

# Heaps

<https://cs.pomona.edu/classes/cs140/>

# Outline

## Topics and Learning Objectives

- Discuss data structure operations
- Cover heap sort
- Discuss heaps

## Exercise

- Heap practice

# Extra Resources

- Introduction to Algorithms, 3rd, chapter 6
- Algorithms Illuminated, Part 2: Chapter 10

# Data Structures

Used in essentially every single programming task that you can think of

- What are some examples of data structures?
- What are some example programs?

What do they do?

- They organize data so that it can be effectively accessed.
- A data structure is not necessarily a method of laying out data in memory
- It is a way of logically thinking about your data.

# The **Heap** Data Structure (**not heap memory**)

A container for objects that have **key values**  
(Sometimes called a “Priority Queue”)

## Operations:

- Insert :  $O(\lg n)$
  - Extract-Min (or max) :  $O(\lg n)$
  
  - Heapify :  $O(n)$  for batched inserts
  - Arbitrary Deletion :  $O(\lg n)$
- 
- Good for continually getting a minimum (or maximum) value

# Heap used to improve algorithm

## Selection sort

- Continually look for the smallest element
- The element currently being considered is in blue
- The current smallest element is in red
- Sorted elements are in yellow

	8
	5
	2
	6
	9
	3
	1
	4
	0
	7

# Heap used to improve algorithm

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	7

# Heap used to improve algorithm

## Selection sort

- Continually look for the smallest element

What is the runtime of selection sort?

How can we make it faster with a heap?

With a heap:  $O(n^2) \rightarrow O(n \lg n)$

- Insert all elements into a heap:  $n$
- Extract each element:  $n * \lg n$

	8
	5
	2
	6
	9
	3
	1
	4
	0
	7

# Example : Event Manager

Uses a **priority queue** (synonym for Heap)

Example: simulation or game

- play sounds
- render animation We can probably delay this without much trouble.
- detect collisions
- register input Probably the most important to get correct. But does it need to be the highest priority?

# Heap Implementation

No pointer following  
Fewer heap-allocations  
No pointer storing

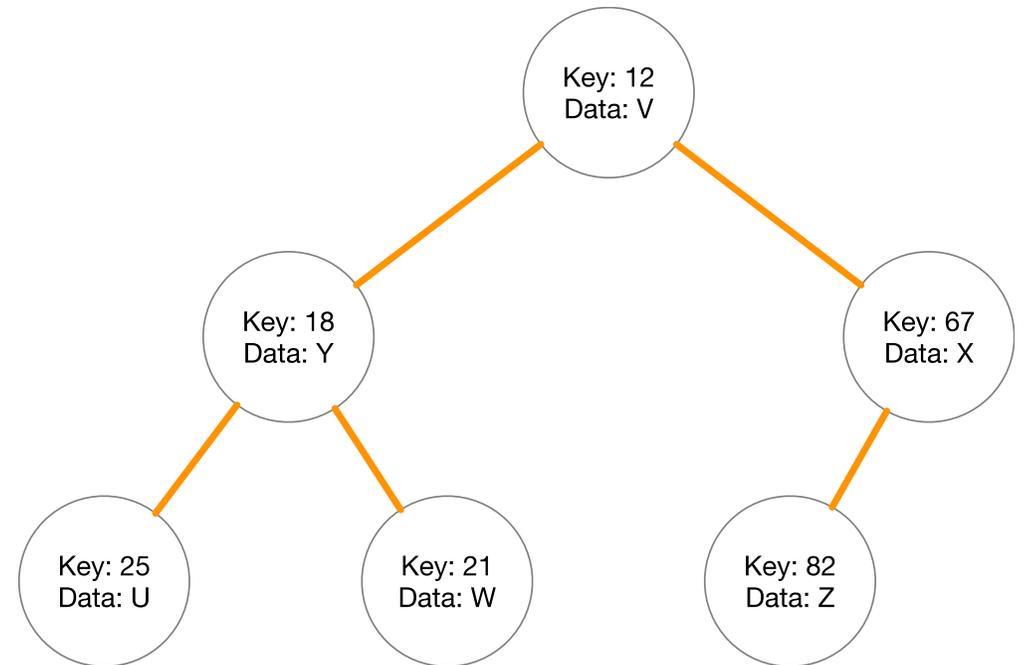
Conceptually you should think of a Heap as a binary tree

But it is implemented using an array (why?)

Heap Property: for any given node  $x$ ,

1.  $\text{key}[x] \leq \text{key}[x\text{'s left child}]$ , and
2.  $\text{key}[x] \leq \text{key}[x\text{'s right child}]$

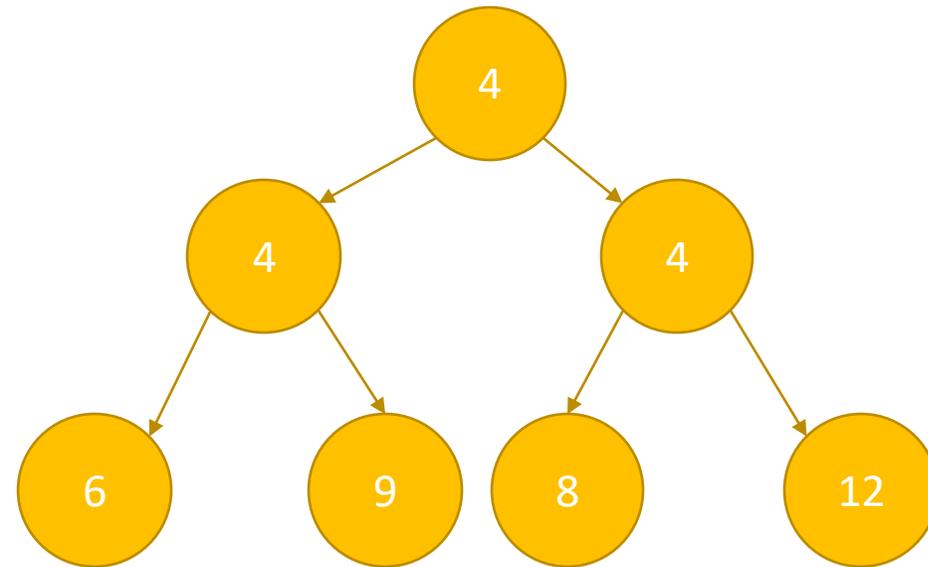
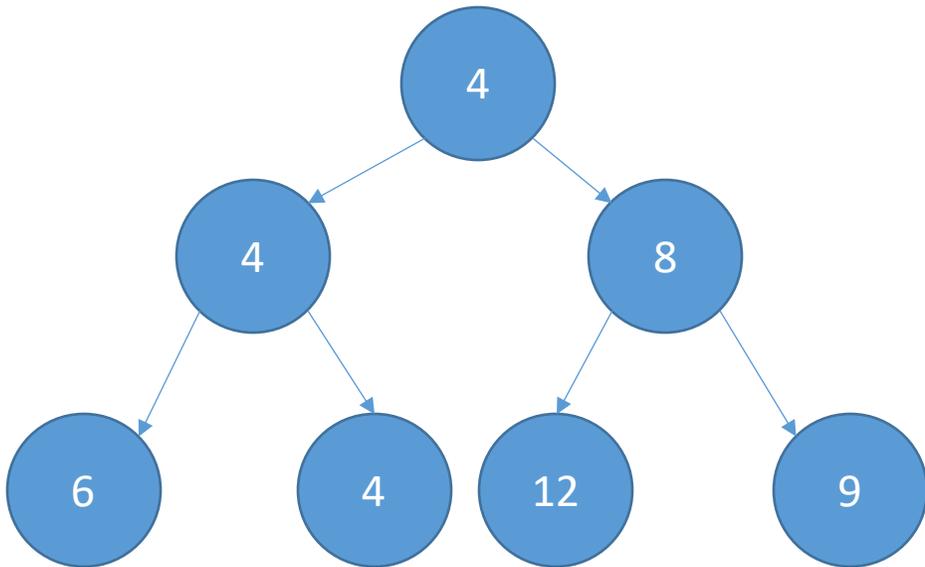
Where is the minimum key?



# Heap Implementation

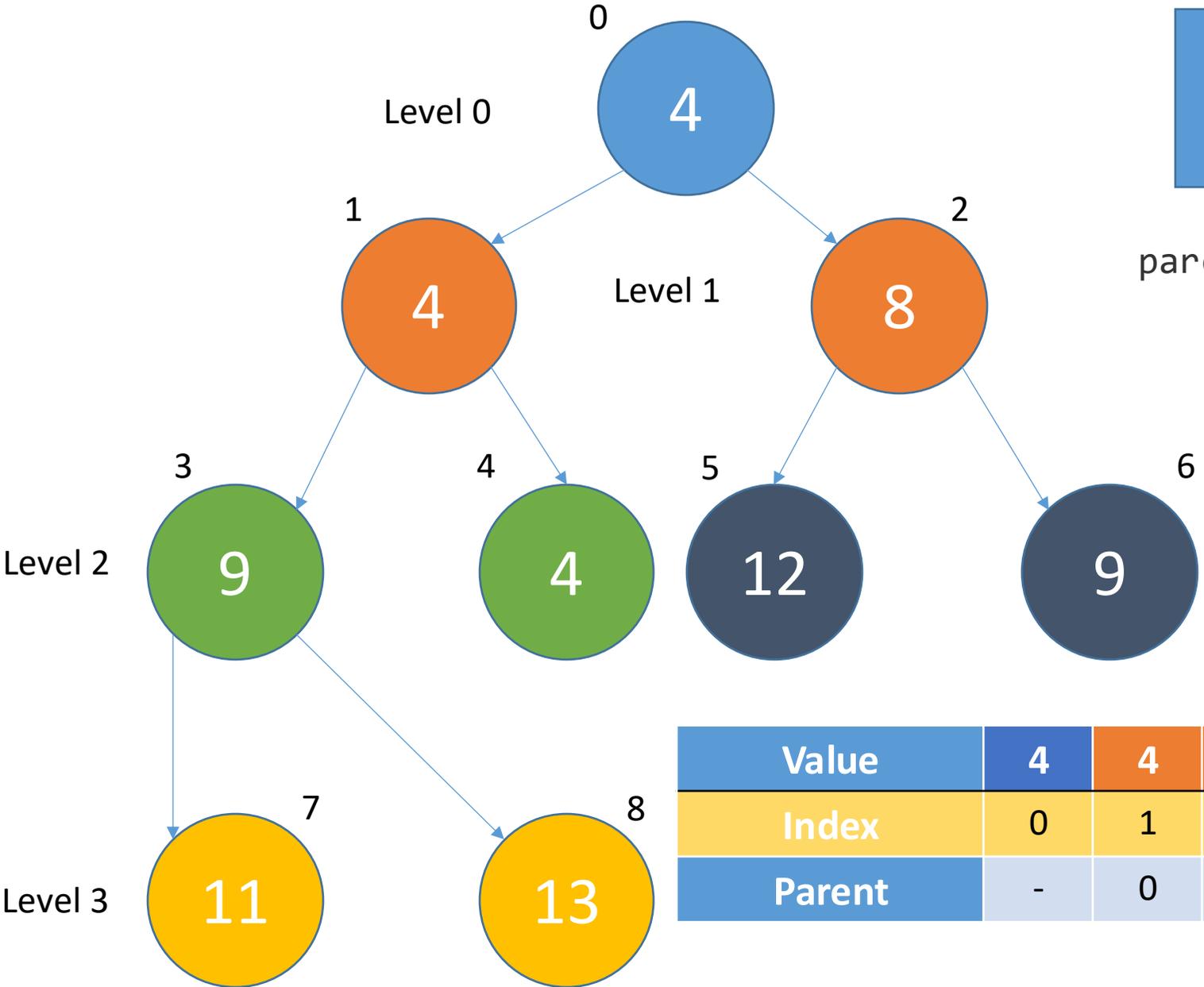
Note: Heaps are **not** unique

*You can have multiple different configurations that hold the same data*



How do you calculate the index of a node's parent?

$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$



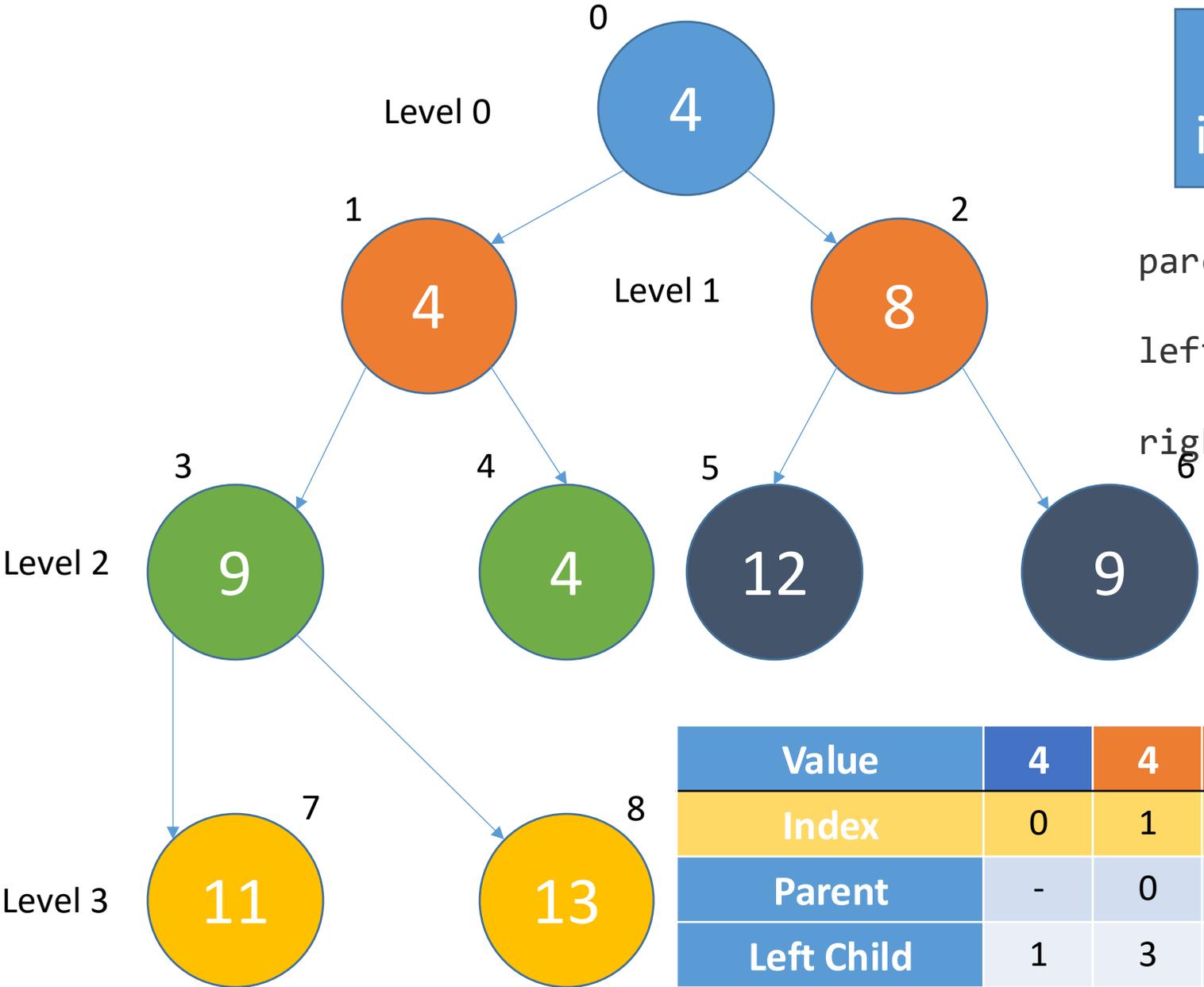
Value	4	4	8	9	4	12	9	11	13
Index	0	1	2	3	4	5	6	7	8
Parent	-	0	0	1	1	2	2	3	3

How do you calculate the indices of a node's children?

