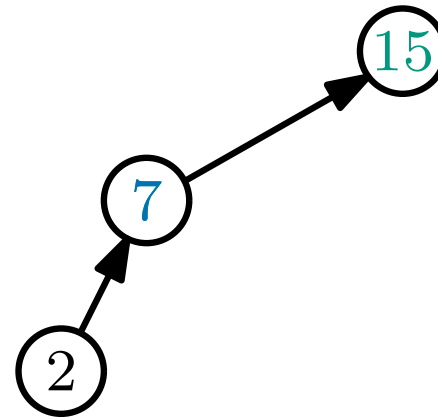
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The UNION operation

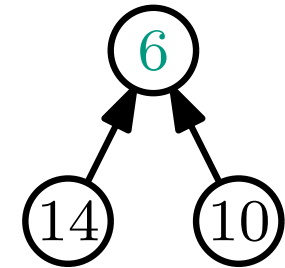
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



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$\{2, 7, 15\}$



$\{6, 10, 14\}$

Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

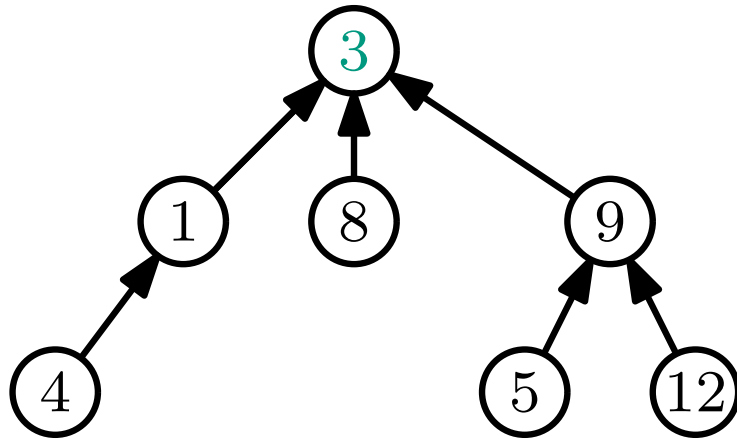
Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

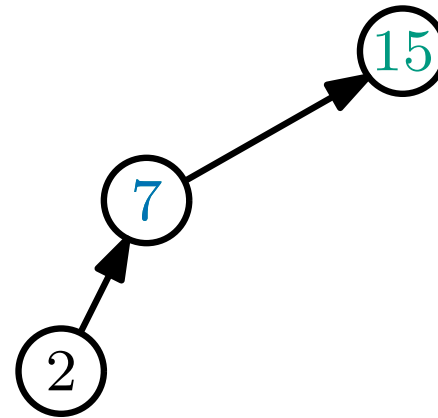
What is the worst-case time complexity of this operation?

The UNION operation

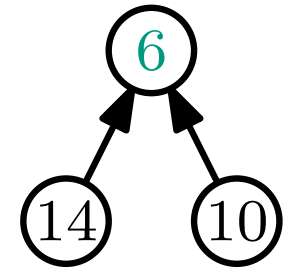
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7, 15\}$



$\{6, 10, 14\}$

$O(h)$ time

Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

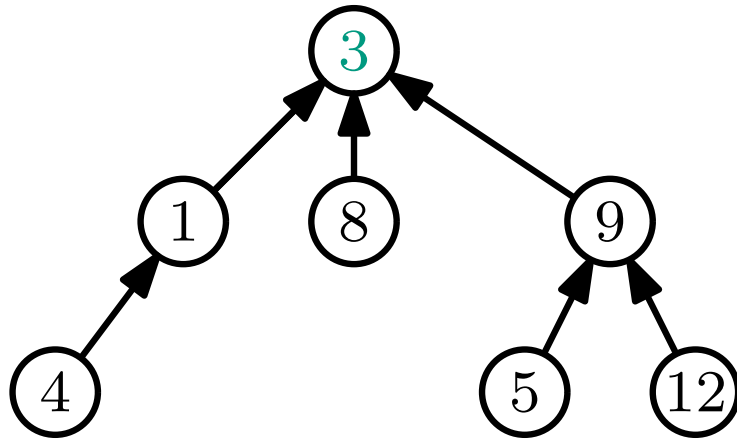
Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

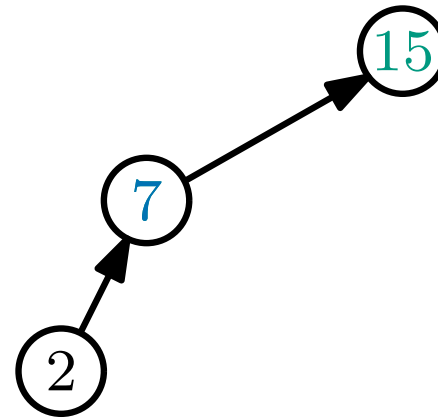
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The UNION operation

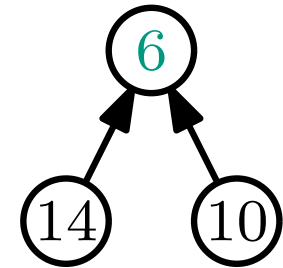
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7, 15\}$



$\{6, 10, 14\}$

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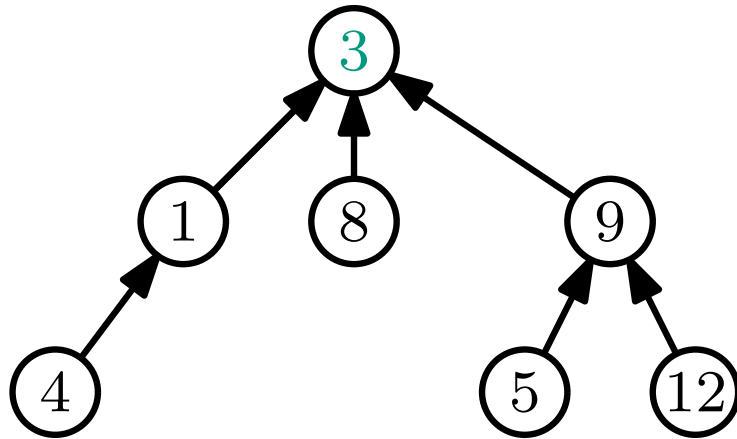
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Step 3: Make r_x a child of r_y (which merges the two trees)

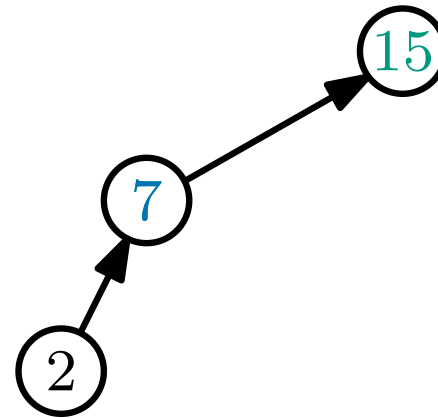
What is the worst-case time complexity of this operation?

The UNION operation

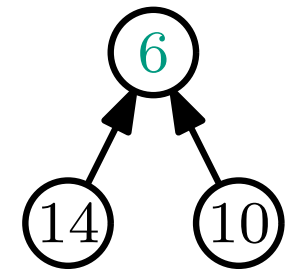
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7, 15\}$



$\{6, 10, 14\}$

$O(h)$ time

Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

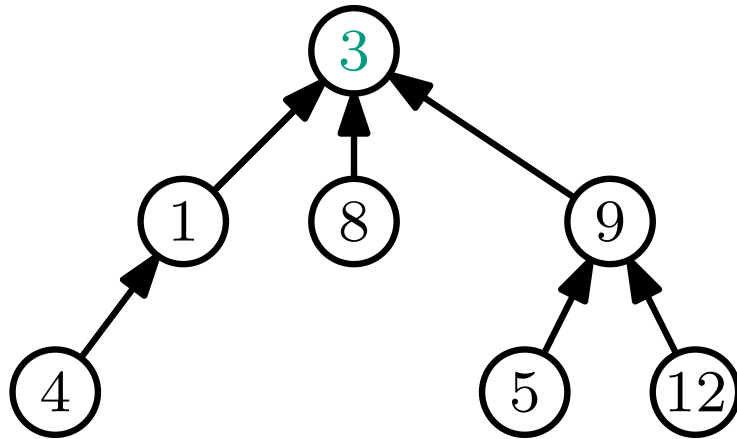
Step 3: Make r_x a child of r_y (which merges the two trees)

$O(1)$ time

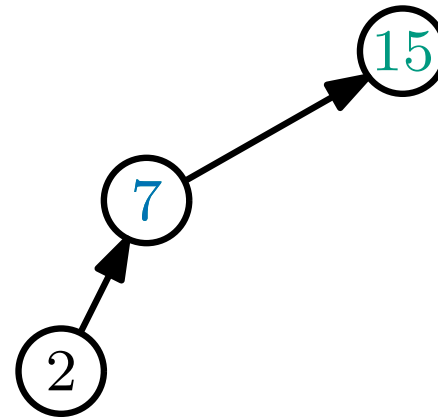
What is the worst-case time complexity of this operation?

The UNION operation

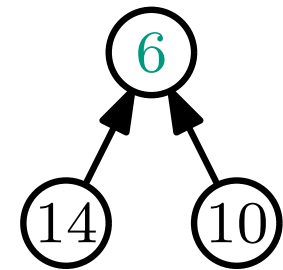
$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



$\{1, 3, 4, 5, 8, 9, 12\}$



$\{2, 7, 15\}$



$\{6, 10, 14\}$

$O(h)$ time

Step 1: Compute $r_x = \text{FINDSET}(x)$ - the root of the tree containing x

Step 2: Compute $r_y = \text{FINDSET}(y)$ - the root of the tree containing y

Step 3: Make r_x a child of r_y (which merges the two trees)

$O(1)$ time

What is the worst-case time complexity of this operation?

it's $O(h)$ again

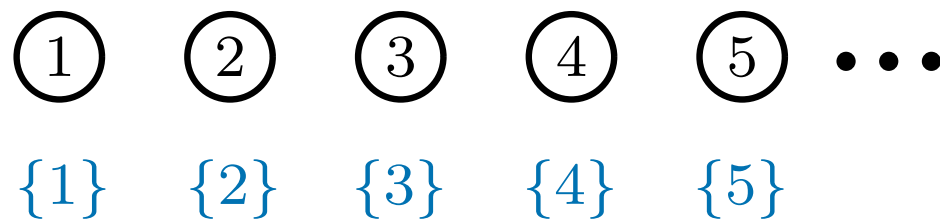
How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

How high does the sycamore grow?

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increase the tallest tree height, h by one. . .

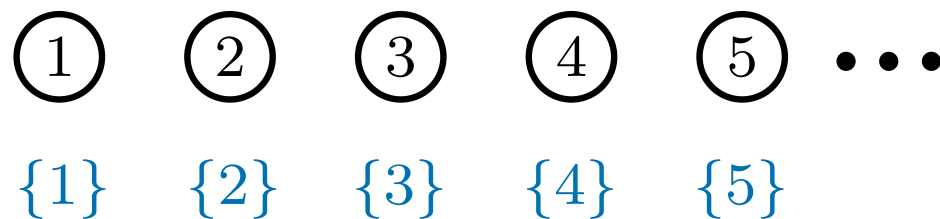
Consider the following sets:



How high does the sycamore grow?

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increase the tallest tree height, h by one. . .

Consider the following sets:

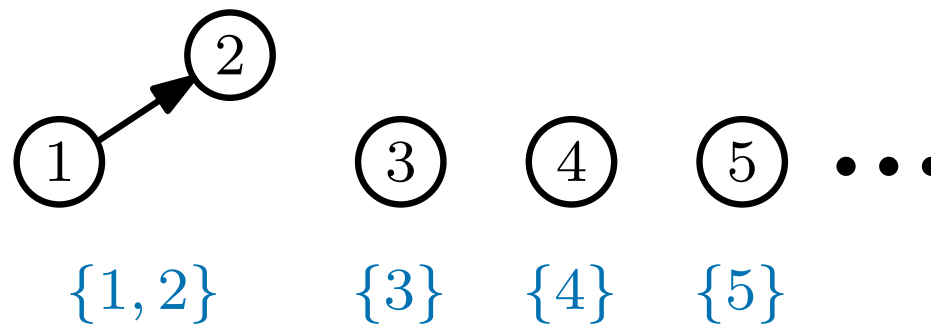


Now perform UNION(1, 2)

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:

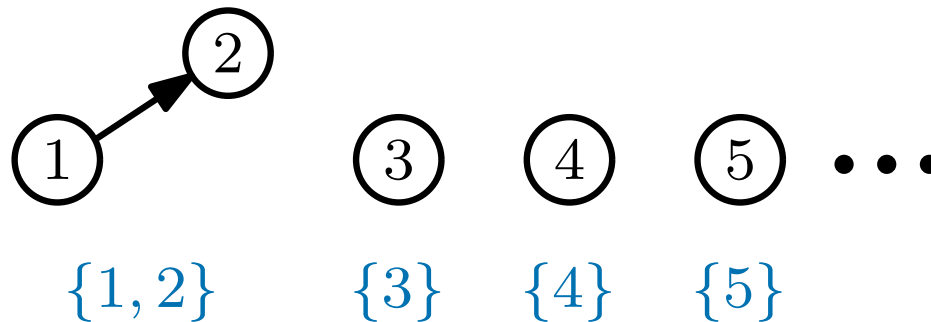


Now perform UNION(1, 2)

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

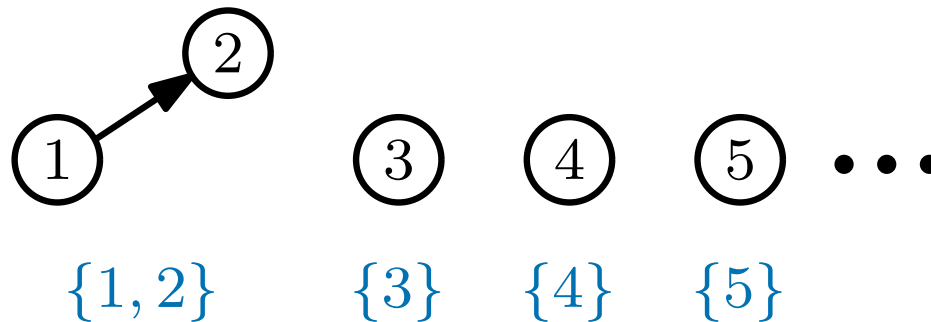
Consider the following sets:



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increase the tallest tree height, h by one. . .

Consider the following sets:

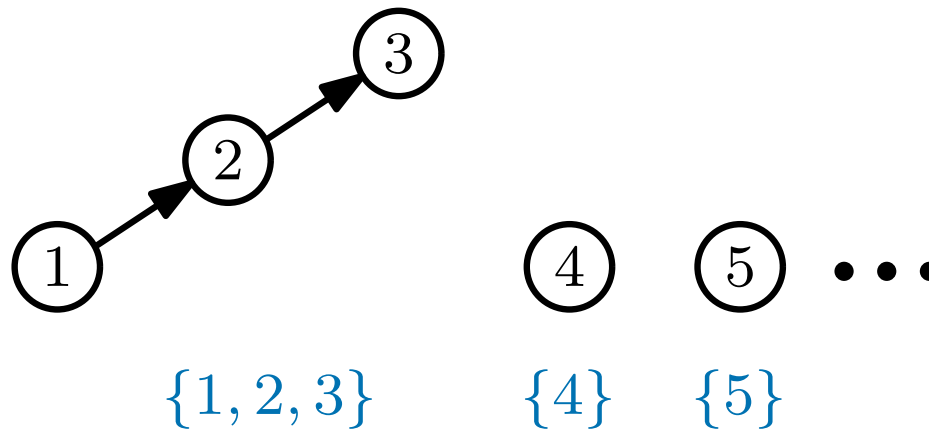


Now perform UNION(1, 3)

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:

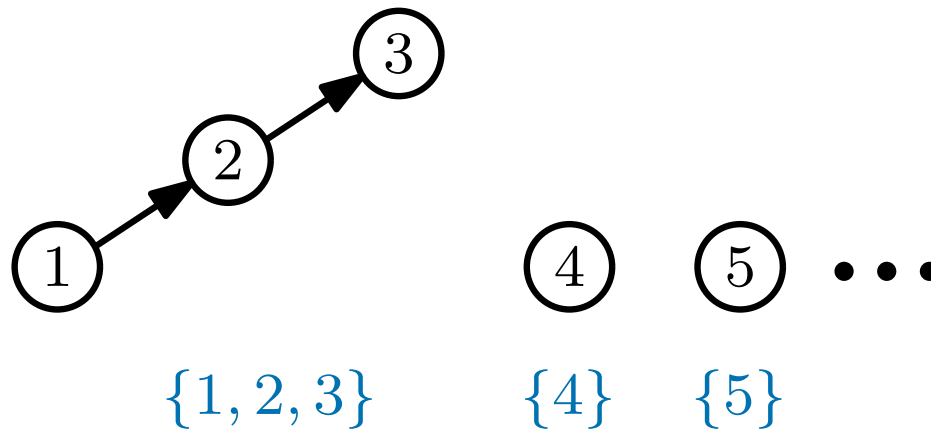


Now perform $\text{UNION}(1, 3)$

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

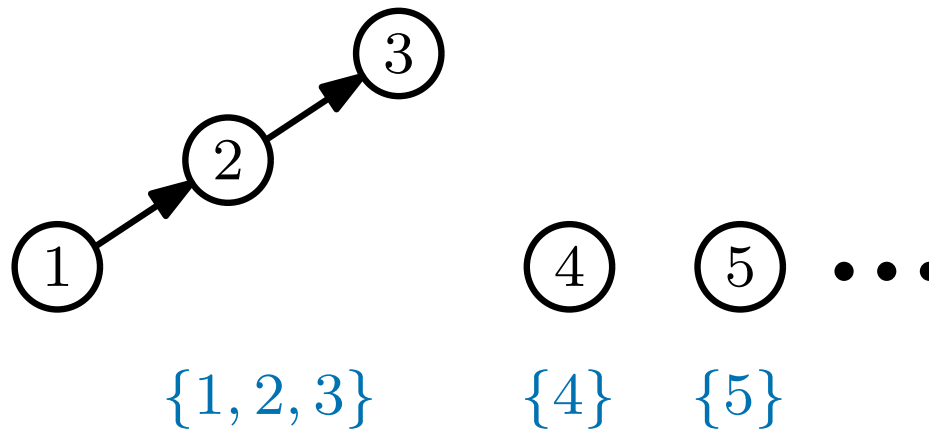
Consider the following sets:



How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:

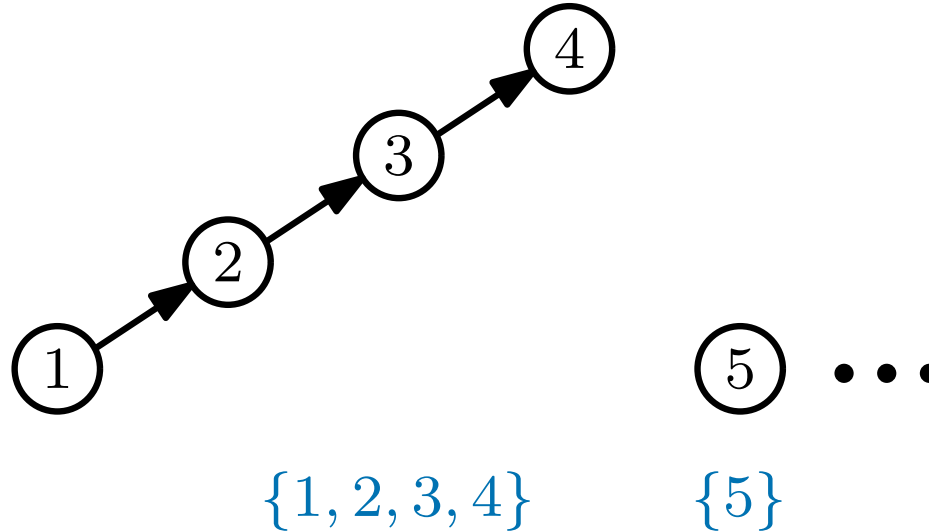


Now perform $\text{UNION}(1, 4)$

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:

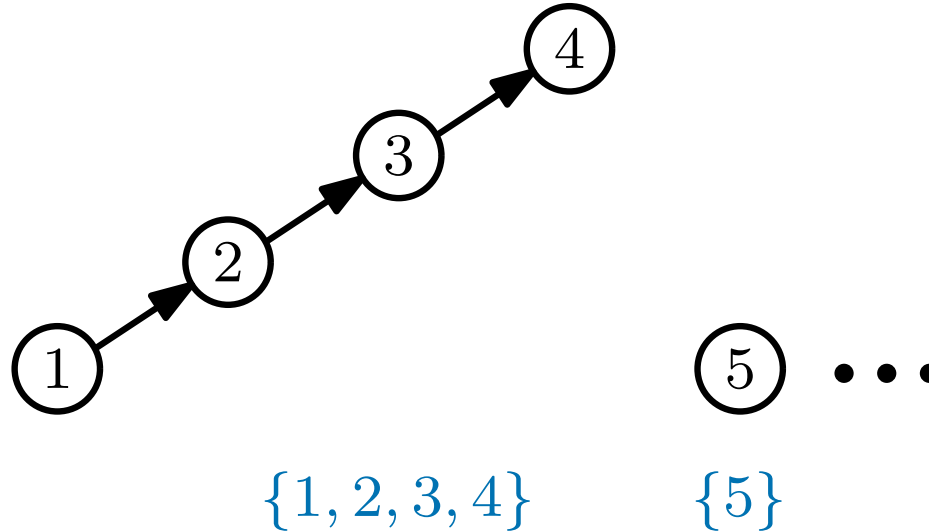


Now perform $\text{UNION}(1, 4)$

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

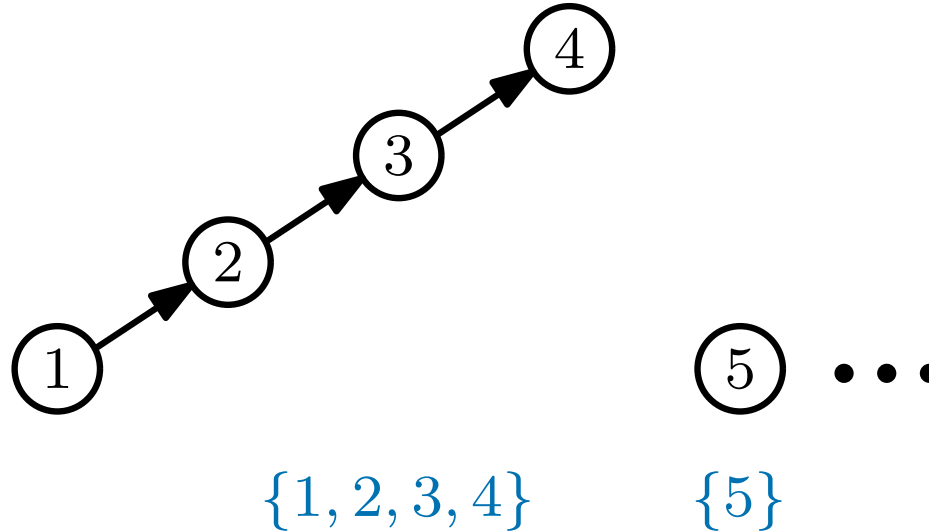
Consider the following sets:



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Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:

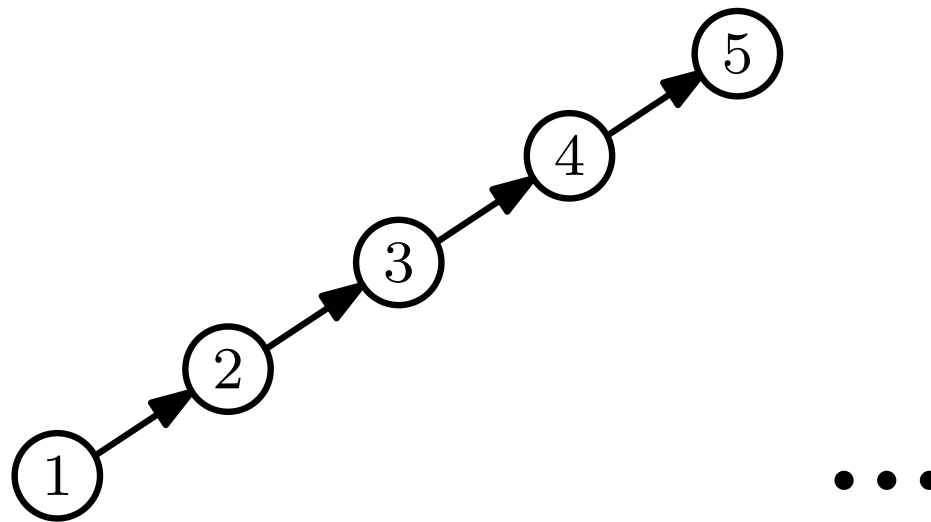


Now perform UNION(1, 5)

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

Consider the following sets:



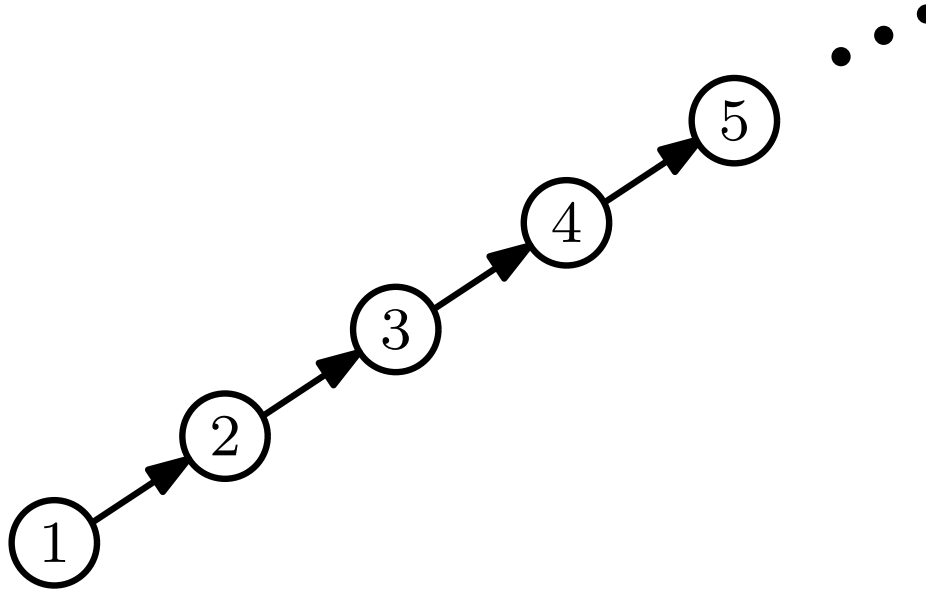
$\{1, 2, 3, 4, 5\}$

Now perform UNION(1, 5)

How high does the sycamore grow?

