

# Overloading in Java

```
int myMethod() { .. }
```

```
int myMethod(String s) { .. }
```

```
int myMethod(int[] arr) { .. }
```

```
Int x = myMethod();
```

```
Int y = myMethod("Dog");
```

```
Int arr[] = {1,2,3};
```

```
Int z = myMethod(arr);
```

# Overloading in Java

```
void inOrderTraversal() { .. }
```

```
void inOrderTraversal(DLinkedList dll) {  
    ..  
    this.getLeftChild().inOrderTraversal(dll);  
    ..  
}
```

```
RootNode.inOrderTraversal();
```

```
DLinkedList dll = new DLinkedList();
```

```
RootNode.inOrderTraversal(dll);
```

Software Development 3 (F27SG)

Lecture 17

# Efficient Sorting

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# Outline

- By the end of this lecture you should
  - know about the **Divide-and-Conquer** pattern
  - understand the  $n\text{-log-}n$  function,  $O(n \log n)$ , and analyse the difference with  $O(n)$
  - understand the  $O(n \log n)$  Divide-and-Conquer sorting algorithm called **Merge-Sort**
  - be able to compare the Insertion-Sort and Merge-Sort algorithms using Big-O
  - be familiar with the **Quick-Sort** algorithm

# Analysis of Insertion Sort

- **Insertion-Sort** has 2 nested loops
  - worst case each loop iterates over every element
  - thus it is  $O(n^2)$
- For ***search*** we saw that we could
  - **decrease** growth rate from  $O(n)$  to  $O(\log n)$ 
    - by **halving the problem in each step**
- Can a similar approach be applied to ***sorting***?

# The Divide and Conquer Pattern

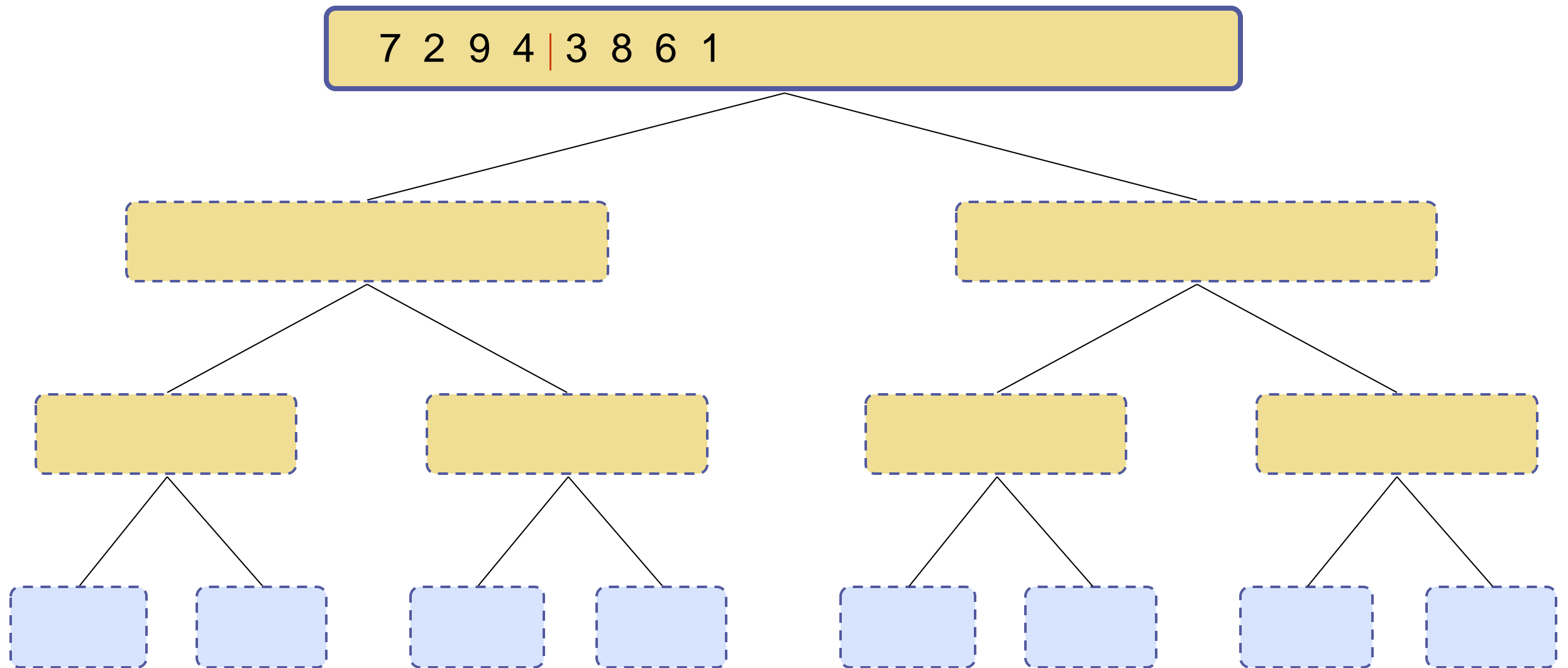
- The **Divide-and-Conquer** design pattern is recursively defined as follows:
  - **Base case:** If the size is smaller than a certain threshold then solve the problem in a straightforward manner
  - **Step case:** the step case consists of 3 steps:
    1. **Divide:** divide the data into 2 disjoint subsets
    2. **Recur:** solve the subproblems
    3. **Conquer:** combine the solutions of the subproblems into a solution of the main problem

# Merge-Sort

- **Merge-Sort is a Divide-and-Conquer algorithm:**
  - **Base case:** if size is less than 1 return
    - or use e.g. *Insertion-Sort* when the list is “small” (e.g.  $< 10$ )
  - **Step case:**
    1. split input list into two equal halves
    2. *recursively* sort left half, *recursively* sort right half
    3. merge two sorted halves into original list
      - while both lists have any items
        - » compare first element of left and right
        - » delete smallest and insert it into first free place of original list
      - one of the lists will still have elements left
        - » move them to the end of the original list