$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$

$$\text{left\_child\_index} = 2 * \text{node\_index} + 1$$

$$\text{right\_child\_index} = 2 * (\text{node\_index} + 1)$$

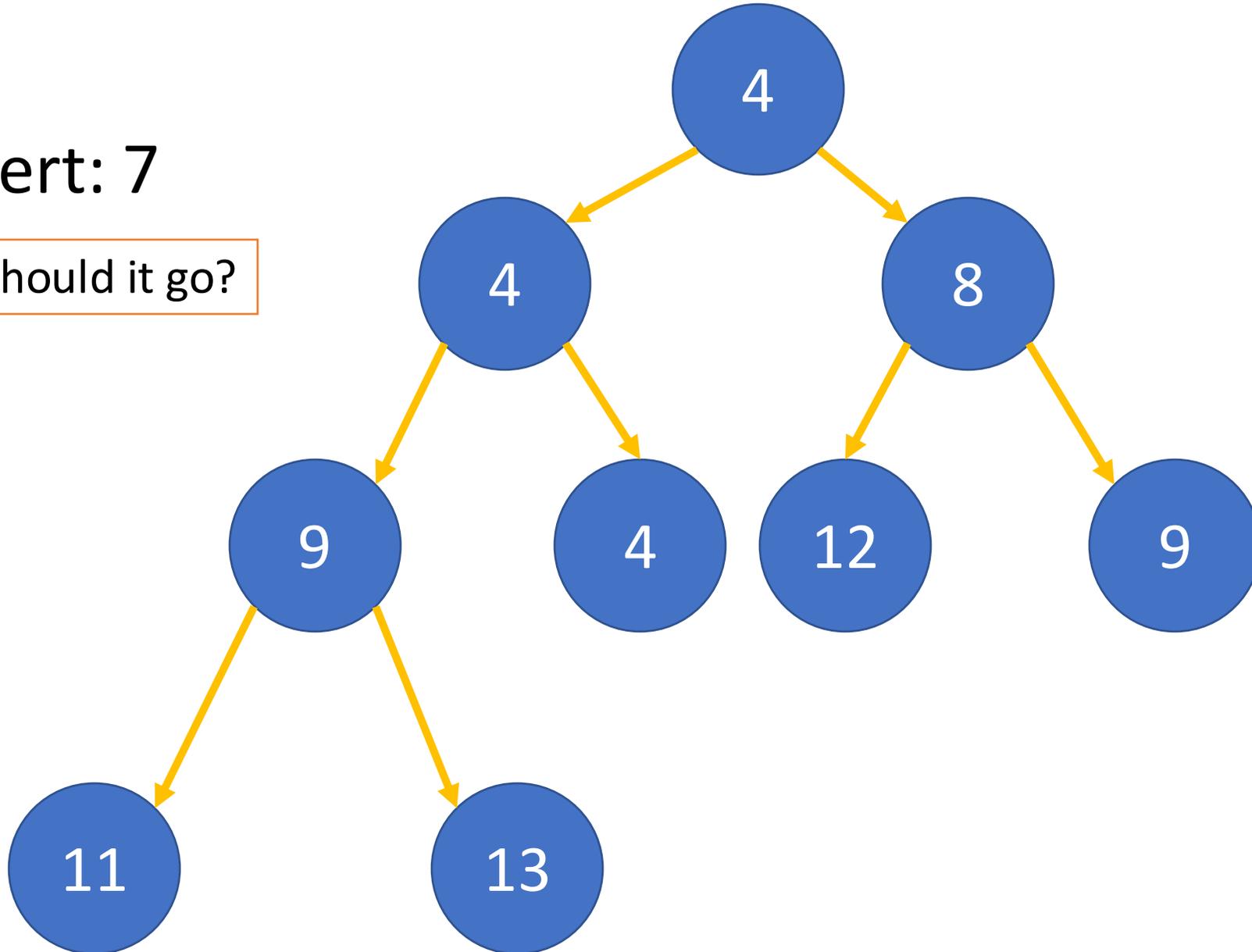


Value	4	4	8	9	4	12	9	11	13
Index	0	1	2	3	4	5	6	7	8
Parent	-	0	0	1	1	2	2	3	3
Left Child	1	3	5	7	9	11	13	15	17
Right Child	2	4	6	8	10	12	14	16	18

Exercise

Insert: 7

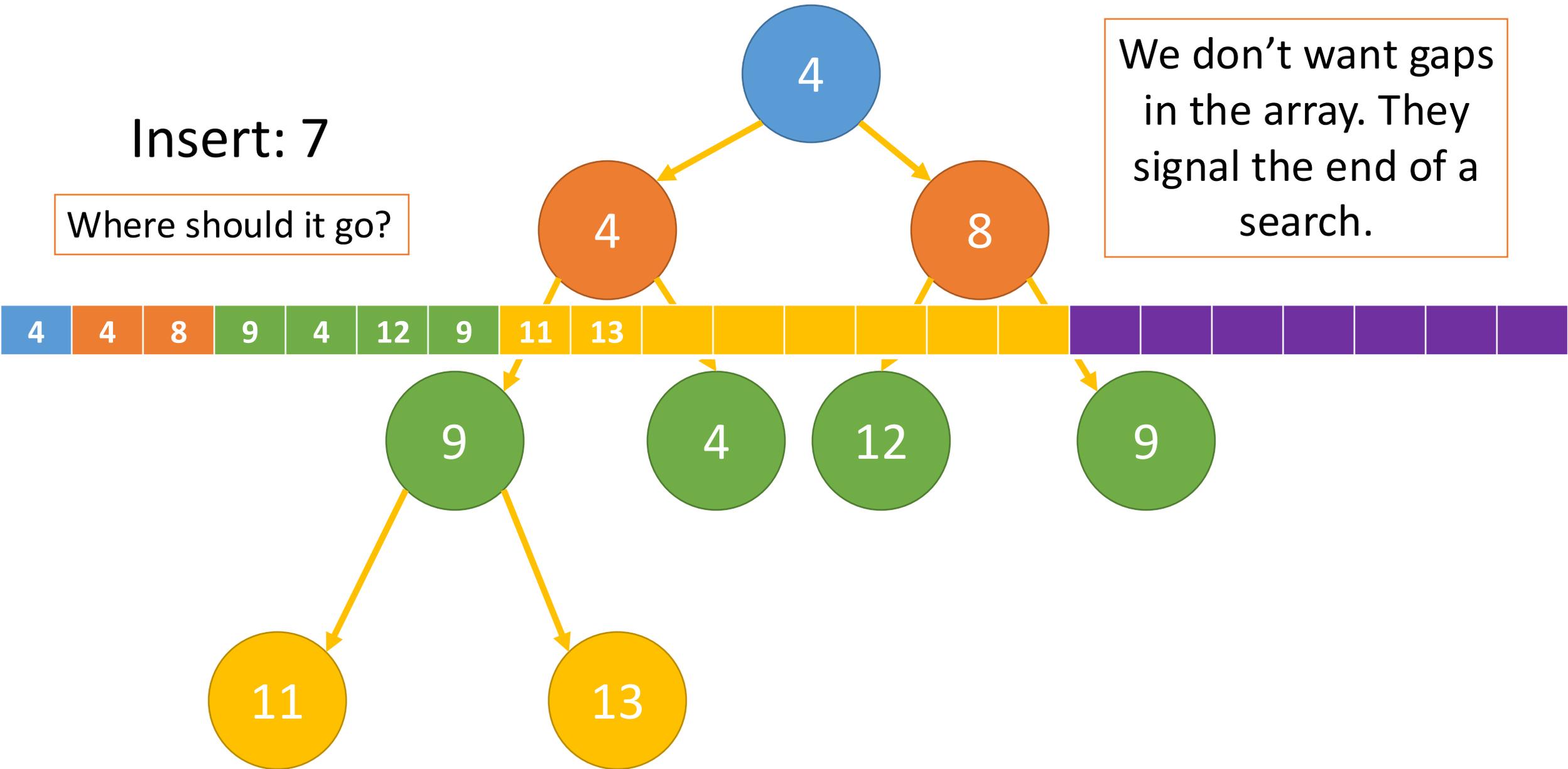
Where should it go?



# Insert: 7

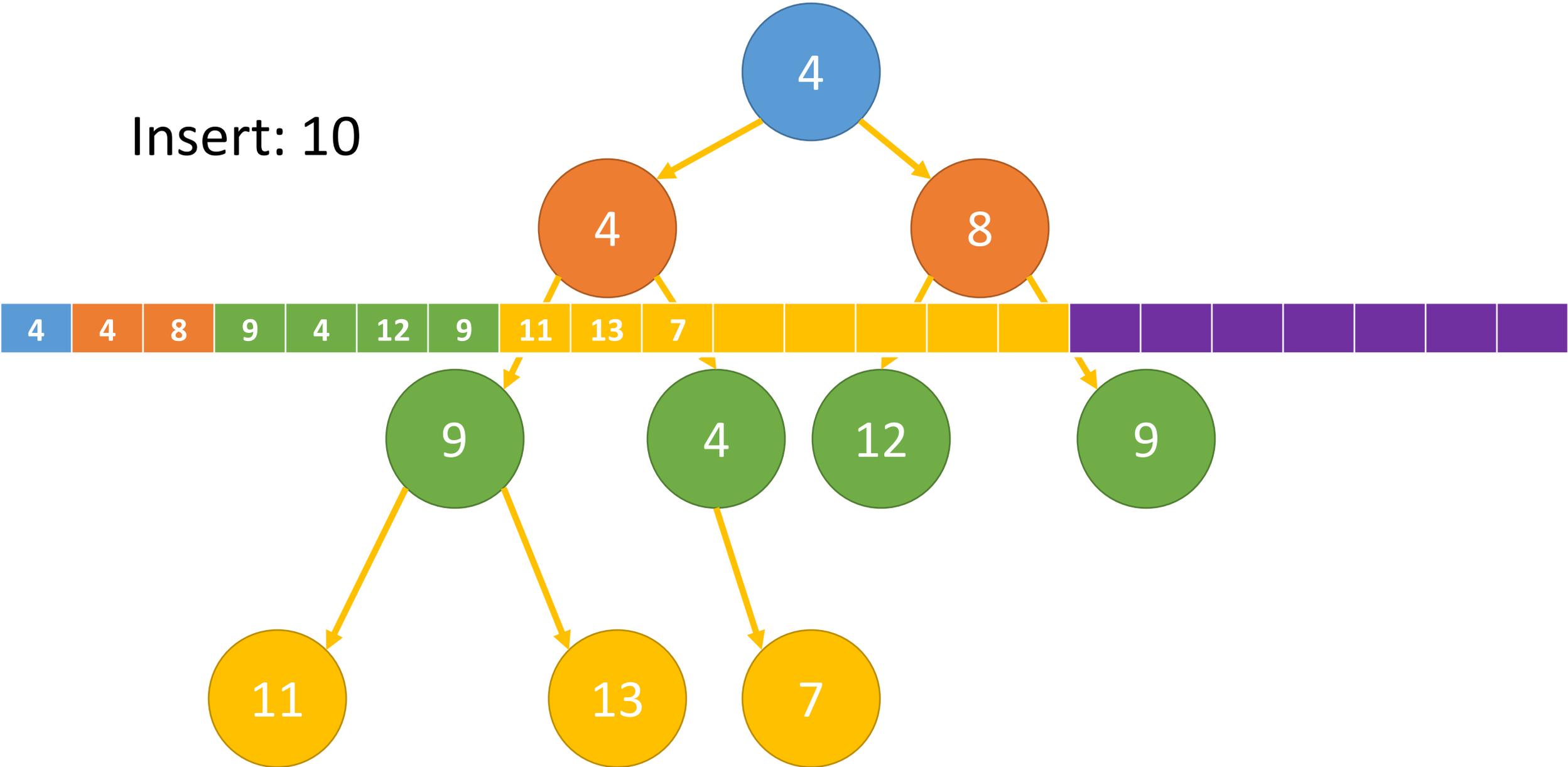
Where should it go?

We don't want gaps in the array. They signal the end of a search.

