

**INFO1105/1905**

**Data Structures**

**Week 11: Sorting**

see textbook section 12.1, 12.2, 12.4

Professor Alan Fekete

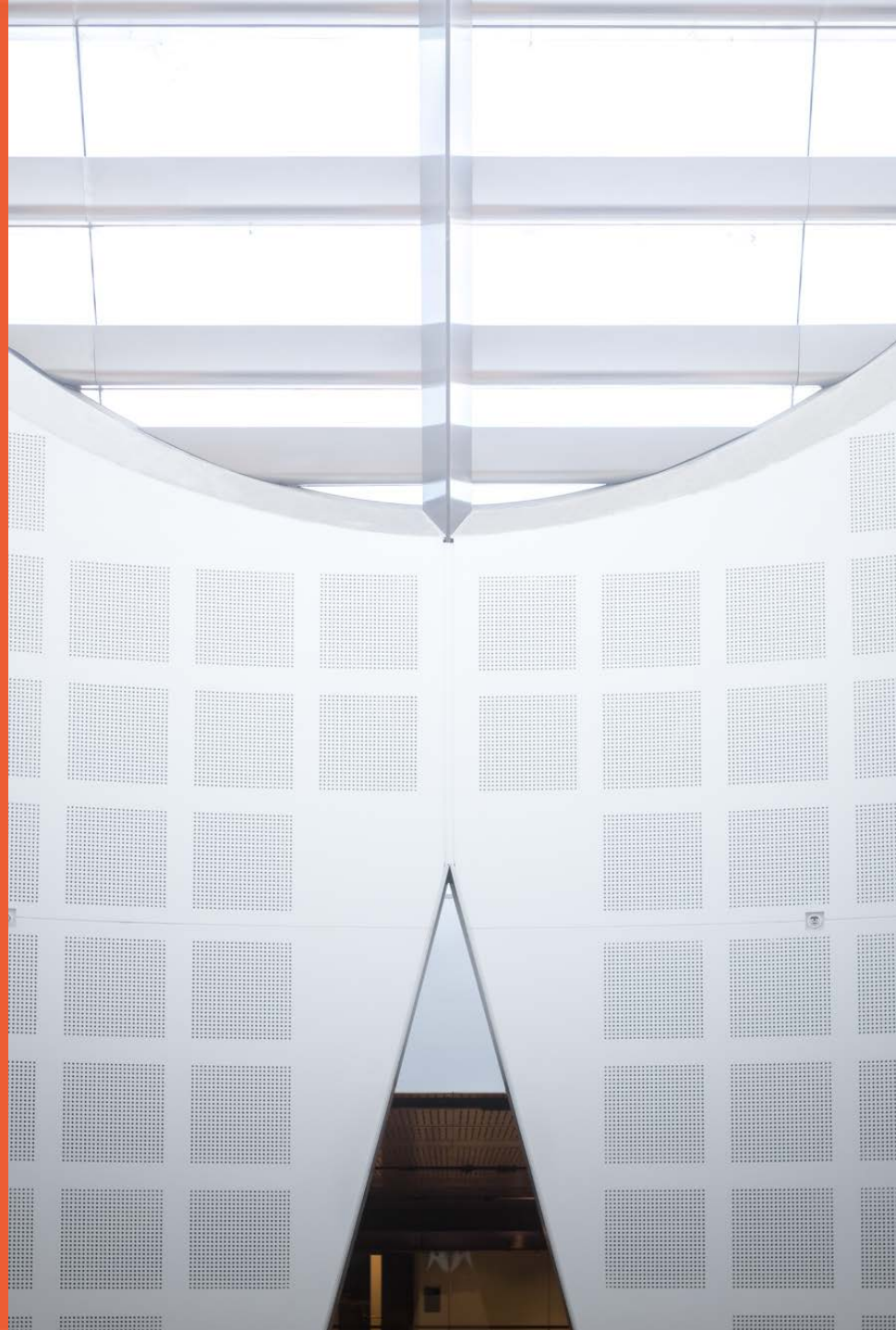
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using material from the textbook  
and A/Prof Kalina Yacef, Dr Taso Viglas



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- These slides contain material from the textbook (Goodrich, Tamassia & Goldwasser)
  - Data structures and algorithms in Java (5<sup>th</sup> & 6<sup>th</sup> edition)
- With modifications and additions from the University of Sydney
- The slides are a guide or overview of some big ideas
  - Students are responsible for knowing what is in the referenced sections of the textbook, not just what is in the slides

## Reminder! Quiz 5

- Quiz 5 will take place during lab in week 12
- Done online, over a 20 minutes duration,
  - during the last 30 minutes of the lab period, or as indicated by your tutor
- A few multiple choice questions,
  - covering material from weeks 9, 10, 11
    - graph traversal (BFS, DFS)
    - directed graphs and algorithms
    - weighted graphs and algorithms
    - sorting: bubblesort, mergesort, quicksort
    - costs (run-time and space) of all algorithms above

## Asst 2

- Due Friday October 27 (week 12); time extended to 11:59pm
- Discussed in second hour today

# Outline

- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: [insertion](#), [selection sort](#), [heap-sort](#)
  - In-place sorting
  - **Bubble sort**
  - **Merge-sort**
  - **Quick-sort**

# Recall: Priority Queue Sorting (from week 5)

- We can use a priority queue to sort a set of comparable elements
  1. Insert the elements one by one with a series of insert operations
    - element is used as key
    - null is the value (never considered, just goes along to fit the priority queue API)
  2. Remove the elements one-by-one with a series of removeMin operations
    - elements come out in sorted order
- The running time of this sorting method depends on the priority queue implementation

## Algorithm *PQ-Sort*(*S*, *C*)

**Input** list *S*, comparator *C* for the elements of *S*

**Output** list *S* sorted in increasing order according to *C*

*P* ← priority queue with comparator *C*

**while** (*!S.isEmpty* ())

*e* ← *S.removeFirst* ()

*P.insert* (*e*, null)

**while** (*!P.isEmpty* ())

*e* ← *P.removeMin* ().*getKey* ()

*S.addLast* (*e*)

## Recall: PQ-sorts (from week 5)

- PQ implemented as unsorted list
  - called **selection-sort**
  - worst-case runtime cost is  **$O(n^2)$ : removeMin  $O(n)$**
- PQ implemented as sorted list
  - called **insertion-sort**
  - worst-case runtime cost is  **$O(n^2)$ : insert  $O(n)$**
- PQ implemented as heap (either heap-ordered complete tree, or as heap-in-array)
  - called **heap-sort**
  - worst-case runtime cost is  **$O(n \log n)$ : insert/removeMin  $O(\log n)$**



# Outline

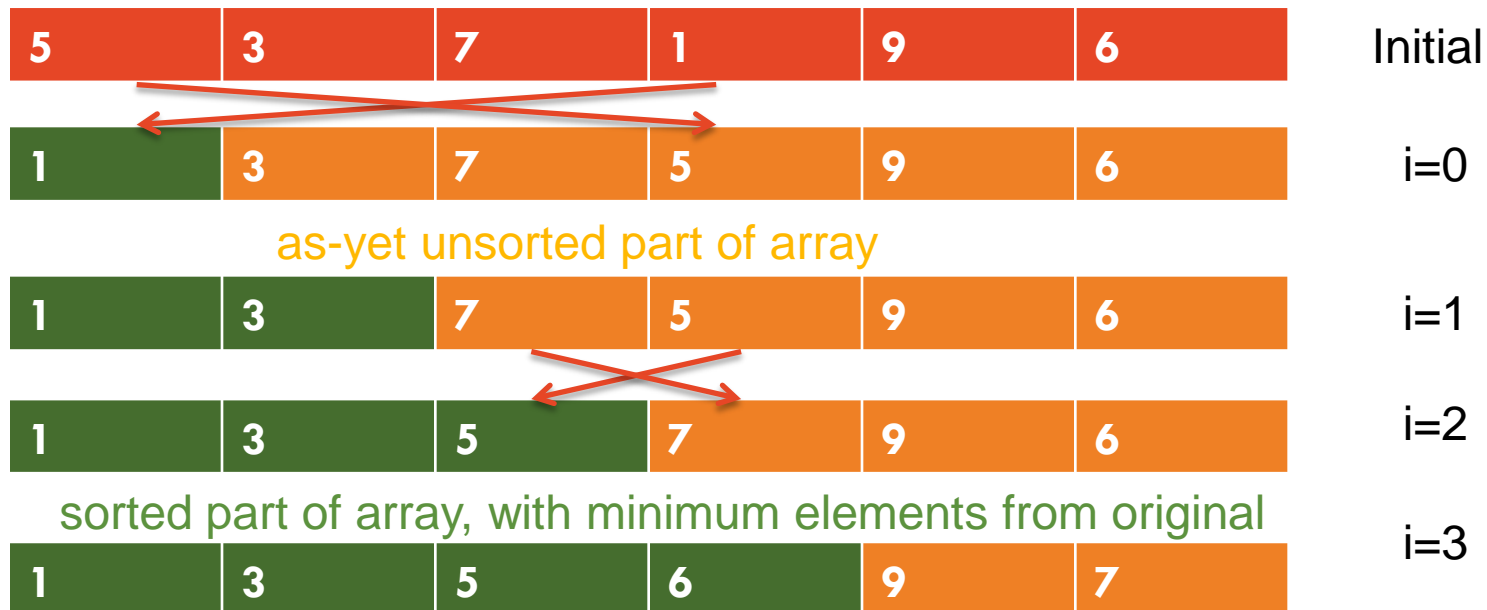
- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: insertion, selection sort, heap-sort
  - **In-place sorting**
  - Bubble sort
  - Merge-sort
  - Quick-sort

# In-place sorting

- the simplest form of each PQ-sort keeps an extra data structure whose space is  $O(n)$ 
  - *in addition to* the space used for the input data itself
  - we move elements into the **pq**, and perhaps move them around in the **pq** as more are inserted, and then move them from the **pq** to the output
- We can code similar algorithmic techniques, so that they move elements only within the array itself, which ends up being sorted
  - it may keep a small amount of extra space, for indices, local maximum, etc; this extra space should be  $O(1)$

## In-place selection sort

- The part of the array from index 0 to i keeps the smallest elements of the original array, sorted
- The part of the array from index  $i+1$  to  $N-1$  keeps the rest of the original contents
  - Each inner loop takes min of unsorted part, and swaps it into index  $i+1$
- At the end of the algorithm  $i = N-1$ , so all the array is sorted



## In-place insertion sort

- The part of the array from index 0 to  $i$  keeps a priority queue as a sorted array, that is a sorted form of the original contents
- The part of the array from index  $i+1$  to  $N-1$  has its original contents
- At the end of the algorithm  $i = N-1$ , so all the array is sorted

5	3	7	1	9	6	Initial
5	3	7	1	9	6	$i=0$
as-yet unsorted part of array, in original locations						
3	5	7	1	9	6	$i=1$
3	5	7	1	9	6	$i=2$
sorted part of array						
1	3	5	7	9	6	$i=3$