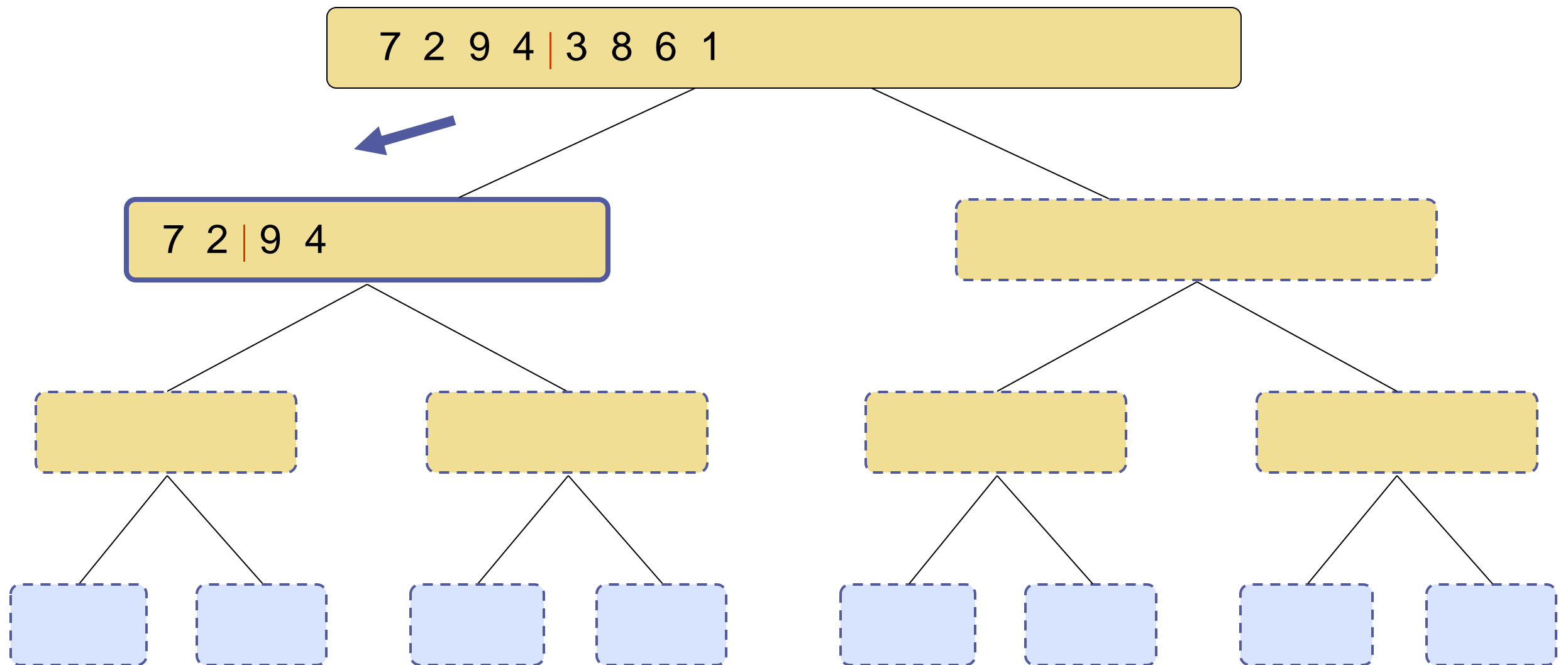
# Merge-Sort Example (1)

- Here, we will sort the list **[7,2,9,4,3,8,6,1]** :