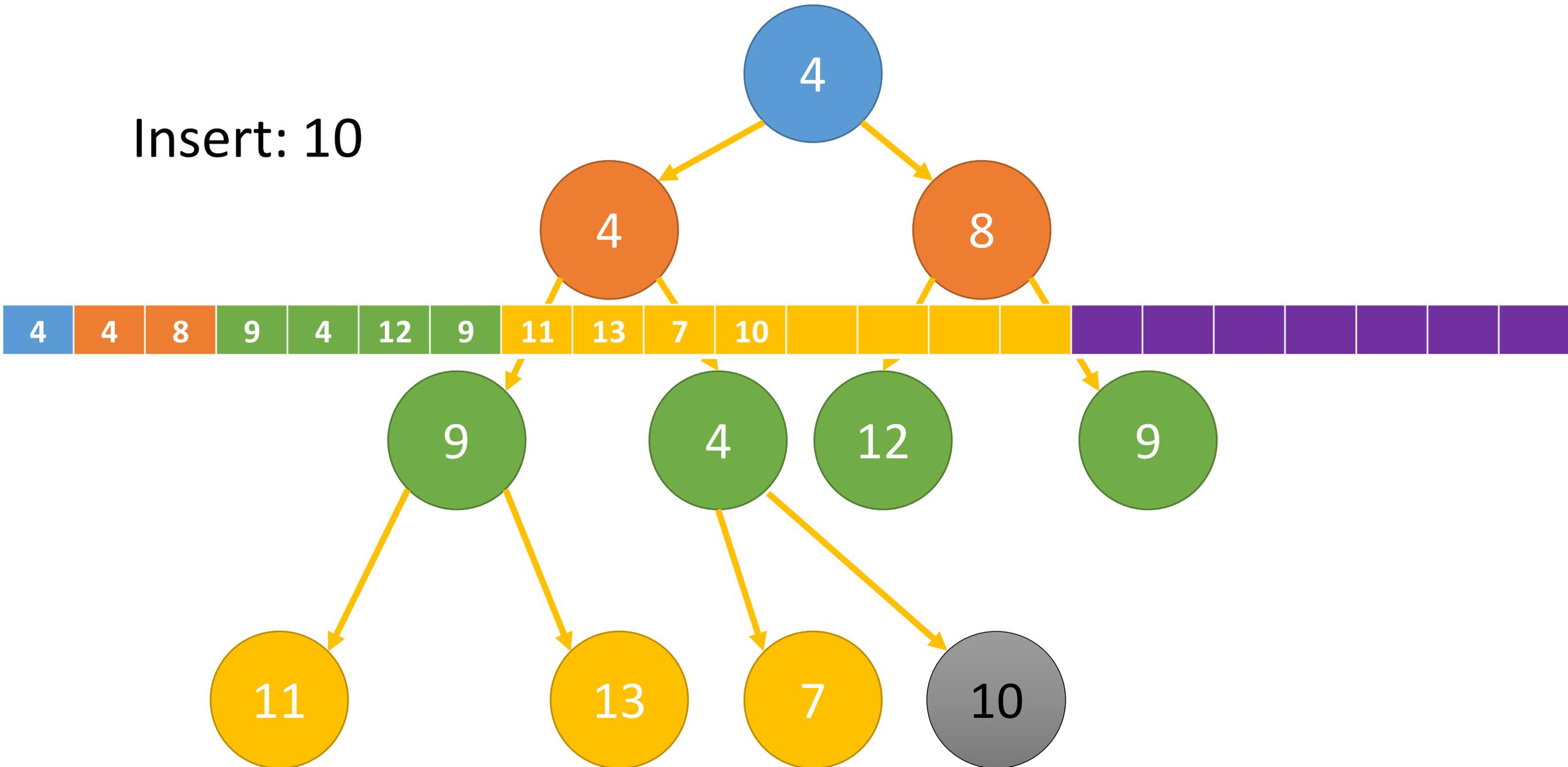




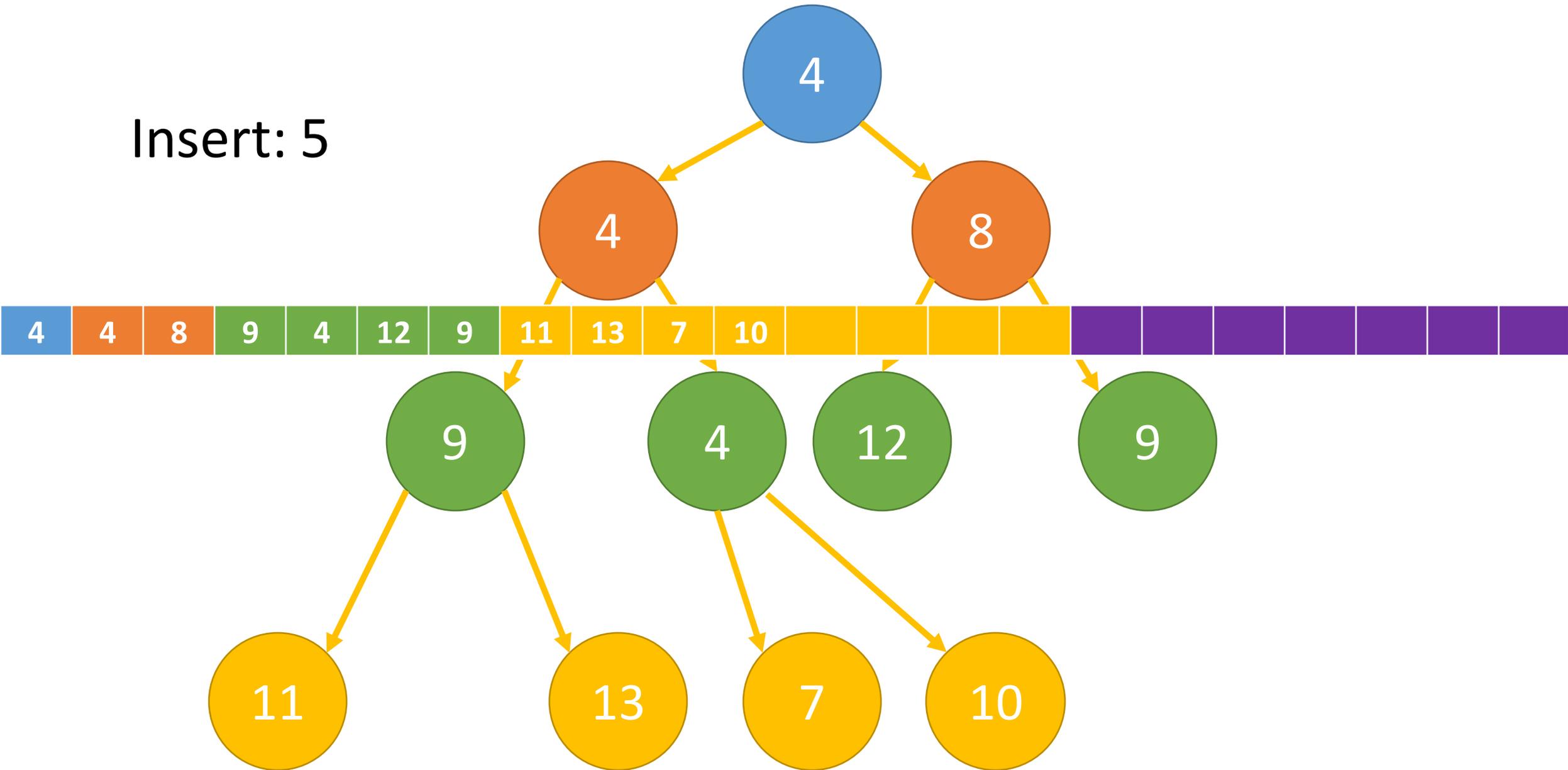
Insert: 10



Insert: 10

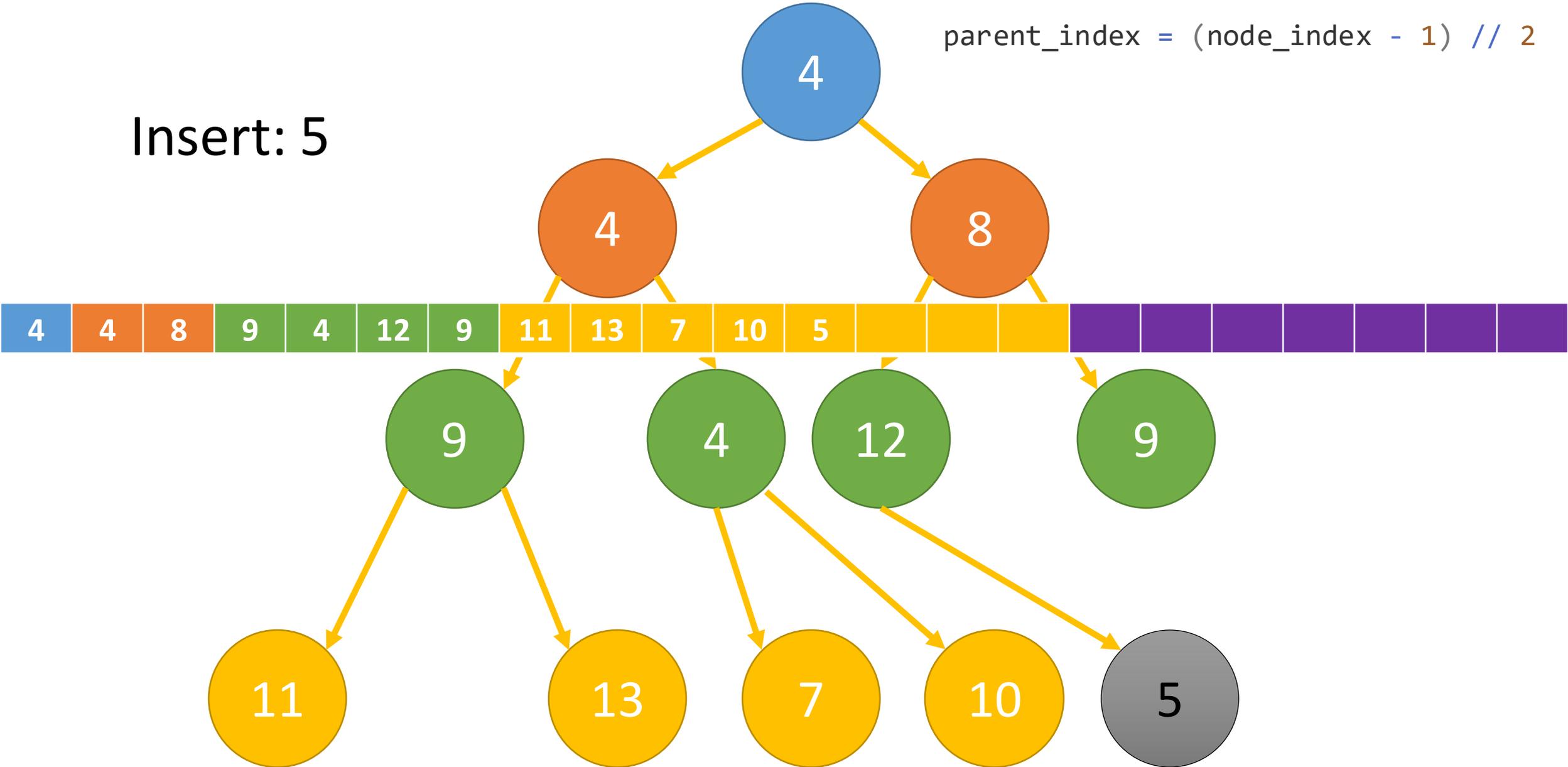


Insert: 5



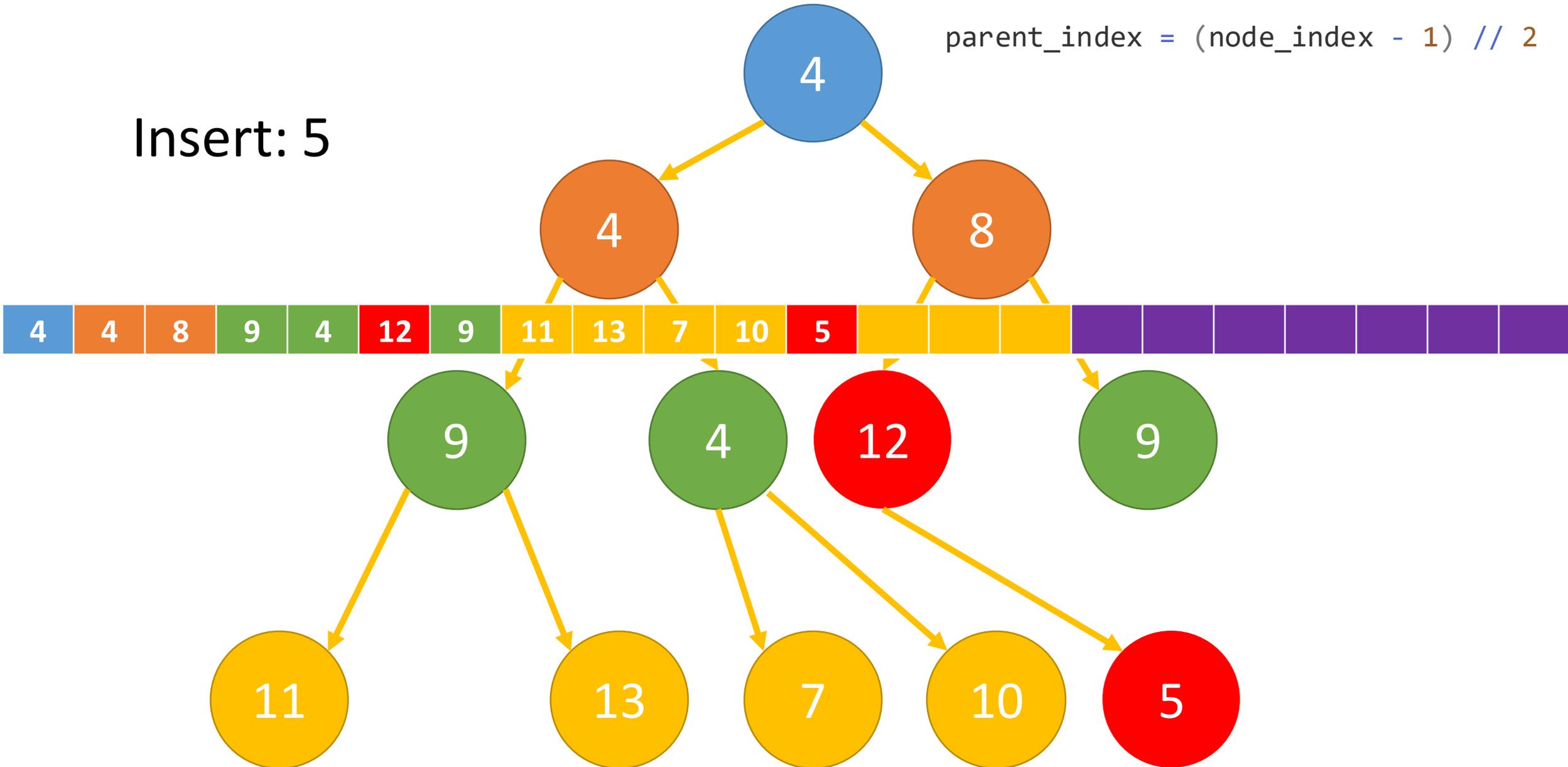
Insert: 5

$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$



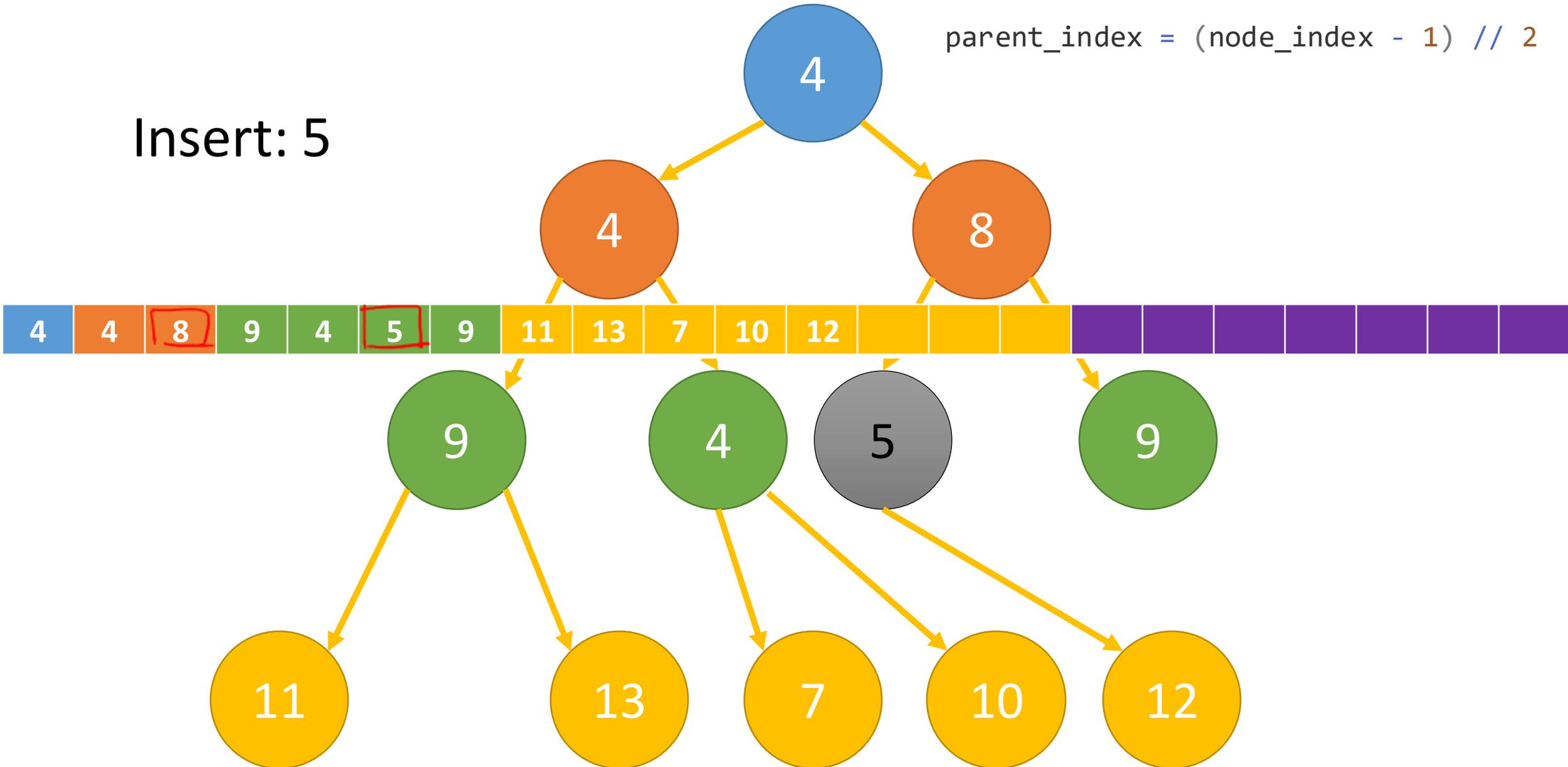