## In-place heapsort

- Clever technique to build the **maxheap-in-array**, working in-place (Section 9.4.2)
- Repeat: removing max among remaining elements and put into index  $N-1$ 
  - remove the max (which is at index 0)
  - restructure the remaining maxheap-in-array in index 0 to  $N-1$
  - put the removed **max** in index  $N-1$
- Time complexity:  $O(n \log n)$

# Outline

- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: insertion, selection sort, heap-sort
  - In-place sorting
  - **Bubble sort**
  - Merge-sort
  - Quick-sort

# Bubble-sort

- in-place algorithm for data in an array
- Very easy to code
- Variants can be parallelized well
- But: performance is usually slow, or (even) medium data
- Key idea: To sort a sequence of  $n$  comparable elements
  - Scan the sequence  $n-1$  times
  - In each step of a scan, compare the current element with the next and swap them if they are out of order
    - note that if a swap occurs, the next step will compare the larger element in its new position, with its successor in the array
    - by the end of the scan, the largest element has reached the last position
    - so the next scan works on a slightly shorter part of the array

# Example Bubble-sort

\_\_: comparison

swap

sorted

First Pass:

( 5 1 4 2 8 ) → ( 1 5 4 2 8 )

( 1 5 4 2 8 ) → ( 1 4 5 2 8 )

( 1 4 5 2 8 ) → ( 1 4 2 5 8 )

( 1 4 2 5 8 ) → ( 1 4 2 5 8 )

Second Pass:

( 1 4 2 5 8 ) → ( 1 4 2 5 8 )

( 1 4 2 5 8 ) → ( 1 2 4 5 8 )

( 1 2 4 5 8 ) → ( 1 2 4 5 8 )

Third Pass:

( 1 2 4 5 8 ) → ( 1 2 4 5 8 )

( 1 2 4 5 8 ) → ( 1 2 4 5 8 )

Fourth Pass:

( 1 2 4 5 8 ) → ( 1 2 4 5 8 )



## Bubble-sort

```
array elements[1..N]  
for j := 1 to N-1 do  
    for k := 1 to N-j do  
        if elements[k] > elements[k+1]  
            then swap(k, k+1, elements)
```

## Iterations of bubble-sort

- Each full iteration through the array (inner for loop) will place **at least one element at its final resting position**
  - The last element
- Therefore in the next iteration we do not need to check if the last element needs to be swapped
  - Inner for loop stops at  $N-j$  in the  $j$ -th iteration

# Bubble-sort running time

*array elements[1..N]*

*for j := 1 to N-1 do*

*for k := 1 to N-j do*

*if elements[k] > elements[k+1]*

*then swap(k, k+1, elements)*

outer loop executes  $N-1$  times:  $O(N)$

inner loop executes a number of times  
that depends on  $j$   
worst case is to execute  $N-1$  times:  $O(N)$

inside the inner loop:  
time doesn't grow no matter  
how much data in array  
cost:  $O(1)$

total worst-case running time:  $O(N) * O(N) * O(1) = O(N^2)$

## Variant bubble-sort

// may stop early as soon as one pass finds no swaps needed

array elements[1..N]

swapDone = **true**

**while** swapDone **do**

    swapDone = **false**

**for** k := 1 **to** N-1 **do**

**if** elements[k] > elements[k+1] **then**

            swap(k,k+1, elements)

        swapDone = **true**

# Variant Bubble-sort running time bound

- Is this a faster solution than nested for loops ?
  - In many cases, it takes less time, because it ends early
- But what is the worst-case running time?
- General analysis: Each complete iteration of the inner loop will place at least one new element in its final position
  - A maximum of  $N$  iterations of the while loop are needed
  - This implies a bound of no more than  $O(n^2)$
- Is there a lower bound?
  - or, does the variant still need all  $N-1$  iterations of the inner loop in its worst-case?
- What input makes the algorithm use all  $N-1$  iterations of inner loop?
  - When the [input is in reverse sorted order](#)
  - Eg 8,7,6,5,4,3,2,1
  - Indeed any input where the smallest element is at the end
- Variant bubble-sort has  **$O(n^2)$  worst-case runtime**

# Outline

- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: insertion, selection sort, heap-sort
  - In-place sorting
  - Bubble sort
  - **Merge-sort**
  - Quick-sort

# Divide-and-Conquer Algorithm

- **Divide-and conquer** is a general algorithm design paradigm:
  - Divide: divide the input data  $S$  in two disjoint subsets  $S_1$  and  $S_2$
  - Recur: solve the subproblems associated with  $S_1$  and  $S_2$
  - Conquer: combine the solutions for  $S_1$  and  $S_2$  into a solution for  $S$
- The base case for the recursion are subproblems of size 0 or 1
- **Merge-sort** is a sorting algorithm based on the **divide-and-conquer** paradigm
  - $O(n \log n)$  running time
    - It does not use an auxiliary priority queue
    - It accesses data in a sequential manner (suitable to sort data stored on a disk)

# Merge-Sort

- Merge-sort on an input sequence  $S$  with  $n$  elements consists of three steps:
  - **Divide**: partition  $S$  into two sequences  $S_1$  and  $S_2$  of about  $n/2$  elements each
  - **Recur**: recursively sort  $S_1$  and  $S_2$
  - **Conquer**: merge  $S_1$  and  $S_2$  into a unique sorted sequence

## Algorithm *mergeSort*( $S$ )

**Input** sequence  $S$  with  $n$  elements

**Output** sequence  $S$  sorted  
(according to a comparator function)

**if**  $S.size() > 1$

$(S_1, S_2) \leftarrow partition(S, n/2)$

*mergeSort*( $S_1$ )

*mergeSort*( $S_2$ )

$S \leftarrow merge(S_1, S_2)$

## Merging Two Sorted Sequences

- The conquer step of merge-sort consists of merging two sorted sequences  $A$  and  $B$  into a sorted sequence  $S$  containing the union of the elements of  $A$  and  $B$
- Merging two sorted sequences, each with  $n/2$  elements and implemented by means of a doubly linked list, takes  $O(n)$  time

Algorithm *merge*( $A, B$ )

**Input** sorted sequences  $A, B$  with  $n/2$  elements each

**Output** sorted sequence  $S$  of  $A \cup B$

$S \leftarrow$  empty sequence

**while**  $!A.isEmpty() \ \&\& \ !B.isEmpty()$

**if**  $A.first().element() < B.first().element()$

$S.addLast(A.remove(A.first()))$

**else**

$S.addLast(B.remove(B.first()))$

**while**  $!A.isEmpty()$

$S.addLast(A.remove(A.first()))$

**while**  $!B.isEmpty()$

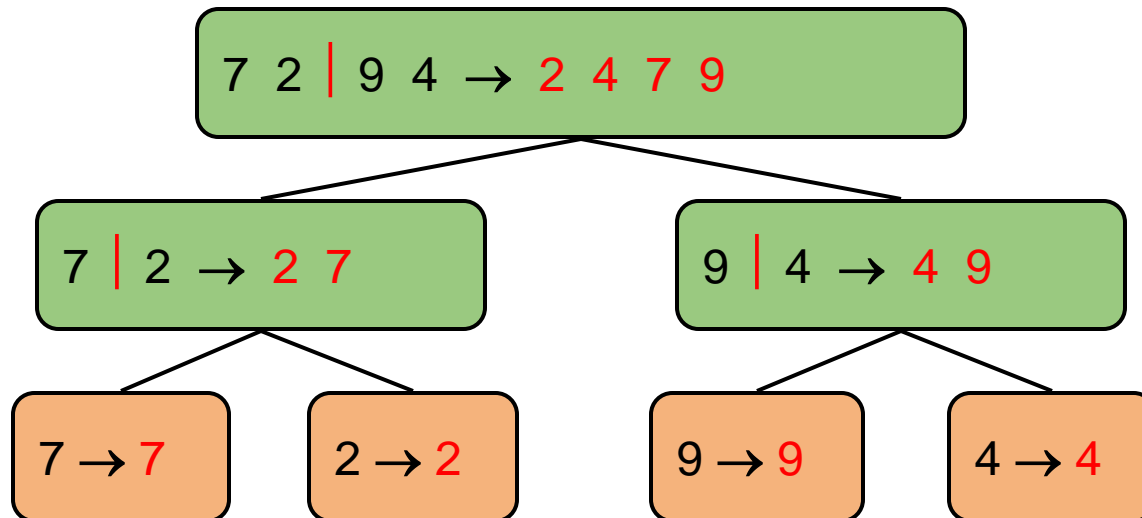
$S.addLast(B.remove(B.first()))$

**return**  $S$

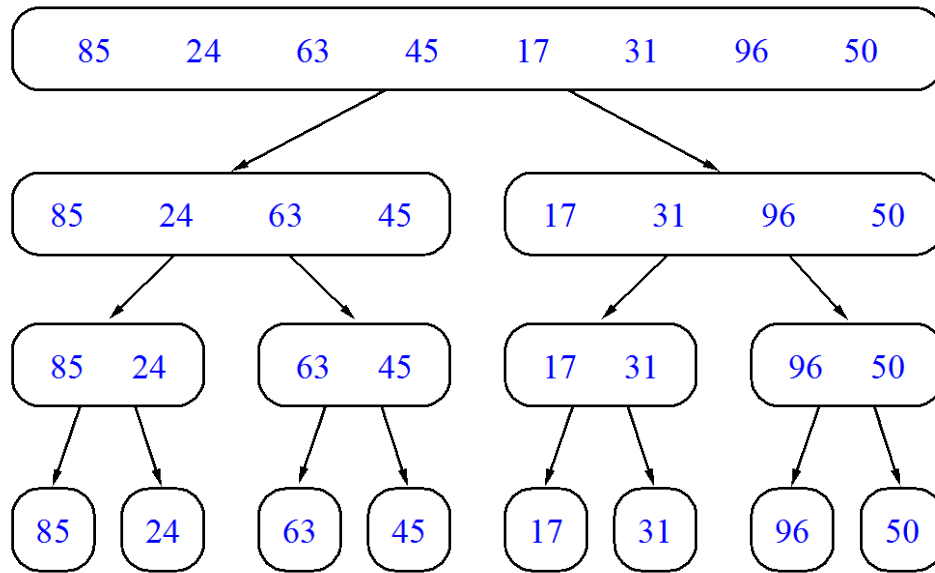


# Merge-Sort Tree

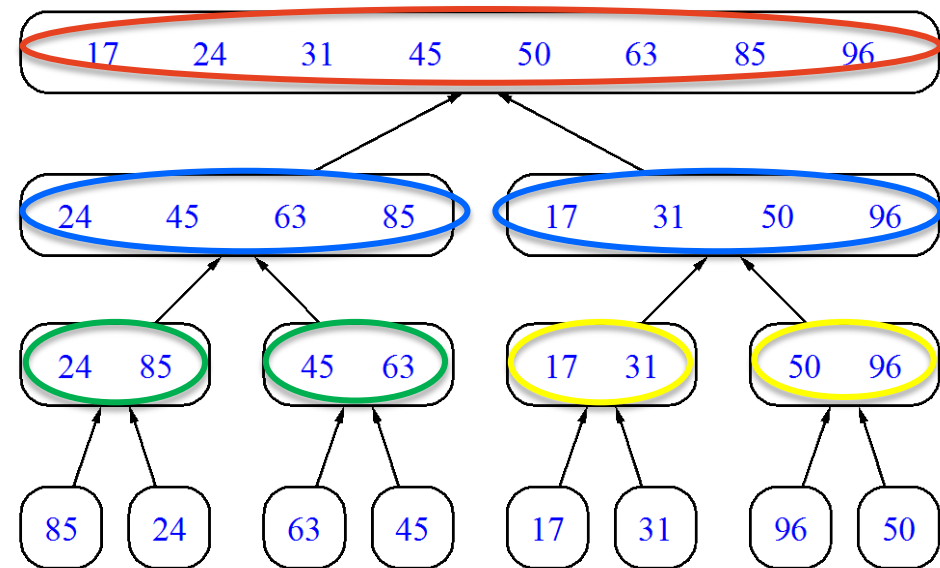
- An execution of merge-sort is depicted by a [binary tree](#)
  - each node represents a recursive call of merge-sort and stores
    - unsorted sequence before the execution and its partition
    - sorted sequence at the end of the execution
  - the [root](#) is the initial call
  - the [leaves](#) are calls on subsequences of size 0 or 1