Unfortunately, every UNION operation could
increase the tallest tree height, h by one. . .

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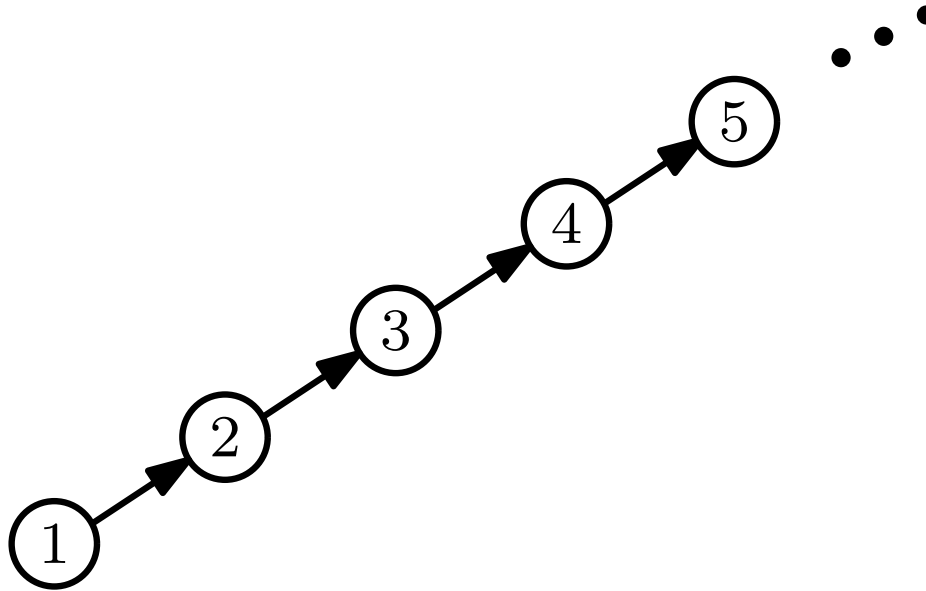
Now perform UNION(1, 5) . . .

So in the worst case the height of the tallest tree is n

How high does the sycamore grow?

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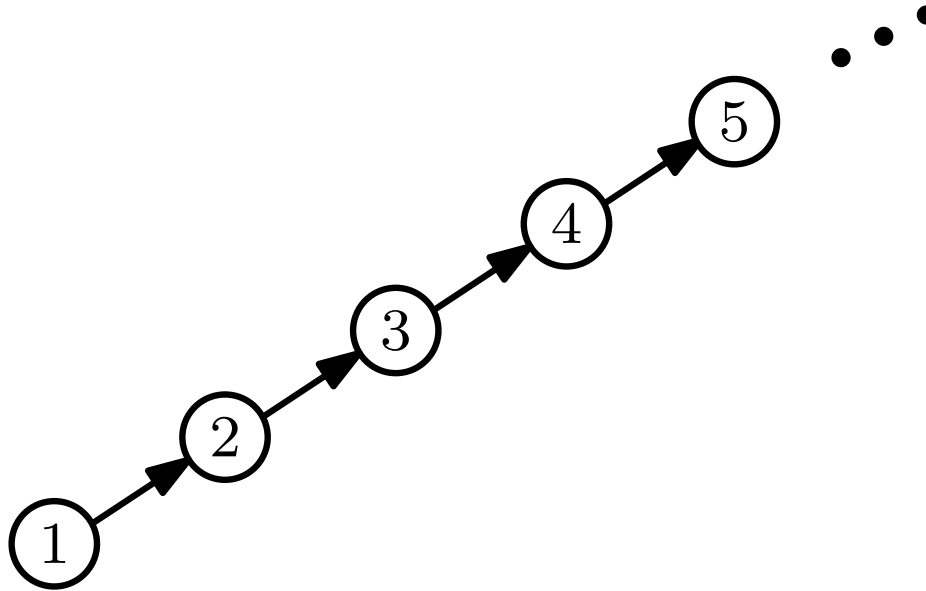


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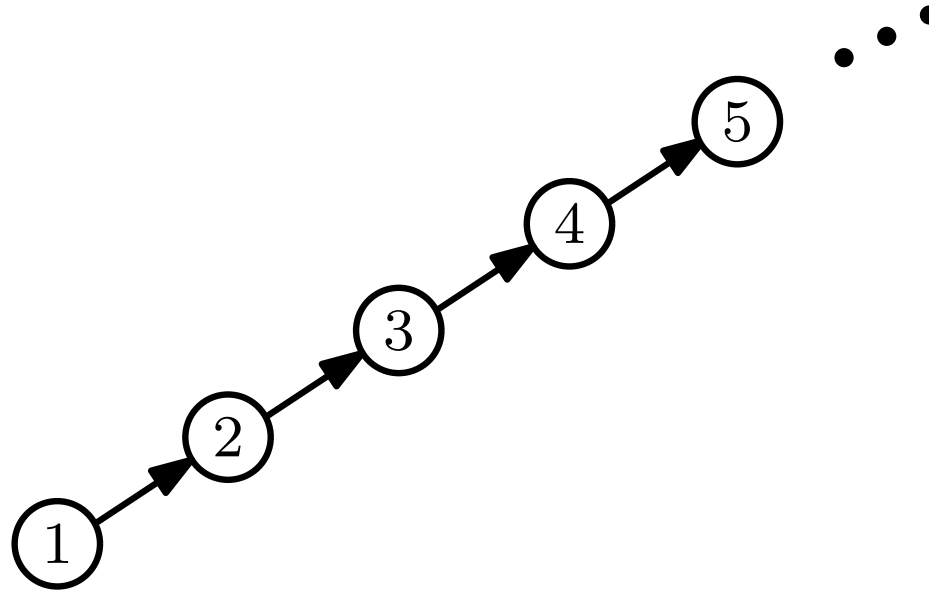
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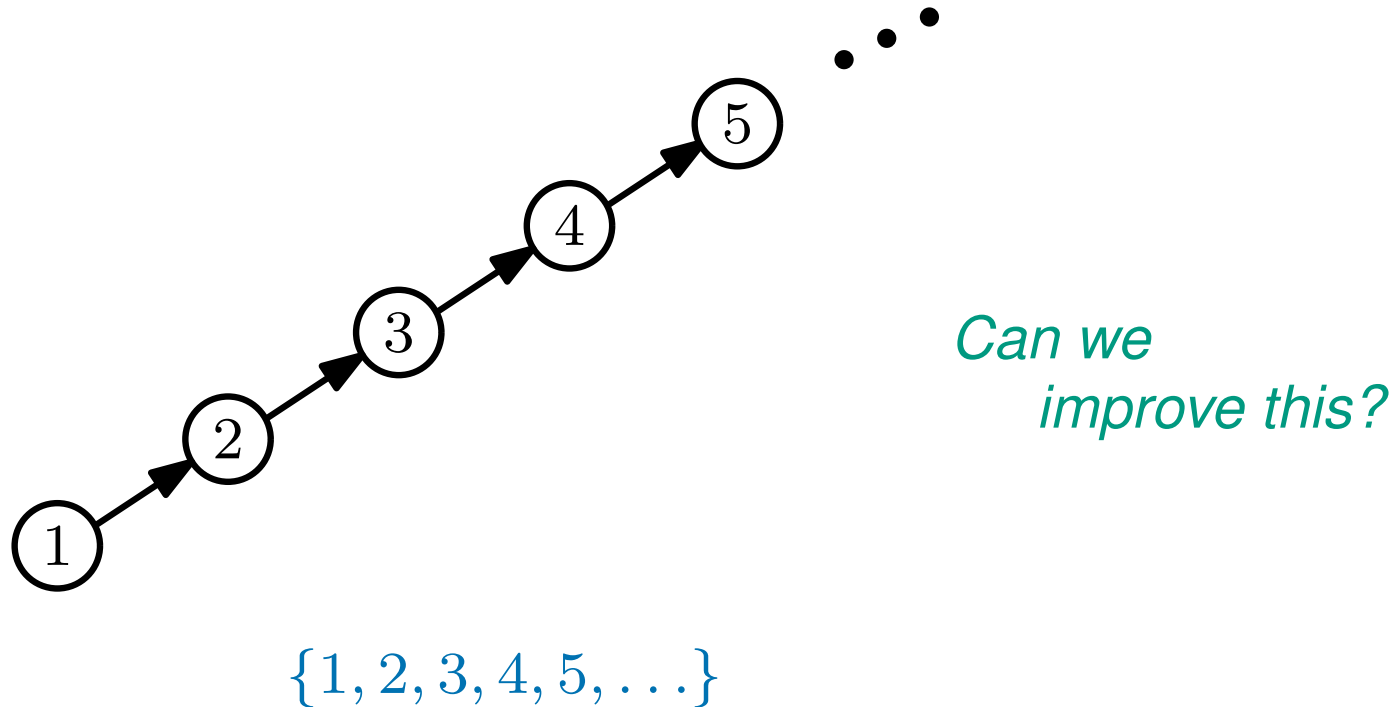
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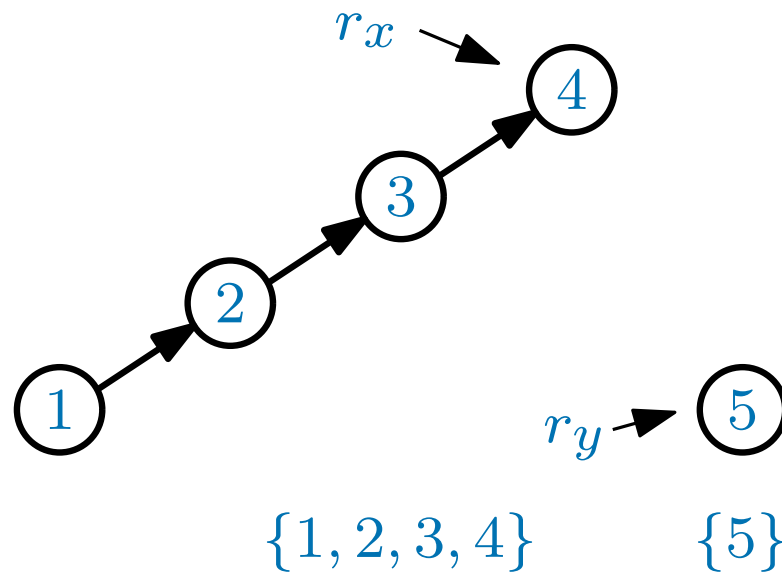
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What's bad about the UNION operation?

$\text{UNION}(x, y)$ - merge the sets containing x and y into a single set



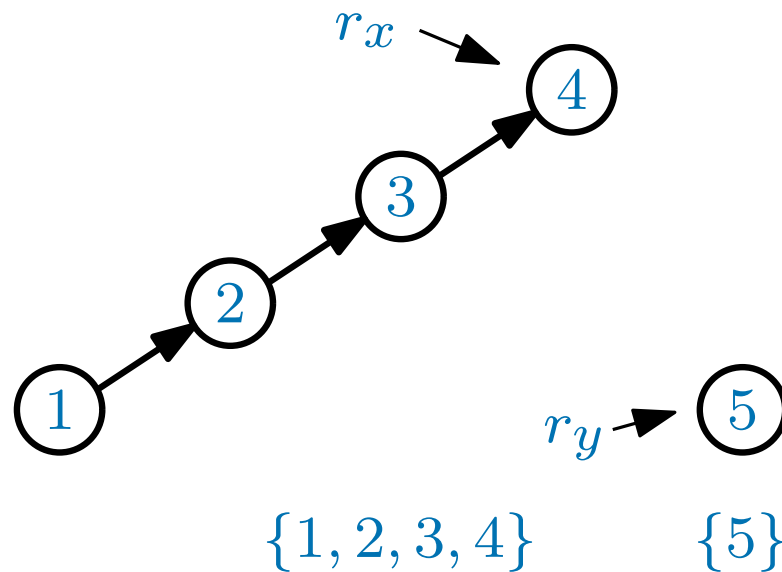
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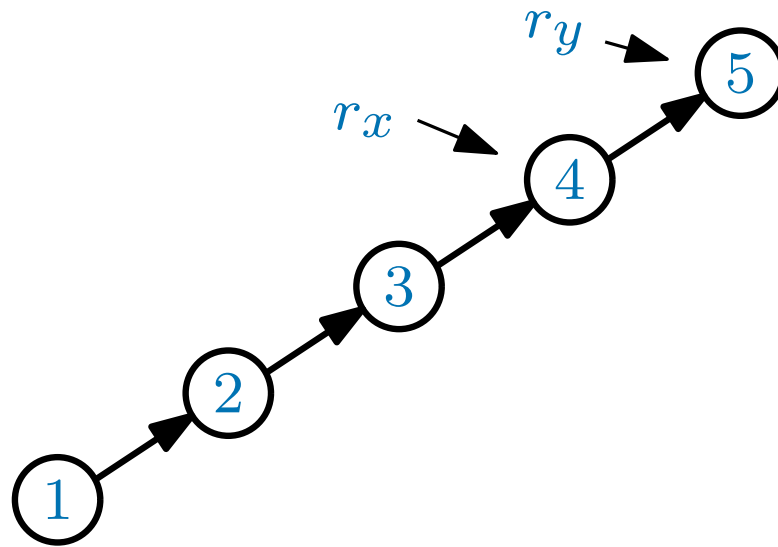
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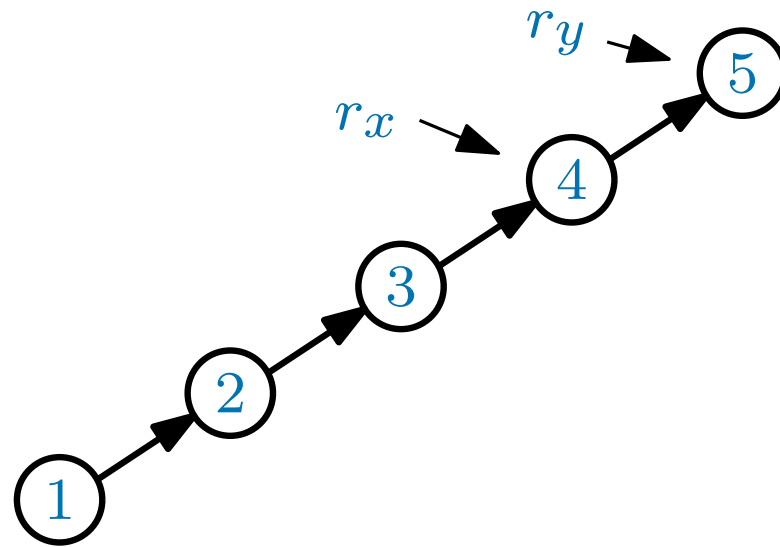
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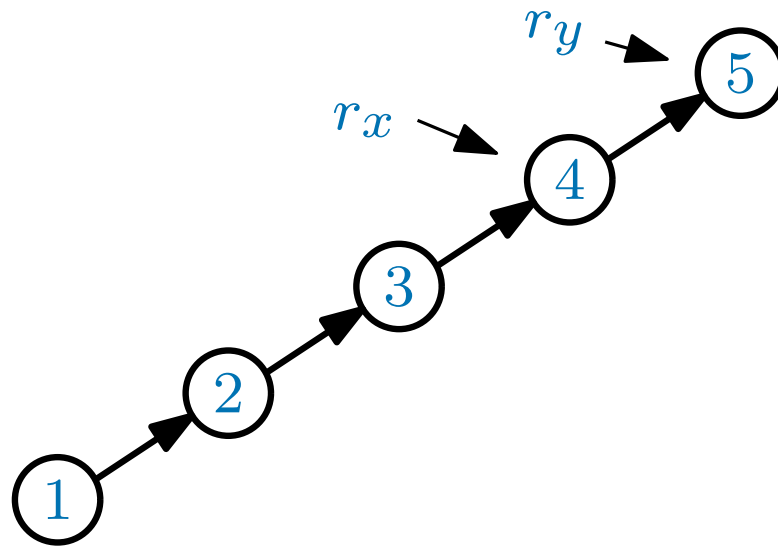
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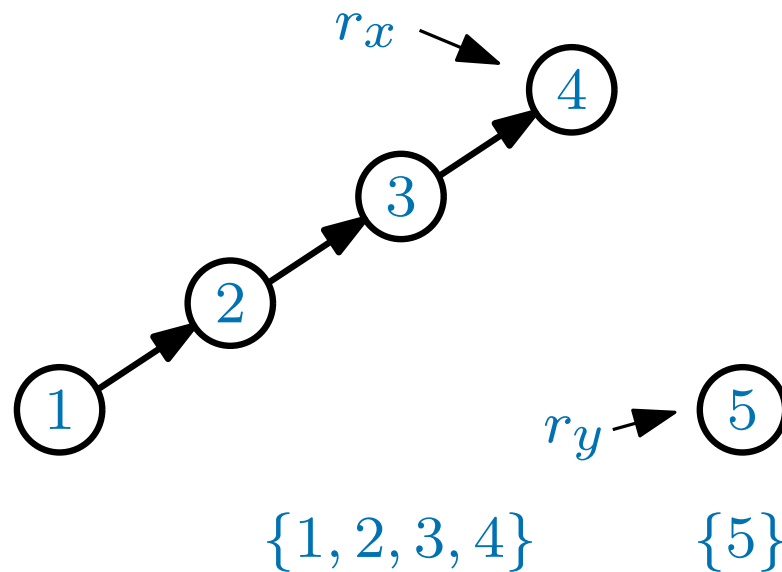
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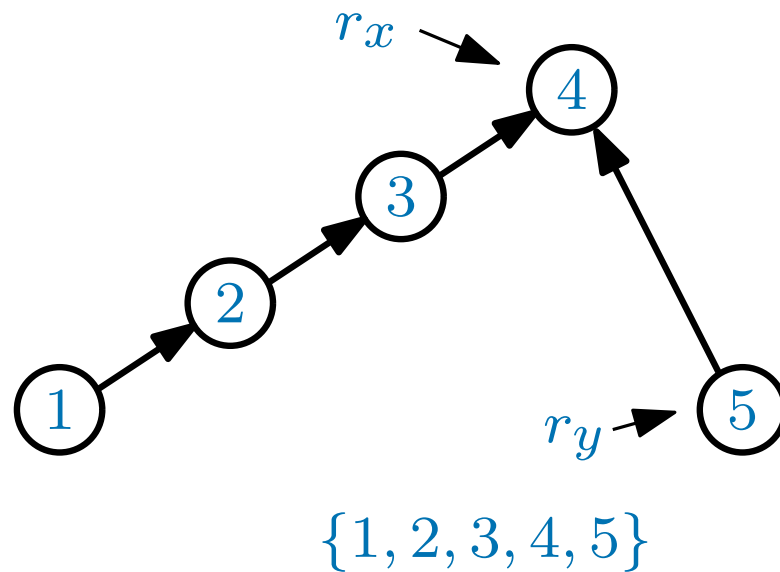
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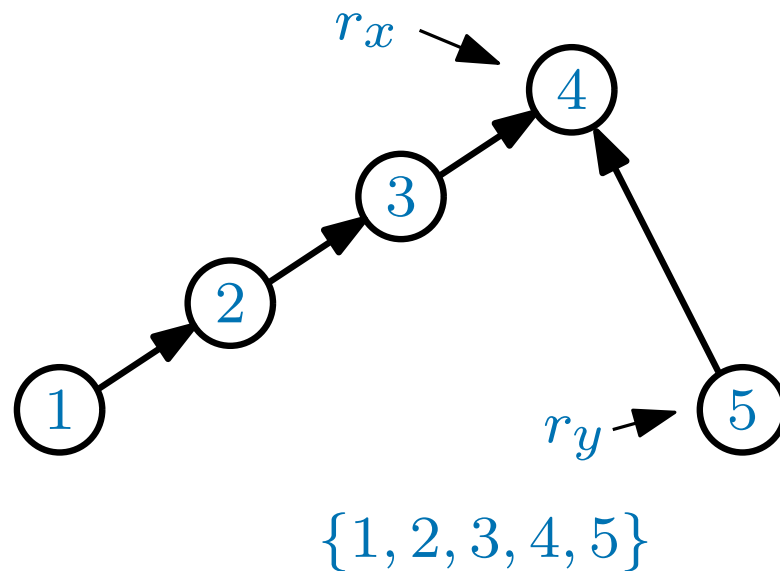
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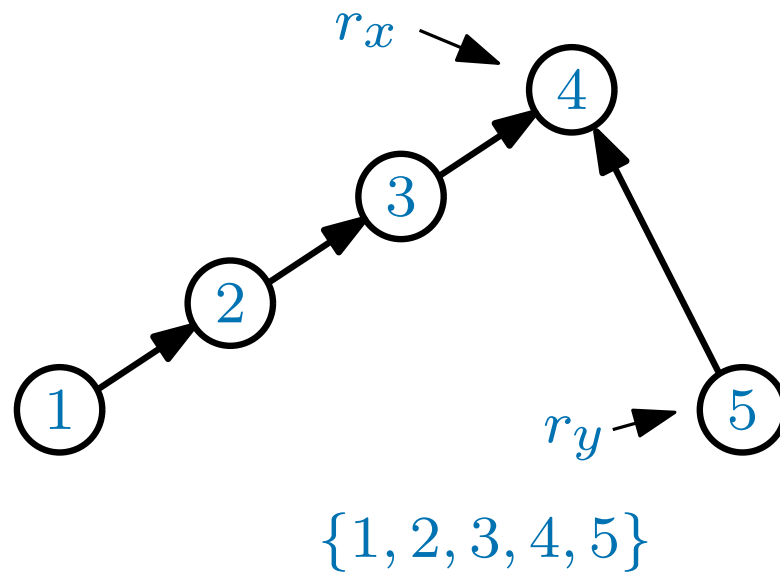
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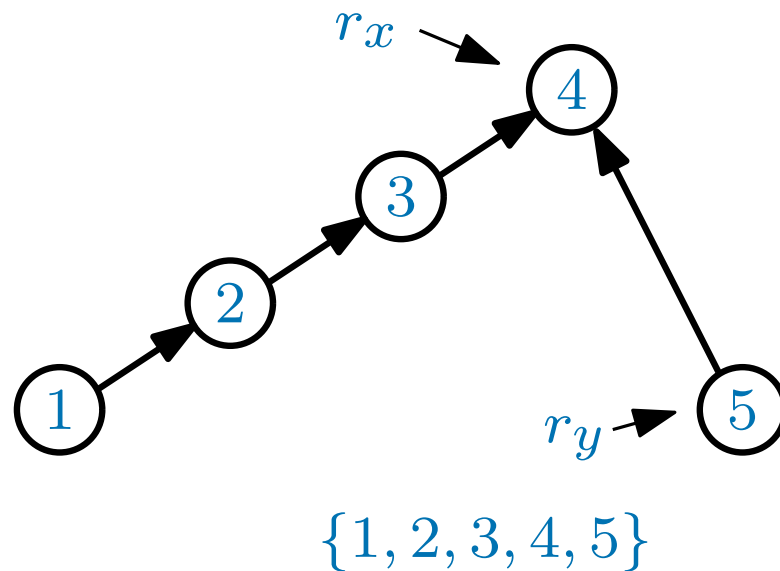
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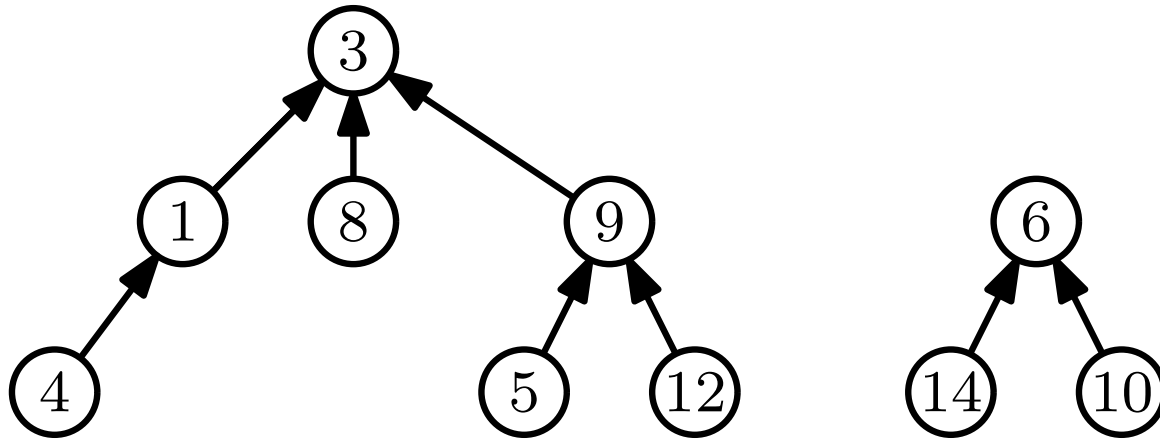
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Key Idea always make the shorter tree the child of the taller tree