$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$

Insert: 5



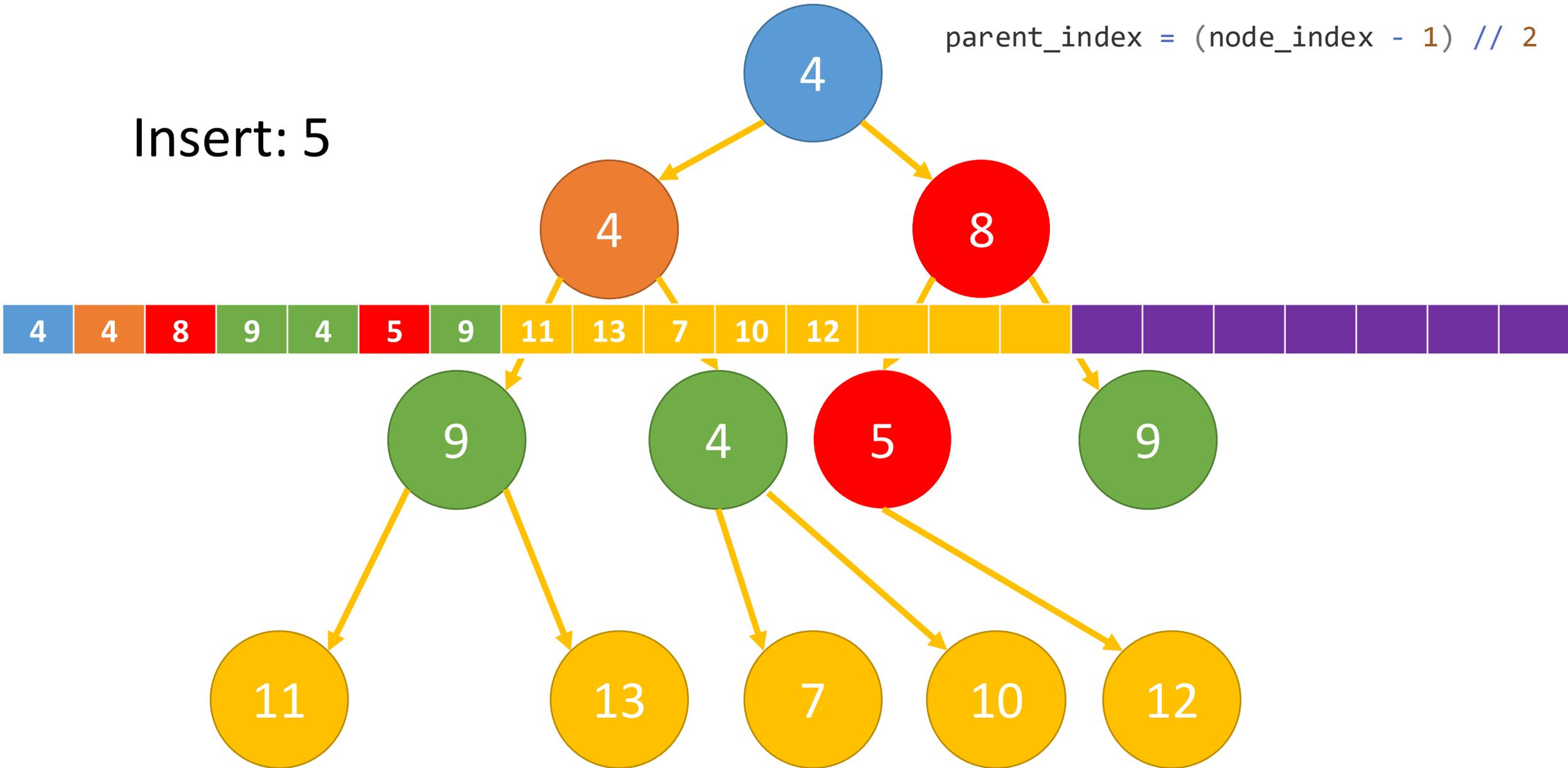
Insert: 5

$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$



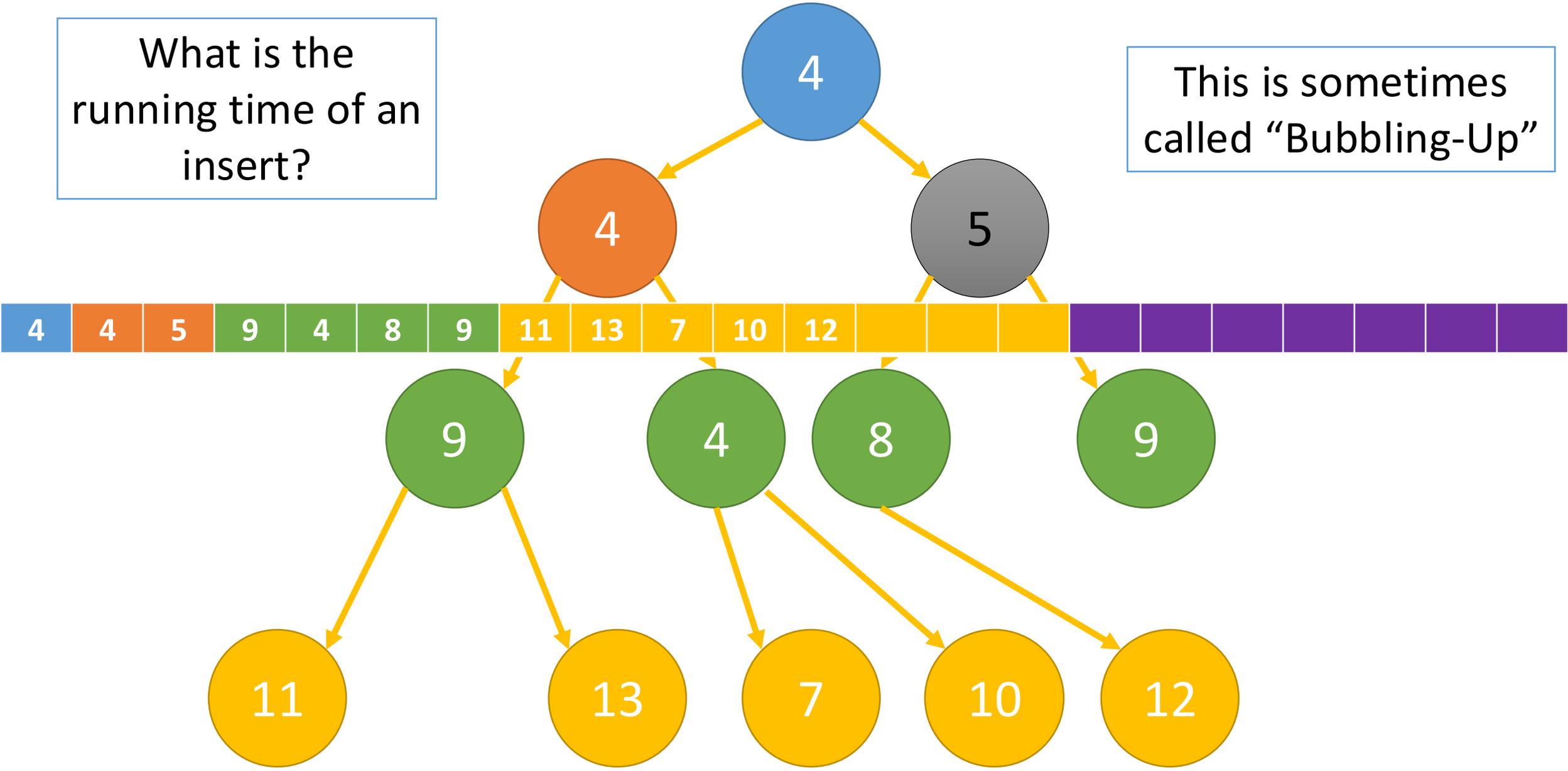
Insert: 5

$$\text{parent\_index} = (\text{node\_index} - 1) // 2$$



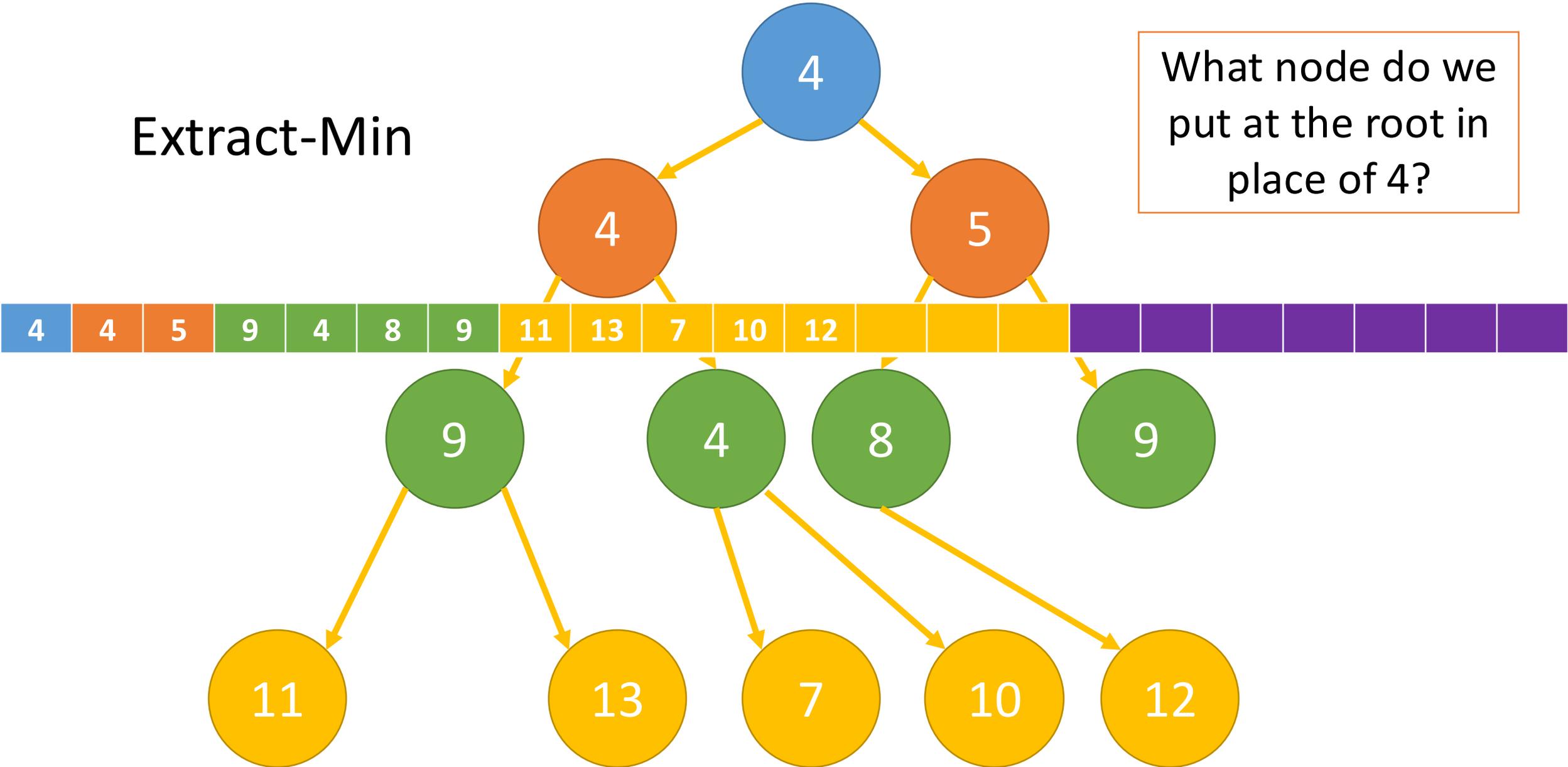
What is the running time of an insert?