# Merge sort trees (input and output sequences)



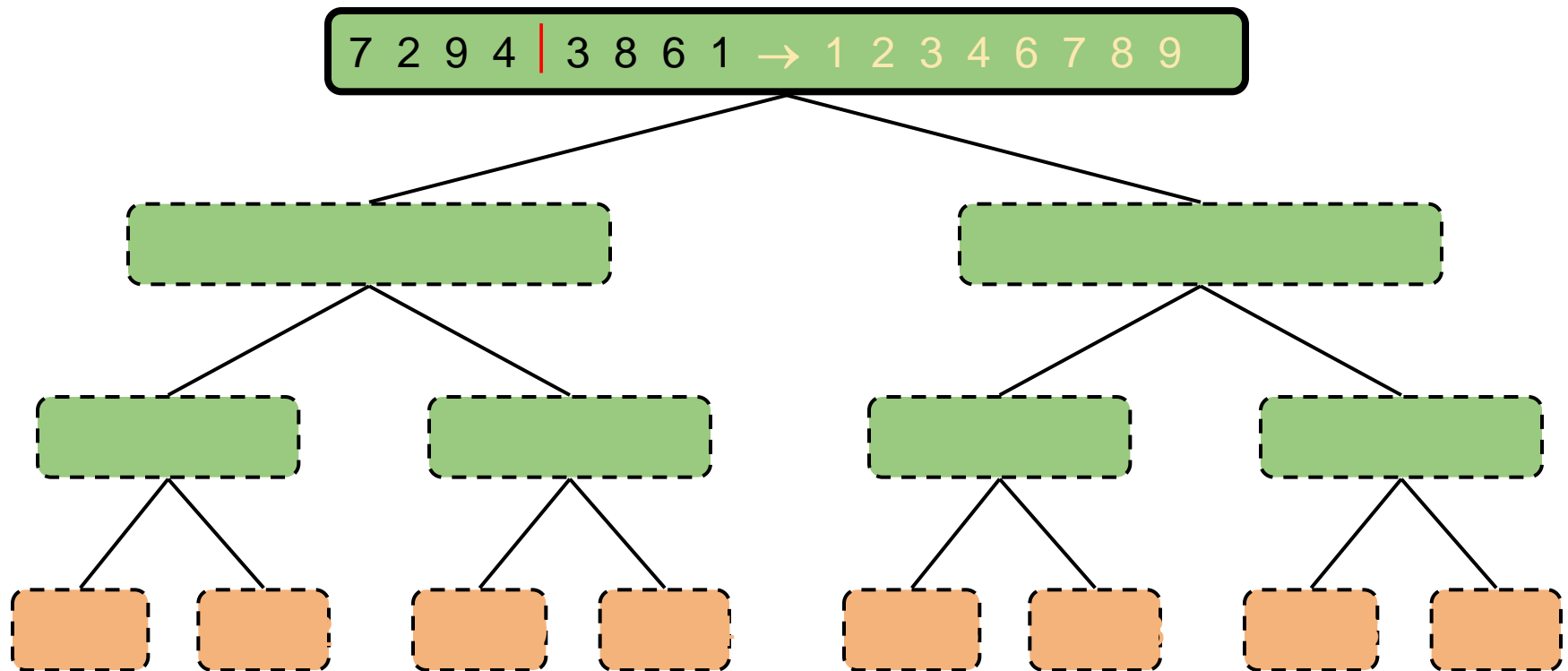
Divide



Merge

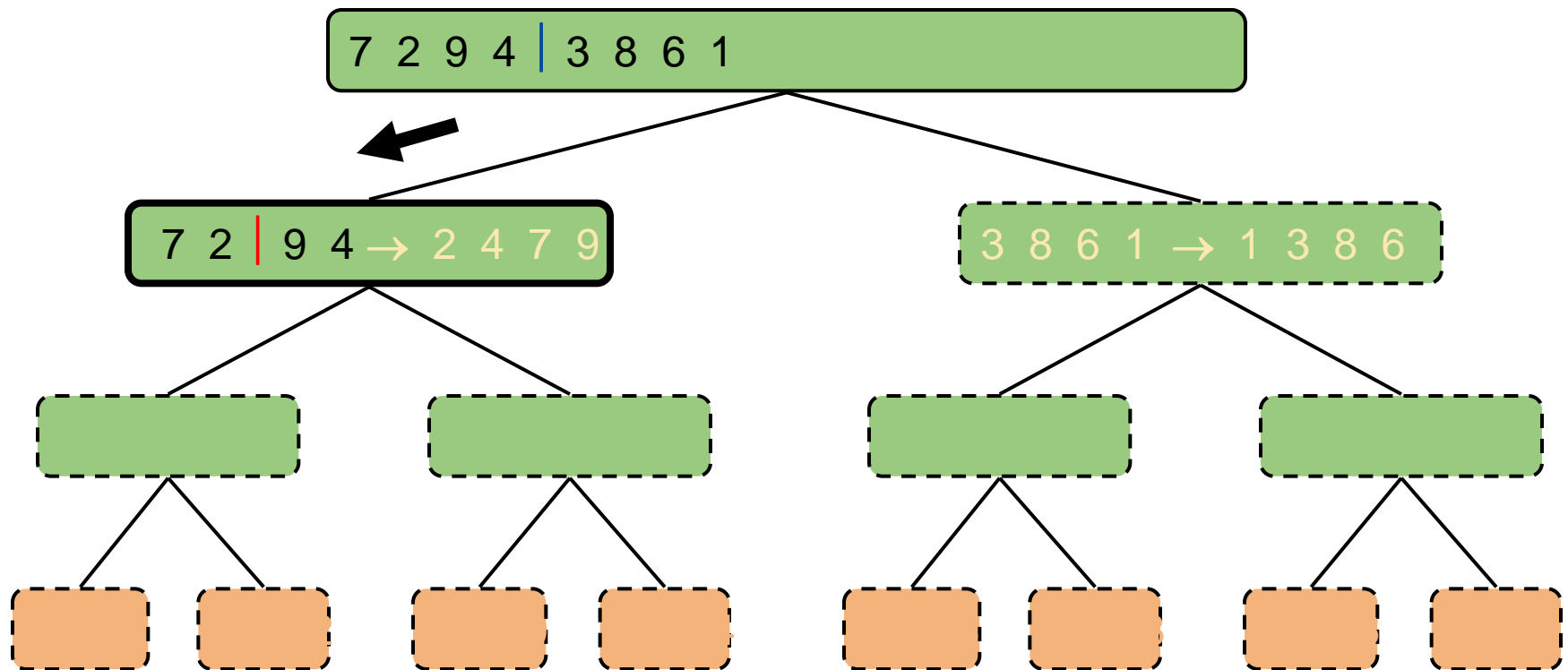
# Execution Example

## — Partition



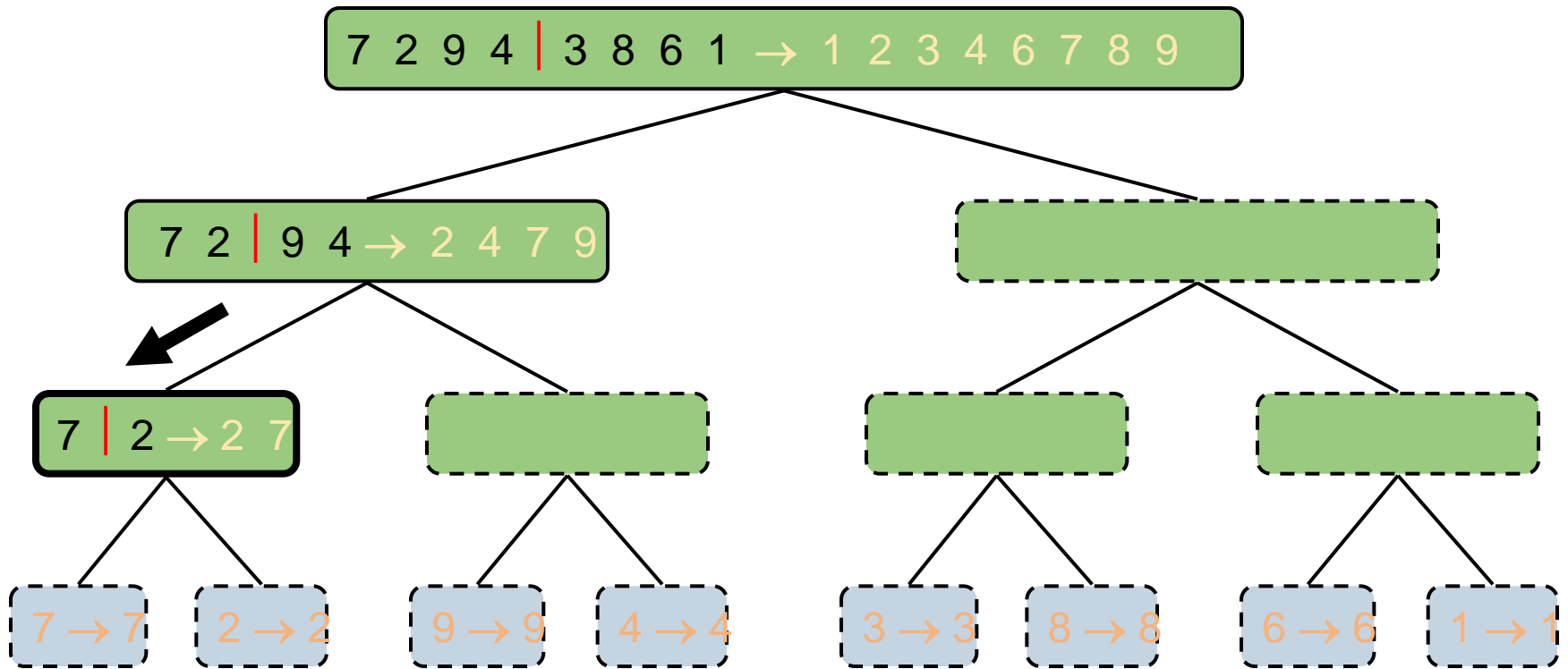
## Execution Example (cont.)