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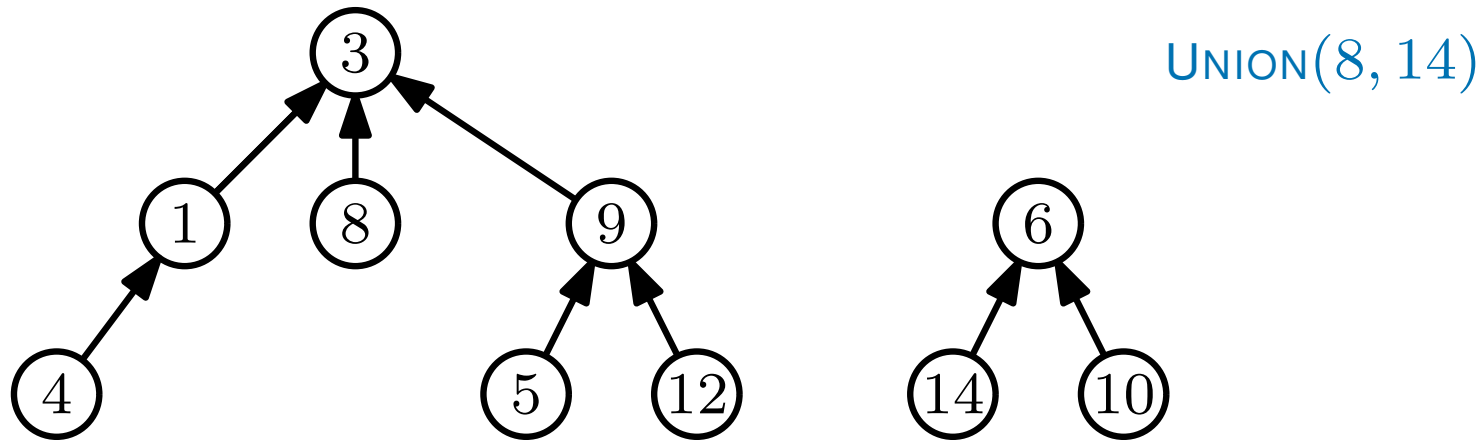
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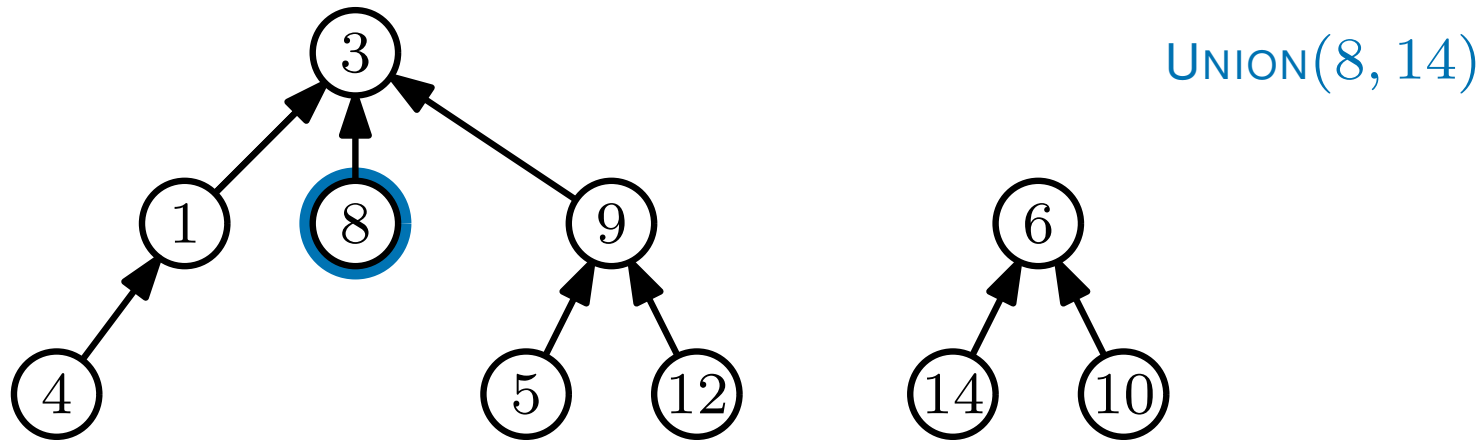
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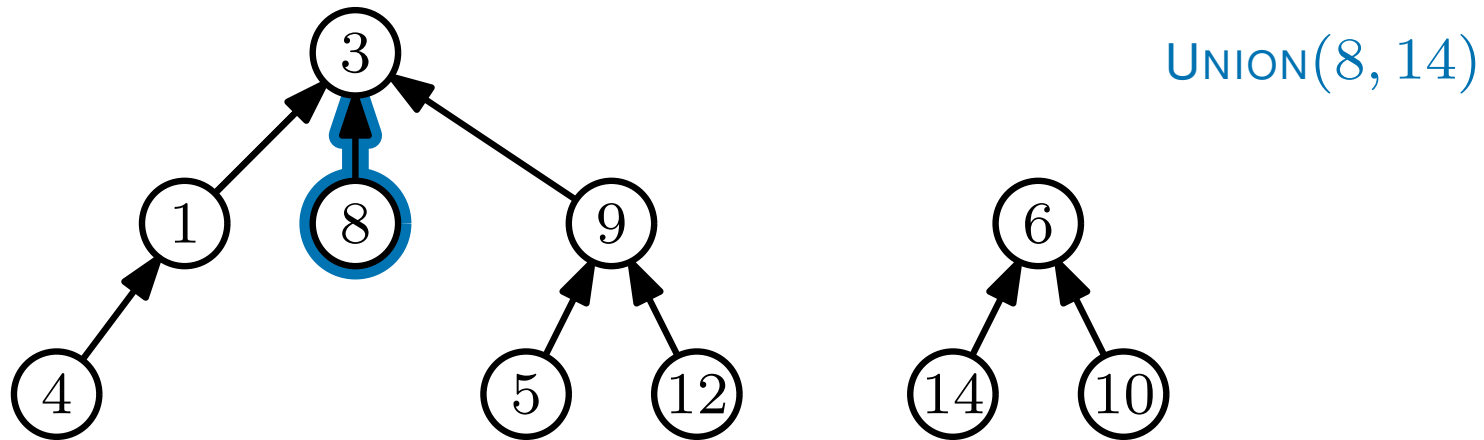
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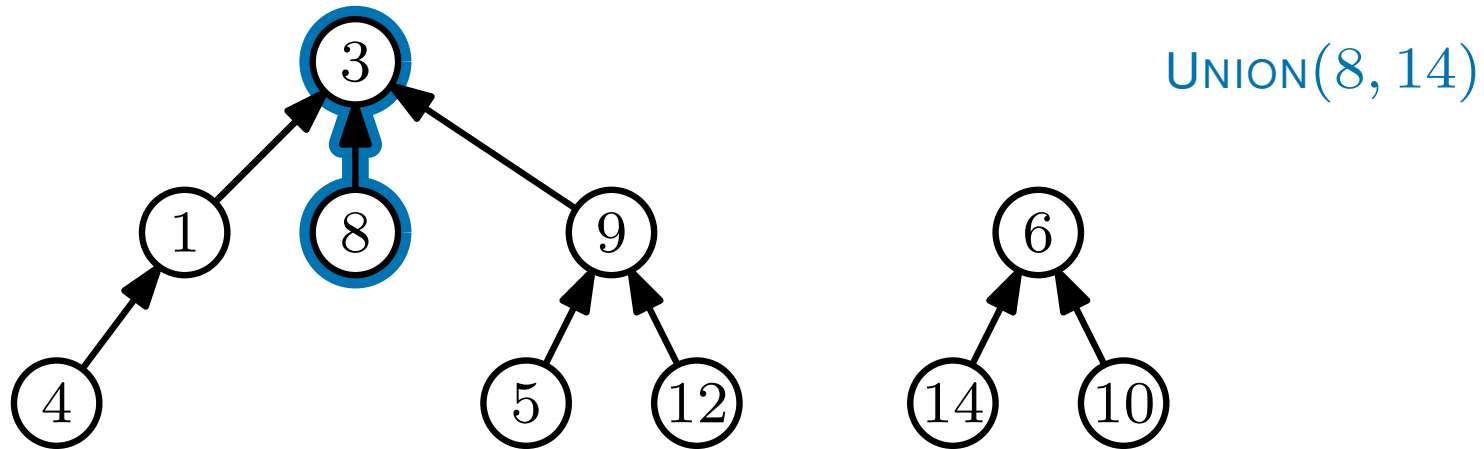
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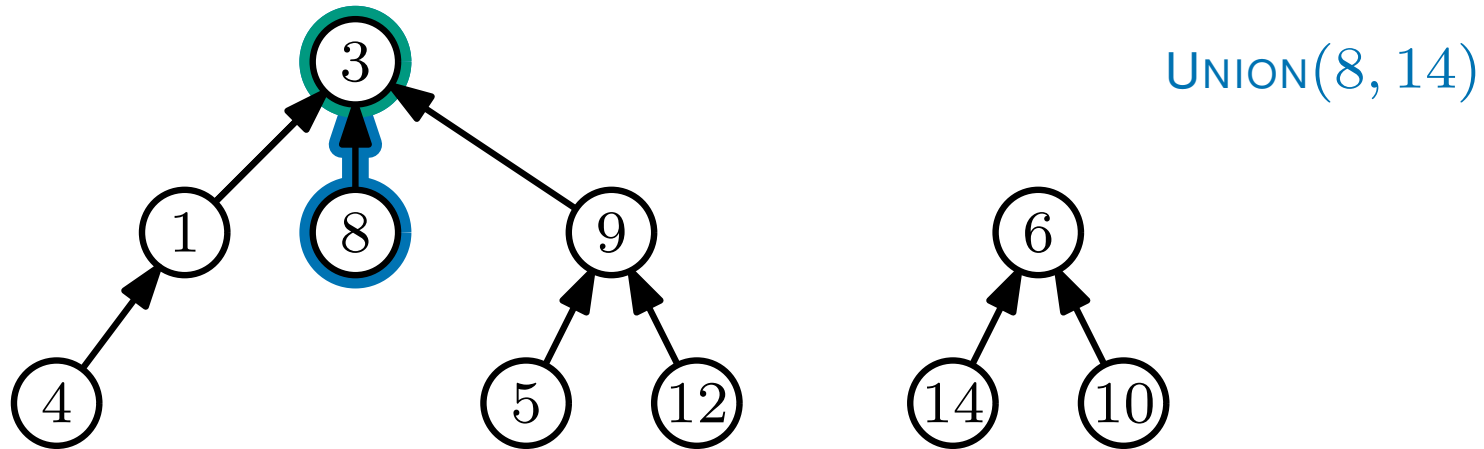
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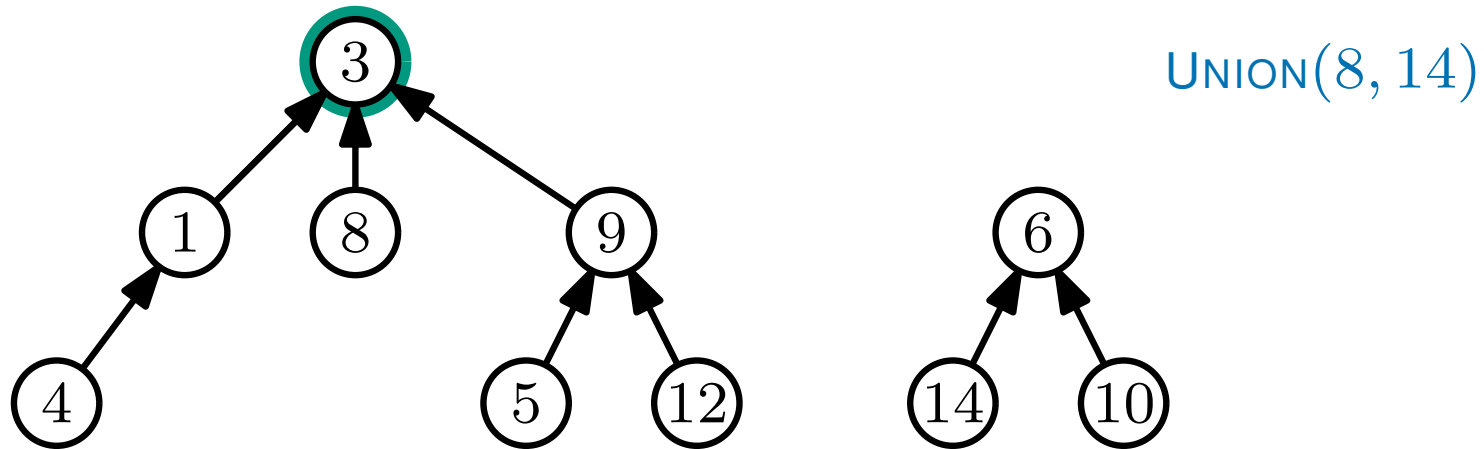
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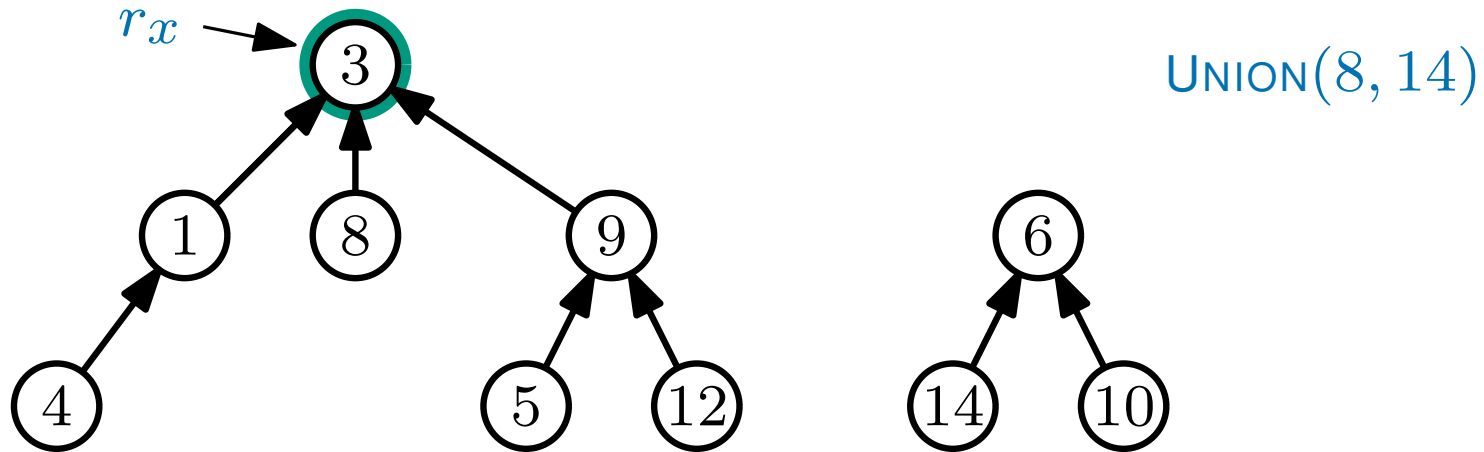
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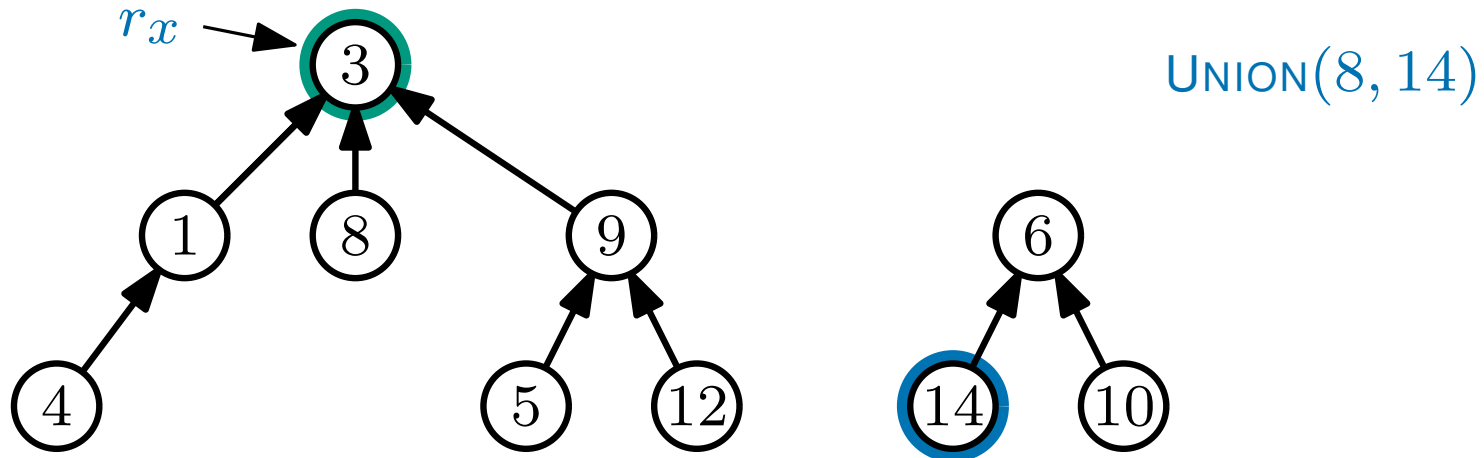
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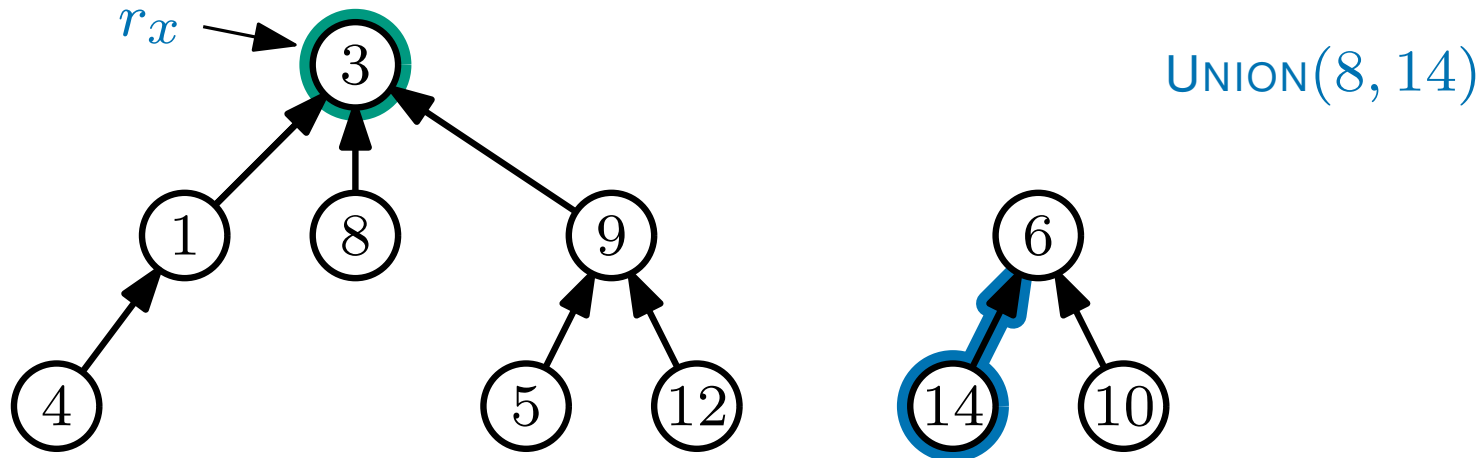
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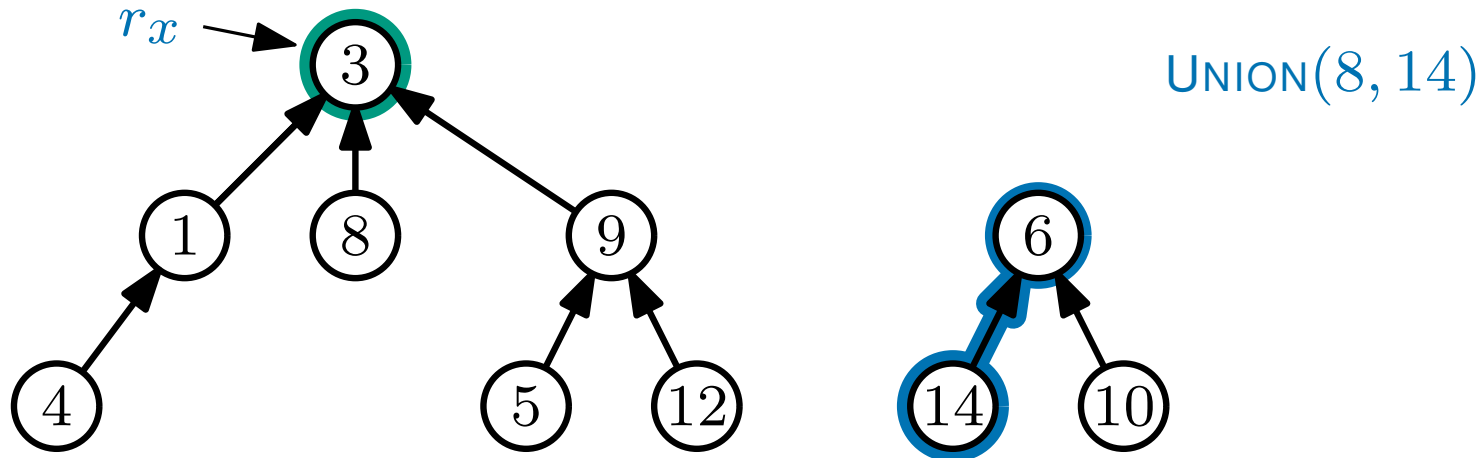
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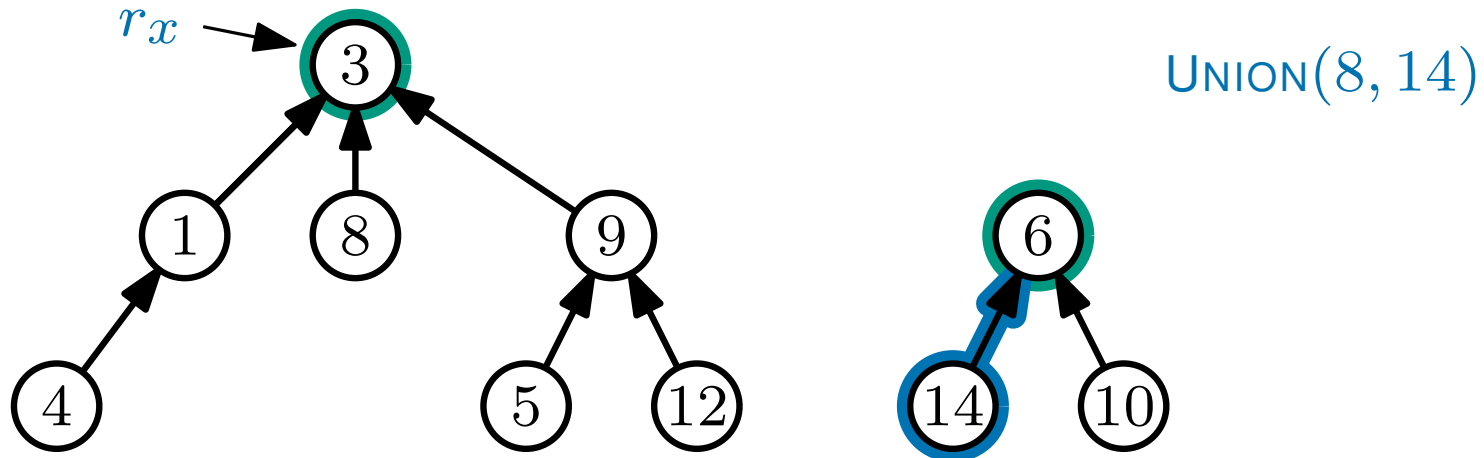
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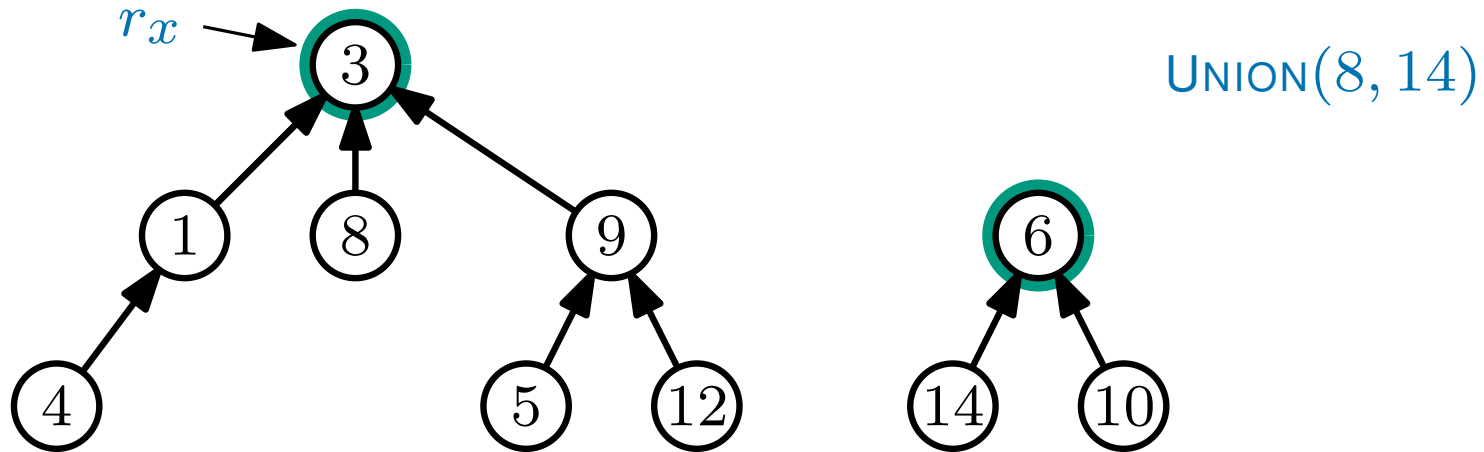
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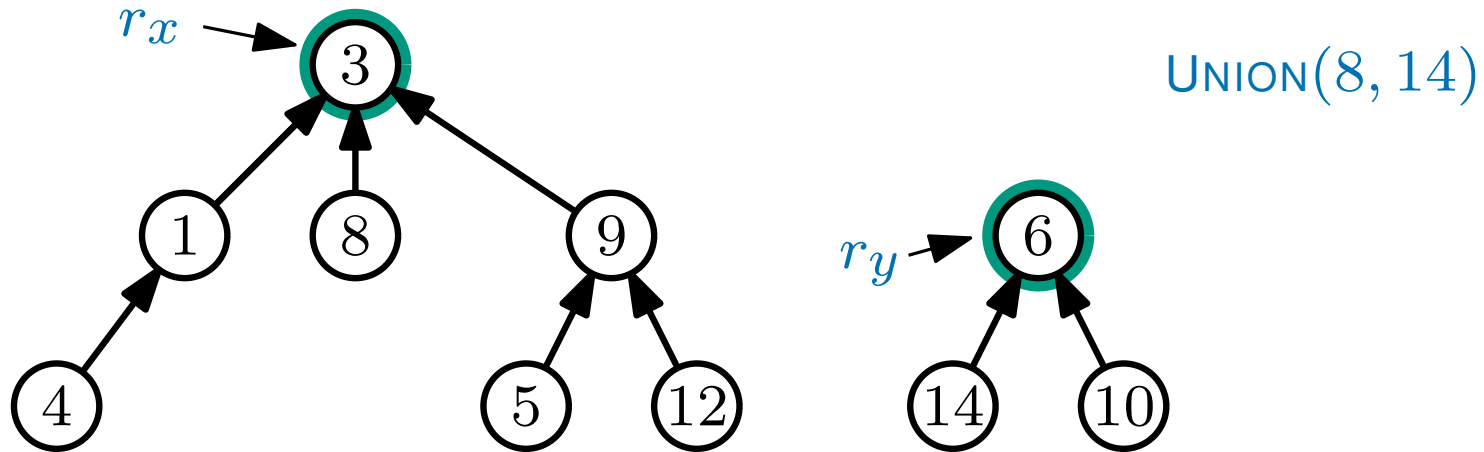
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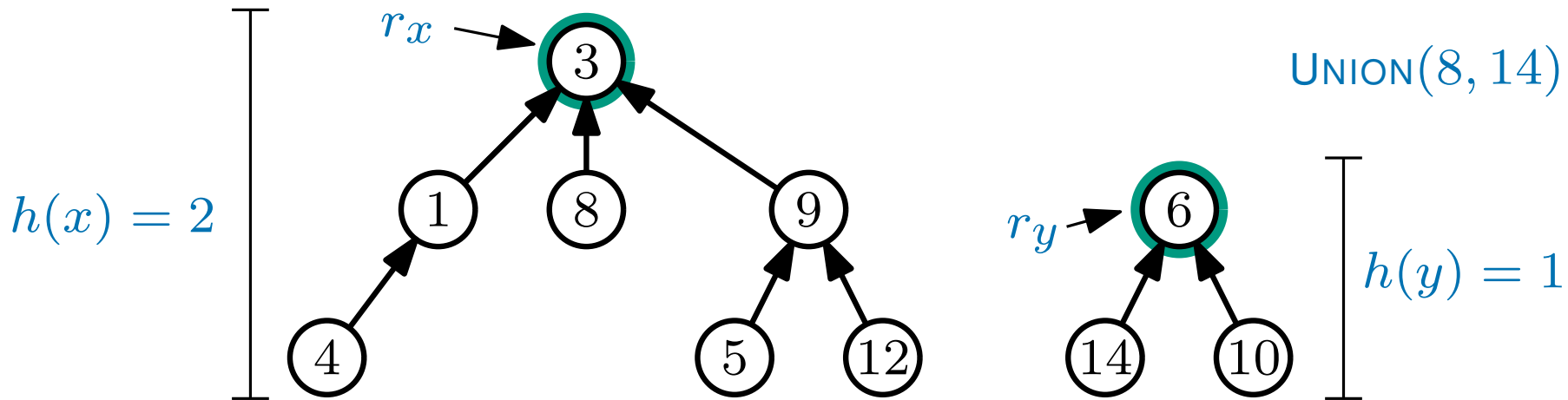
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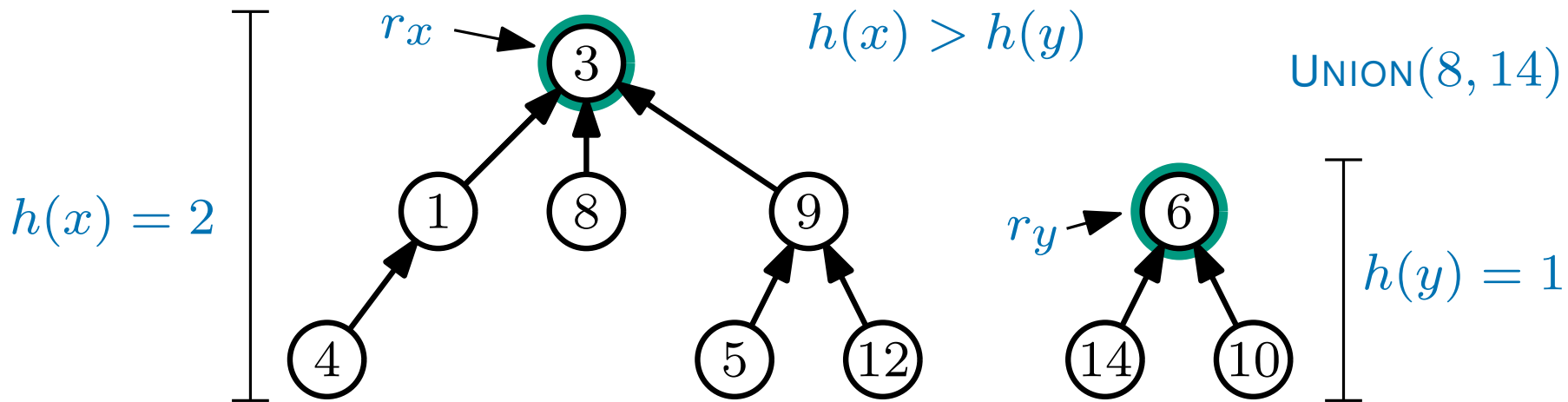
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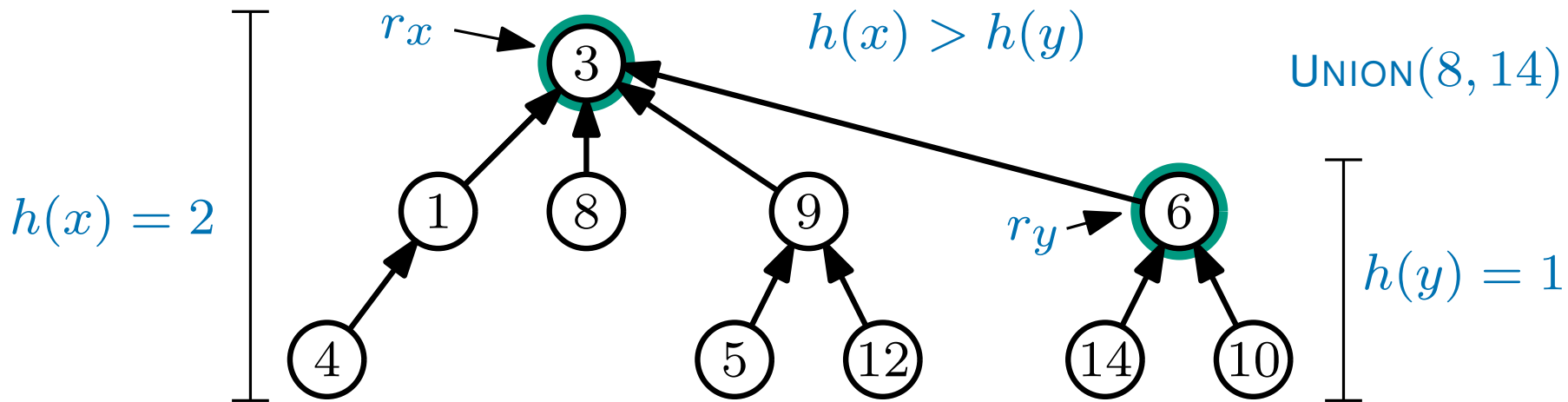
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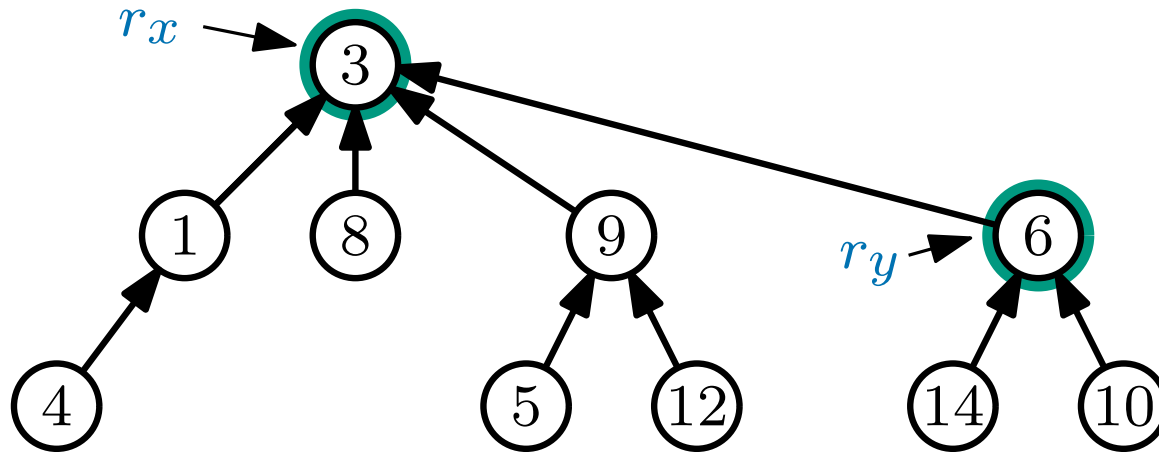
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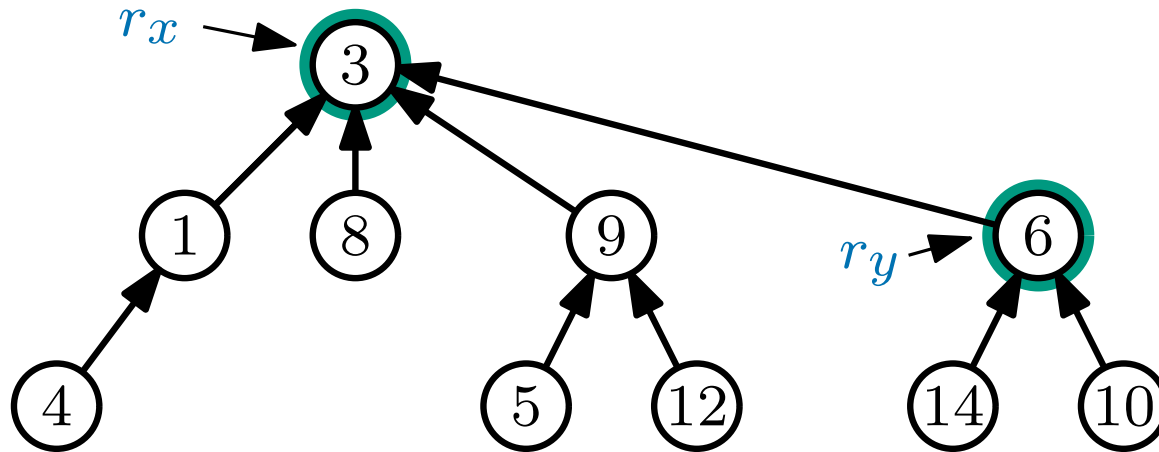
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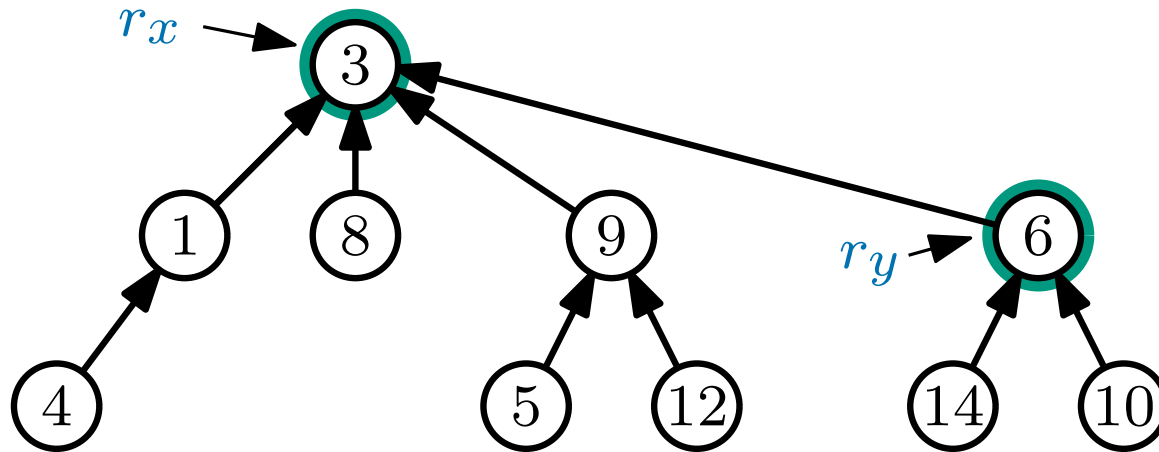
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This still takes $O(h)$ time ... *but the height only increases when $h(x) = h(y)$*