This is sometimes called "Bubbling-Up"



# Extract-Min

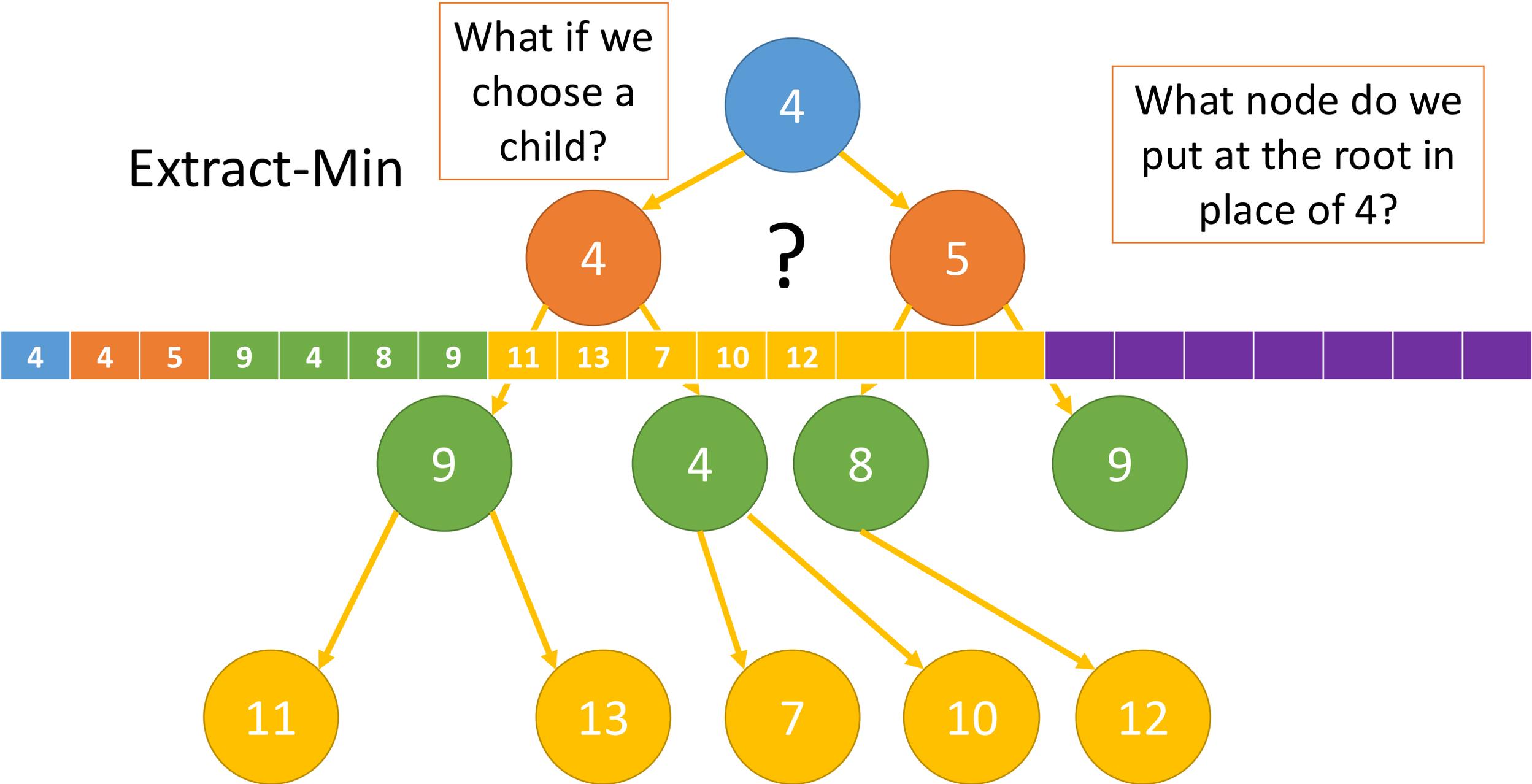
What node do we put at the root in place of 4?



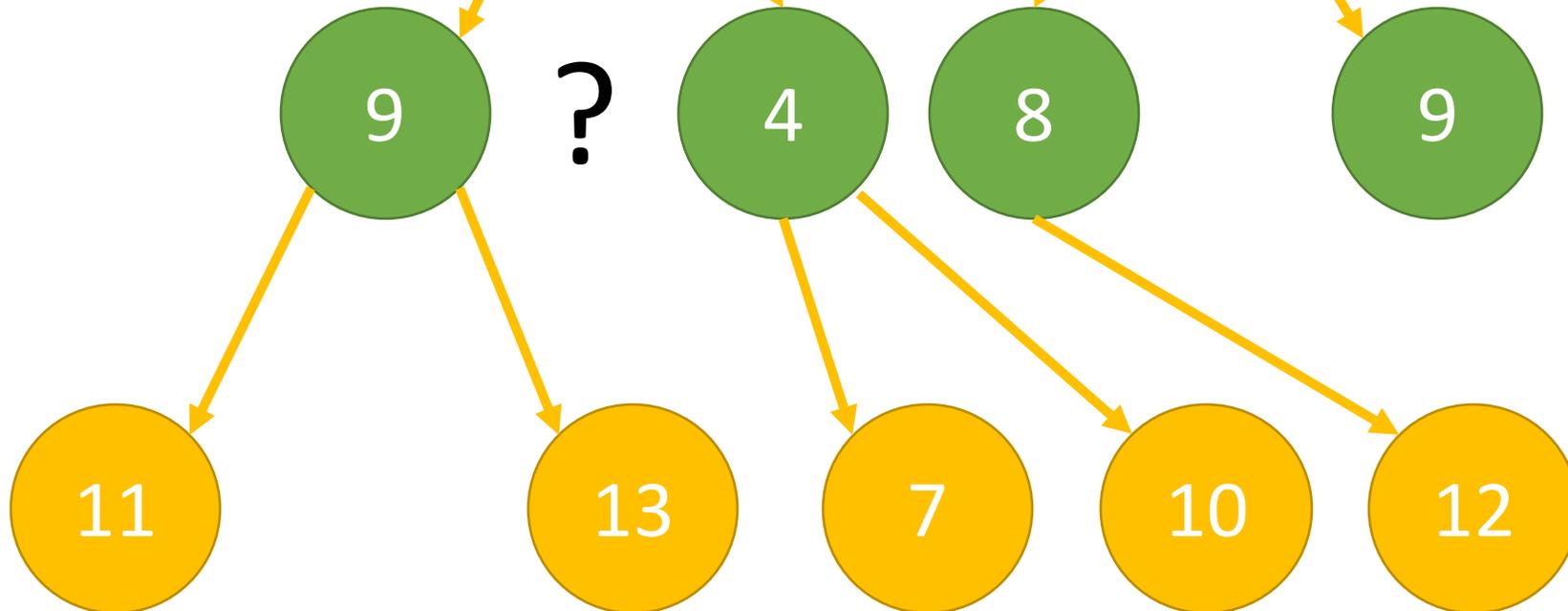
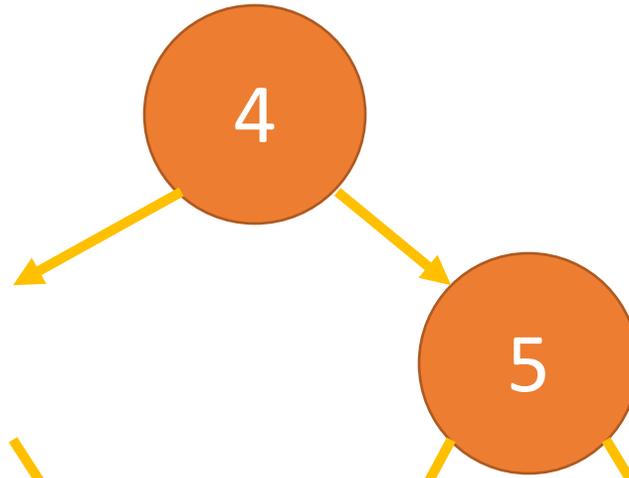
# Extract-Min

What if we choose a child?

What node do we put at the root in place of 4?



# Extract-Min



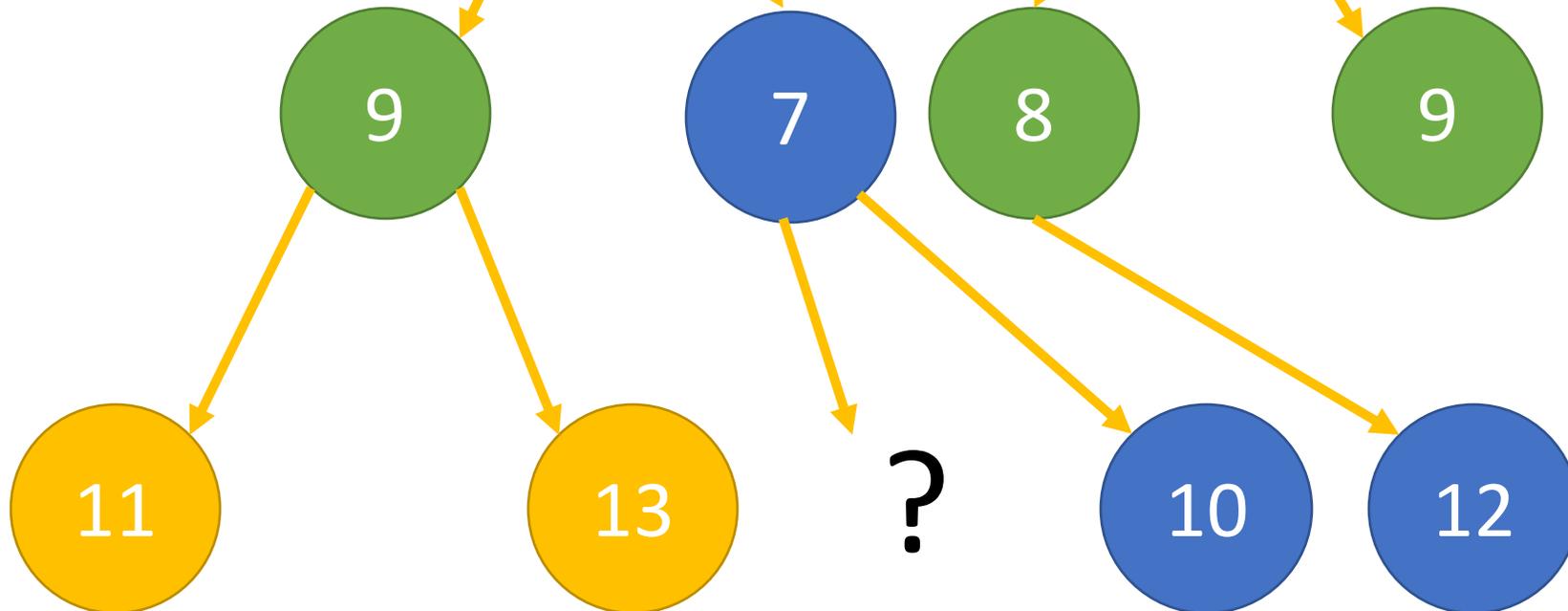
# Extract-Min



?