- Recursive call, partition



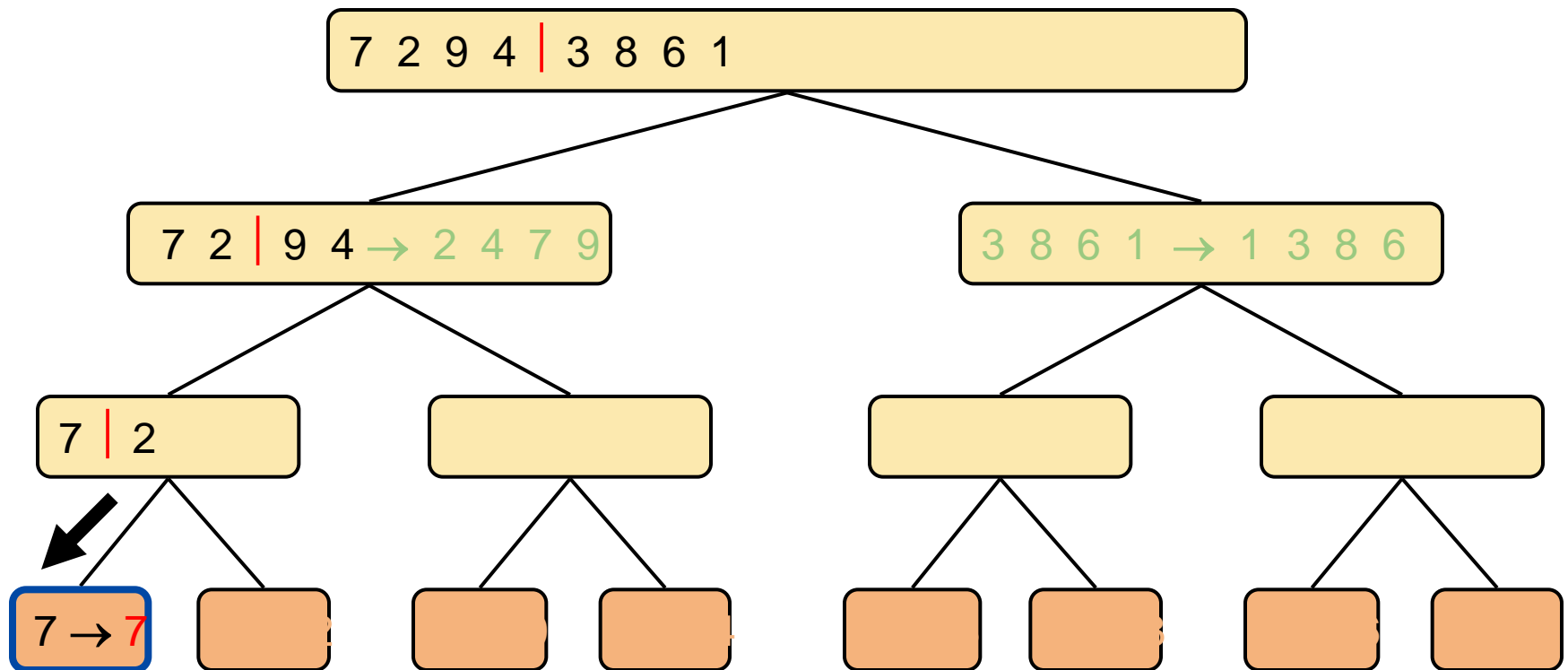
## Execution Example (cont.)

- Recursive call, partition



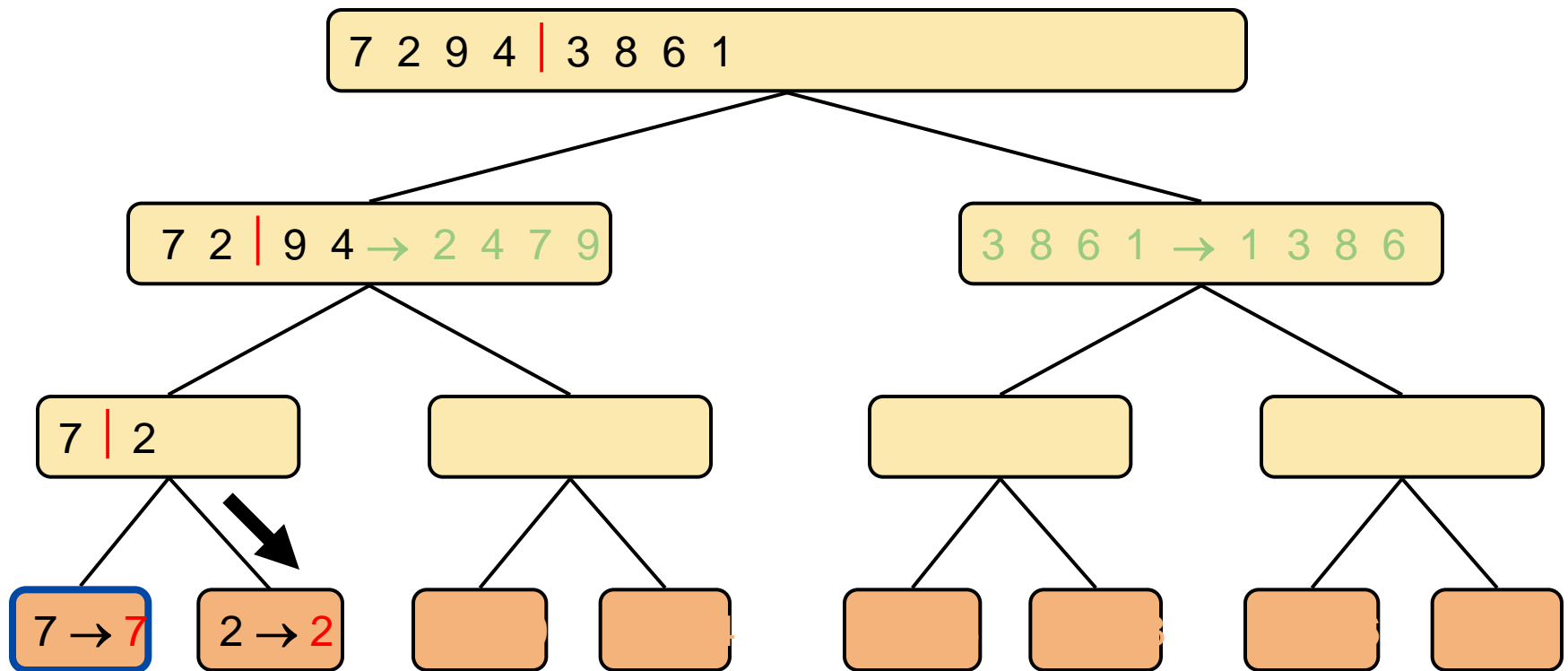
## Execution Example (cont.)

- Recursive call, base case



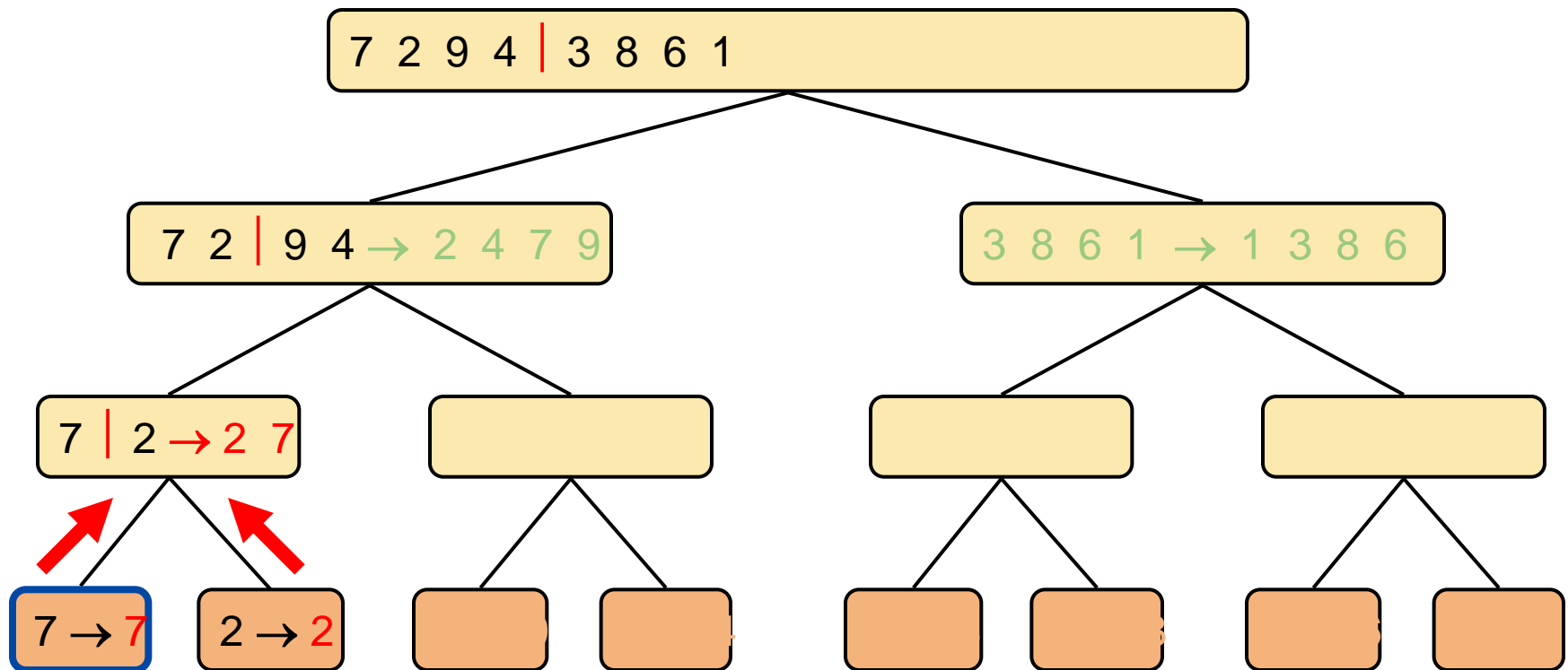
## Execution Example (cont.)