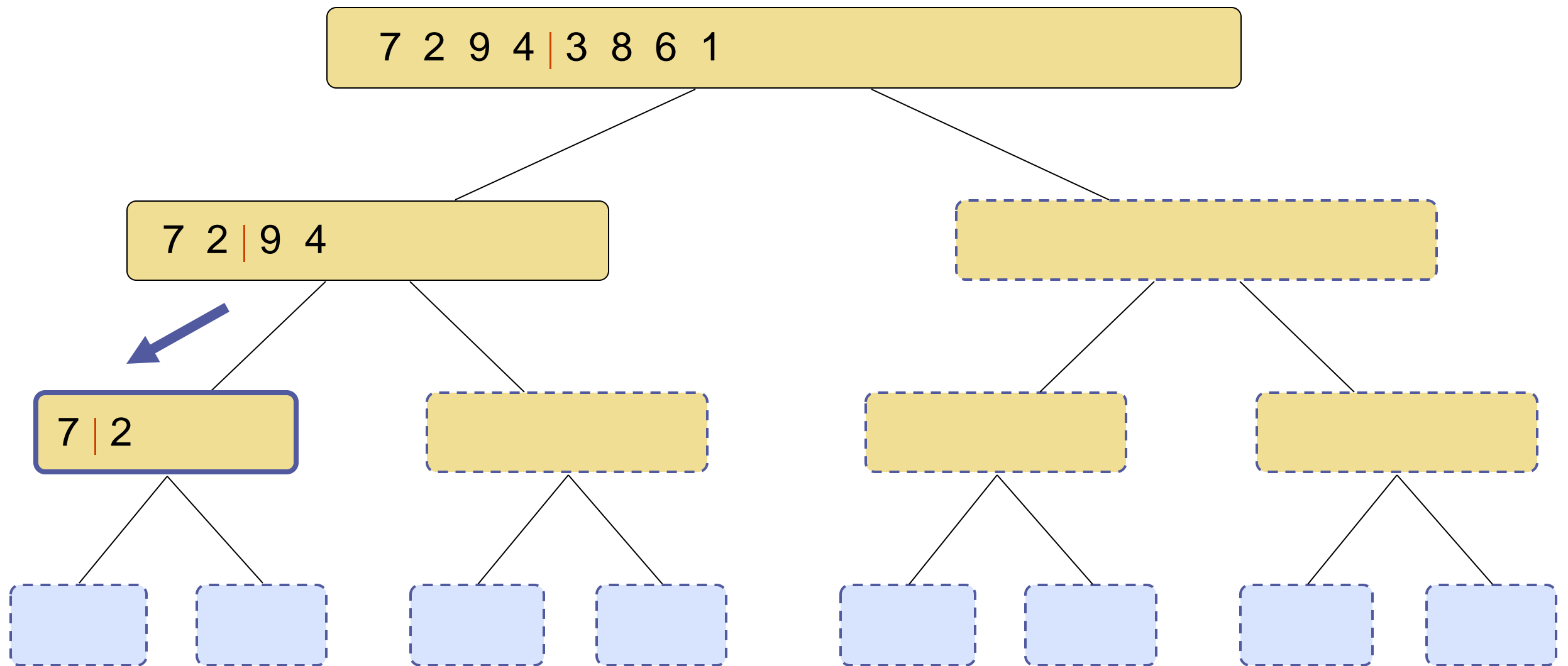




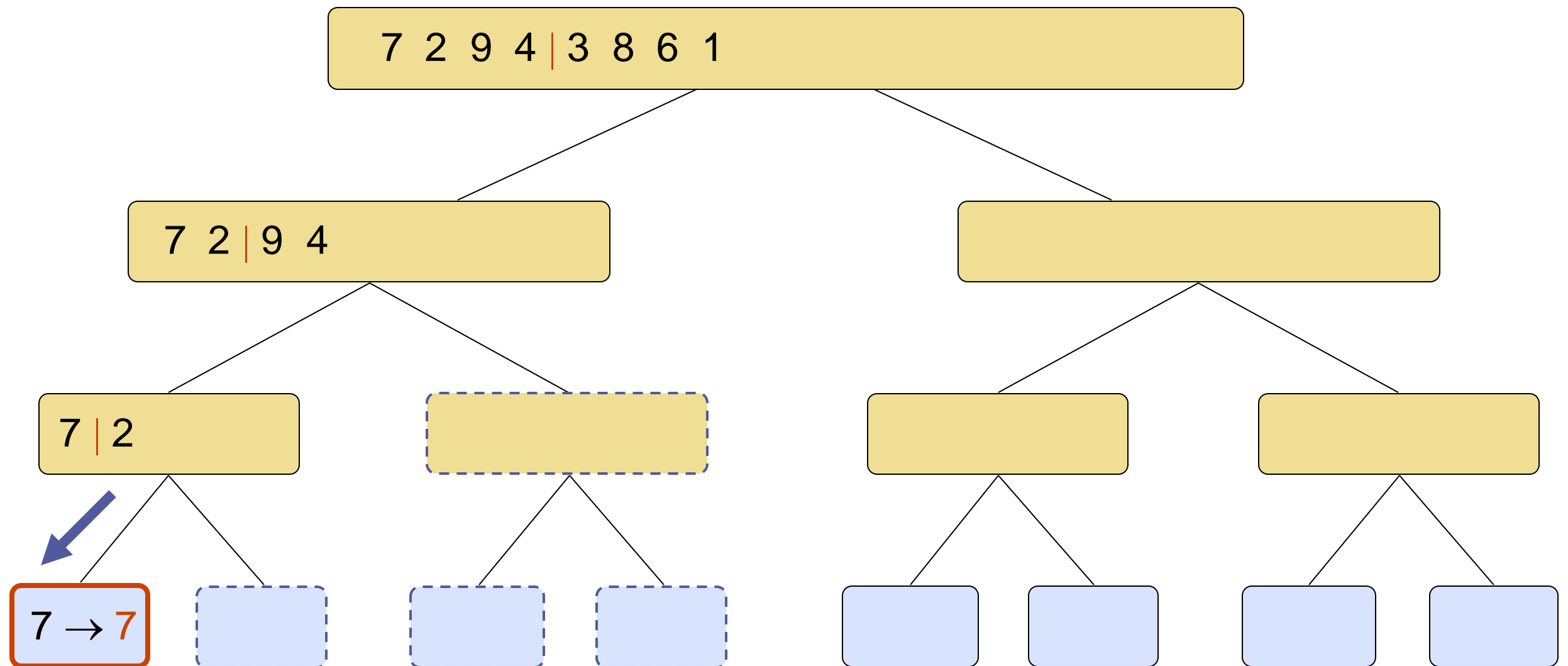
# Merge-Sort Example (2)



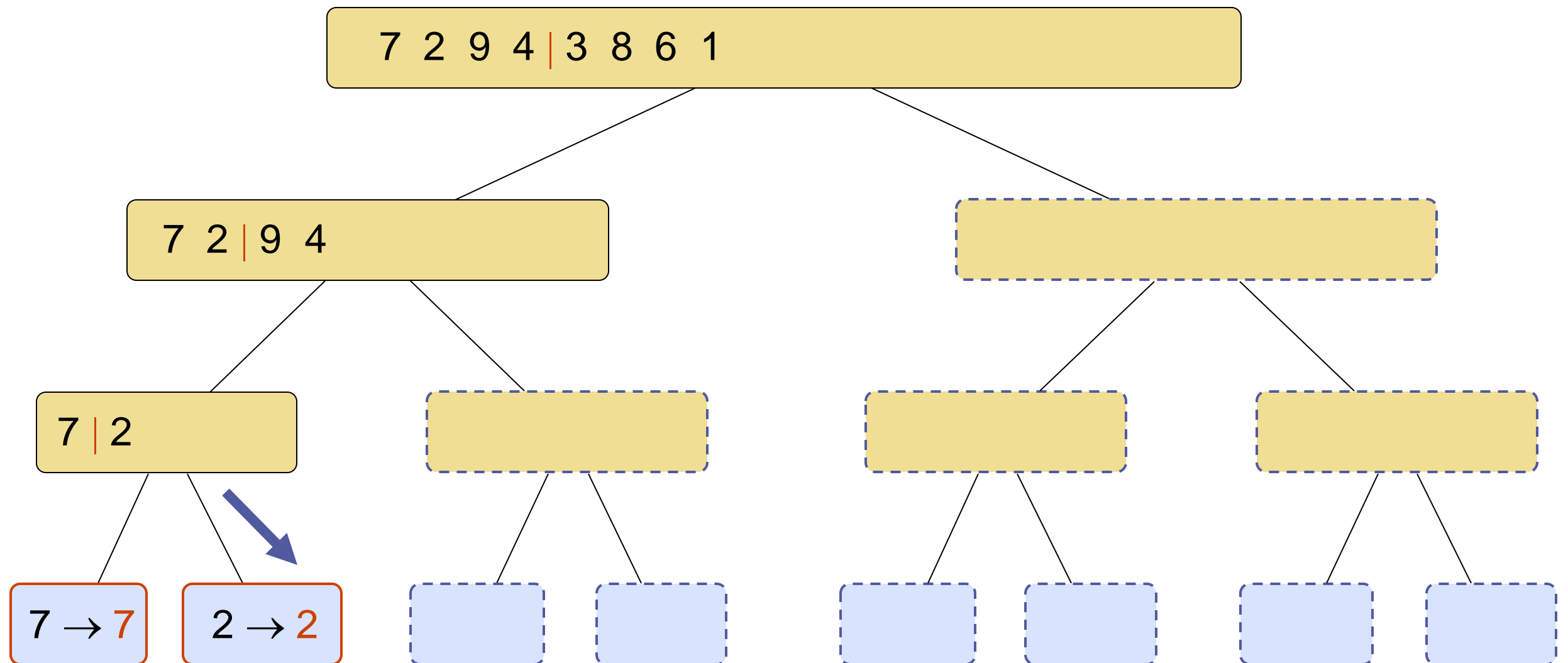
# Merge-Sort Example (3)