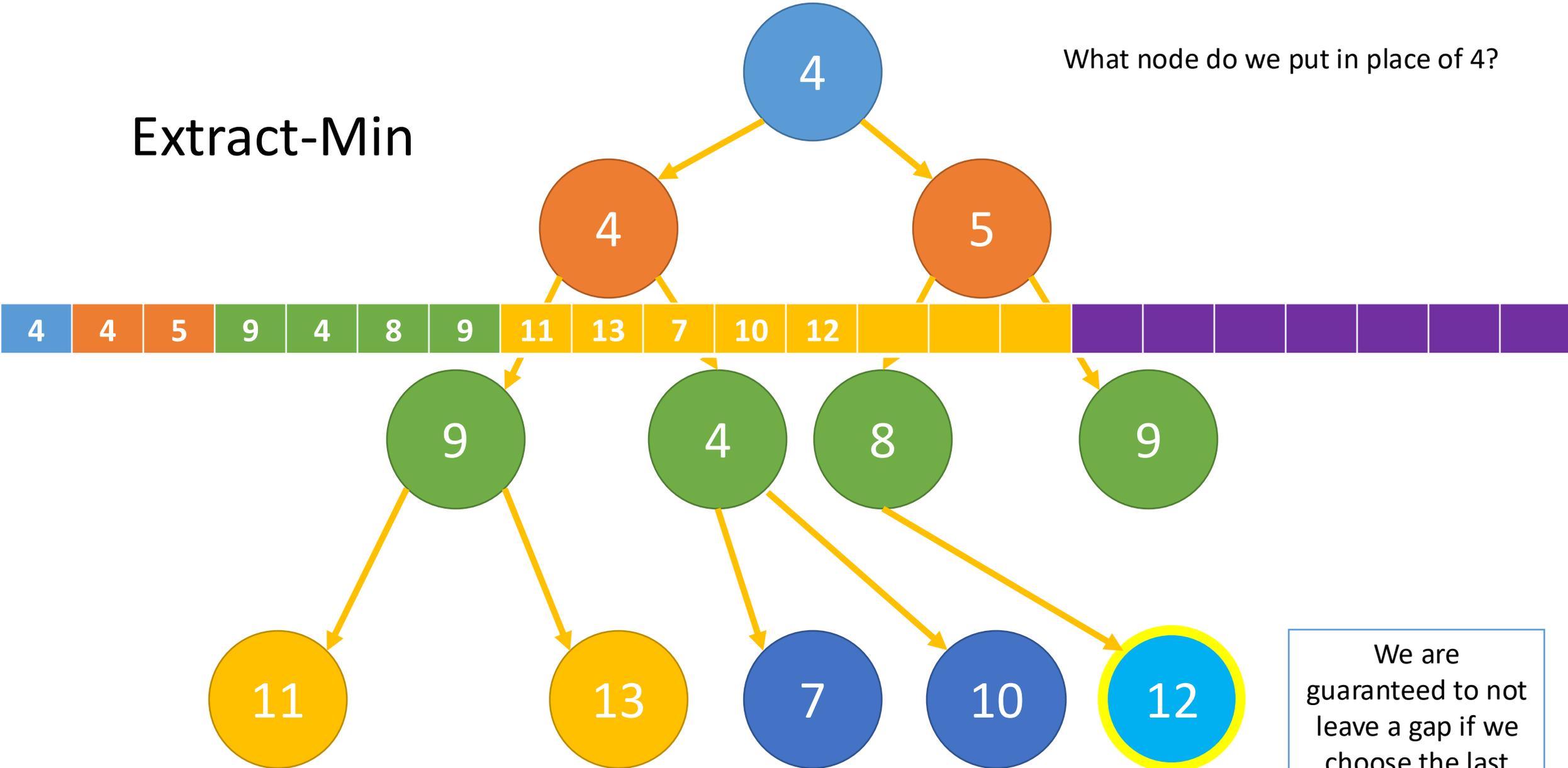


# Extract-Min



# Extract-Min

What node do we put in place of 4?

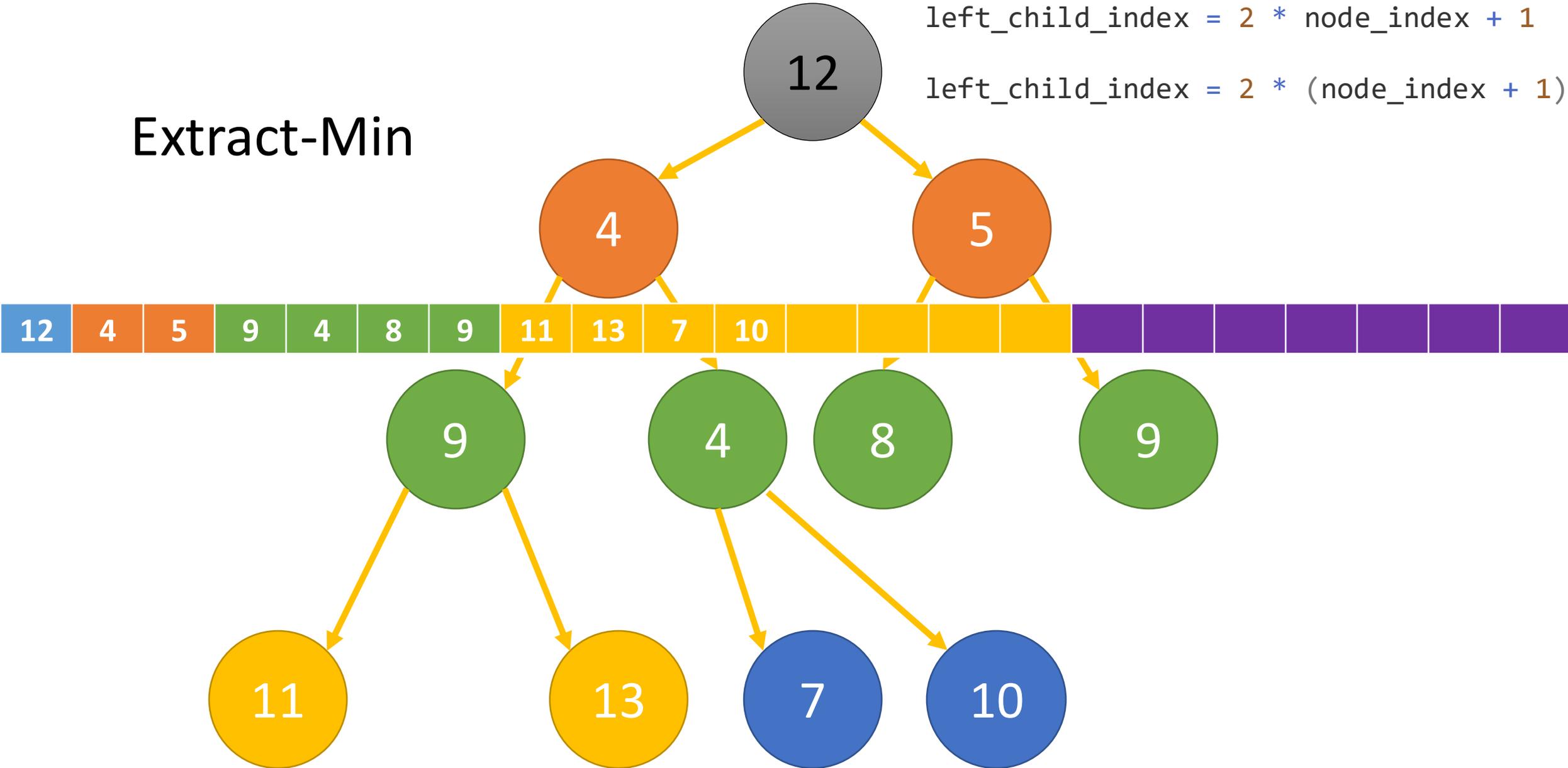


We are guaranteed to not leave a gap if we choose the last node.

# Extract-Min

$$\text{left\_child\_index} = 2 * \text{node\_index} + 1$$

$$\text{left\_child\_index} = 2 * (\text{node\_index} + 1)$$

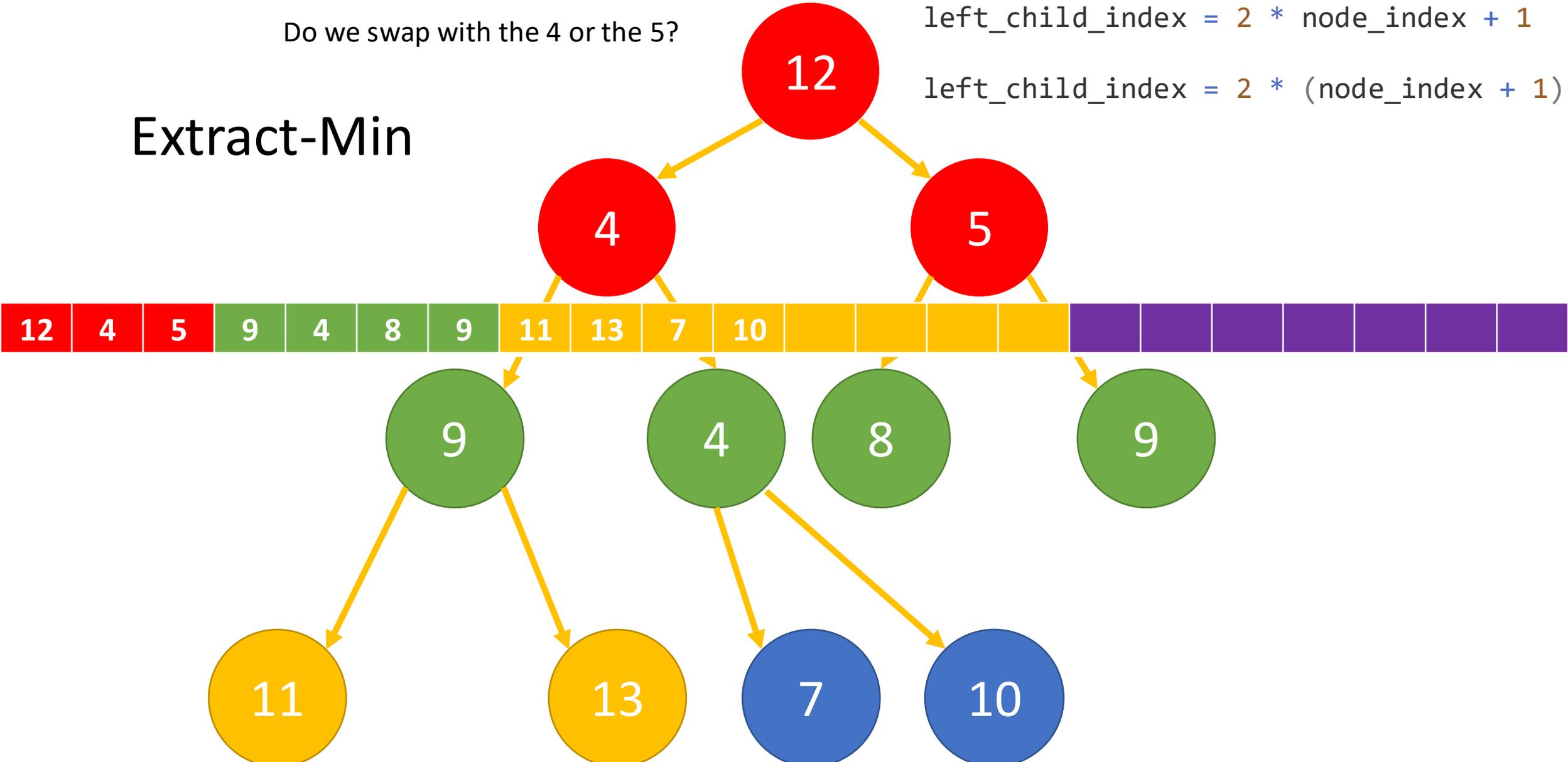


Do we swap with the 4 or the 5?