- Recursive call, base case



## Execution Example (cont.)

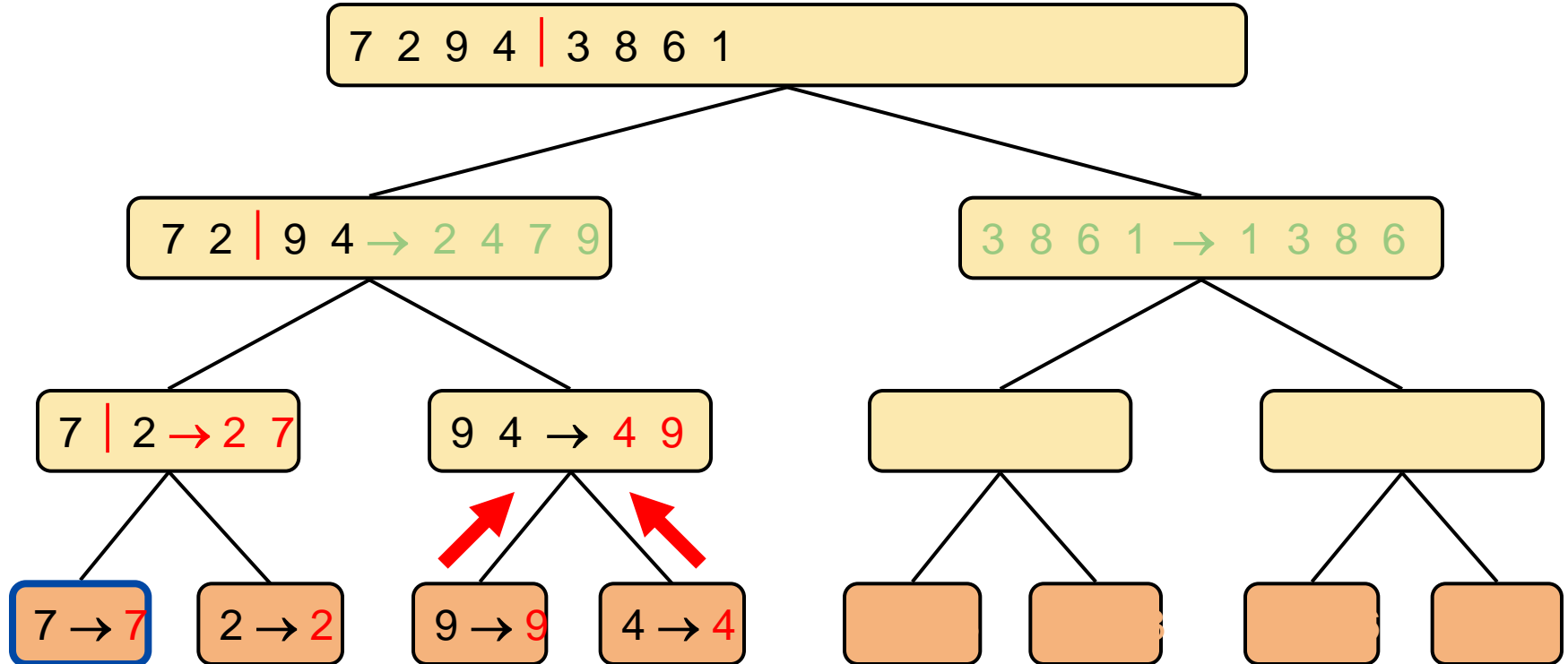
– Merge





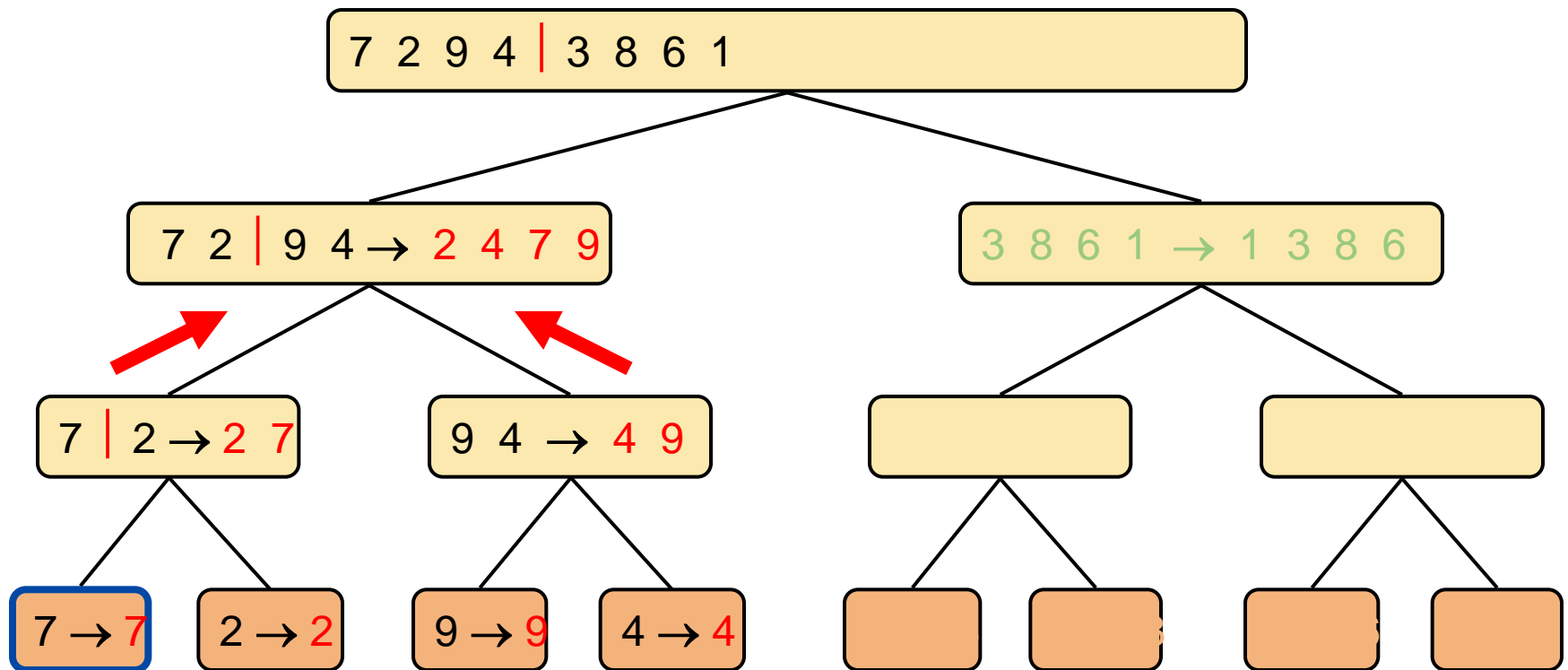
## Execution Example (cont.)

- Recursive call, ..., base case, merge



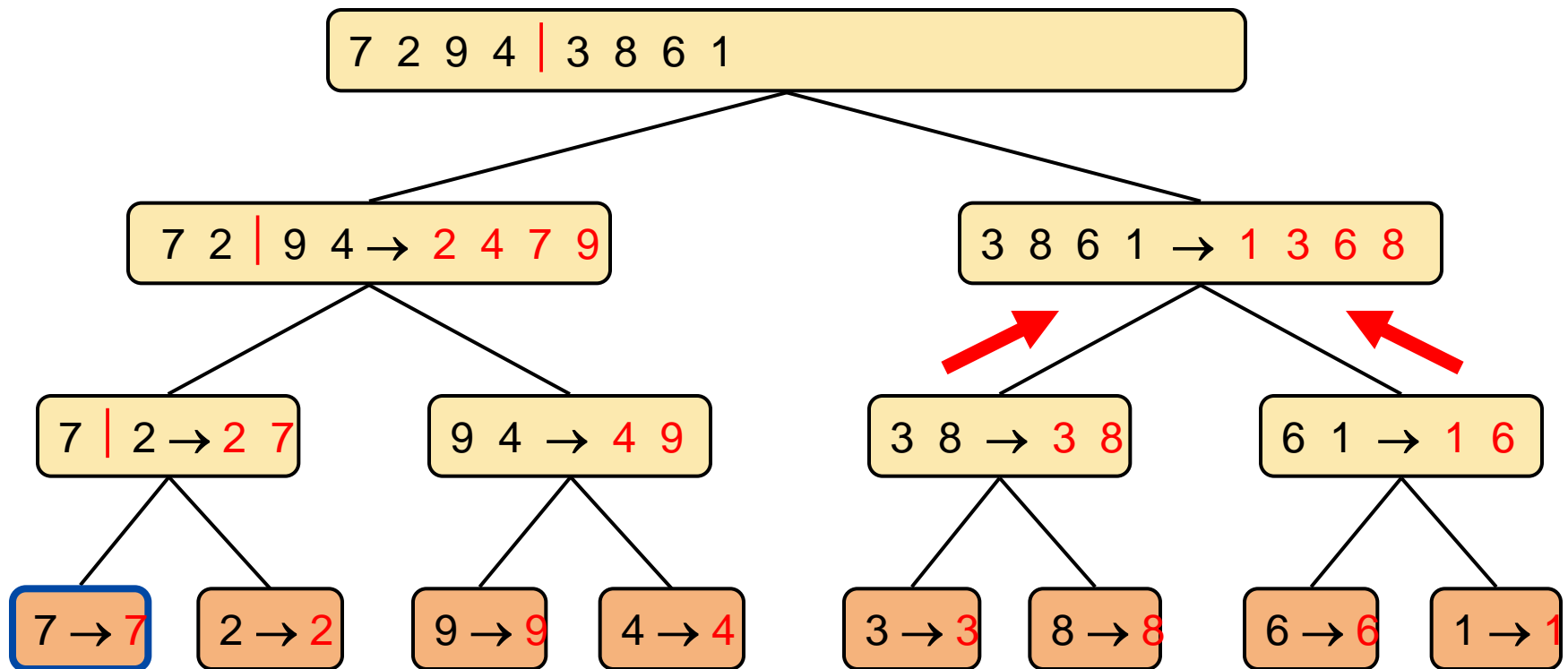
## Execution Example (cont.)

– Merge



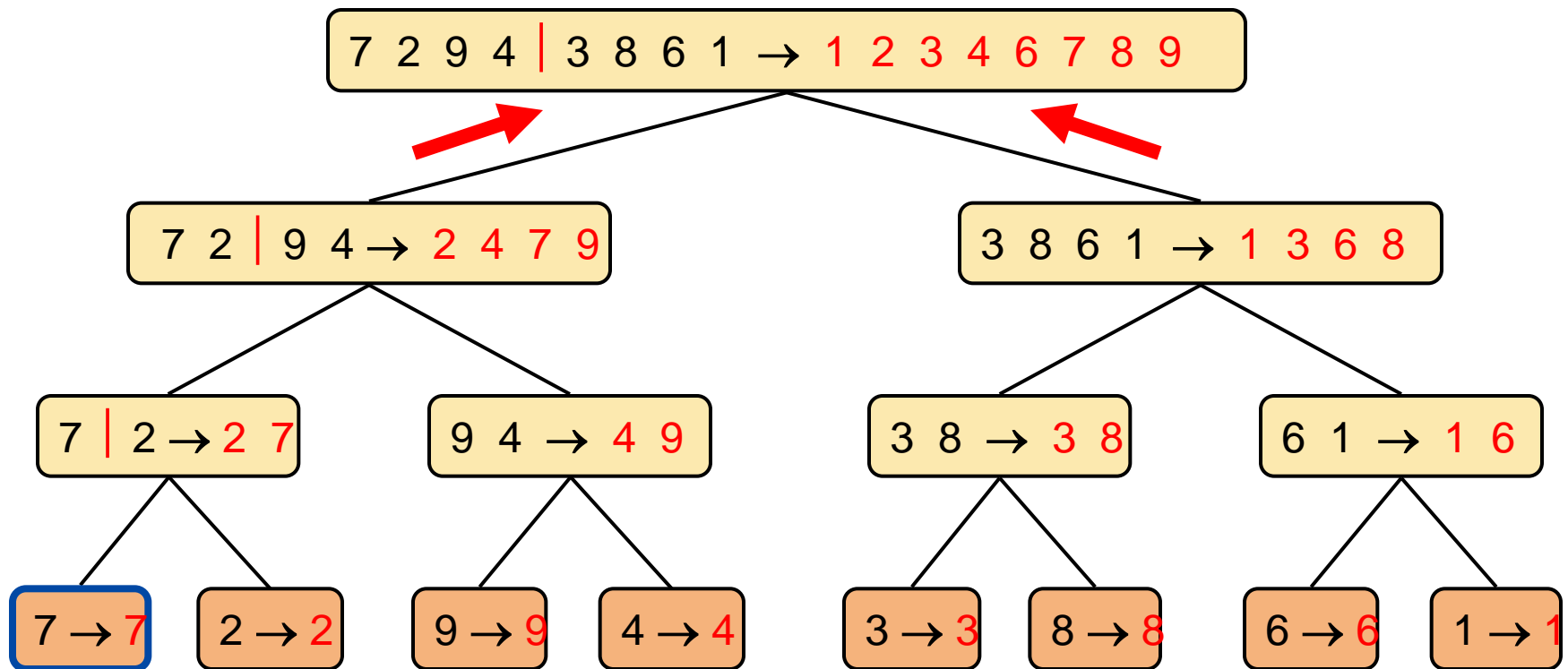
## Execution Example (cont.)