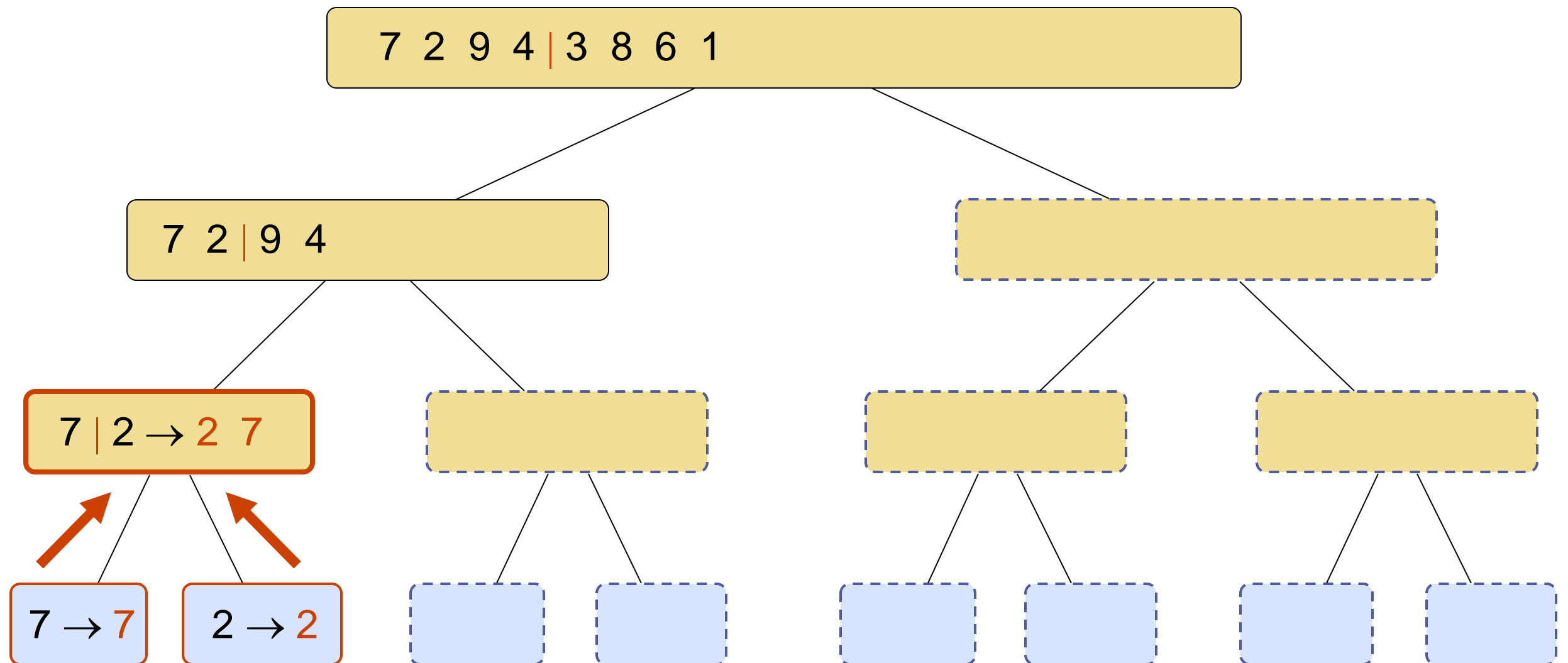
# Merge-Sort Example (4)



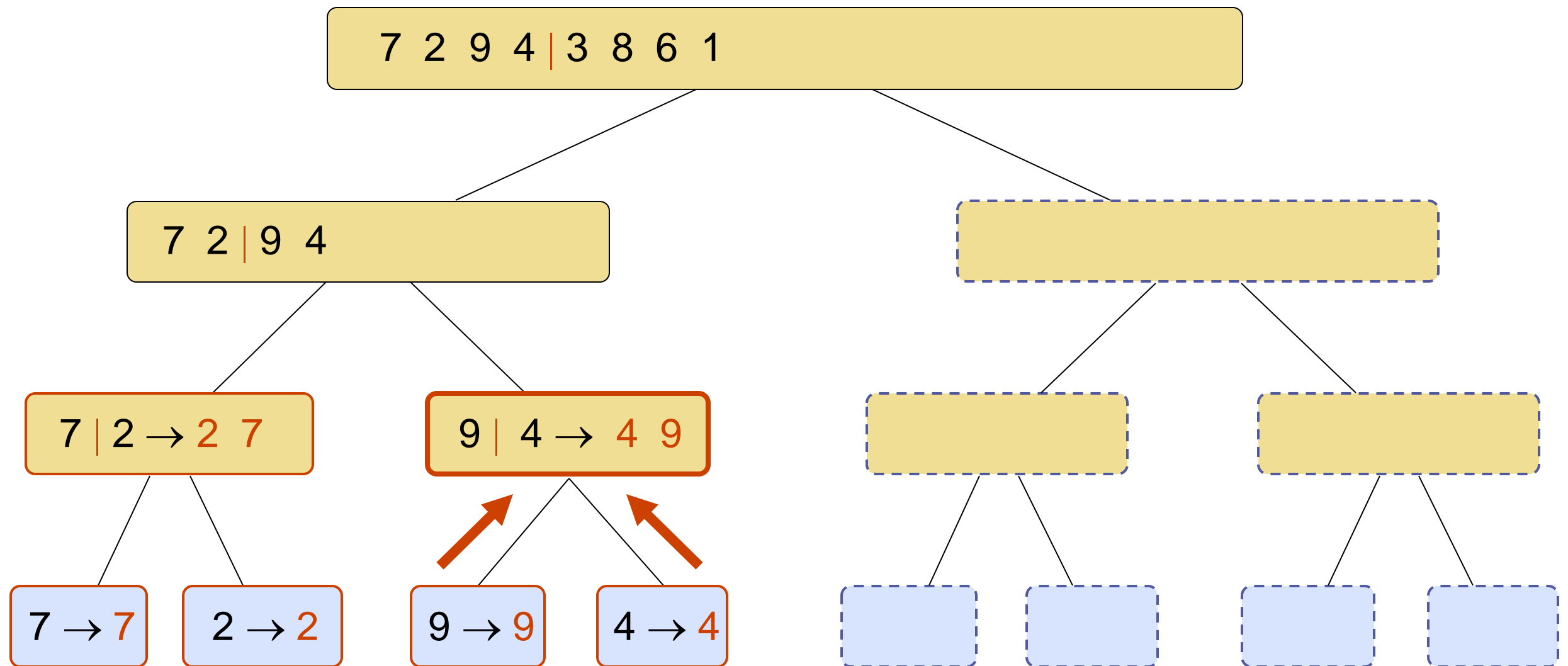
# Merge-Sort Example (5)