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Claim The height, h , of the tallest tree is $O(\log n)$

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We begin by proving that

any tree of height h (created by the data structure)

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Proof by induction on tree height i ,

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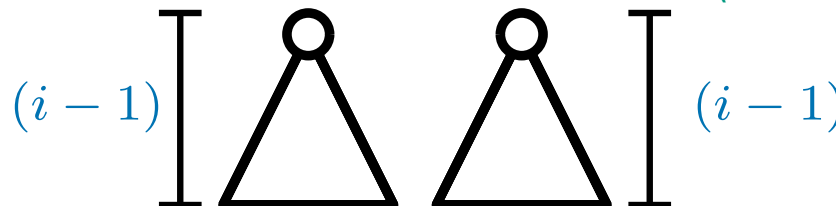
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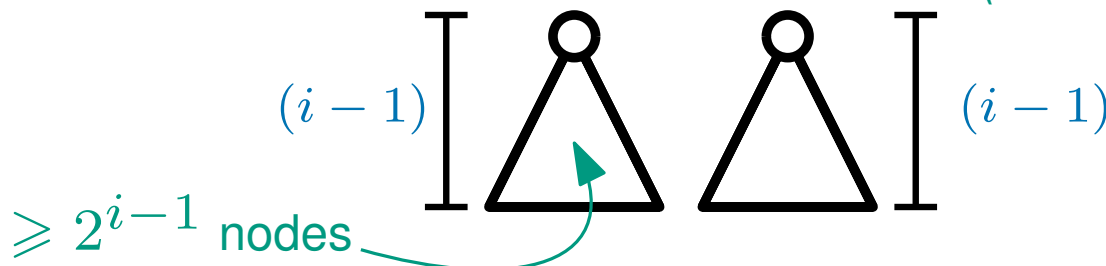
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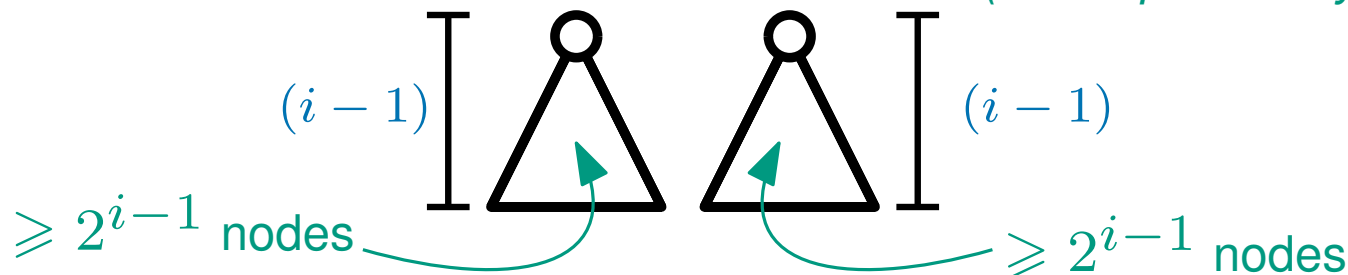
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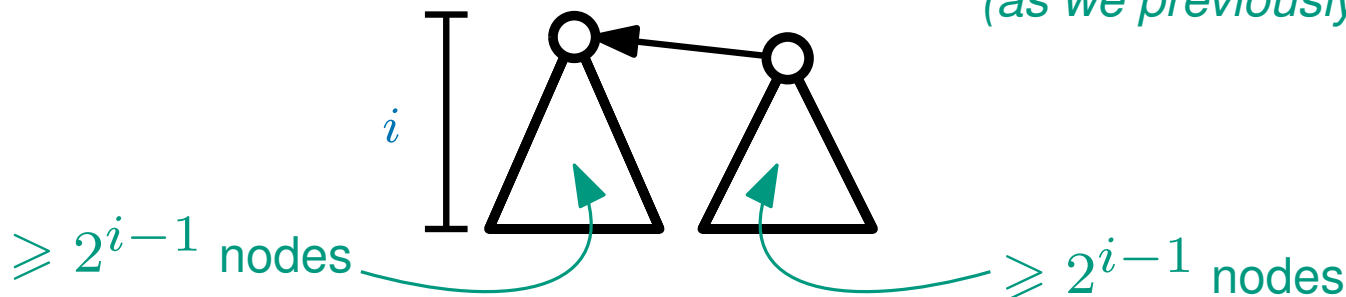
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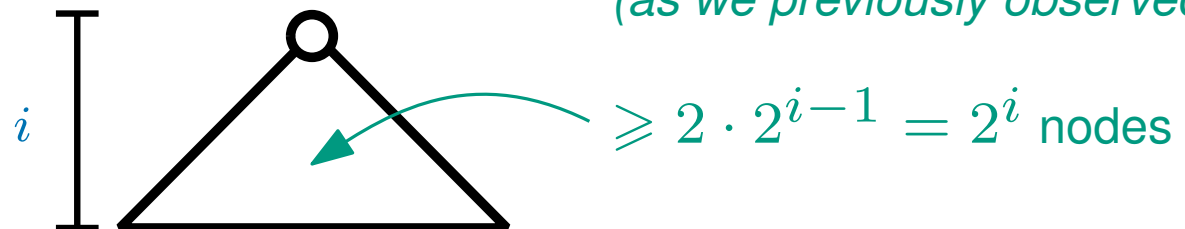
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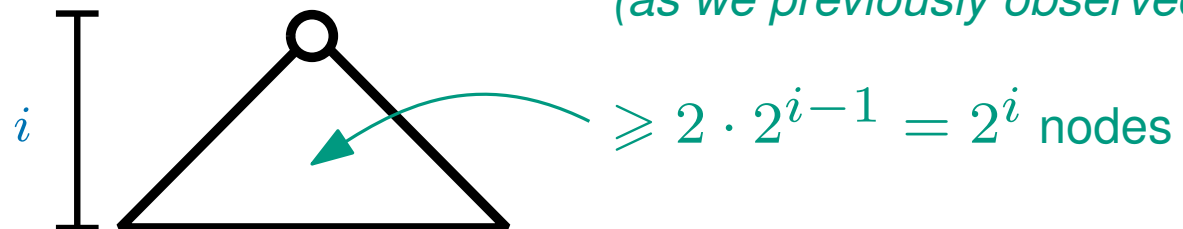
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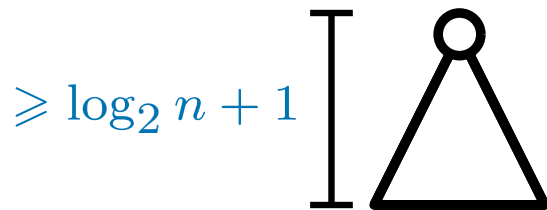
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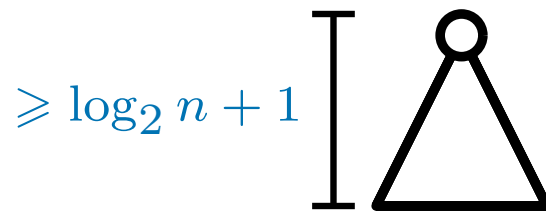
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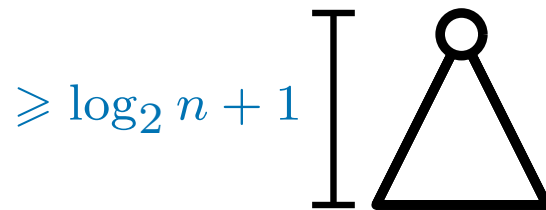
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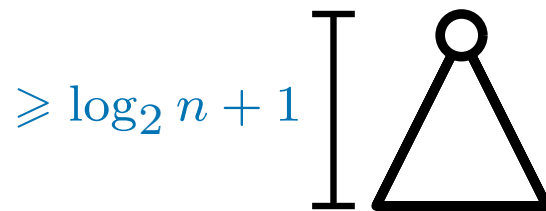
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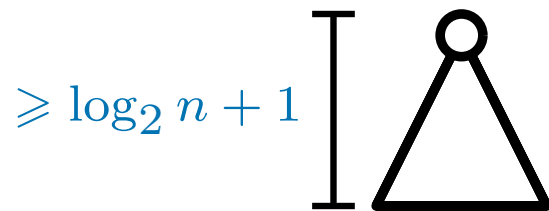
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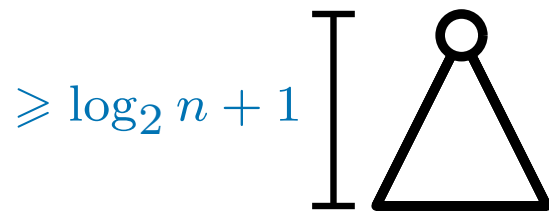
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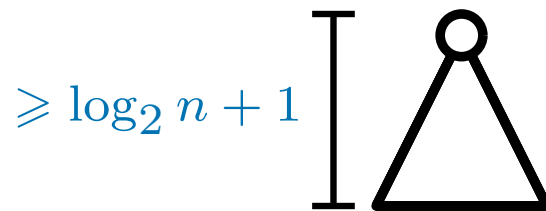
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As, $h \leq \log_2 n$ they both only take $O(\log n)$ time.