$$\text{left\_child\_index} = 2 * \text{node\_index} + 1$$

$$\text{left\_child\_index} = 2 * (\text{node\_index} + 1)$$

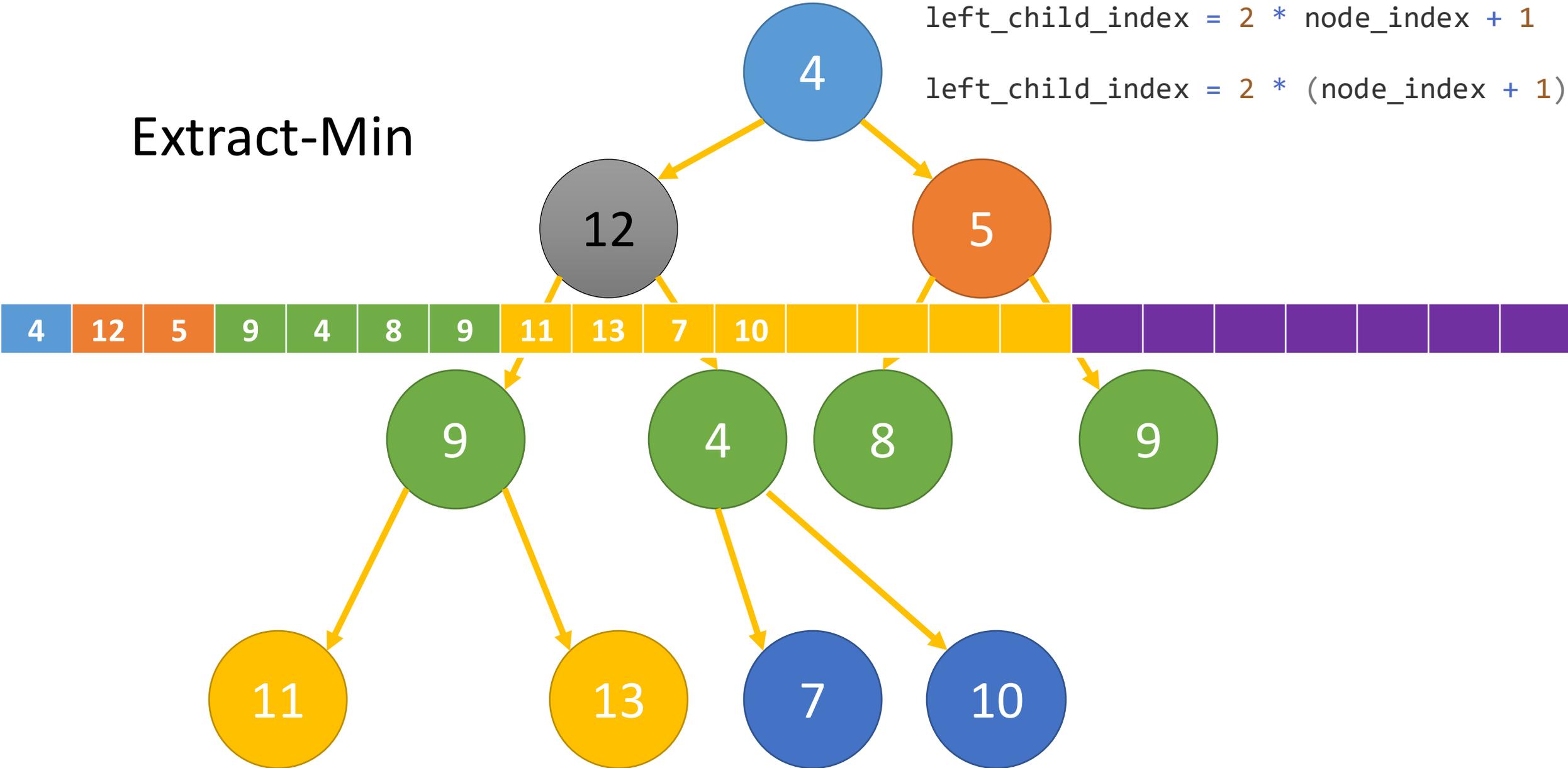
# Extract-Min



# Extract-Min

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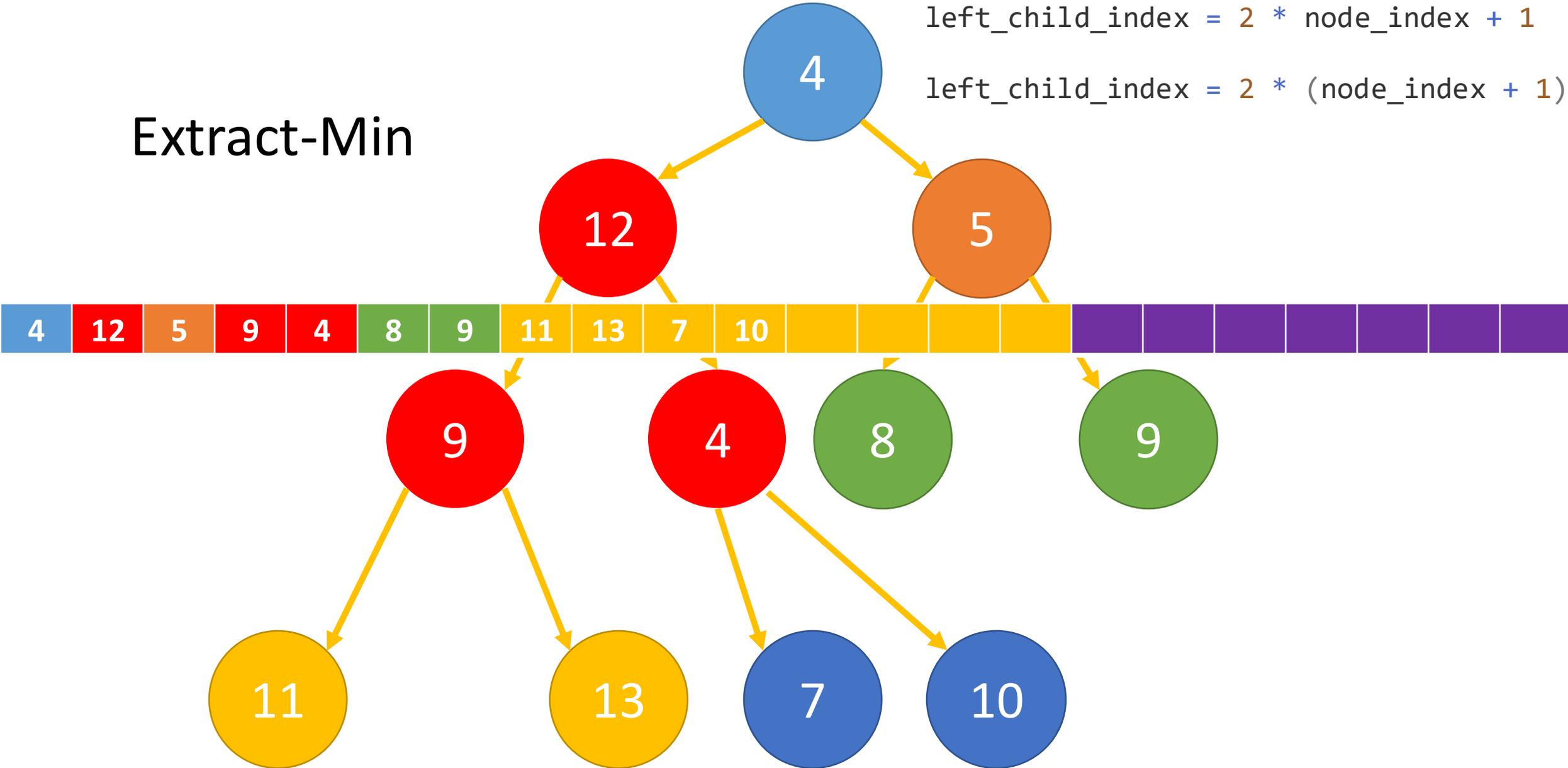
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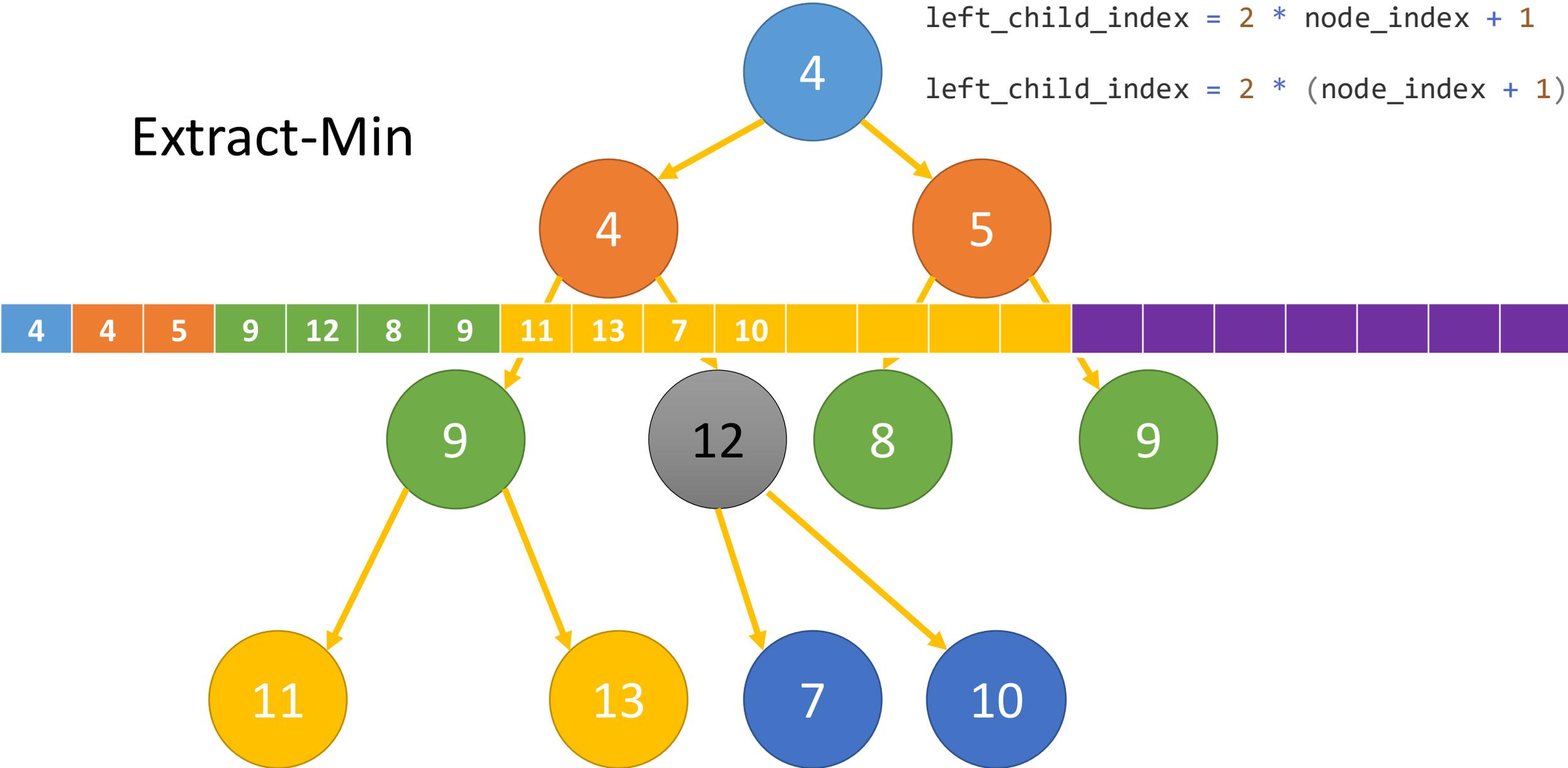
$$\text{left\_child\_index} = 2 * (\text{node\_index} + 1)$$



# Extract-Min

$$\text{left\_child\_index} = 2 * \text{node\_index} + 1$$

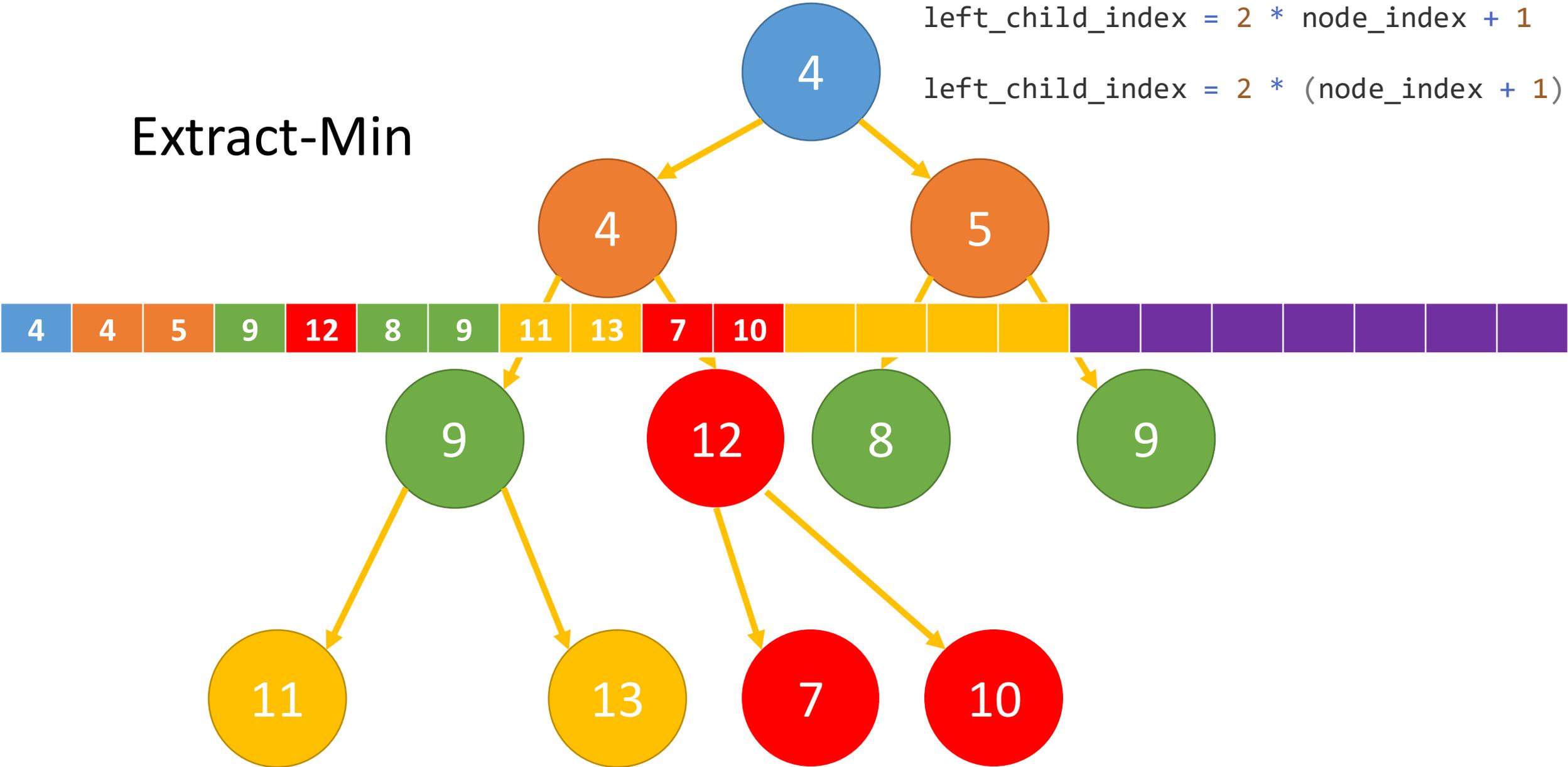
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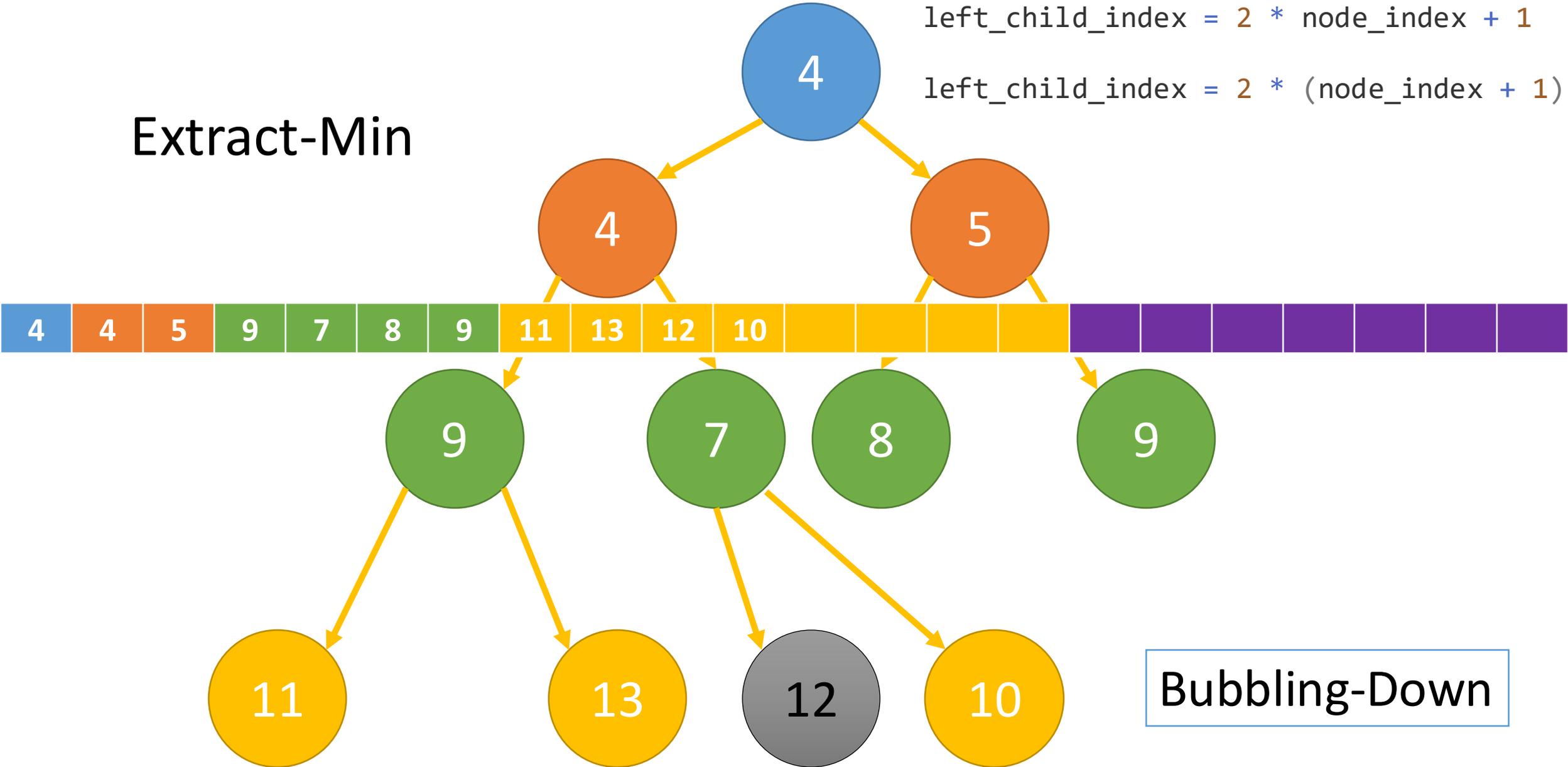
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# Extract-Min

$$\text{left\_child\_index} = 2 * \text{node\_index} + 1$$

$$\text{left\_child\_index} = 2 * (\text{node\_index} + 1)$$



```
FUNCTION Dijkstra(G, start_vertex)
```

```
found = {}
```

```
lengths = {v: INFINITY FOR v IN G.vertices}
```

```
found.add(start_vertex)
```

```
lengths[start_vertex] = 0
```

```
WHILE found.length != G.vertices.length
```

```
  FOR v IN found
```

```
    FOR vOther, weight IN G.edges[v]
```

```
      IF vOther NOT IN found
```

```
        vOther_length = lengths[v] + weight
```

```
        IF vOther_length < min_length
```

```
          min_length = vOther_length
```

```
          vMin = vOther
```

```
found.add(vMin)
```

```
lengths[vMin] = min_length
```

```
RETURN lengths
```

What is the running time?