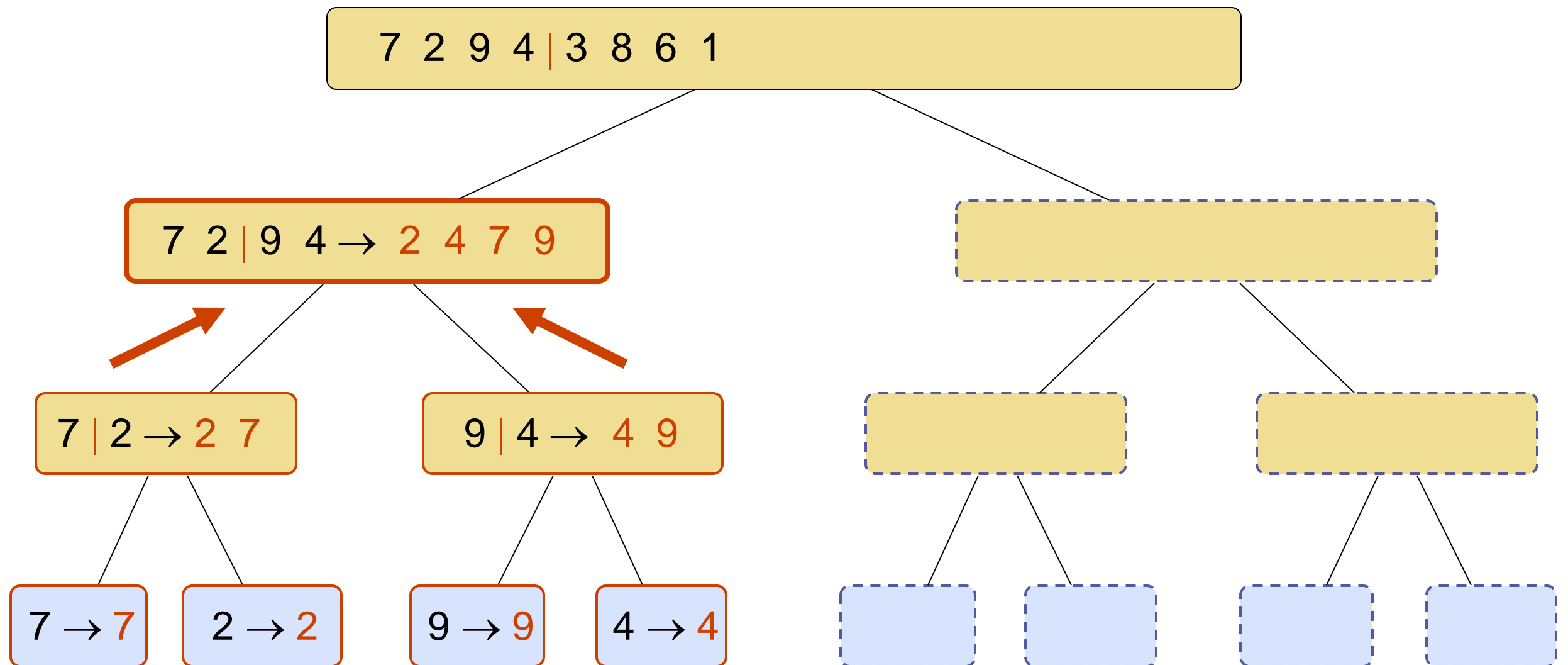
# Merge-Sort Example (6)



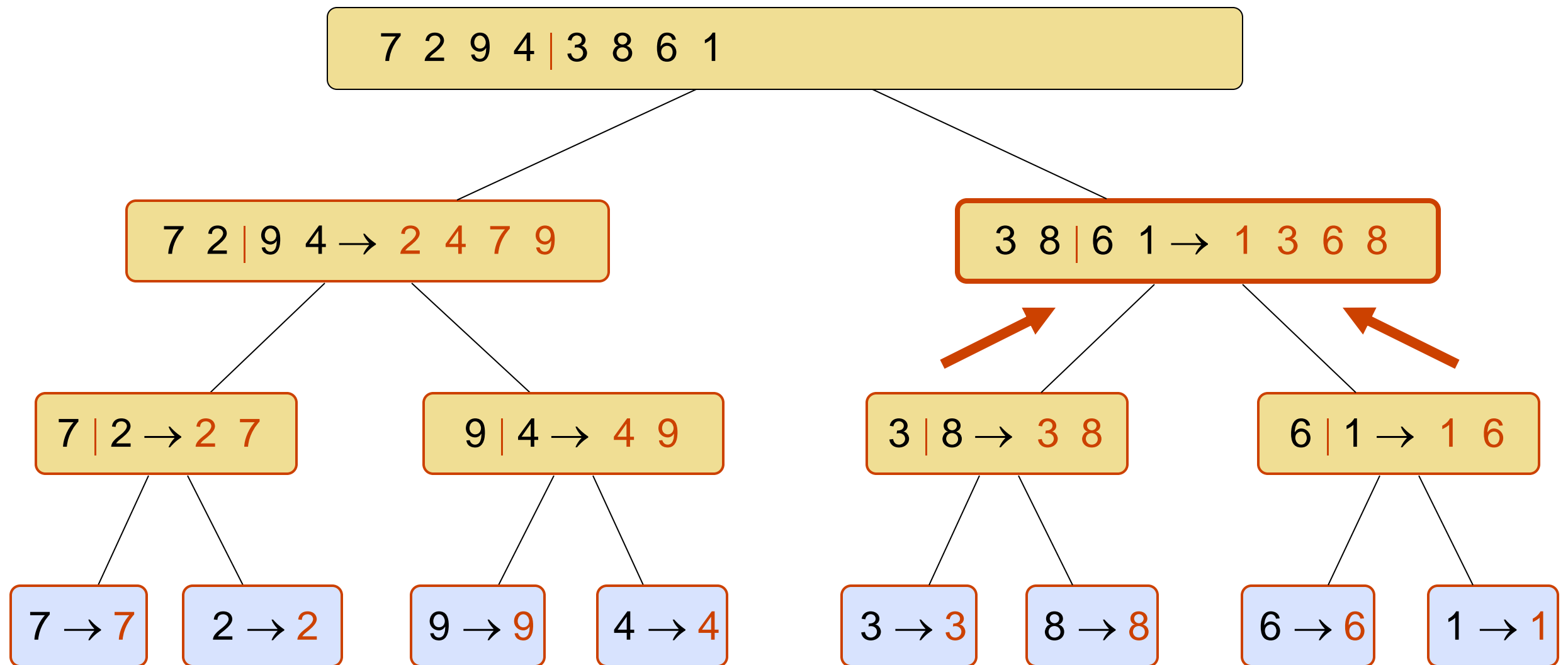
# Merge-Sort Example (7)