Disjoint Set Summary

We have seen a data structure which
stores a collection of disjoint sets

The elements of the sets are numbers from $\{1, 2, \dots, n\}$

The following operations are supported:

MAKESET(x) - make a new set containing only x
 x cannot be a member of any existing set

UNION(x, y) - merge the sets containing x and y into a single set

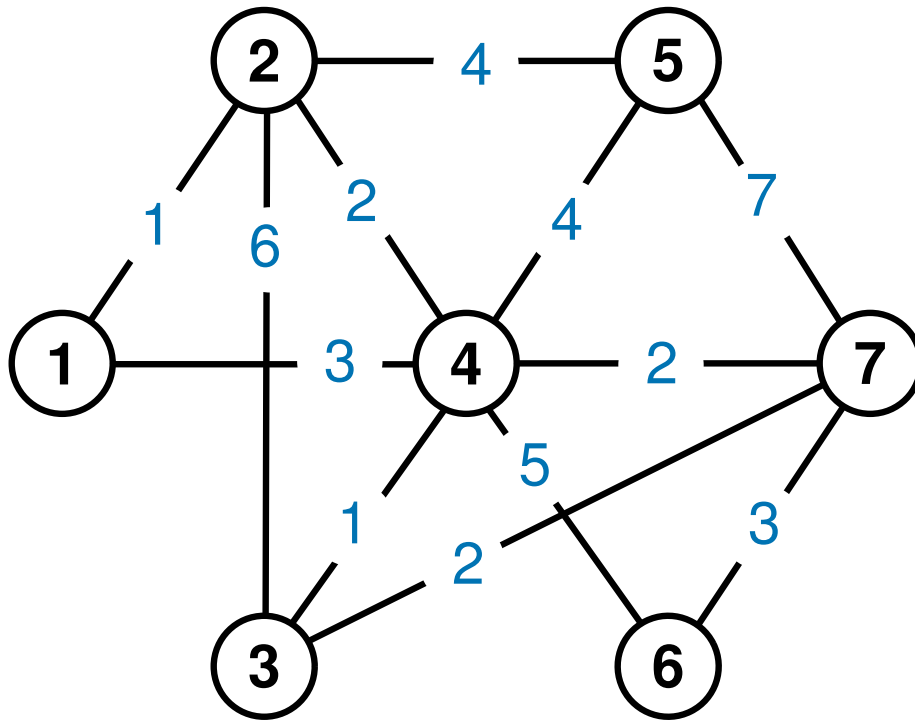
FINDSET(x) - returns the *identity* of the set containing x
the identity of a set is any unique identifier of the set.

The operations **UNION** and **FINDSET** take $O(\log n)$ time.

The operation **MAKESET** runs in $O(1)$ time.

Minimum Spanning Trees

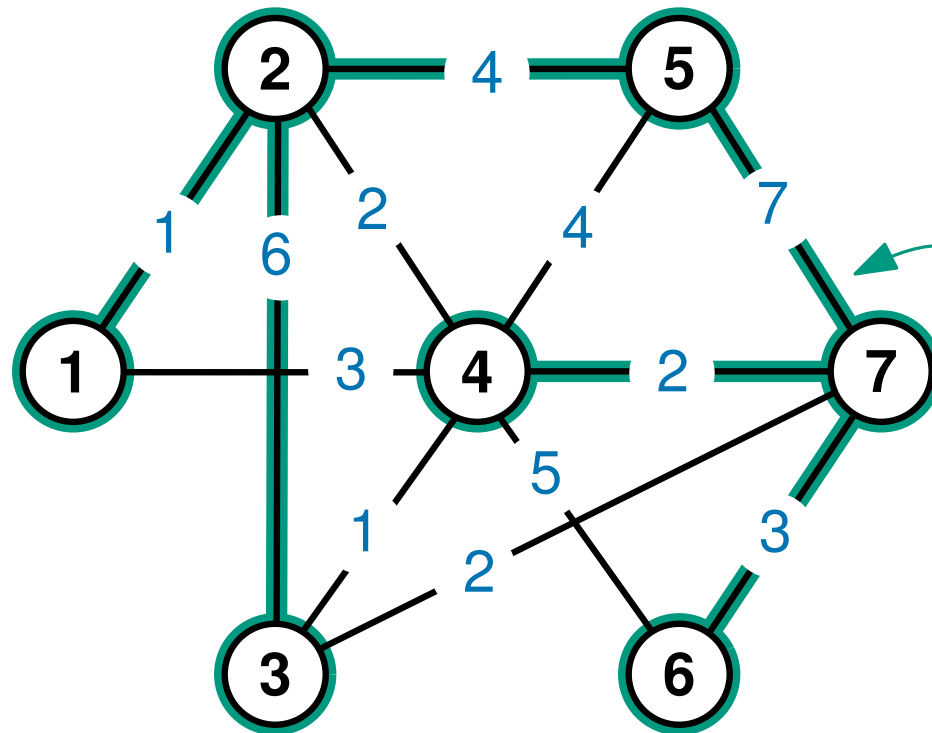
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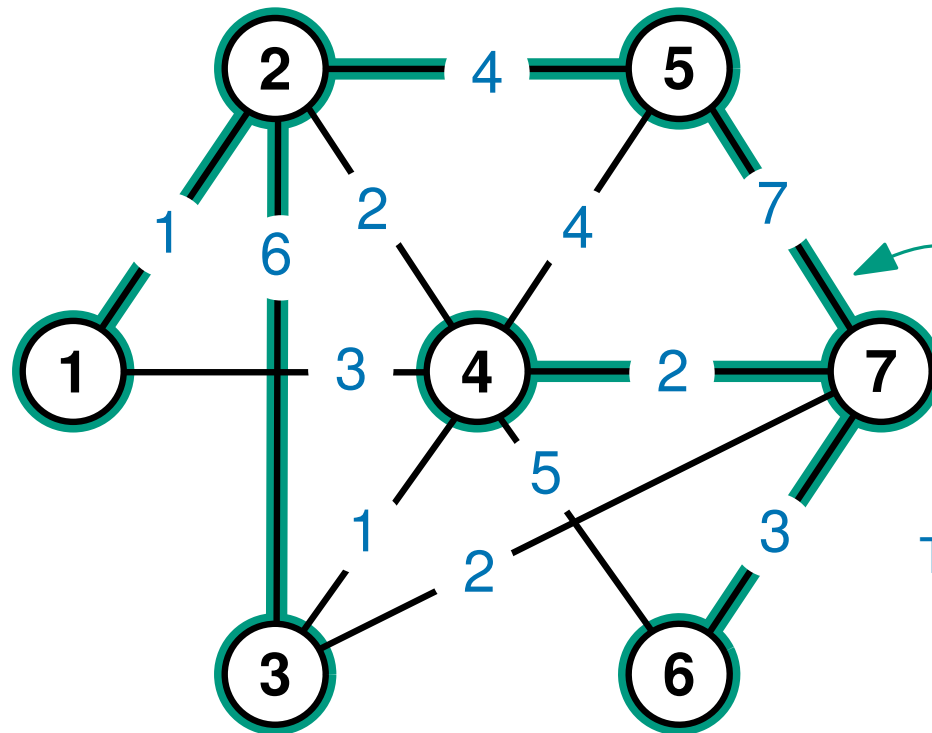


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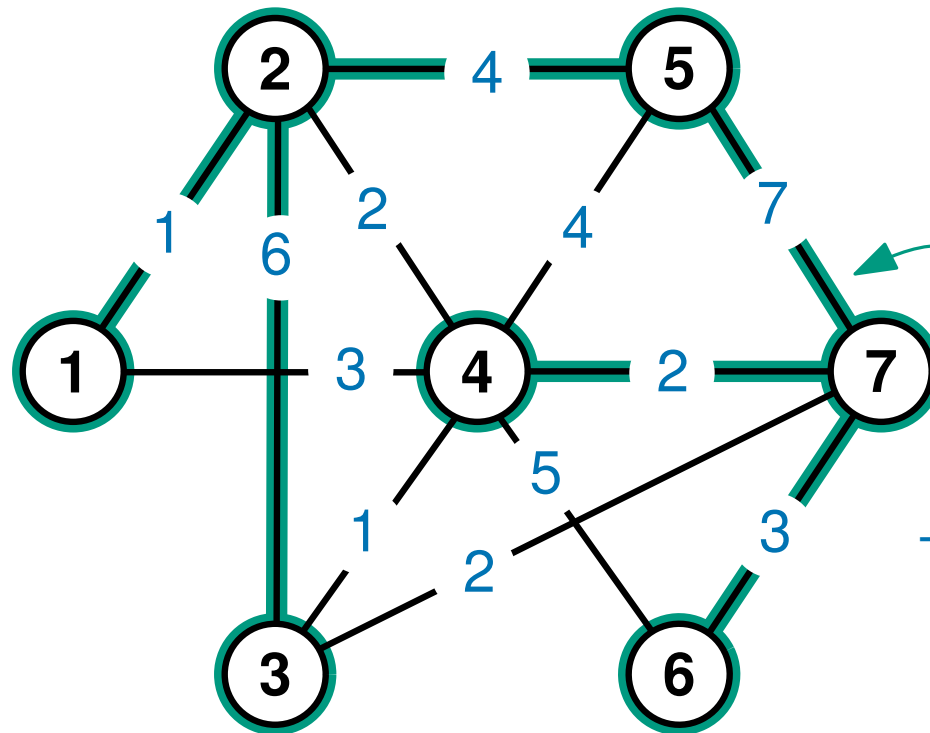
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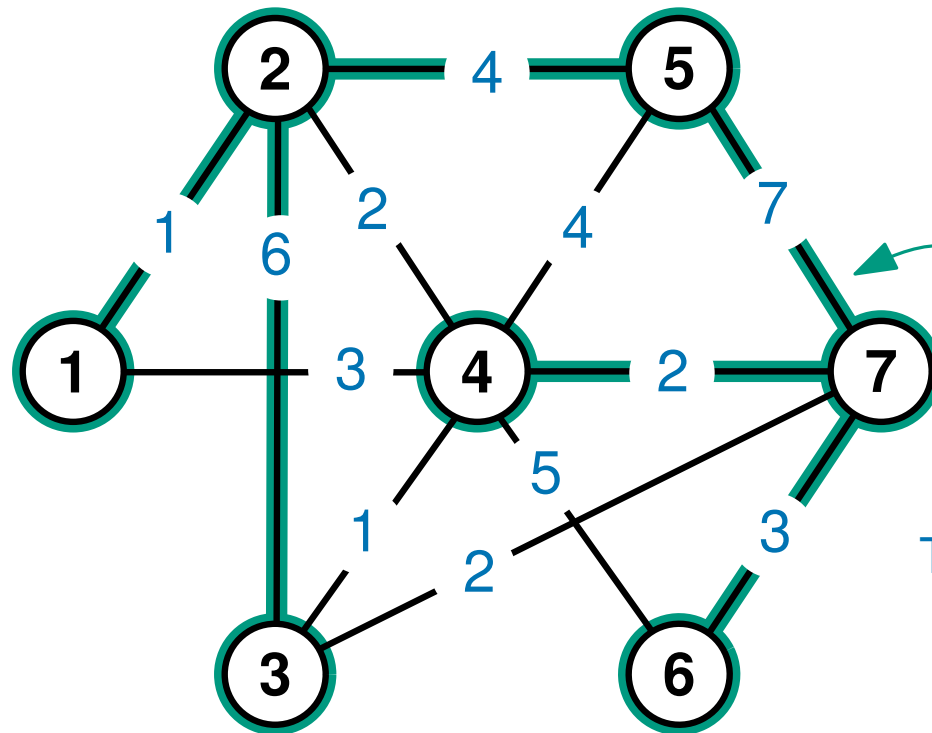
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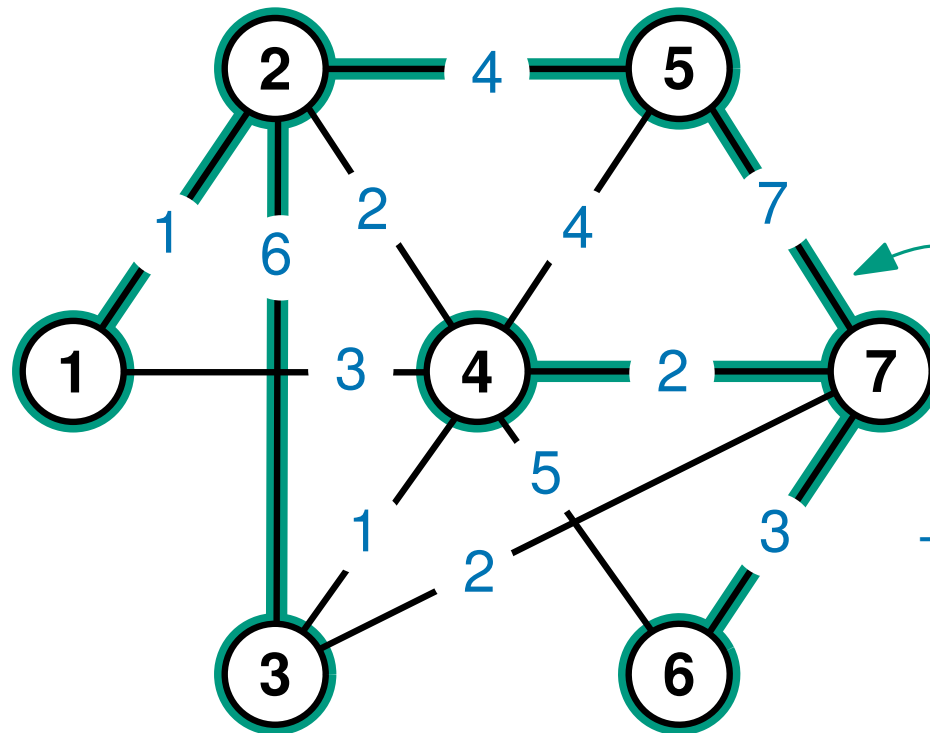
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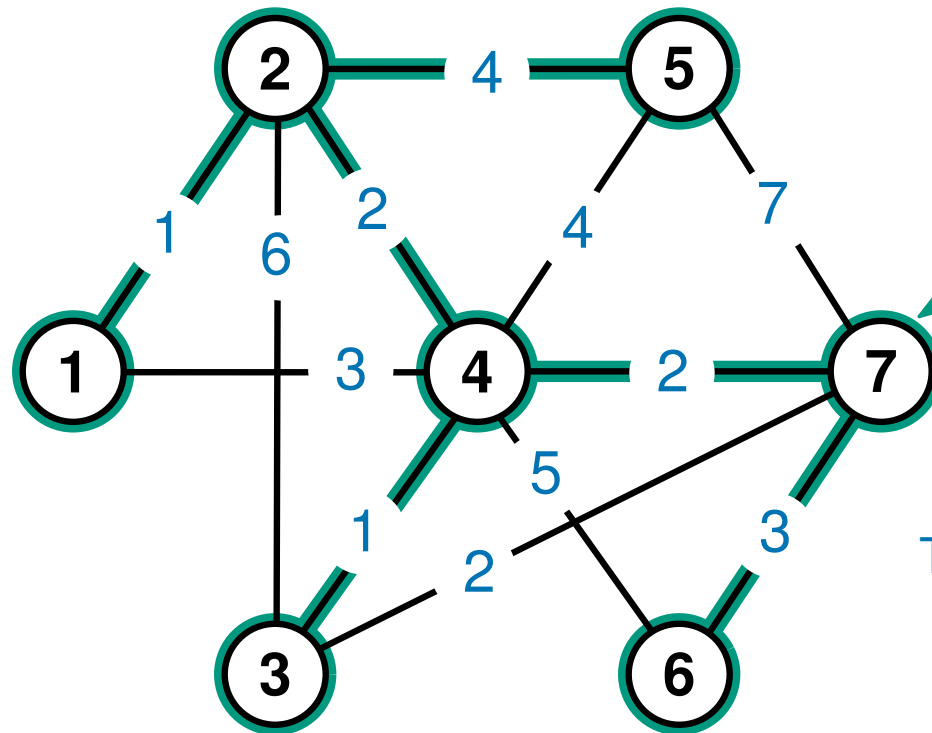
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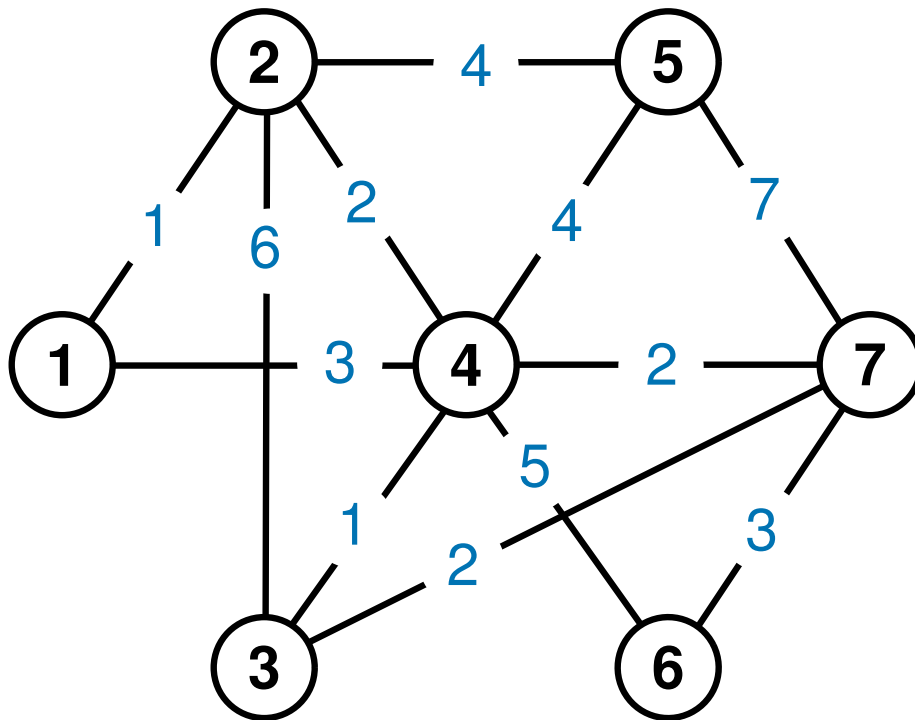
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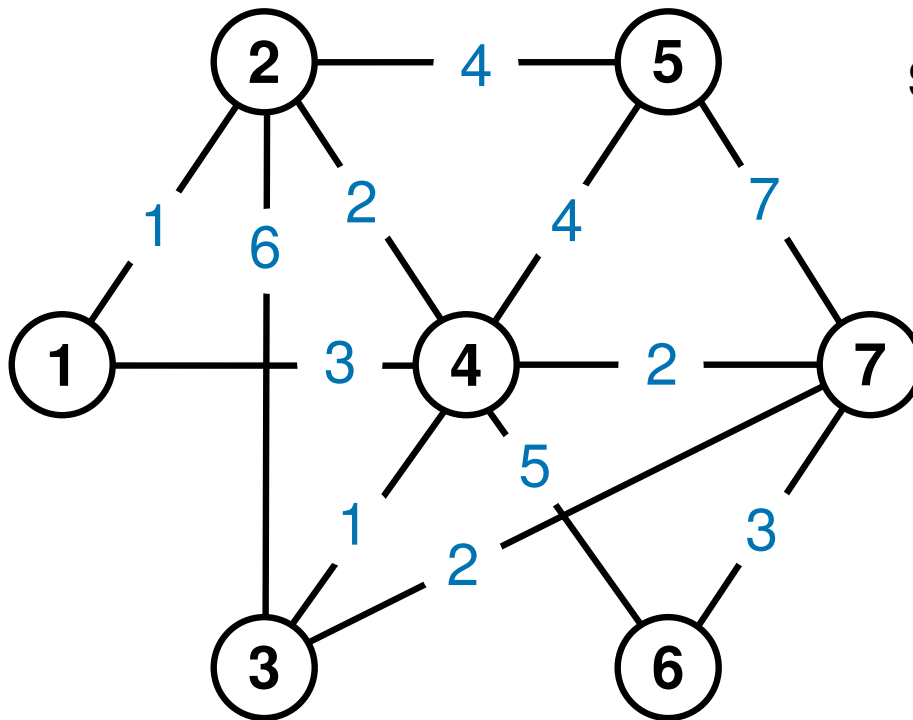
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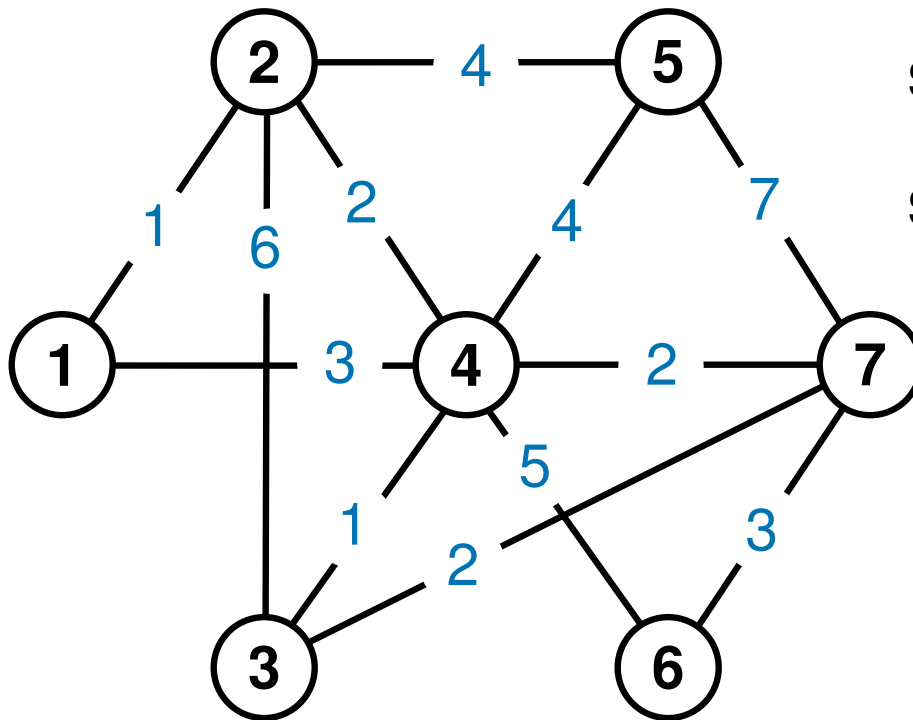
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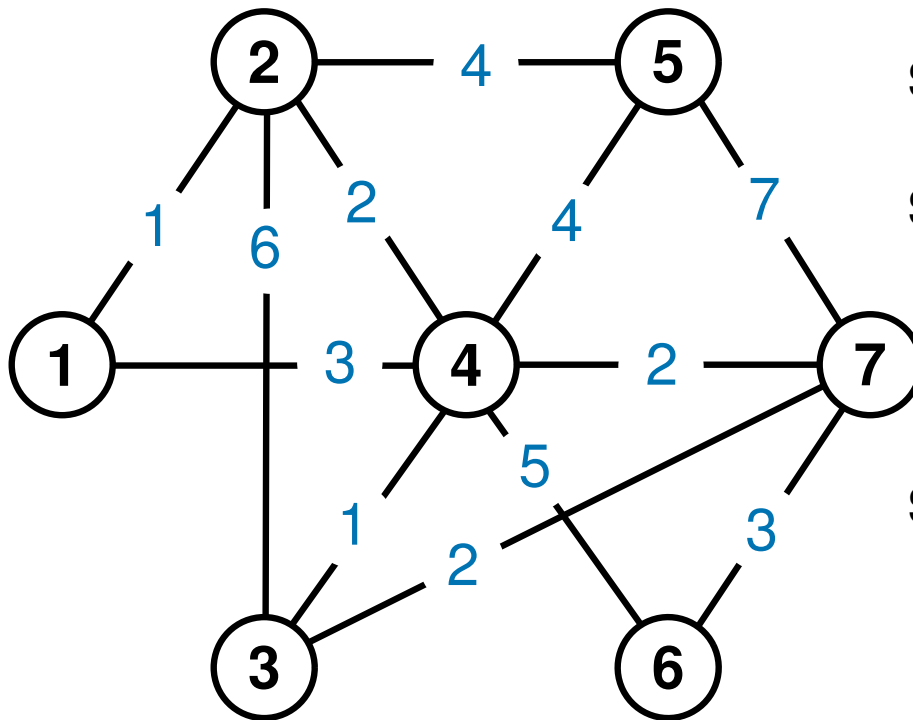