- Recursive call, ..., merge, merge



## Execution Example (cont.)

- Recursive call, ..., merge, merge



## Recall: Recurrence Analysis of Mergesort (from week 8)

MergeSort(list):

If the list has one or less elements, return the list	$O(1)$
Otherwise, <b>Divide</b> the list into two halves A and B	$O(n)$
List sortedA = MergeSort(A)	$T(n/2)$
List sortedB = MergeSort(B)	$T(n/2)$
<b>Merge</b> the sorted lists sortedA and sorted	$O(n)$
Return the merged list	

Let  $T(n)$  be the time to run MergeSort on a list of size  $n$

Then  $T(n) = 2 * T(n/2) + O(n)$

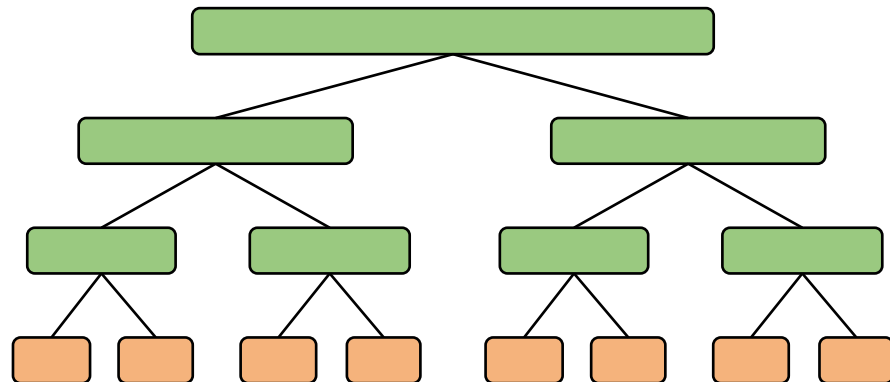
## Solve recurrence for worst-case runtime of MergeSort

- We saw:  $T(n) = 2 * T(n/2) + O(n)$
- From **week 8**, we know this has solution  $T(n) = O(n \log n)$

# Direct Analysis of Merge-Sort

- The height  $h$  of the merge-sort tree is  $O(\log n)$ 
  - at each recursive call we divide in half the sequence,
- The overall amount of work done at the nodes of depth  $i$  is  $O(n)$ 
  - we partition and merge  $2^i$  sequences of size  $n/2^i$
  - we make  $2^{i+1}$  recursive calls
- Thus, the total running time of merge-sort is  $O(n \log n)$

depth	#seqs	size	Time per level
0	1	$n$	$O(n)$
1	2	$n/2$	$O(n)$
$i$	$2^i$	$n/2^i$	$O(n)$
...	...	...	$O(n)$

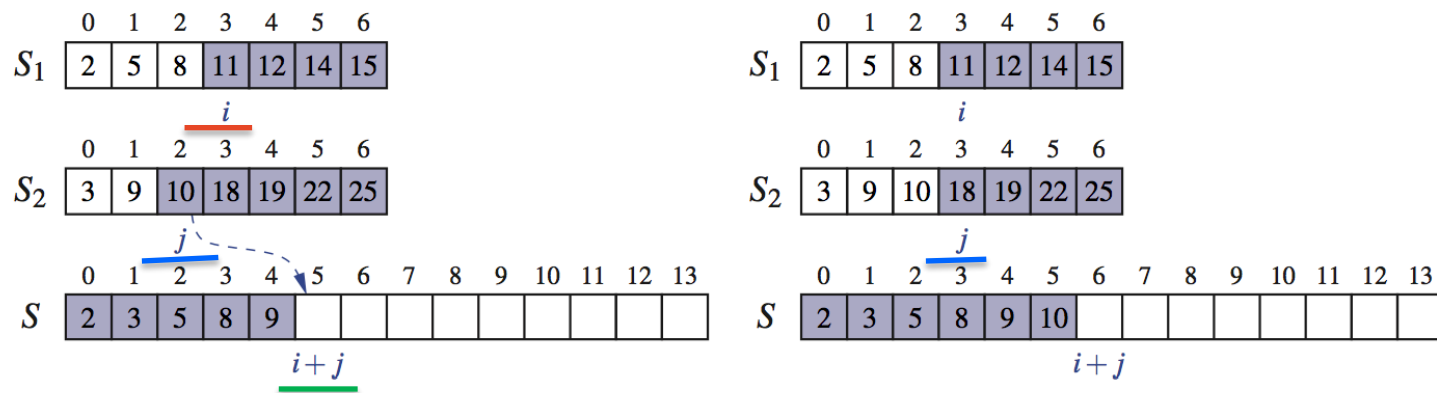


# Java Merge Implementation (using arrays)

```

1  /** Merge contents of arrays S1 and S2 into properly sized array S. */
2  public static <K> void merge(K[ ] S1, K[ ] S2, K[ ] S, Comparator<K> comp) {
3      int i = 0, j = 0;
4      while (i + j < S.length) {
5          if (j == S2.length || (i < S1.length && comp.compare(S1[i], S2[j]) < 0))
6              S[i+j] = S1[i++];           // copy ith element of S1 and increment i
7          else
8              S[i+j] = S2[j++];           // copy jth element of S2 and increment j
9      }
10 }

```



# Java Merge-Sort Implementation

```
1  /** Merge-sort contents of array S. */
2  public static <K> void mergeSort(K[ ] S, Comparator<K> comp) {
3      int n = S.length;
4      if (n < 2) return; // array is trivially sorted
5      // divide
6      int mid = n/2;
7      K[ ] S1 = Arrays.copyOfRange(S, 0, mid); // copy of first half
8      K[ ] S2 = Arrays.copyOfRange(S, mid, n); // copy of second half
9      // conquer (with recursion)
10     mergeSort(S1, comp); // sort copy of first half
11     mergeSort(S2, comp); // sort copy of second half
12     // merge results
13     merge(S1, S2, S, comp); // merge sorted halves back into original
14 }
```



## In-place mergesort

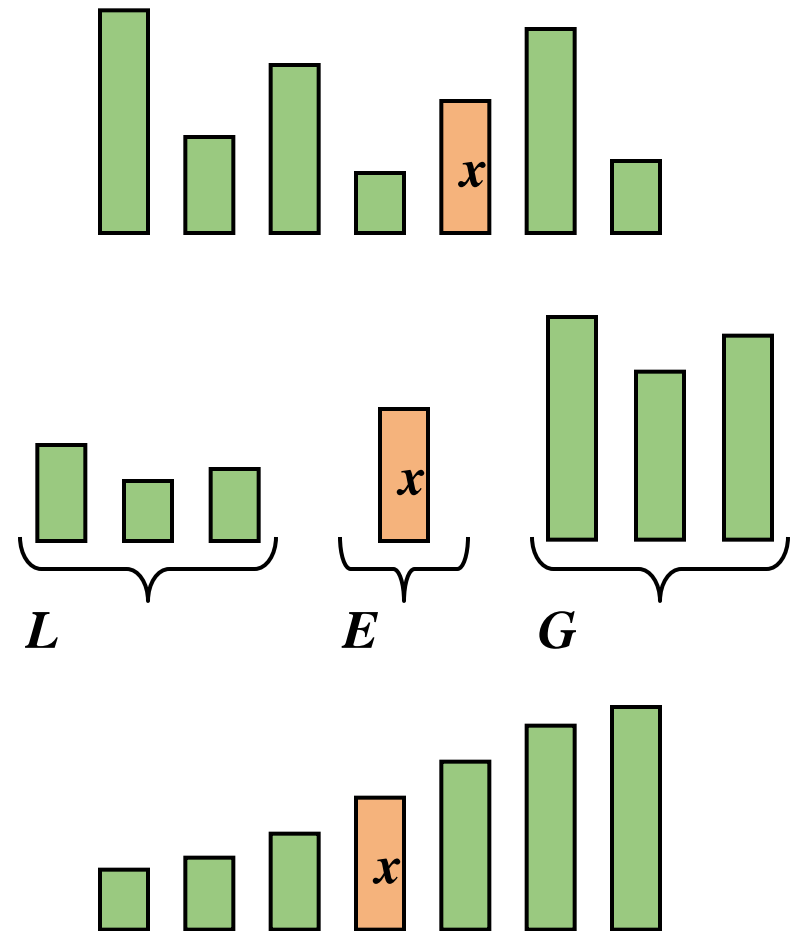
- The algorithm we just saw is not in-place
- It is easy to partition  $S$  in-place
- It is not easy to **merge two sorted lists in-place**, but there are complicated ways to get an algorithm that does so, allowing one to mergesort an array in-place

# Outline

- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: insertion, selection sort, heap-sort
  - In-place sorting
  - Bubble sort
  - Merge-sort
  - **Quick-sort**

# Quick-Sort

- **Quick-sort** is a randomized sorting algorithm based on the divide-and-conquer paradigm:
  - **Divide**: pick a **random element**  $x$  (called pivot) and partition  $S$  into
    - $L$  elements less than  $x$
    - $E$  elements equal  $x$
    - $G$  elements greater than  $x$
  - **Recur**: sort  $L$  and  $G$
  - **Conquer**: join  $L$ ,  $E$  and  $G$
- Unlike merge-sort, hard work done *before* the recursive calls



# Partition

- We partition an input sequence as follows:
  - We remove, in turn, each element  $y$  from  $S$  and
  - We insert  $y$  into  $L, E$  or  $G$ , depending on the result of the comparison with the pivot  $x$
- Each **insertion** and **removal** is at the beginning or at the end of a sequence, and hence takes  $O(1)$  time
- Thus, the **partition step** of quick-sort takes  $O(n)$  time

Algorithm *partition*( $S, p$ )