How many times does the outer loop run?

$O(n)$

How many times do the inner two loops run?

$O(m)$

```

FUNCTION Dijkstra(G, start_vertex)
    found = {}
    lengths = {v: INFINITY FOR v IN G.vertices}

    found.add(start_vertex)
    lengths[start_vertex] = 0

    WHILE found.length != G.vertices.length
        FOR v IN found
            FOR vOther, weight IN G.edges[v]
                IF vOther NOT IN found
                    vOther_length = lengths[v] + weight
                    IF vOther_length < min_length
                        min_length = vOther_length
                        vMin = vOther
            found.add(vMin)
            lengths[vMin] = min_length

    RETURN lengths

```

What is the  
running time?

We can bring this  
down to  $O(m \lg m)$   
with a simple change.

State of the art of Dijkstra's:  
 $O(m + n \lg n)$   
(uses Fibonacci heap)

```

def dijkstras_heap(adjacency_list, start_vertex):
    """Dijkstra's Algorithm implemented with all vertices placed in a heap.

    This version of Dijkstra's Algorithm has a running time of  $O(m \lg m)$ .
    """

    n = len(adjacency_list)

    path_lengths = {v: inf for v in adjacency_list}
    predecessors = {v: None for v in adjacency_list}

    path_lengths[start_vertex] = 0
    predecessors[start_vertex] = None

    found = set()
    vertex_min_heap = [(path_lengths[start_vertex], start_vertex)]

    while len(found) != n:
        vfrom_length, vfrom = heappop(vertex_min_heap)
        found.add(vfrom)

        for vto, weight in adjacency_list[vfrom]:
            path_length = vfrom_length + weight
            if path_length < path_lengths[vto]:
                path_lengths[vto] = path_length
                predecessors[vto] = vfrom

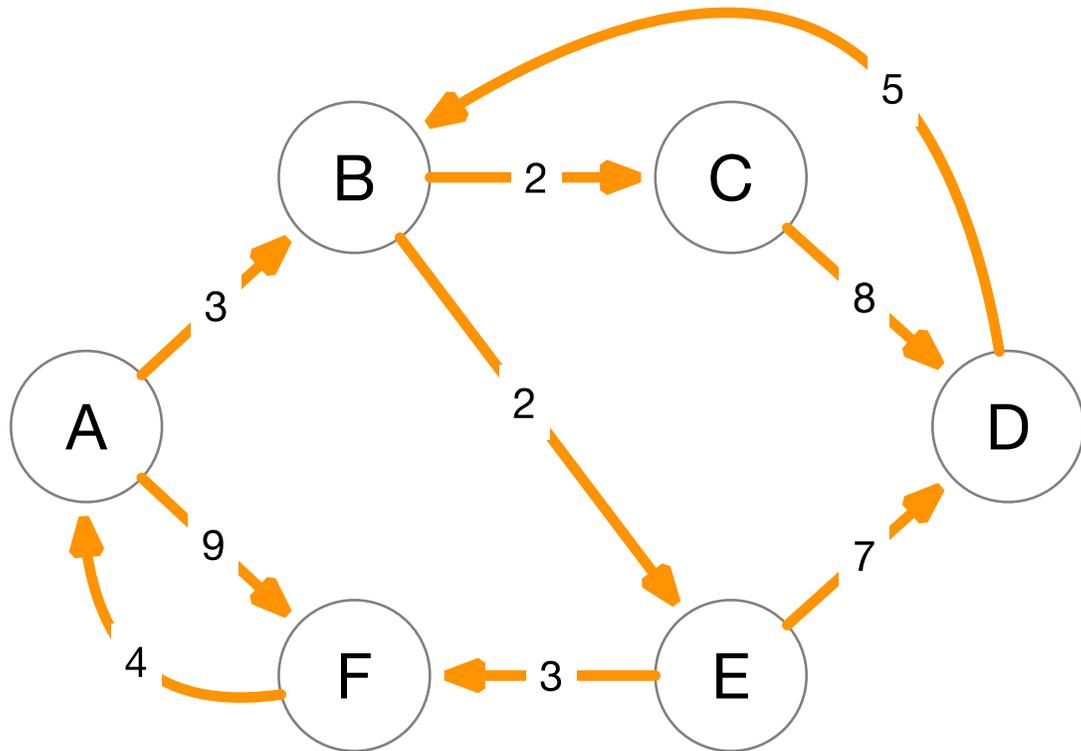
                heappush(vertex_min_heap, (path_lengths[vto], vto))

    return path_lengths, predecessors

```

Not optimal but works  
very well in practice.

```
while len(found) != n:  
    vfrom_length, vfrom = heappop(vertex_min_heap)  
    found.add(vfrom)  
  
    for vto, weight in adjacency_list[vfrom]:  
        path_length = vfrom_length + weight  
        if path_length < path_lengths[vto]:  
            path_lengths[vto] = path_length  
            predecessors[vto] = vfrom  
  
            heappush(vertex_min_heap, (path_lengths[vto], vto))
```



```
def print_path(end_vertex, predecessors):  
  
    path = [end_vertex]  
    pred = predecessors[end_vertex]  
  
    while pred is not None:  
  
        path.append(pred)  
        pred = predecessors[pred]  
  
    print(" -> ".join([str(v) for v in reversed(path)]))
```

# Dijkstra's Algorithm Correctness

## Theorem:

- Dijkstra's algorithm will find the shortest path from the start vertex to every other vertex on any graph with non-negative weights.

## Proof using a loop invariant. Loop predicate:

- At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

# Initialization

- Initially, the found set is empty.  
So, the invariant is trivially true.

Loop predicate/invariant: At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

```
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found = set()
...
while len(found) != n:
    vfrom_length, vfrom = heappop(vertex_min_heap)
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            heappush(vertex_min_heap,
                    (path_lengths[vto], vto))
```

# Maintenance (1)

- Assume all previous iterations have produced the correct shortest path for all vertices in the found set.
- For purposes of a contradiction, assume that when a vertex **u** is added to the found set its path length is **not** optimal.
- At the time **u** is found we must have some path to **u**

Loop predicate/invariant: At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

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while len(found) != n:
    vfrom_length, vfrom = heappop(vertex_min_heap)
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    for vto, weight in adjacency_list[vfrom]:
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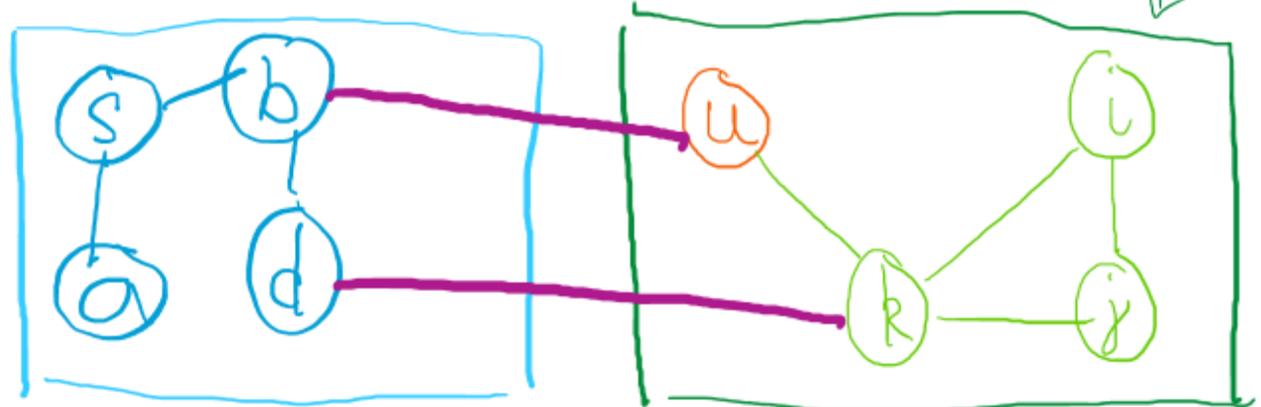
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found ↘

Loop predicate/invariant: At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

Not found ↙



```
while len(found) != n:
    vfrom_length, vfrom = heappop(vertex_min_heap)
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    for vto, weight in adjacency_list[vfrom]:
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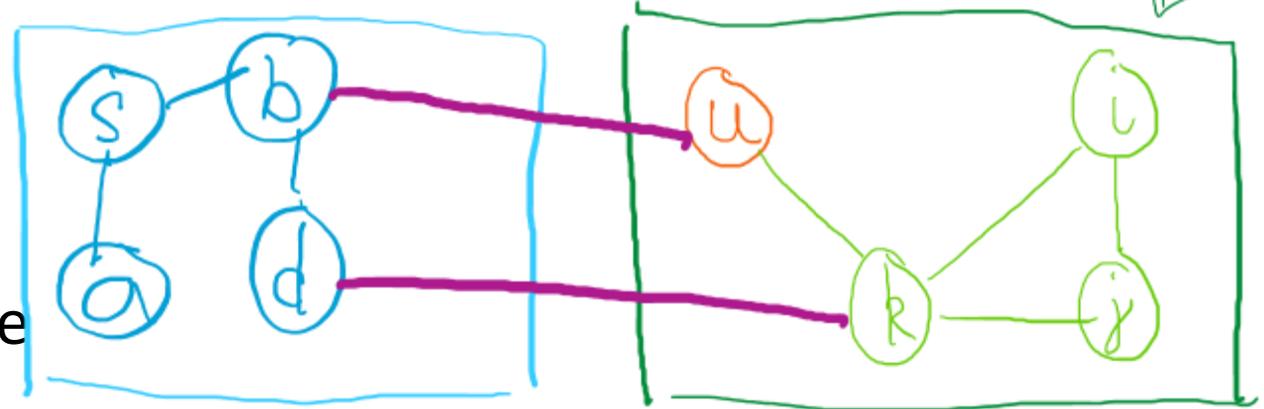
    heappush(vertex_min_heap,
             (path_lengths[vto], vto))
```

# Maintenance (1)

- For purposes of a contradiction, assume that when a vertex **u** is added to the found set its path length is **not** optimal.
- At the time **u** is found we must have some path to **u**
- To have a shorter path to **u**, it must go through some vertex **k** not in found.
- But since we only have positive edges, a shorter path going through **k**, means that **k** must have been chosen before **u**. **Contradiction.**

Loop predicate/invariant: At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

Not found



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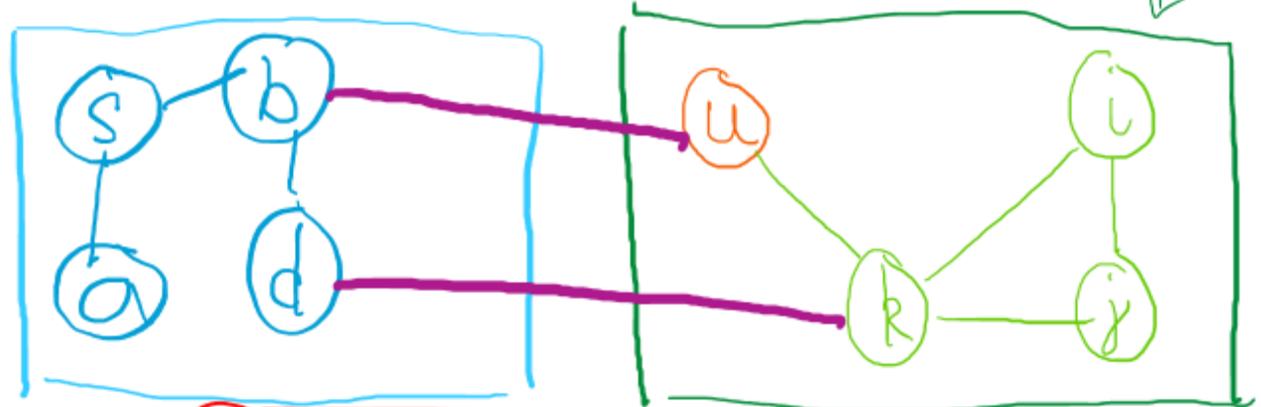
# Termination

- The loop terminates when all vertices have been added to the found set.
- Given the loop invariant the shortest path to all vertices have been calculated.

found ↘

Loop predicate/invariant: At the start of each iteration of the while loop, the shortest path has been found for every vertex in the found set

Not found ↙



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                (path_lengths[vto], vto))
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