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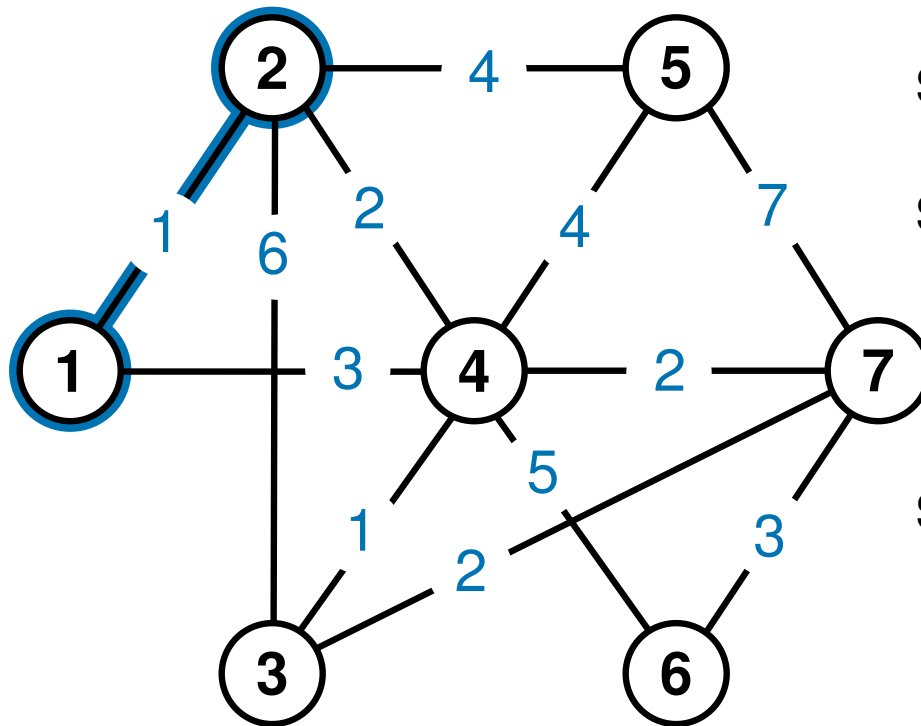
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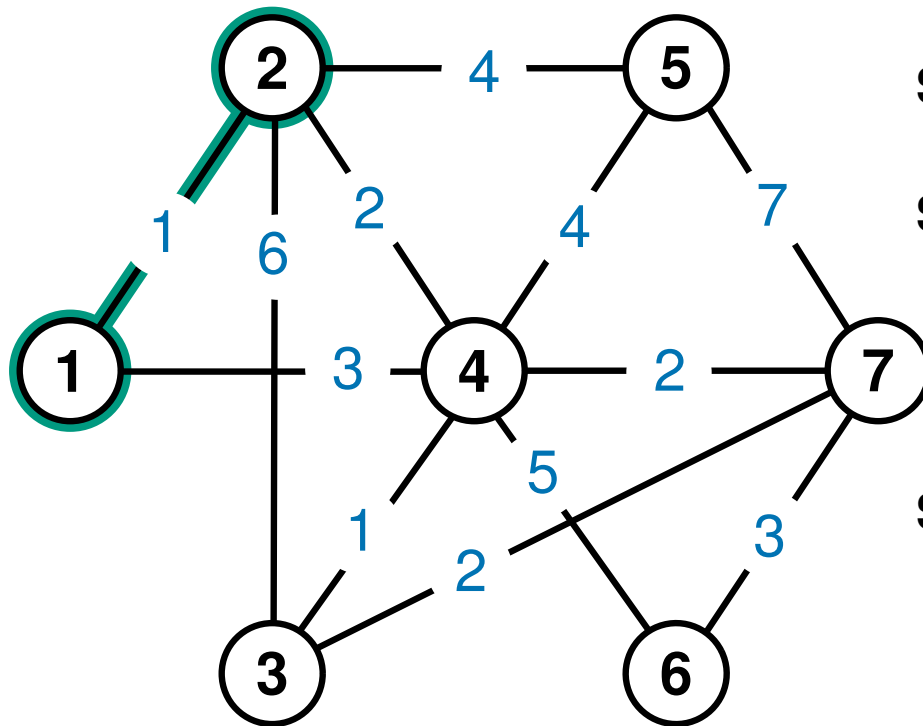
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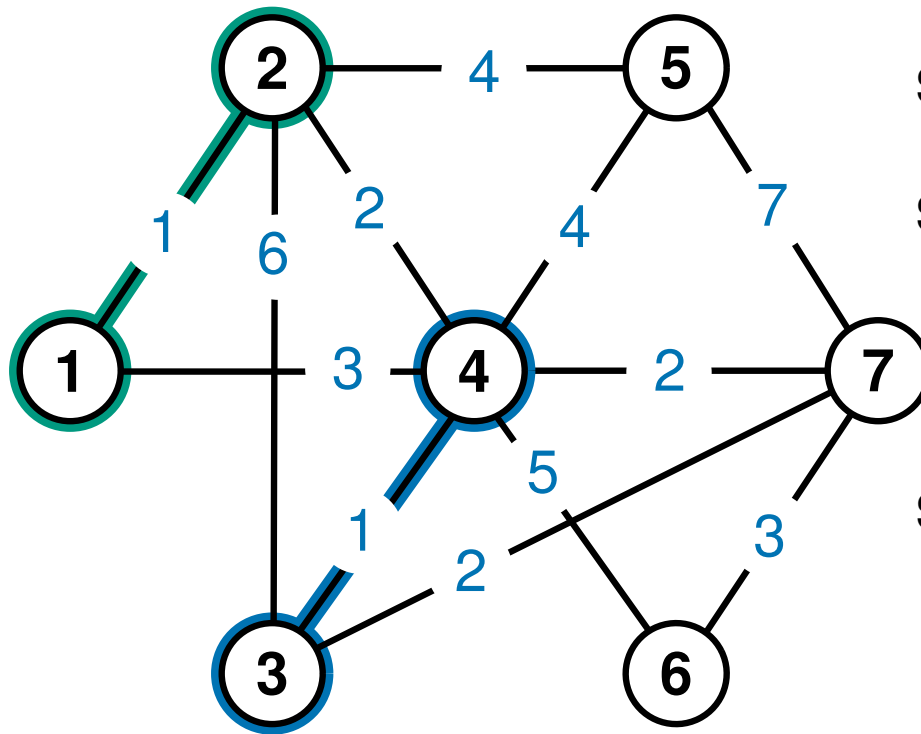
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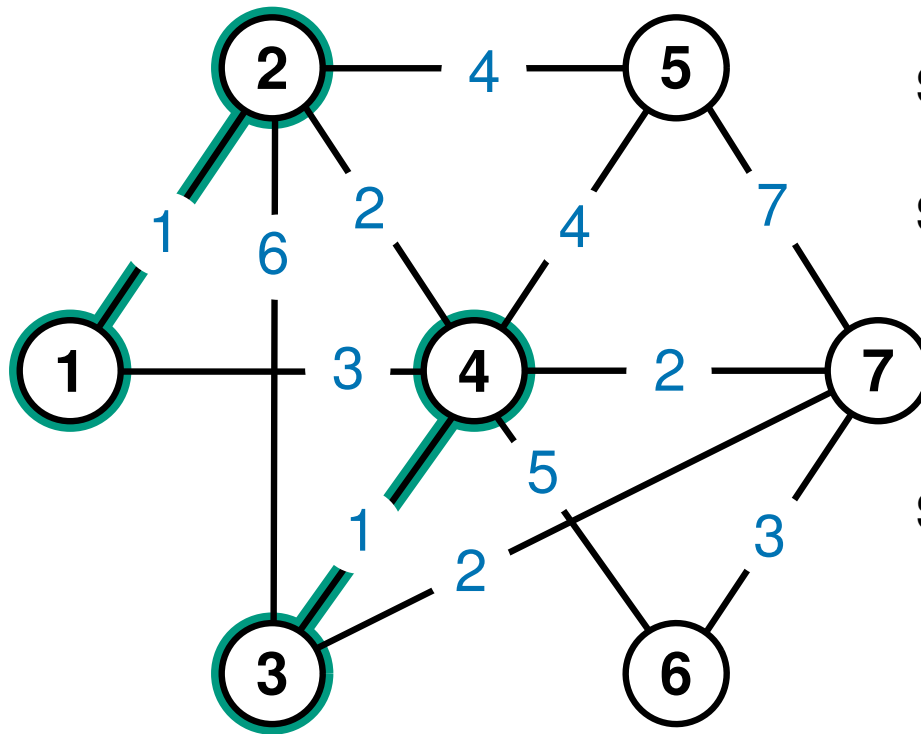
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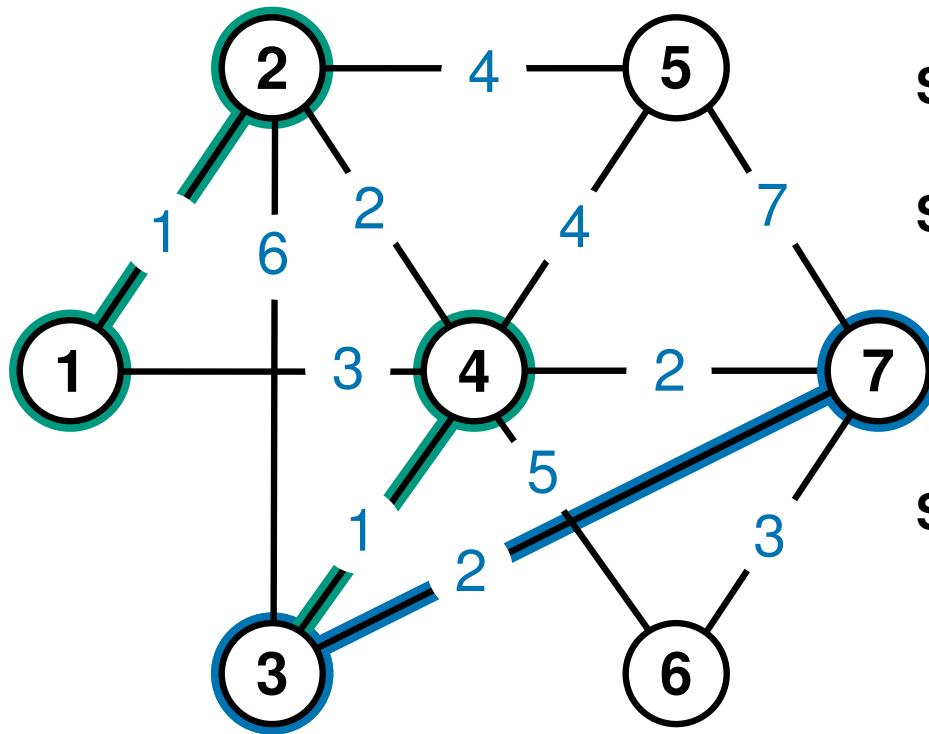
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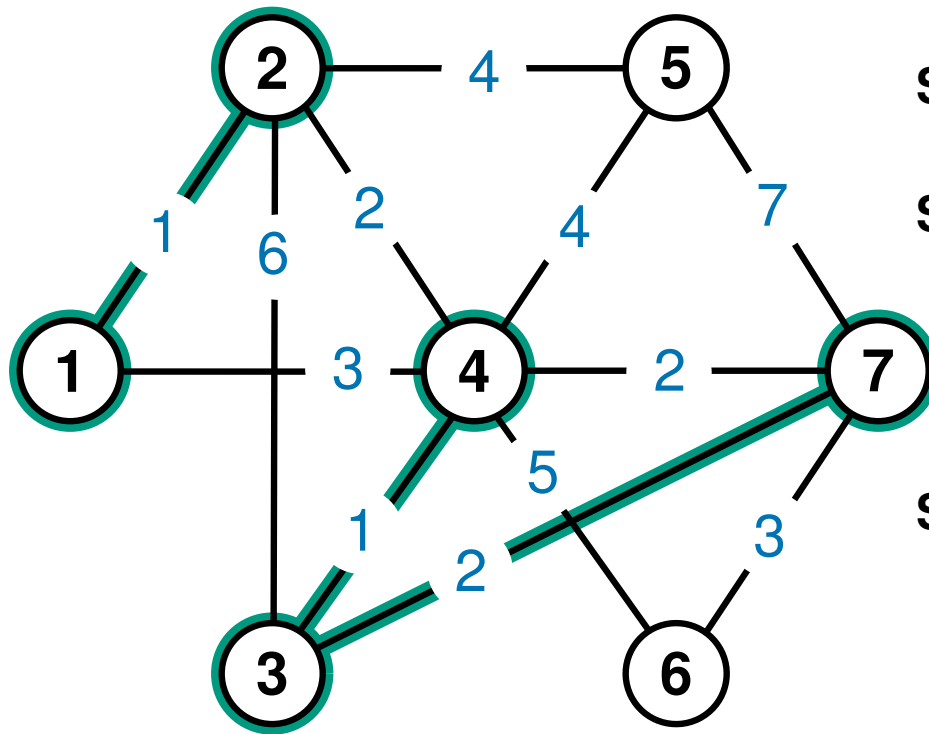
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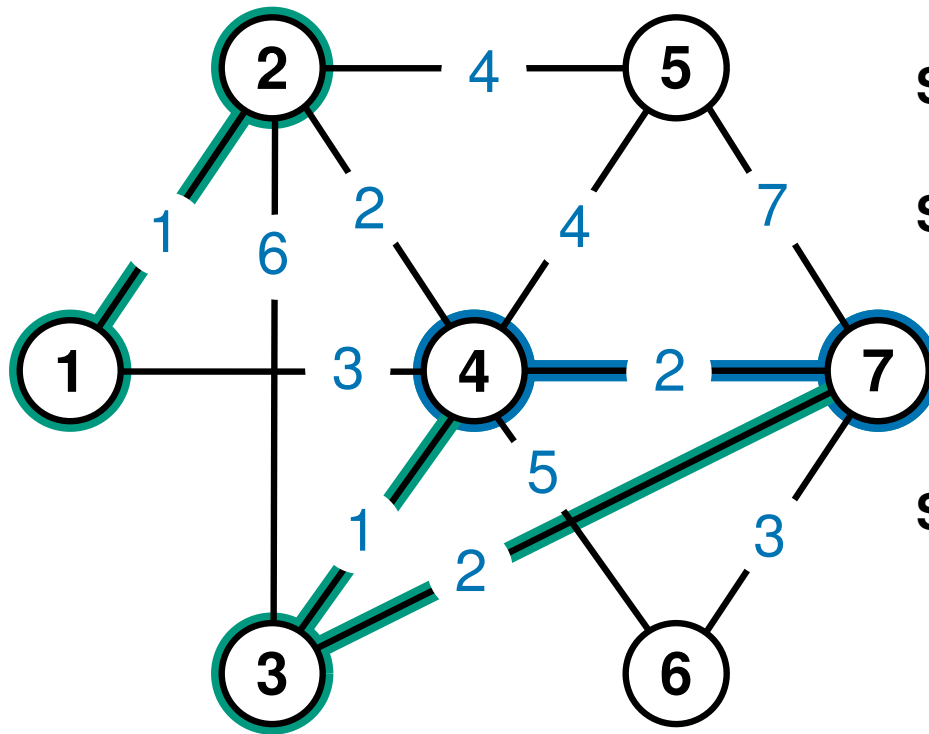
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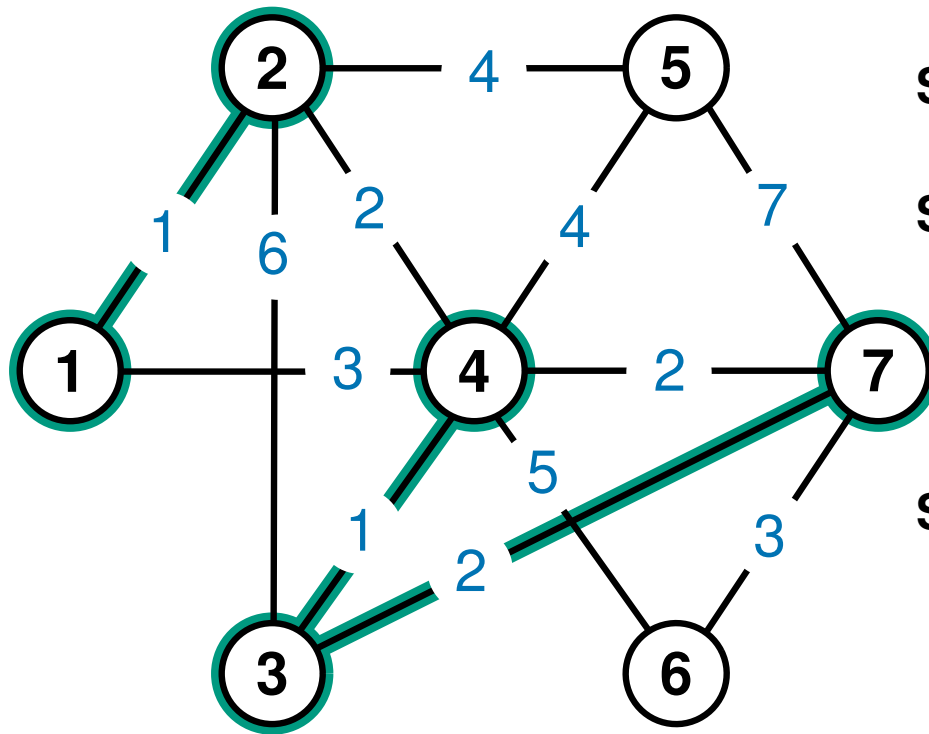
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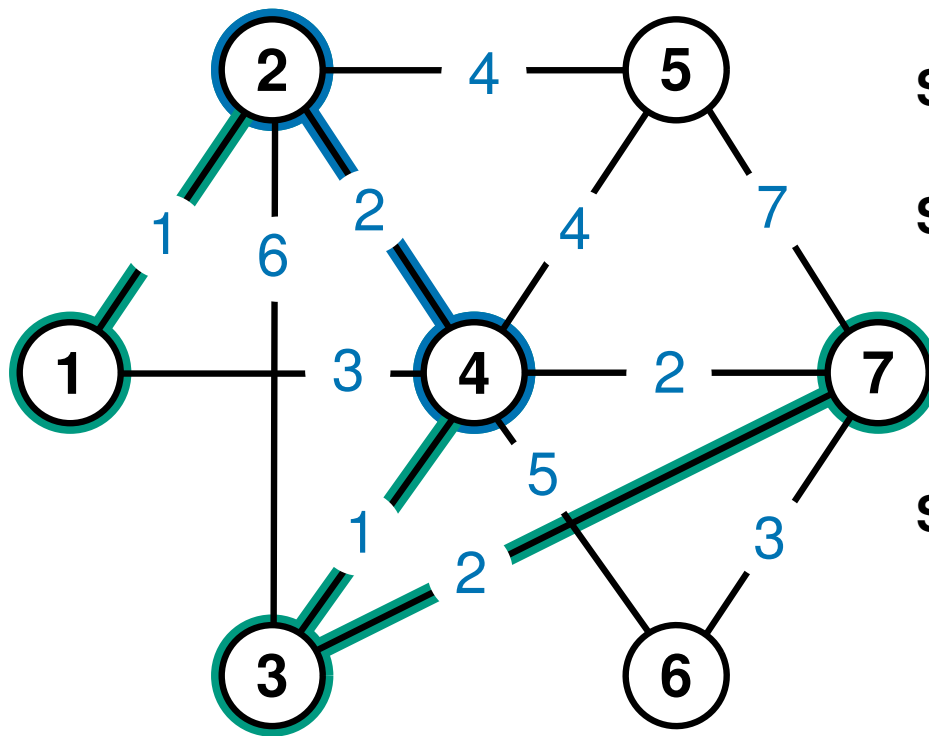
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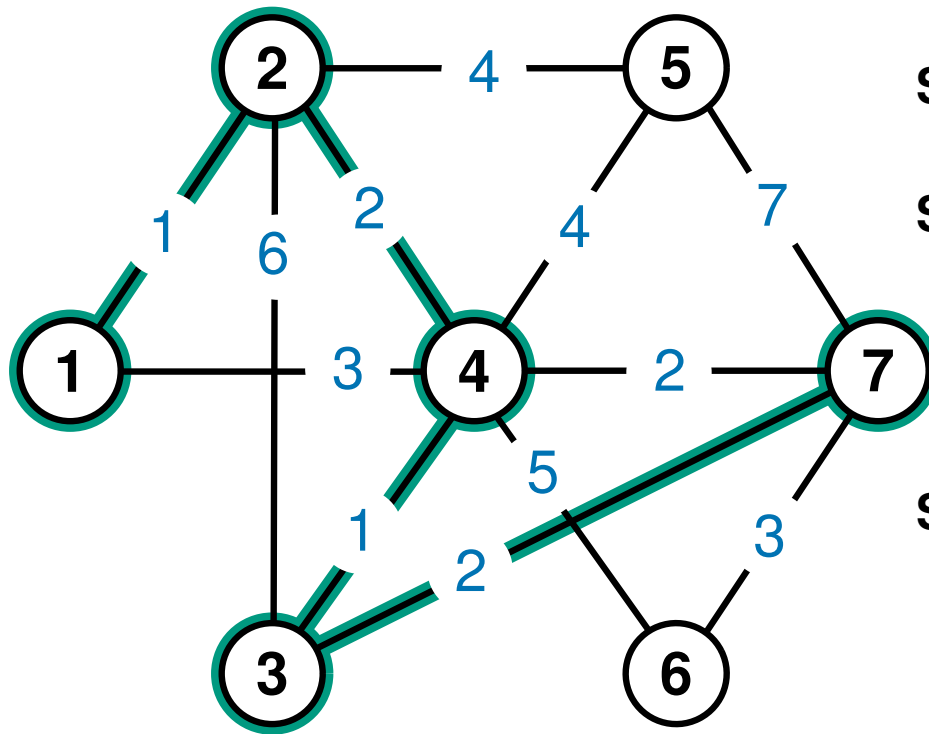
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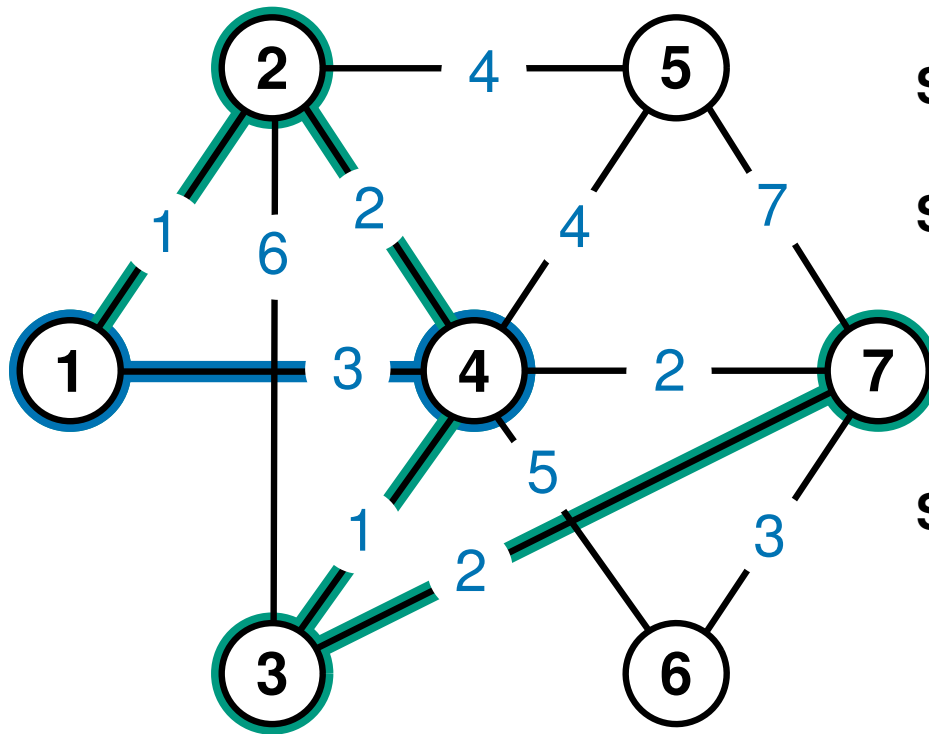
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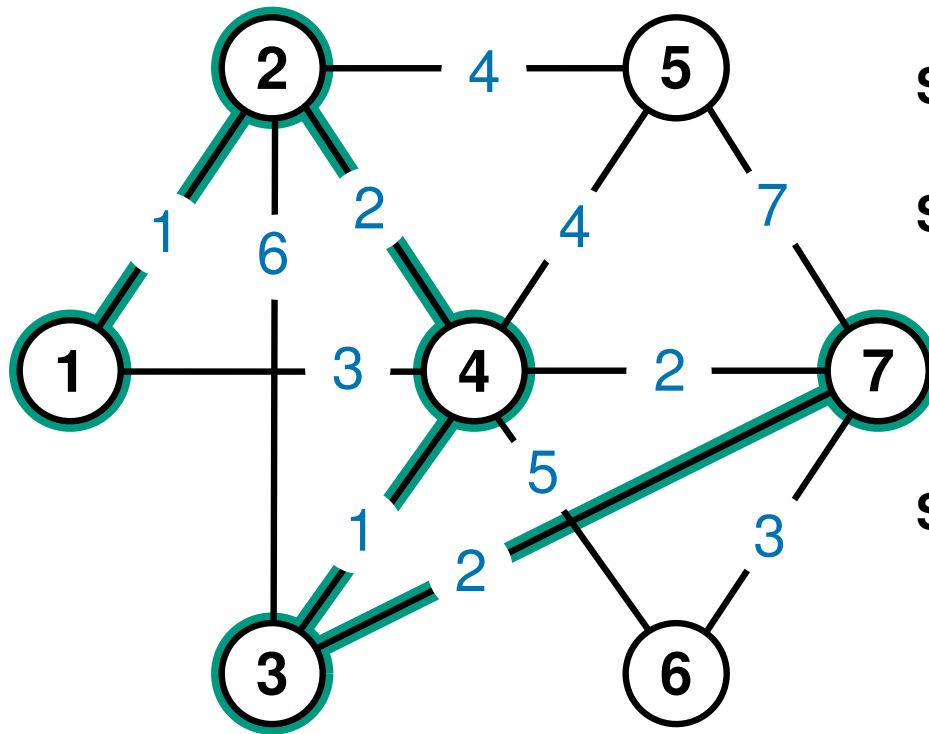
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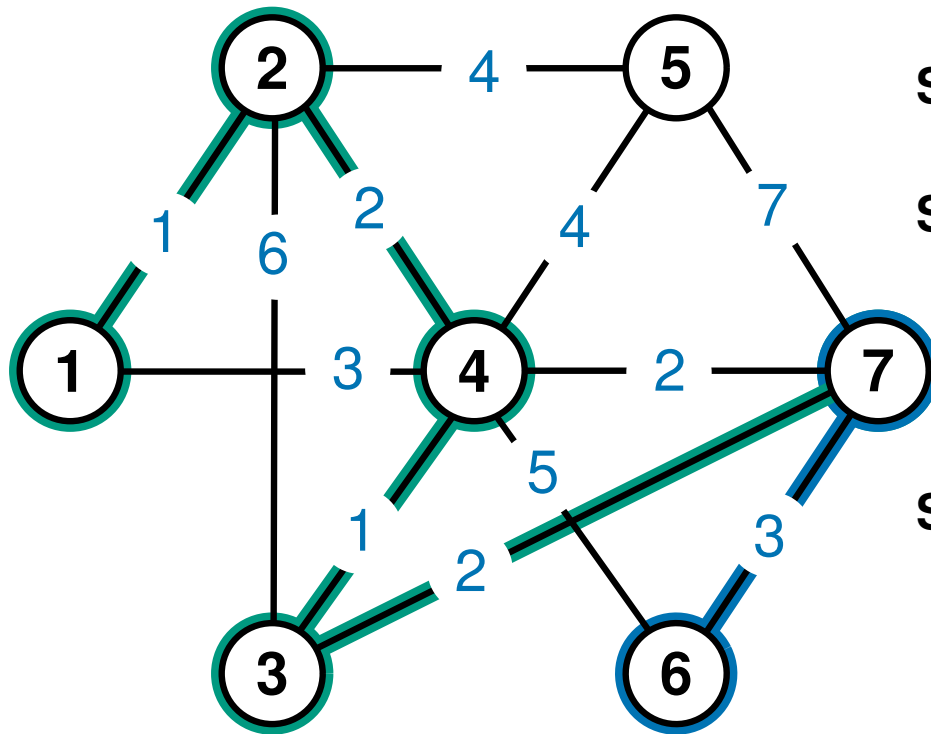
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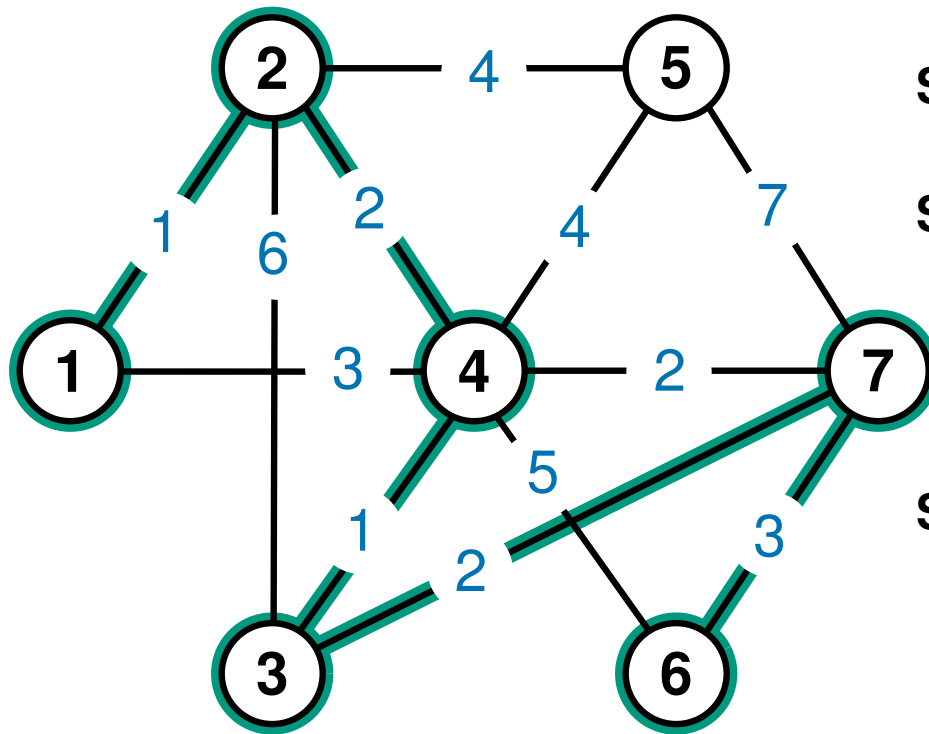