**Input** sequence  $S$ , position  $p$  of pivot

**Output** subsequences  $L, E, G$  of the elements of  $S$  less than, equal to, or greater than the pivot, resp.

$L, E, G \leftarrow$  empty sequences

$x \leftarrow S.remove(p)$

**while**  $\neg S.isEmpty()$

$y \leftarrow S.remove(S.first())$

**if**  $y < x$

$L.addLast(y)$

**else if**  $y = x$

$E.addLast(y)$

**else**  $\{ y > x \}$

$G.addLast(y)$

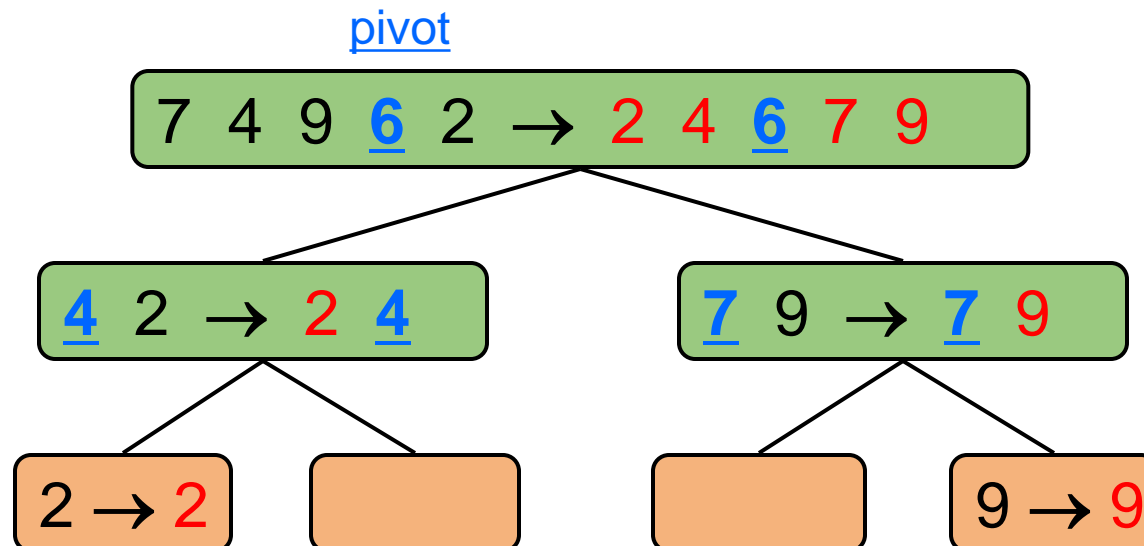
**return**  $L, E, G$

# Java Implementation

```
1  /** Quick-sort contents of a queue. */
2  public static <K> void quickSort(Queue<K> S, Comparator<K> comp) {
3      int n = S.size();
4      if (n < 2) return;           // queue is trivially sorted
5      // divide
6      K pivot = S.first();         // using first as arbitrary pivot
7      Queue<K> L = new LinkedList<>();
8      Queue<K> E = new LinkedList<>();
9      Queue<K> G = new LinkedList<>();
10     while (!S.isEmpty()) {       // divide original into L, E, and G
11         K element = S.dequeue();
12         int c = comp.compare(element, pivot);
13         if (c < 0)                // element is less than pivot
14             L.enqueue(element);
15         else if (c == 0)           // element is equal to pivot
16             E.enqueue(element);
17         else                      // element is greater than pivot
18             G.enqueue(element);
19     }
20     // conquer
21     quickSort(L, comp);           // sort elements less than pivot
22     quickSort(G, comp);         // sort elements greater than pivot
23     // concatenate results
24     while (!L.isEmpty())
25         S.enqueue(L.dequeue());
26     while (!E.isEmpty())
27         S.enqueue(E.dequeue());
28     while (!G.isEmpty())
29         S.enqueue(G.dequeue());
30 }
```

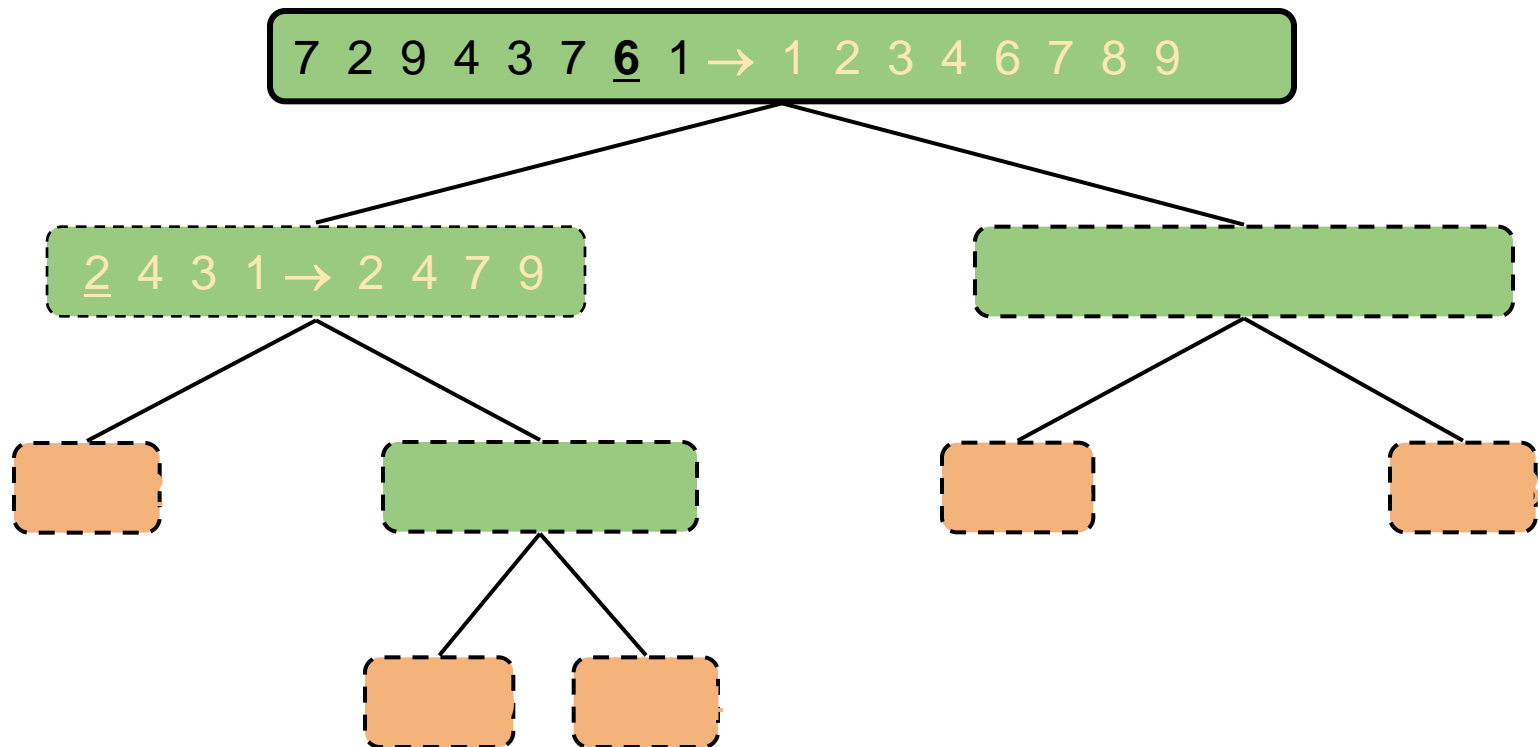
# Quick-Sort Tree

- An execution of quick-sort is depicted by a binary tree
  - Each node represents a recursive call of quick-sort and stores
    - Unsorted sequence before the execution and its pivot
    - Sorted sequence at the end of the execution
  - The root is the initial call
  - The leaves are calls on subsequences of size 0 or 1