# Merge-Sort Example (8)

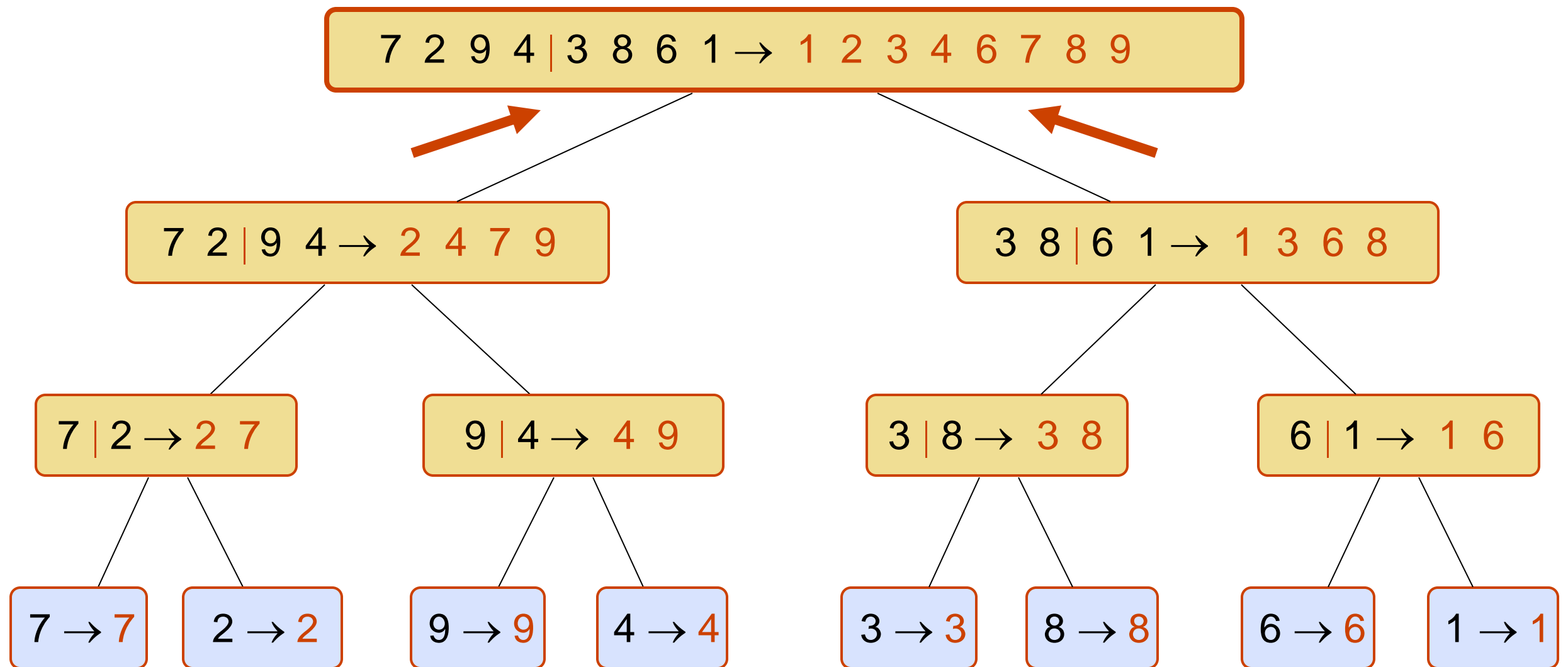


# Merge-Sort Example (9)





# Merge-Sort Example (10)





# Exercise

**Base case:** if the size is less than 1 return

**Step case:**

1. split the input list into two equal halves
2. *recursively* sort the left half  
    *recursively* sort the right half
3. merge the two sorted halves into the original list
  - while both lists have any items
    - » compare the first element of left and right
    - » delete the smallest and insert it into first free place of original list
  - one of the lists will still have elements left
    - » move them to the end of the original list

Use **Merge-Sort** to sort the list {4,3,6,5,2,1}. Show

- the recursion trace,
- and each step of the merge operation.
- To get you started:
  - we start with mergeSort({4,3,6,5,2,1})
  - halfSize = 3, so two new lists are created
    - left = {4,3,6} and right = {5,2,1}
  - we then call mergeSort(left) and mergeSort(right) ...

# The Merge Sort

Eclipse demonstration

# Merge-Sort in Java (1)

- We will now implement Merge-Sort for a list of integers.
- What are the arguments, return value, and base case(s) of the method?

```
public static void mergeSort(int[] list){  
    int length = list.length;  
    if (length < 2) // list of 1 or 0 elements is sorted  
        return;
```

# Merge-Sort in Java (2)

- In the step case we
  - divide the list into two halves
  - recursively sort the left side and the right side
  - then merge the result into the original list
- There are several ways we can implement this
  1. we can sort the list “in place” and create a new temp array we merge into, or
  2. we can create two temp arrays for left and right and use the actual array to merge
- We will follow the 2nd approach. First we create the arrays:

```
int halfSize = length/2;  
int[] left = new int[halfSize];  
int[] right = new int[length-halfSize];
```

# Merge-Sort in Java (3)

- Then we copy the elements to the two new arrays
- We call this **partitioning** the list

```
int index = 0;
while(index < halfSize){
    left[index] = list[index];
    index++;
}
index = 0;
while(index < length-halfSize){
    right[index] = list[index+halfSize];
    index++;
}
```

# Merge-Sort in Java (4)

- Finally we recursively sort each half
- ... and merge the results

```
mergeSort(left);  
mergeSort(right);  
merge(left,right,list);  
}
```

# The Merge Method (1)

- The merge method takes two sorted arrays
  - the left side, and
  - the right side
- ... and *merges* the into a third array (given as an argument):

```
public static void merge(int[] left,int[] right,int[] list){  
    int index = 0; // index of list  
    int lindex = 0; // index of left  
    int rindex = 0; // index of right
```



# The Merge Method (1)

Eclipse demonstration

# The Merge Method (2)

- To merge the arrays we use the fact that both left and right is sorted
- We iterate through the array and compare the “current” element of each list
  - and pick the smallest
- When we reach the end of one array we stop

```
while(index < left.length && rindex < right.length){  
    if (left[lindex] <= right[rindex]) // the “current” left is smallest  
        list[index++] = left[lindex++];  
    else // the “current” right is smallest  
        list[index++] = right[rindex++];  
}
```

# The Merge Method (3)

- At the end
  - one of the arrays will still have elements left
  - these elements are then added to the end of the merged array

```
while(lindex < left.length)
    list[index++] = left[lindex++];
while(rindex < right.length)
    list[index++] = right[rindex++];
}
```

# Analysis of MergeSort

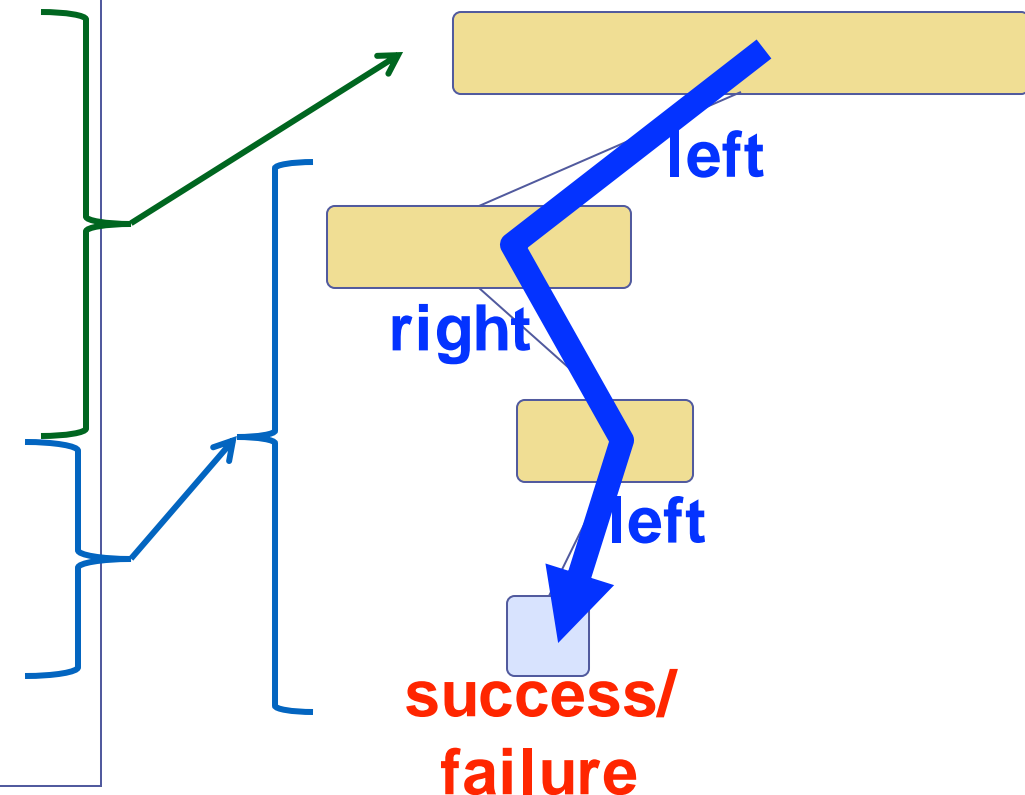
- Complexity of MergeSort
  - Similar to complexity of Binary Search
- So lets revisit the complexity of Binary Search...