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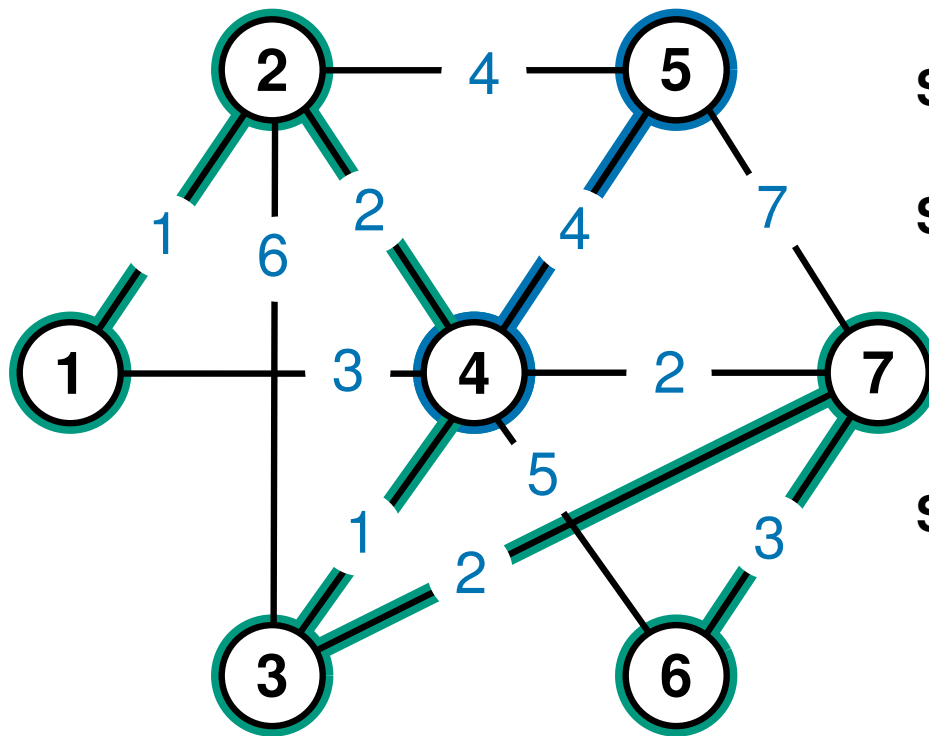
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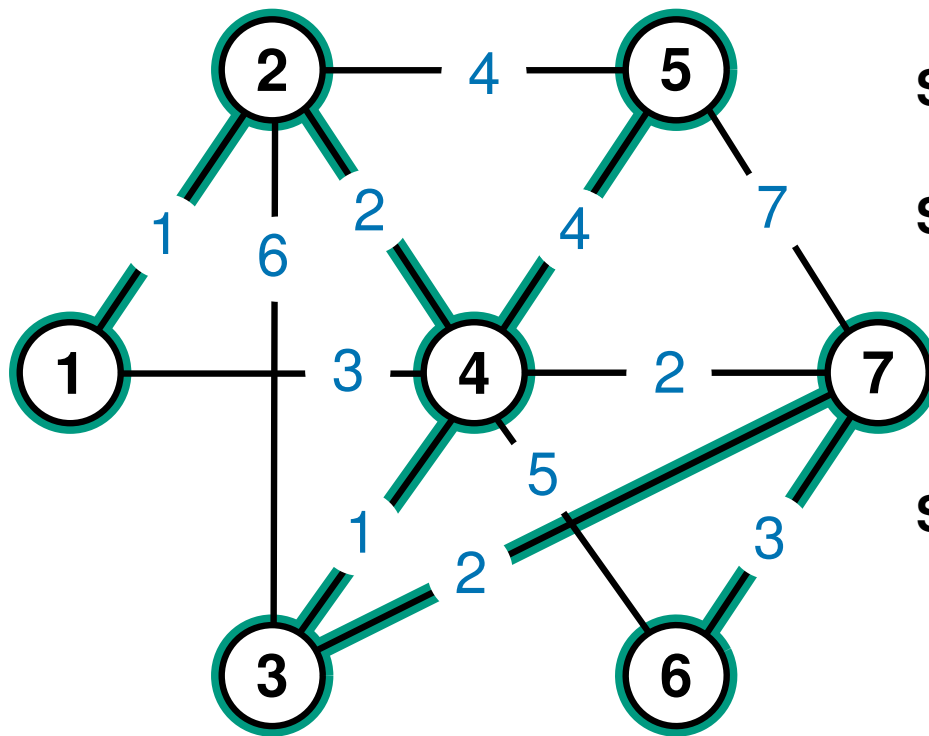
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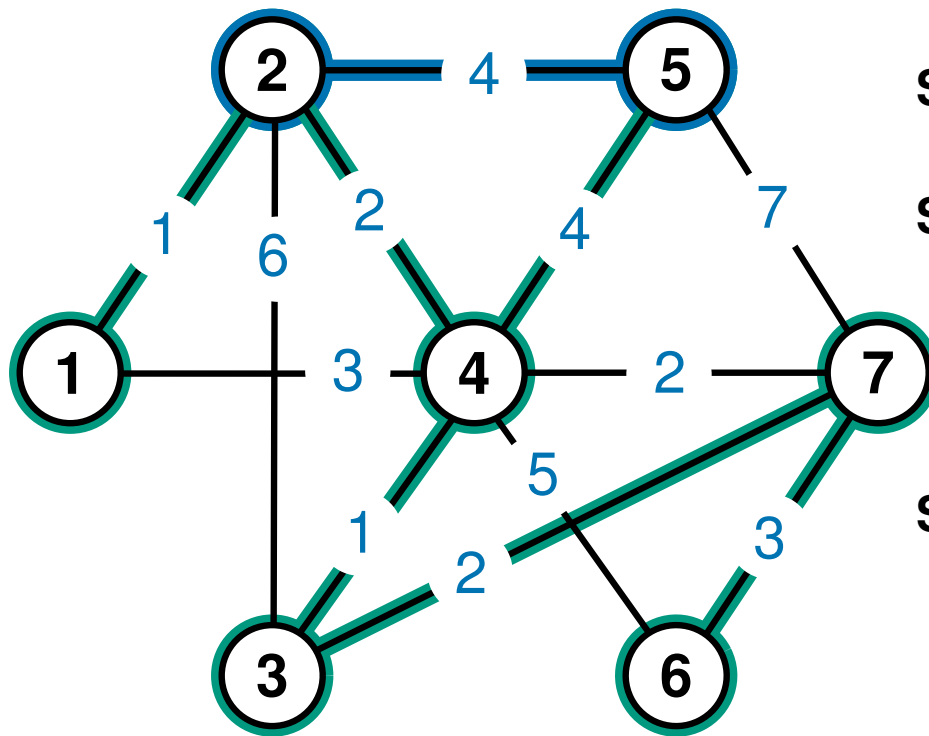
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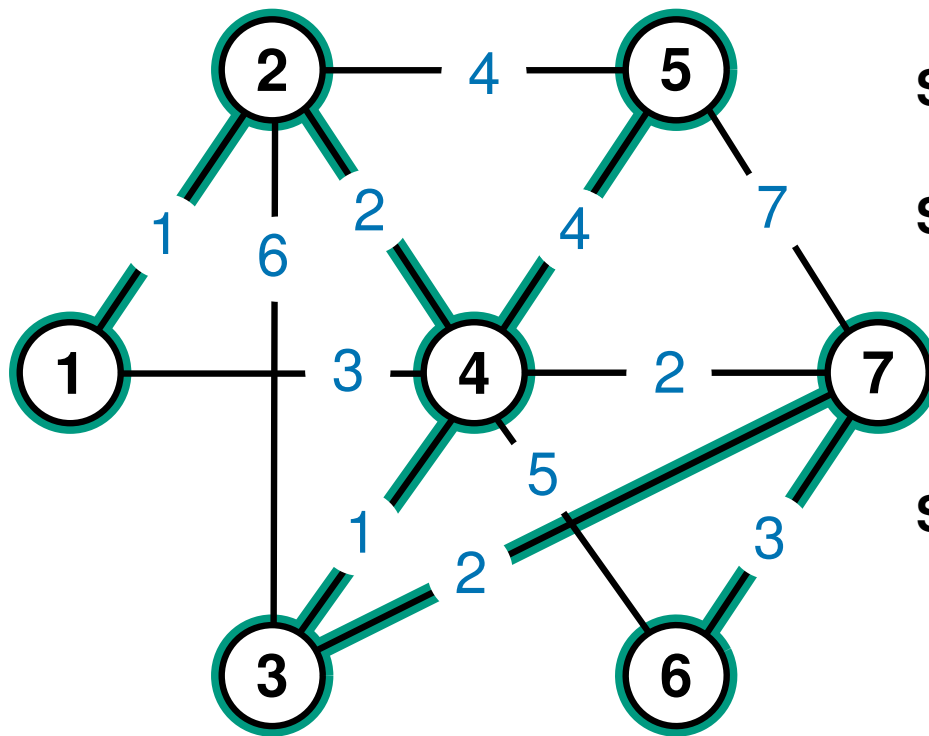
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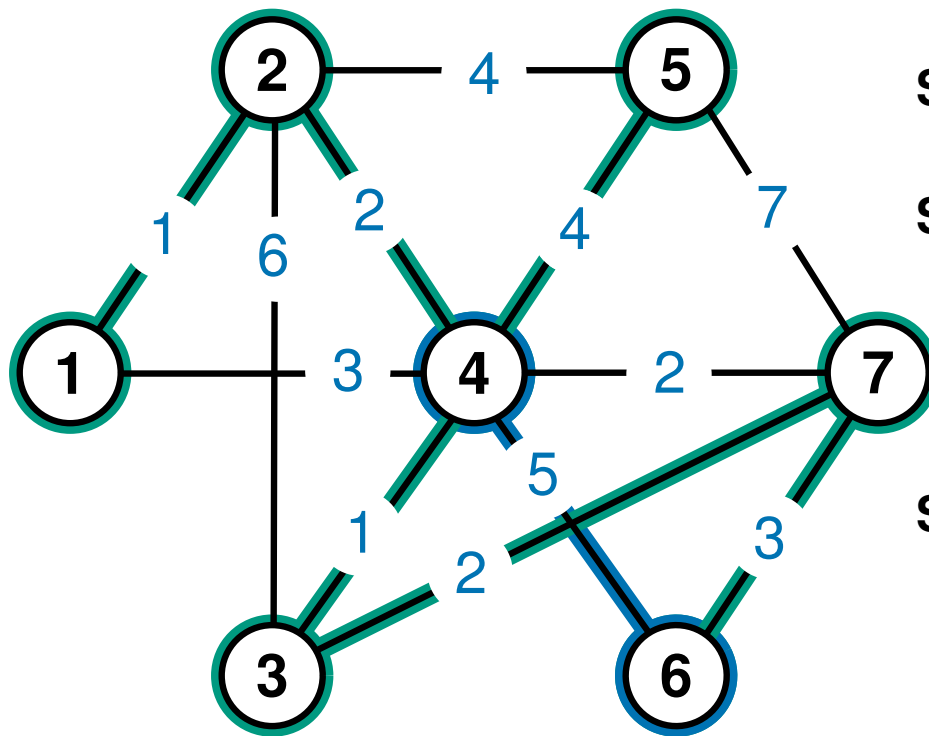
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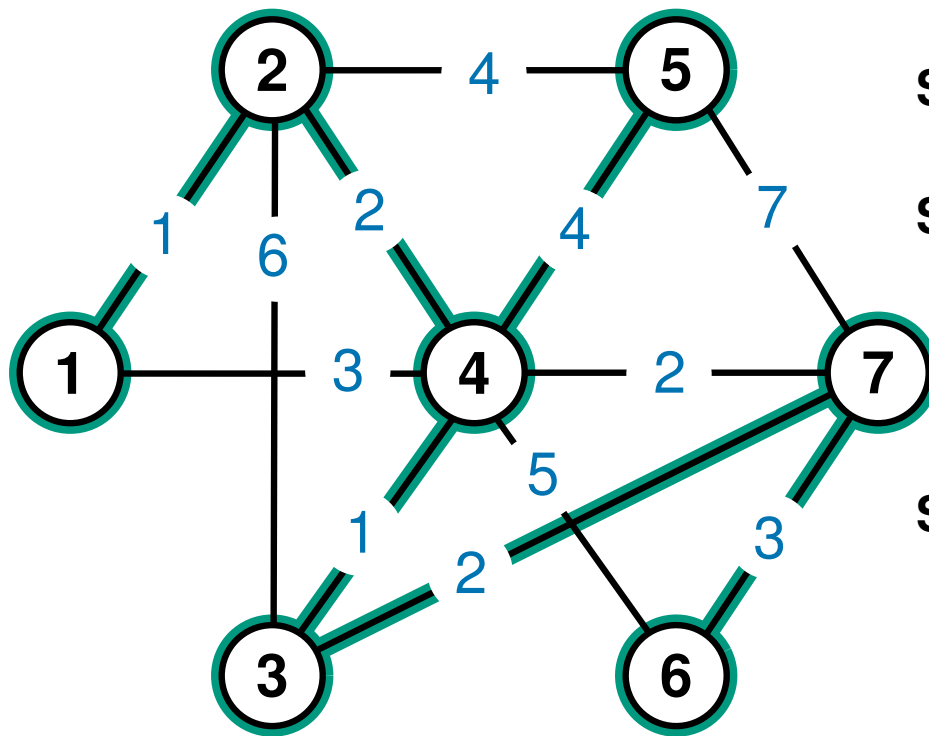
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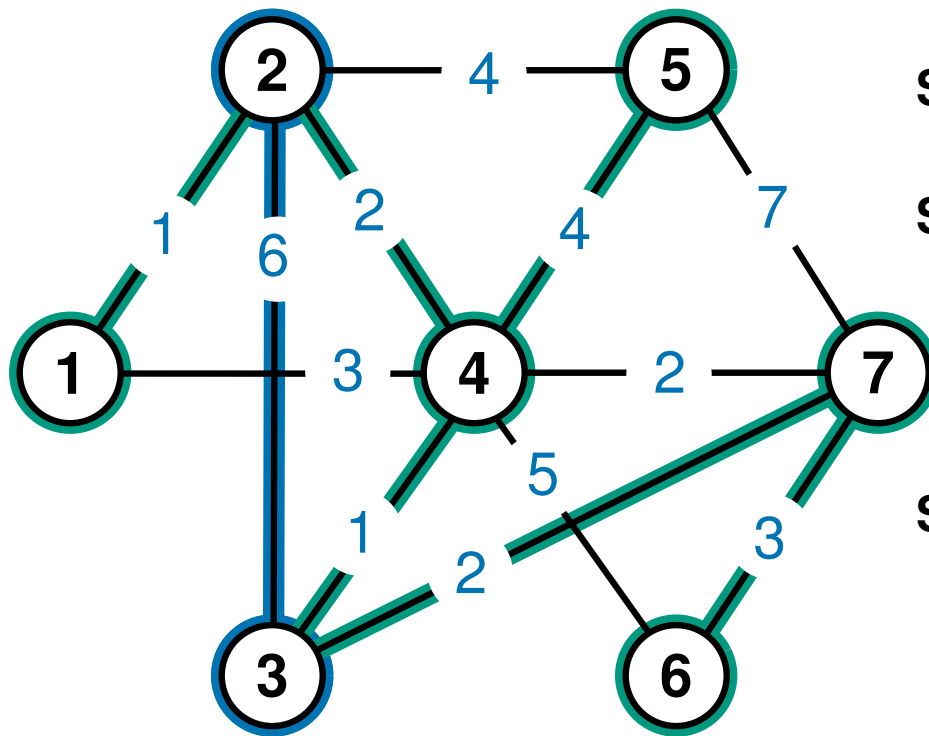
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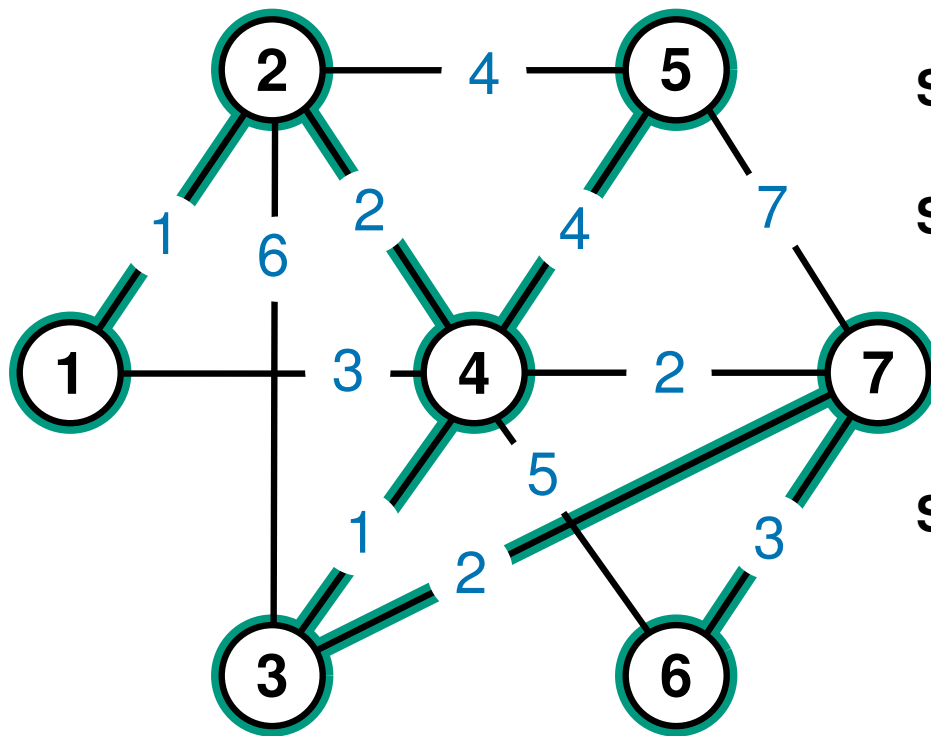
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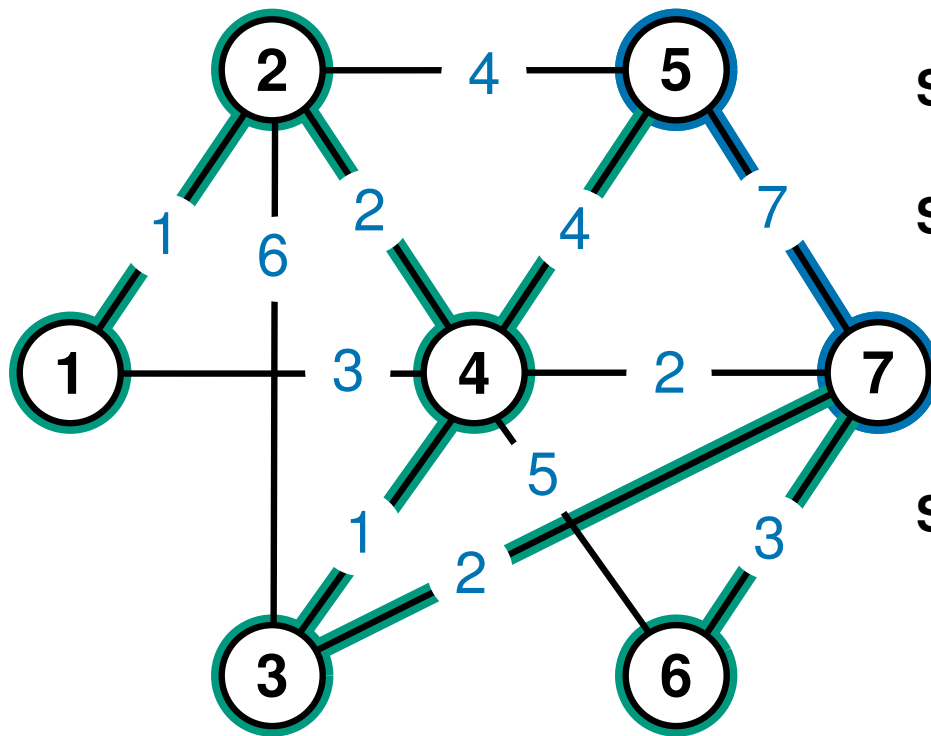
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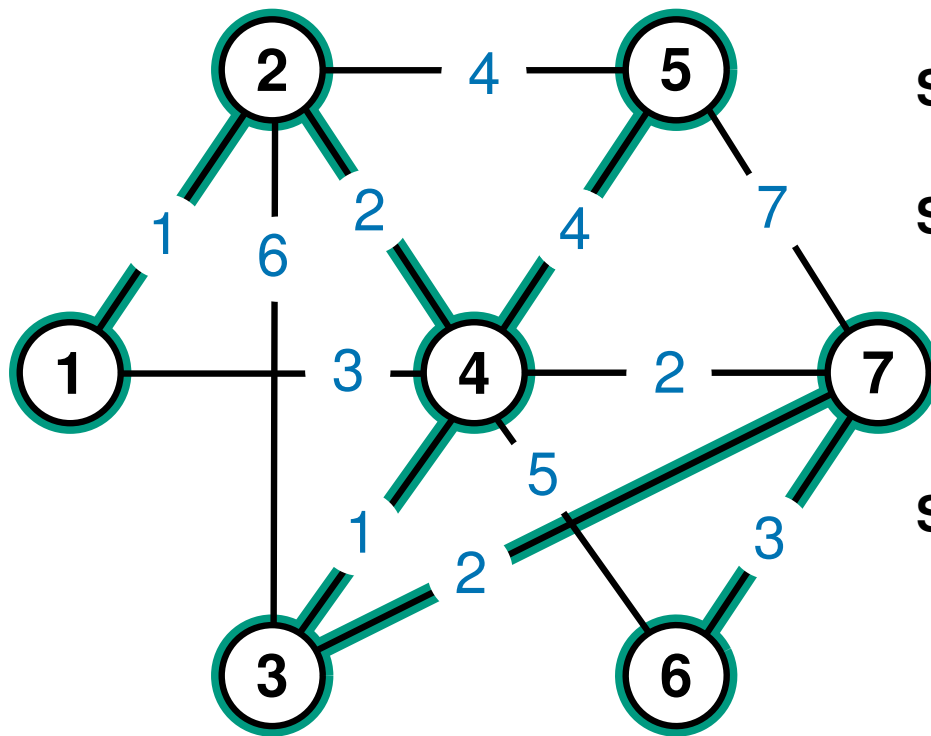
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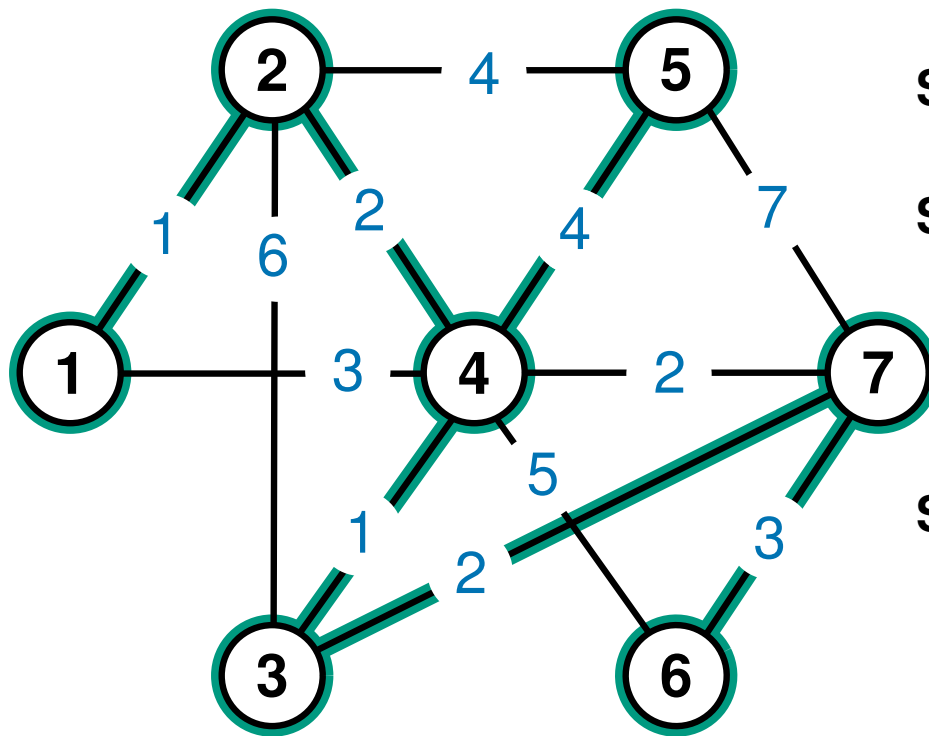
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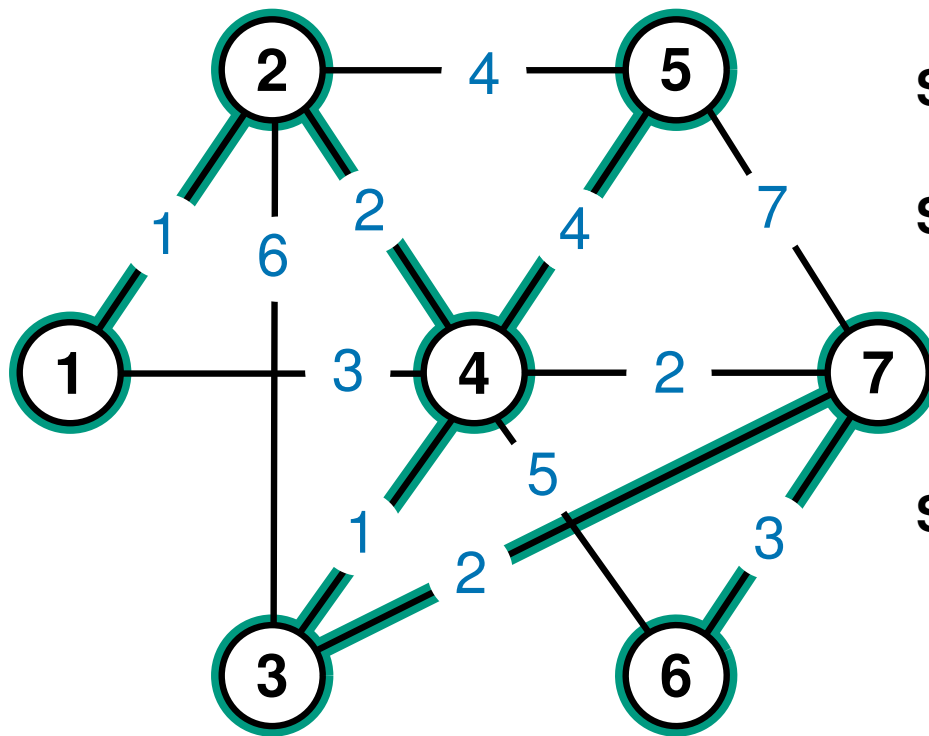
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If we implement the operations as we have seen, they run in $O(\log |V|)$ time