# Analysis of Binary Search

(revisited)

- We can think of the complexity as
  - Complexity of current step  
+ complexity of the recursive calls
  - For current step it is constant
  - Each step halves the size of array
  - Hence complexity **is the number of recursive calls**

```
public int binarySearch (int key, int first, int last)
throws NotFoundException {
    if (first > last)
        throw new NotFoundException ("not found");
    else {
        int middle = (first + last) / 2;
        if (key == arr[middle])
            return arr[middle];
        else if (key < arr[middle])
            return binarySearch (key, first, middle-1);
        else
            return binarySearch (key, middle+1, last);
    }
}
```



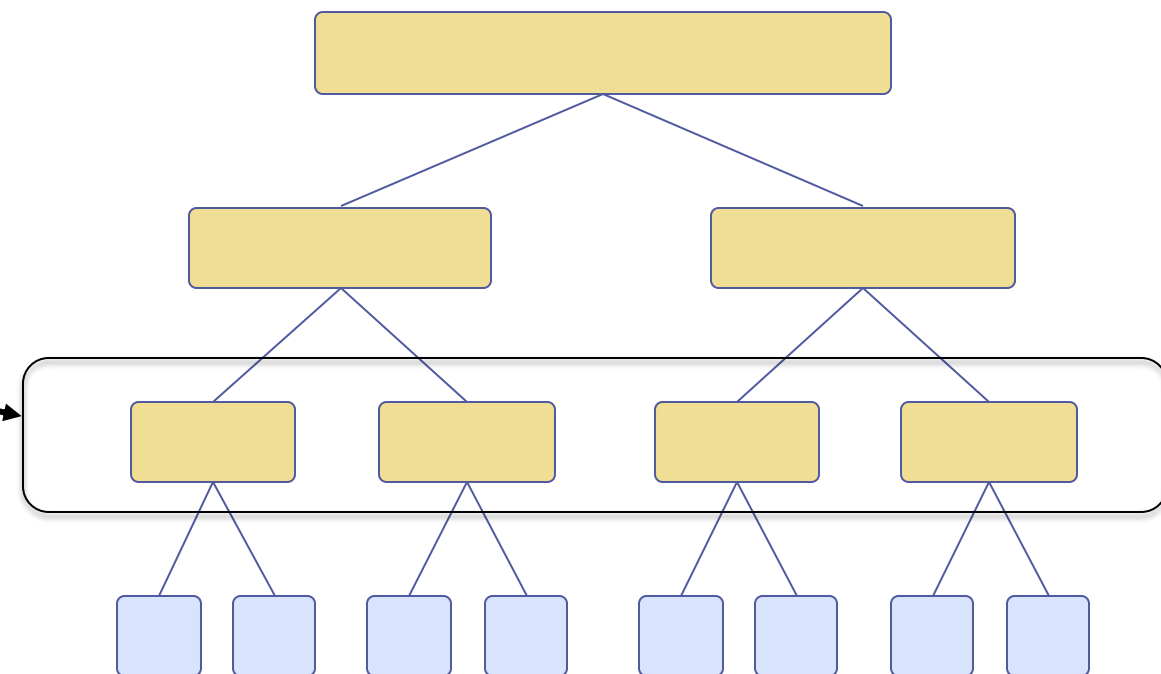
# Binary Search has $\log(N)$ Growth

- If we double input size, the **logarithm** increases by 1, e.g.
  - **$\log 8 = 3$**  (since  $2^3 = 8$ )
  - **$\log 16 = 4$**  (since  $2^4 = 16$ )
- Similar if we double size of array for **binary search**
  - we make one more recursive call
  - ... ***list is halved for each call***
  - E.g. array of size 16
    - will result in a call to one of the half
    - ... which is of size 8
  - Thus the **height of the tree is  $\log(N)$**
  - **Worst case: thing searching for is a leaf**
- Thus **binary search is worst case  $O(\log N)$**

# Analysis of Merge-Sort (1)

- Next we observe some properties **for each *depth* in tree**
  - each node has two child nodes
  - list size of child node is half the size of parent's list
- As a consequence, each **depth  $i$** 
  - has  **$2^i$  nodes**
  - the size of the list for each node is  **$N/2^i$**

- E.g. sorting 8 elements, at depth 2
- Nodes =  $2^2 = 4$
- Size of each list =  $8/4 = 2$



# Analysis of Merge-Sort (2)

- Now analyse the time spent for each node at depth  $i$ 
  - includes
    - *divide* (**partition**)
    - *conquer* (**merge**)
- **Partition** iterates list at a *linear time*  $O(N)$ 
  - **while**(index < halfSize){ .. }
  - **while**(index < length-halfSize){ .. }
- **Merge** two sub lists also iterates list at a *linear time*  $O(N)$ 
  - **while**(lindex < left.length && rindex < right.length){ .. }
- $N + N = 2N$ , remove constant
  - Therefore, partition + merge =  **$O(N)$**  ( $N$  is size of array)
- Since ***size of list*** at this depth is  $N/2^i$ , time spent on a node is  **$O(N/2^i)$**



# Analysis of Merge-Sort (3)

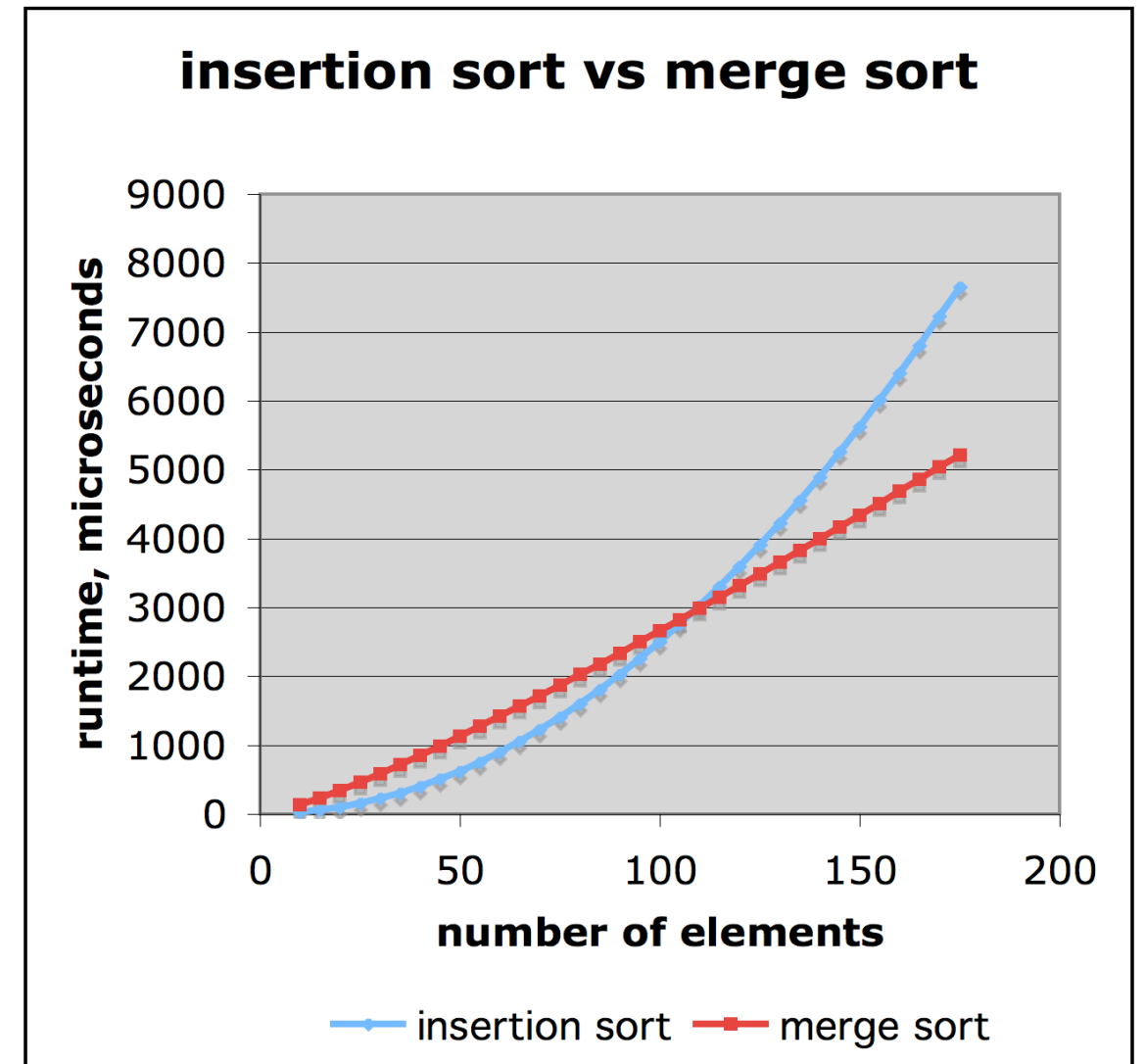
- Total time spent:  
“time spent at each depth” times “depth of tree”
- **Time spent**
  - Time for each node is  $O(N/2^i)$
  - At each depth there are  $2^i$  nodes
  - at depth  $i$  total time is  $O(2^i \cdot N/2^i)$ 
    - $2^i \times N/2^i = 2^i/1 \times N/2^i = (2^i \times N)/(2^i \times 1) = (2^i \times N)/2^i = N$
  - Hence time spent at each depth is  $O(N)$
- The **depth of the tree** is  $O(\log N)$
- Meaning  $O(N) \cdot O(\log N) = O(N \log N)$

# The N-log-N function

- The final function we will see is the *N-log-N* function:  
$$f(n) = n \log n$$
which assigns to input  $n$ 
  - the value of  $n$  times the (base-two) logarithm of  $n$ .
- The growth rate is
  - much better than the *quadratic* function
  - and slightly higher than the *linear* function

# Insertion-Sort vs Merge-Sort

- Insertion-Sort
  - growth rate of  $O(n^2)$
  - performs better than merge-sort when collection
    - is almost sorted, or
    - is small
- Merge-Sort
  - growth rate of  $O(n \log n)$
  - outperforms Insertion-Sort on larger collections



# Quick-Sort

- **Quick-Sort** is another *Divide-and-Conquer* algorithm
- We will work on this in the lab, and will only give a high-level description here for a list  $S$ 
  1. **Divide**: select an element  $x$  from  $S$  which is called the **pivot** (often the last element of  $S$ ).  
Divide  $S$  into 3 sub-lists:
    - $L$  storing elements of  $S$  less than  $x$
    - $E$  storing elements of  $S$  equal to  $x$
    - $G$  storing elements of  $S$  larger than  $x$
  2. **Recur**: Recursively sort  $L$  and  $G$
  3. **Conquer**: Put back elements in the order of first elements of  $L$ , then elements of  $E$ , then elements of  $G$ .



# Exercise

- 1. Divide:** select an element  $x$  from  $S$  which is called the **pivot** (often the last element of  $S$ ).  
Divide  $S$  into 3 sub-lists:
  - $L$  storing elements of  $S$  less than  $x$
  - $E$  storing elements of  $S$  equal to  $x$
  - $G$  storing elements of  $S$  larger than  $x$
- 2. Recur:** Recursively sort  $L$  and  $G$
- 3. Conquer:** Put back elements in the order of first elements of  $L$ , then elements of  $E$ , then elements of  $G$ .

Use **Quick-Sort** to sort the list  $\{4,3,6,5,2,1\}$

- You can choose the pivot yourself
- explain the values of  $L, E, G$  for each recursive call
- To get you started (assuming first element is the pivot):
  - we start with  $\text{quickSort}(\{4,3,6,5,2,1\})$  and  $\text{pivot} = 4$
  - we get the lists  $L = \{3,2,1\}$ ,  $E = \{4\}$  and  $G = \{6,5\}$ .
  - we then recursively call  $\text{quickSort}(L)$  and  $\text{quickSort}(G)$  ...

# Summary

- You have learned about an efficient sorting algorithm called **Merge-Sort**
- You know  $O(N \log N)$  and understand the time complexity of Merge-Sort
- You know how Merge-Sort and Insertion-Sort compares
  - and how to use Big-O to compare them
- You are familiar with **Quick-Sort**
- Visualisation of Bubble-Sort and Quick-Sort:
  - <https://www.youtube.com/watch?v=vxENKlcs2Tw>