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 Therefore the overall running time becomes $O(|E| \log |V|)$

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Let K be the spanning tree outputted by Kruskal

(here we have omitted the proof that Kruskal always outputs a spanning tree)

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so there is a minimum spanning tree with 11 edges in common with K

How do we make M_2 ?

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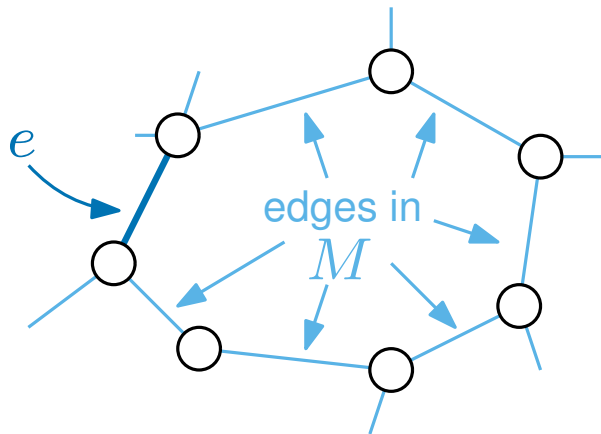
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If we were to add e to M we would introduce a cycle.

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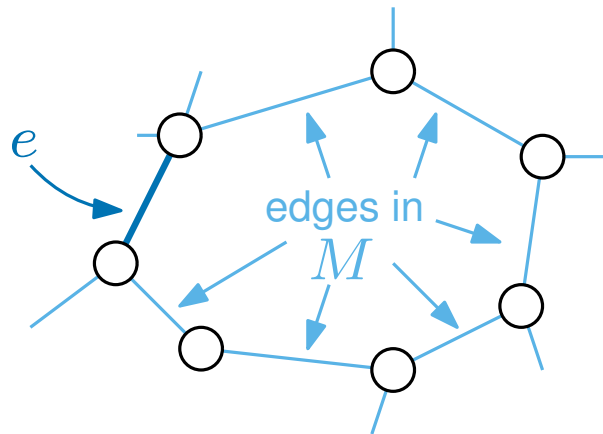


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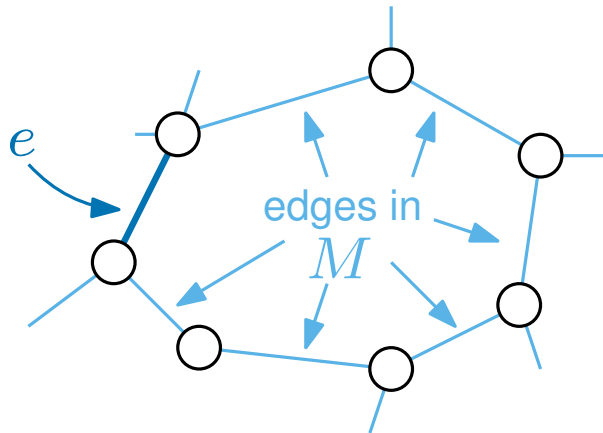


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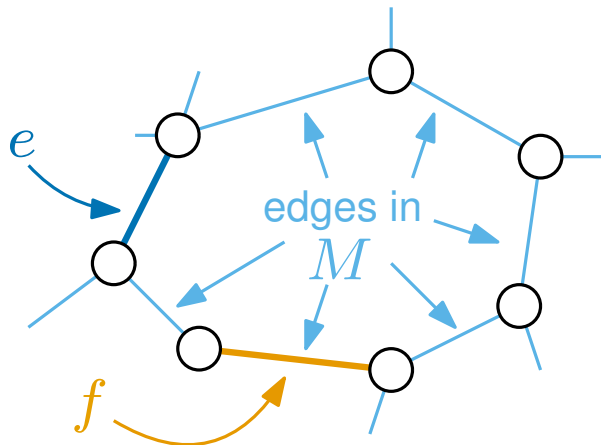
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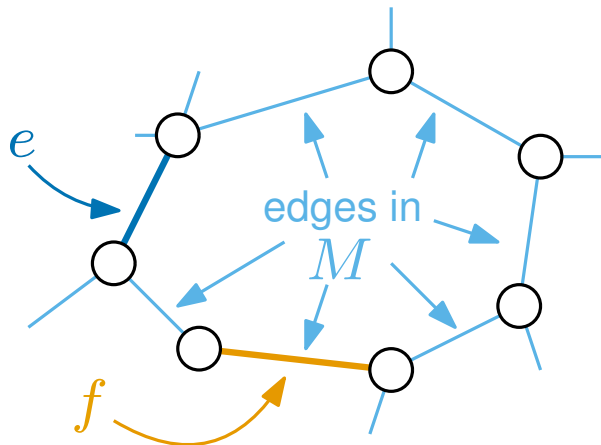
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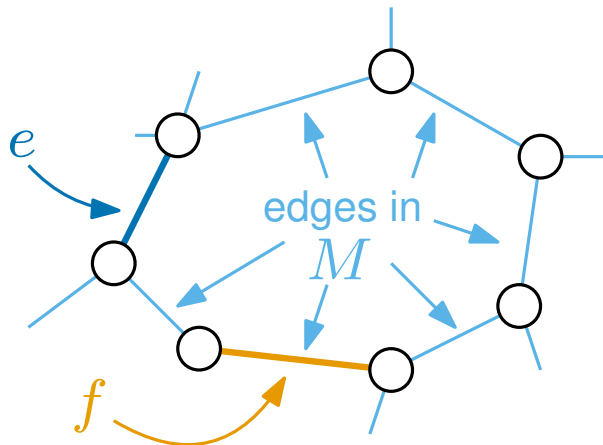
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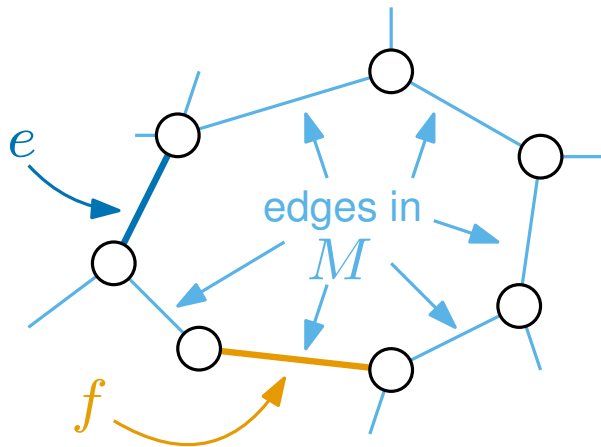
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Let M_2 be M with e added and f removed

How do we make M_2 ?

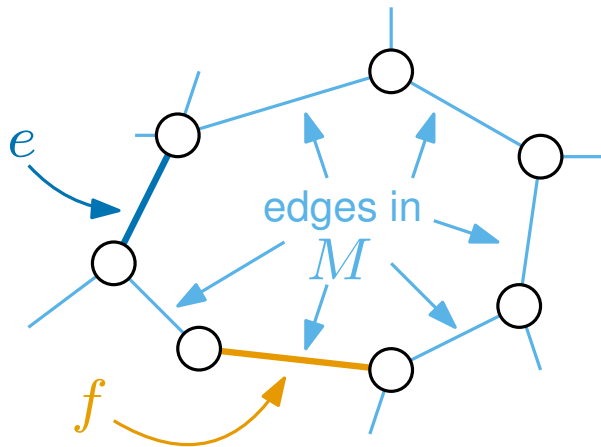
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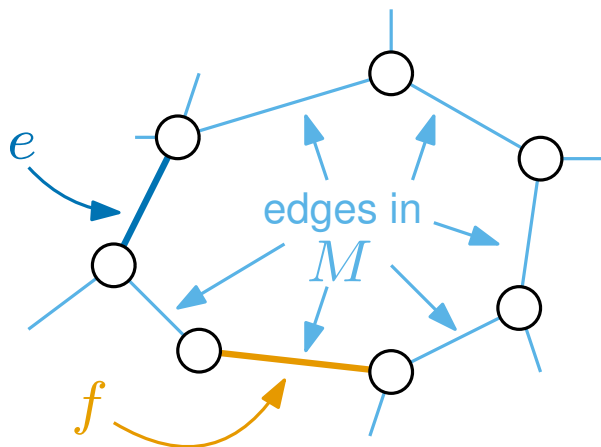
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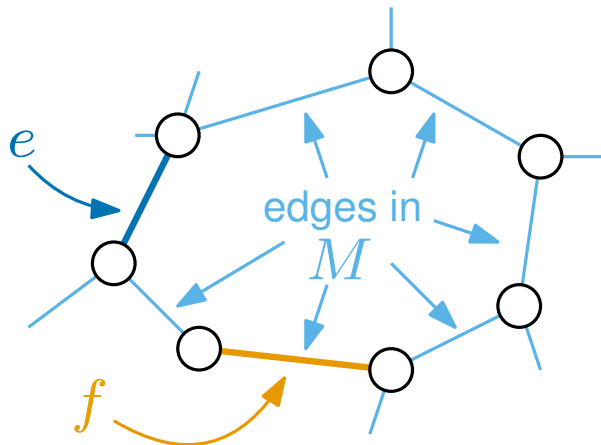


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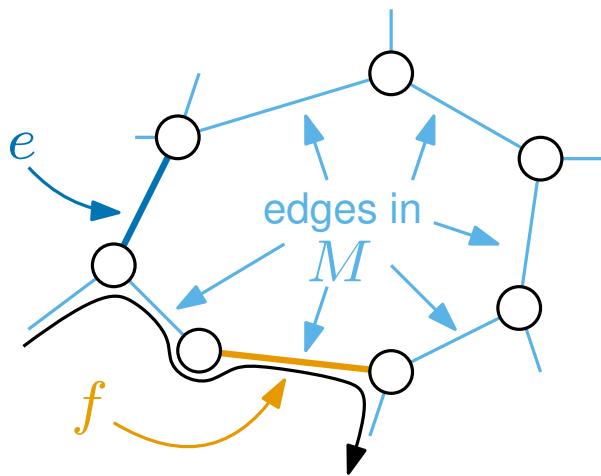
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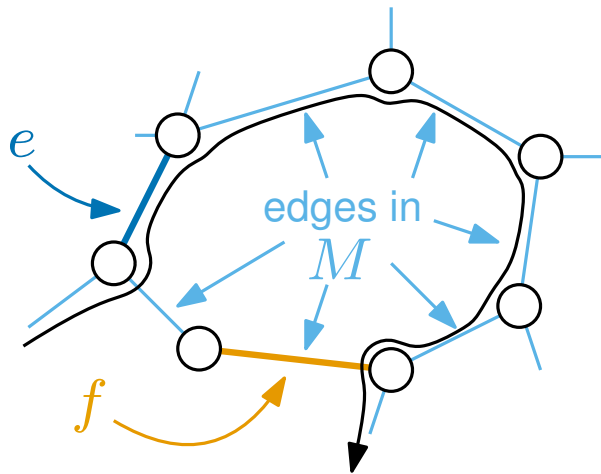


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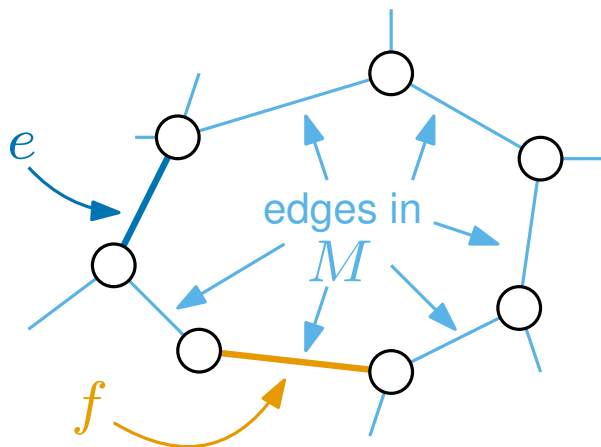


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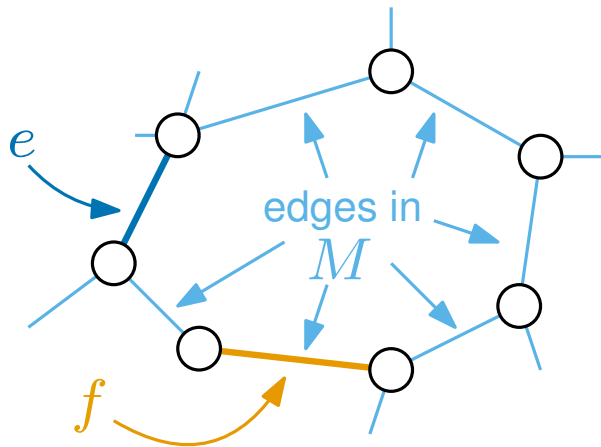
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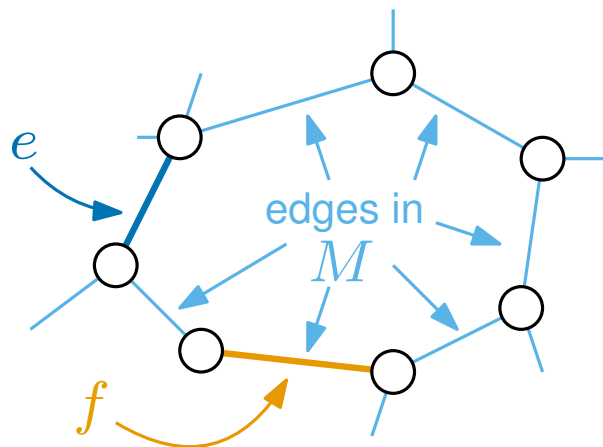
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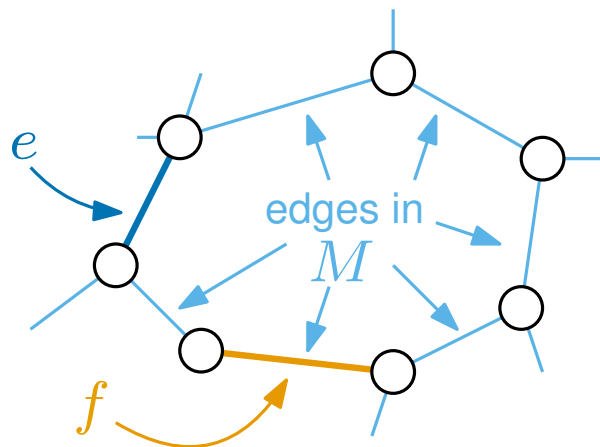
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Post lecture addition

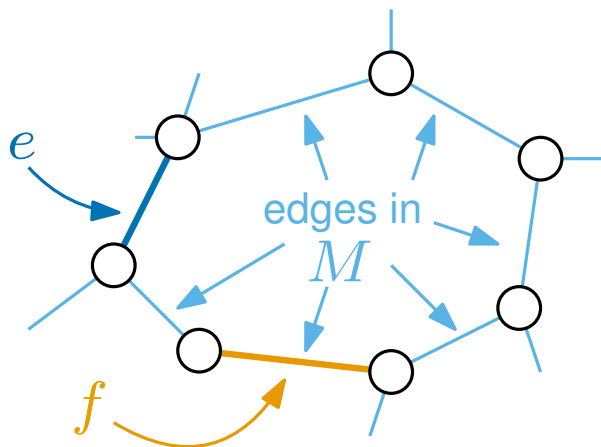
I have added a proof of this claim at the end of the slides

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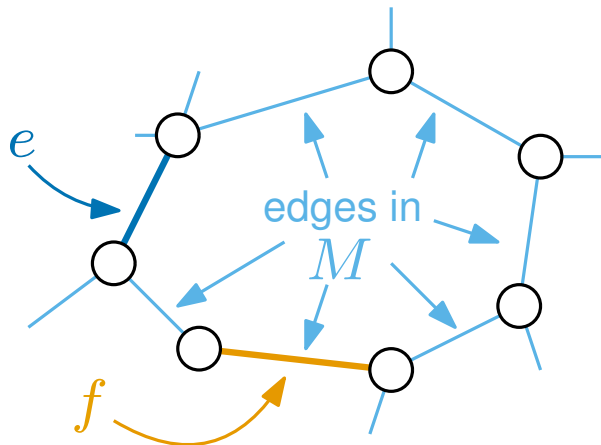
the weight of e is at most the weight of f ...

How do we make M_2 ?

Let K be the spanning tree outputted by Kruskal

and M be any minimum spanning tree such that $M \neq K$

Let M_2 be M with e added and f removed



(adding e introduces exactly one cycle)

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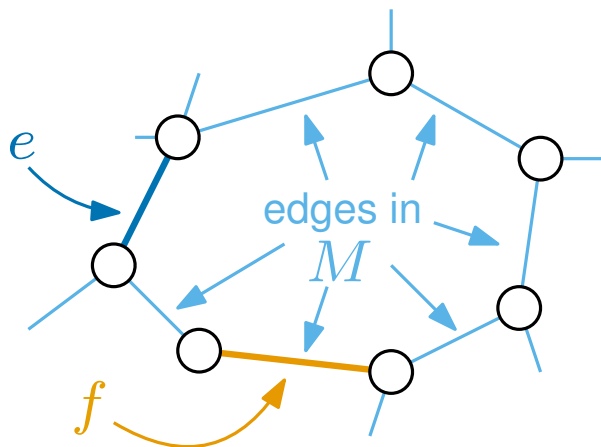
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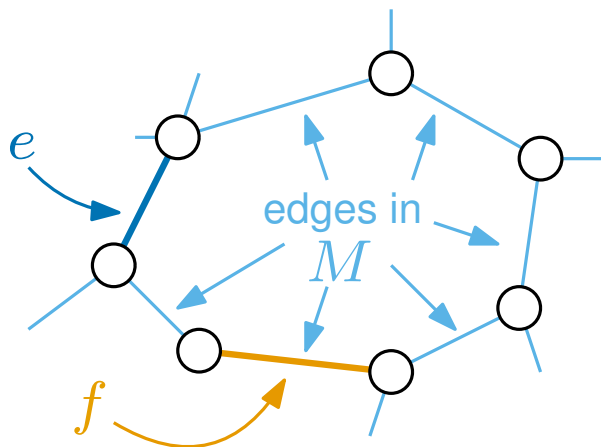
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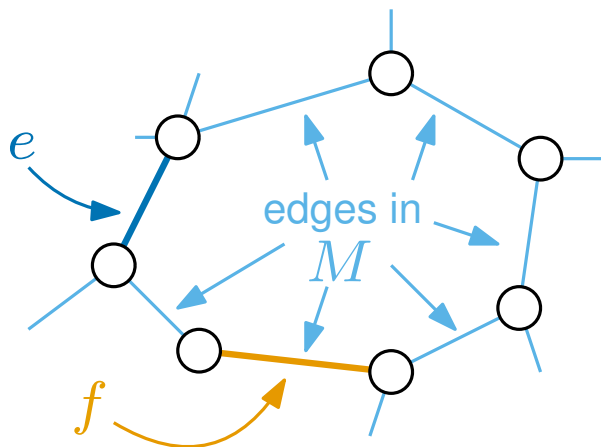
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As we said before, the proof that K is a minimum spanning tree then follows from repeatedly applying this argument

Summary

We first saw a **data structure** which stores a collection of disjoint sets

The elements of the sets are numbers from $\{1, 2, \dots, n\}$

The operations **UNION** and **FINDSET** run in $O(\log n)$ time

and the operation **MAKESET** runs in $O(1)$ time.

We then saw Kruskal's algorithm

which finds a minimum spanning tree in an connected, undirected graph

and runs in $O(|E| \log |V|)$ time

when implemented using the above data structure

Prims algorithm for finding a minimum spanning tree in a connected, undirected graph

also runs in $O(|E| \log |V|)$ time

when the priority queue is implemented using a binary heap

Why is the weight of e at most the weight of f ?

(this slide was added after the lecture)

For a contradiction, assume that f is lighter than e .

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so every edge in K' is no heavier than f which is in turn lighter than e

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Otherwise, there is an edge in K' which is lighter than e and contained in K but not in M .
This would contradict the definition of e .

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So M contains a cycle, *Contradiction!*