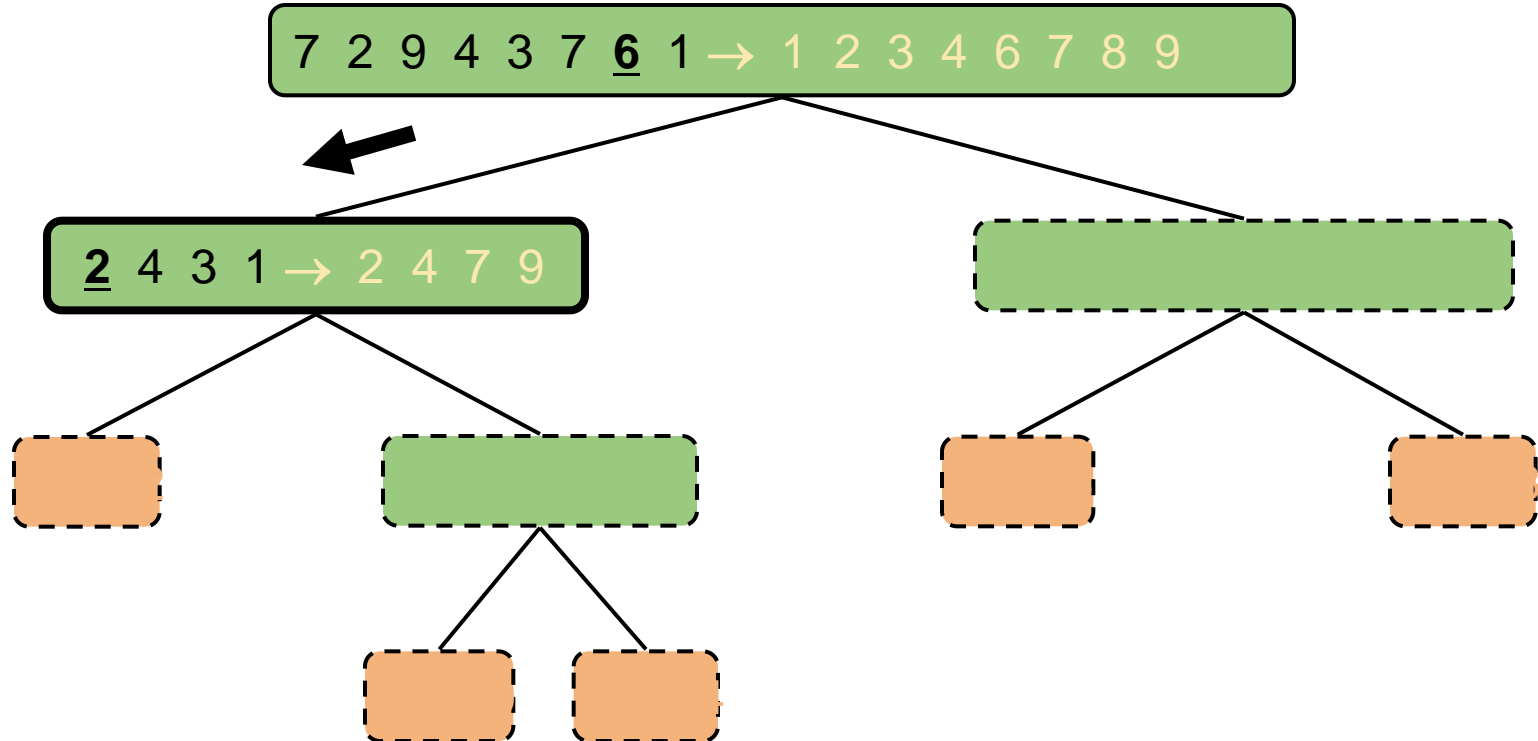
# Execution Example

## — Pivot selection



## Execution Example (cont.)

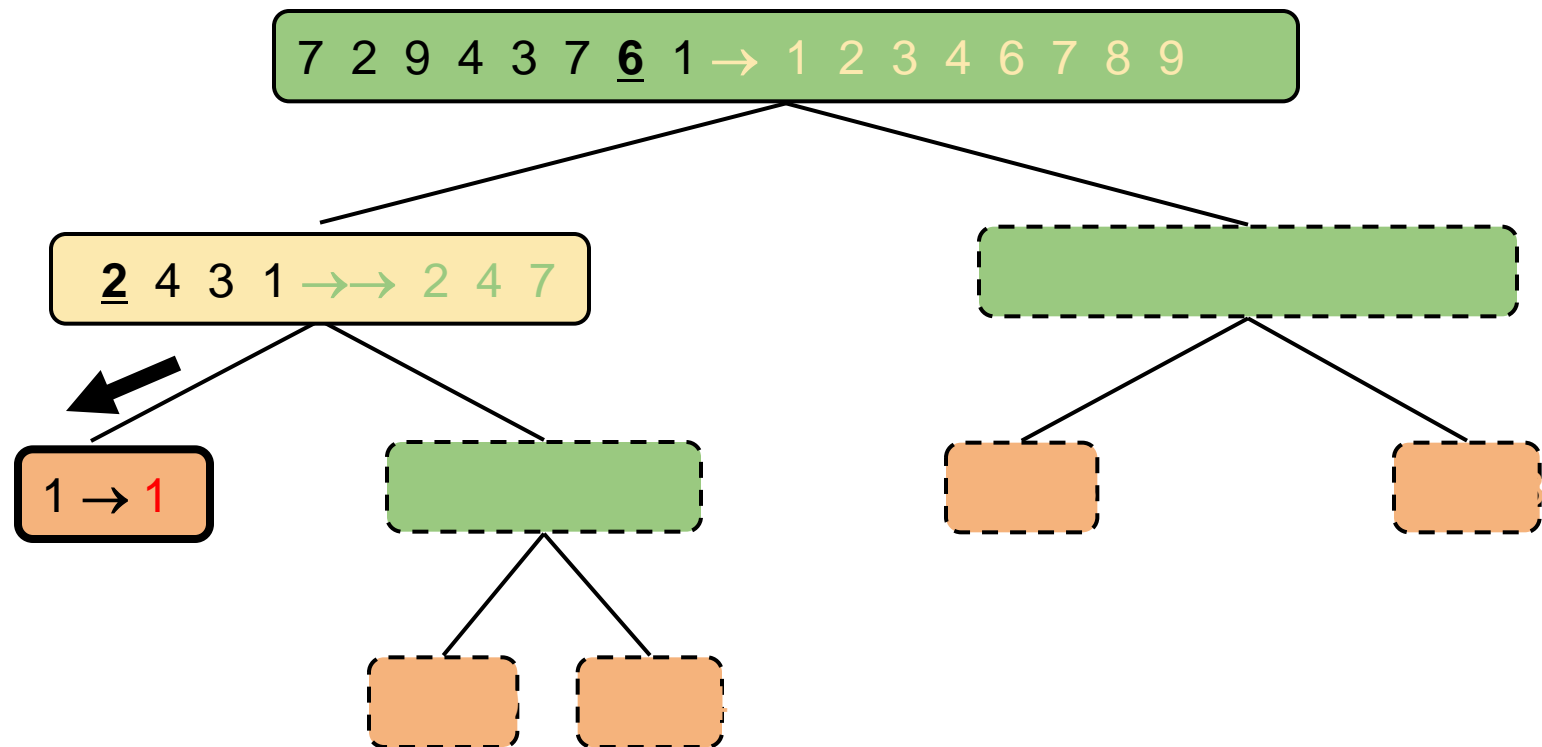
- Partition, recursive call, pivot selection





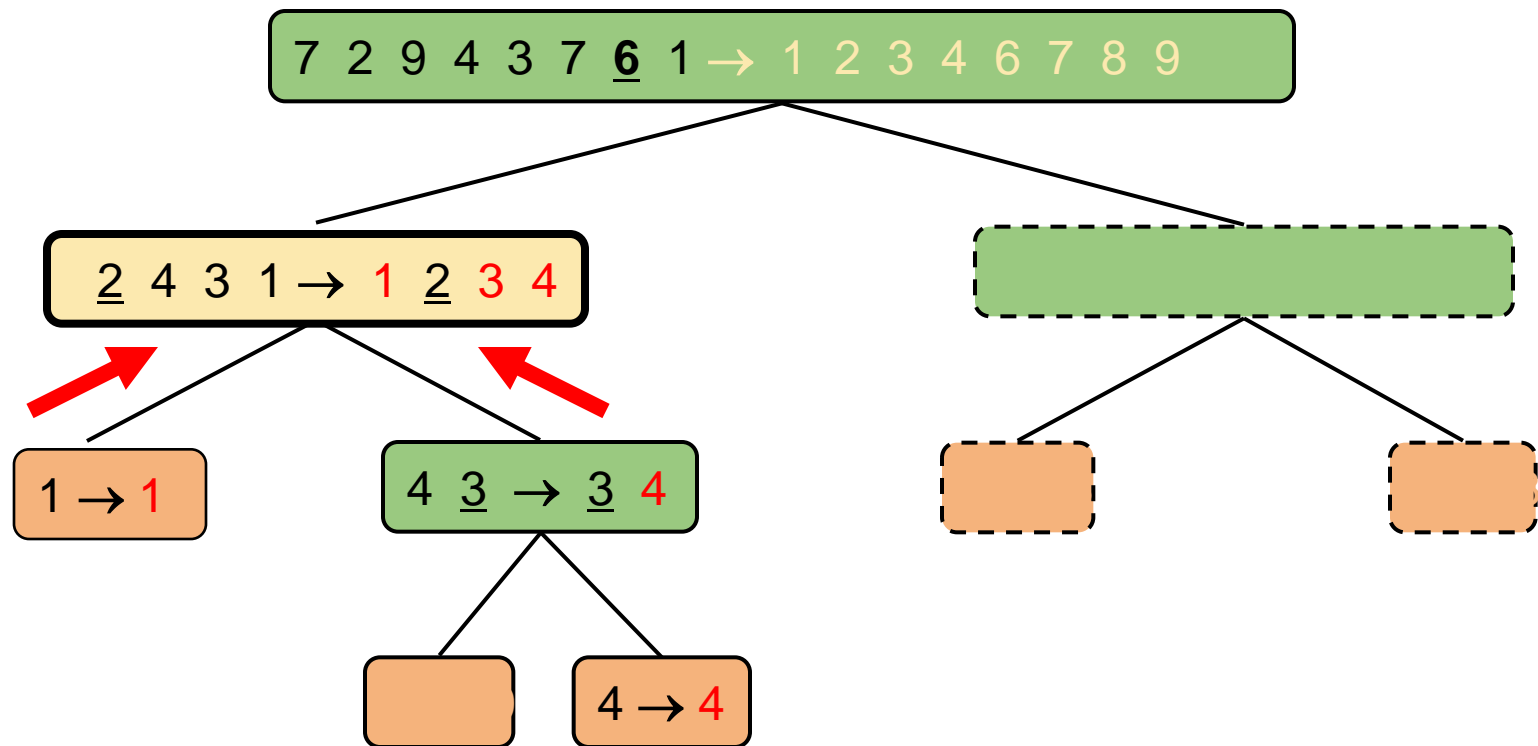
## Execution Example (cont.)

- Partition, recursive call, base case



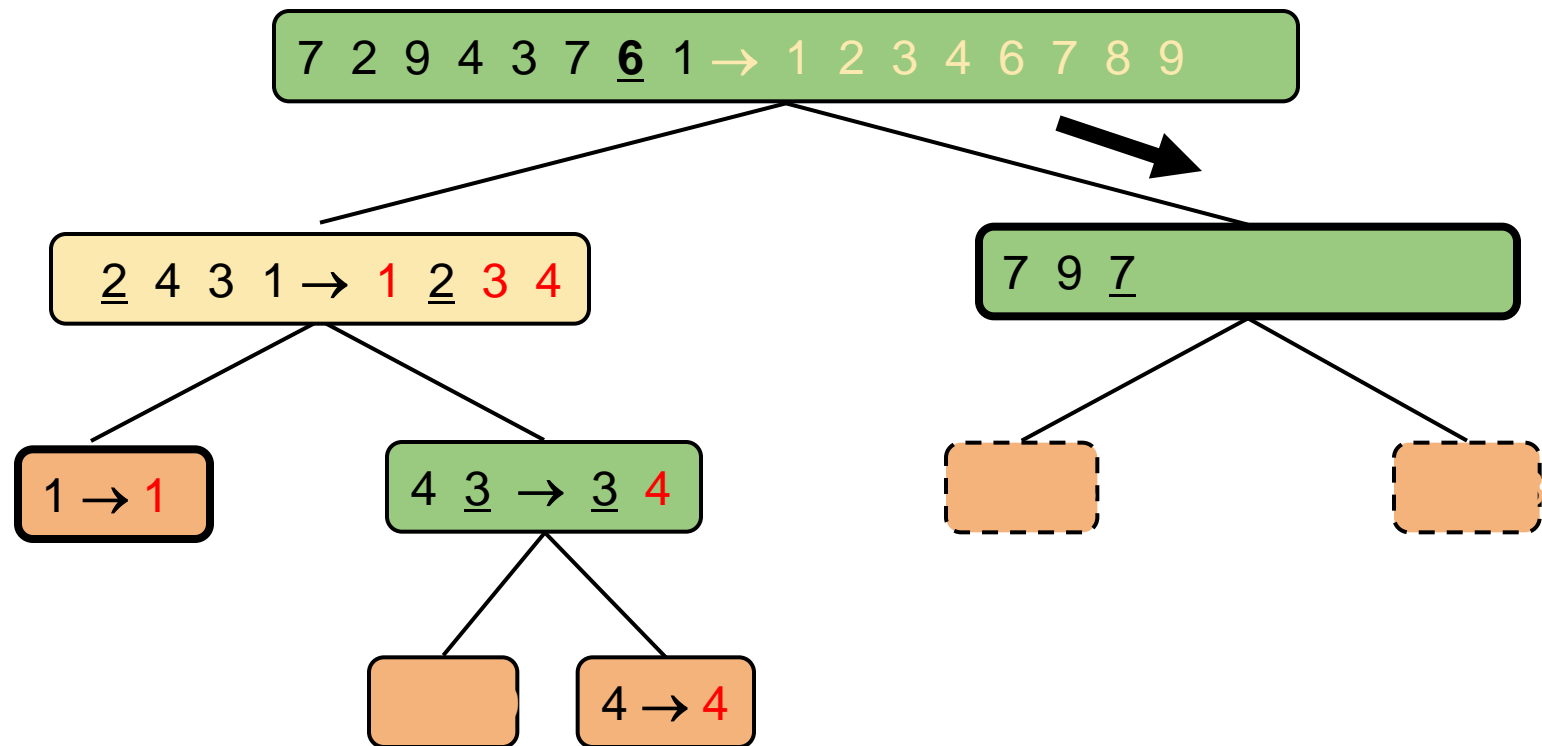
## Execution Example (cont.)

- Recursive call, ..., base case, join



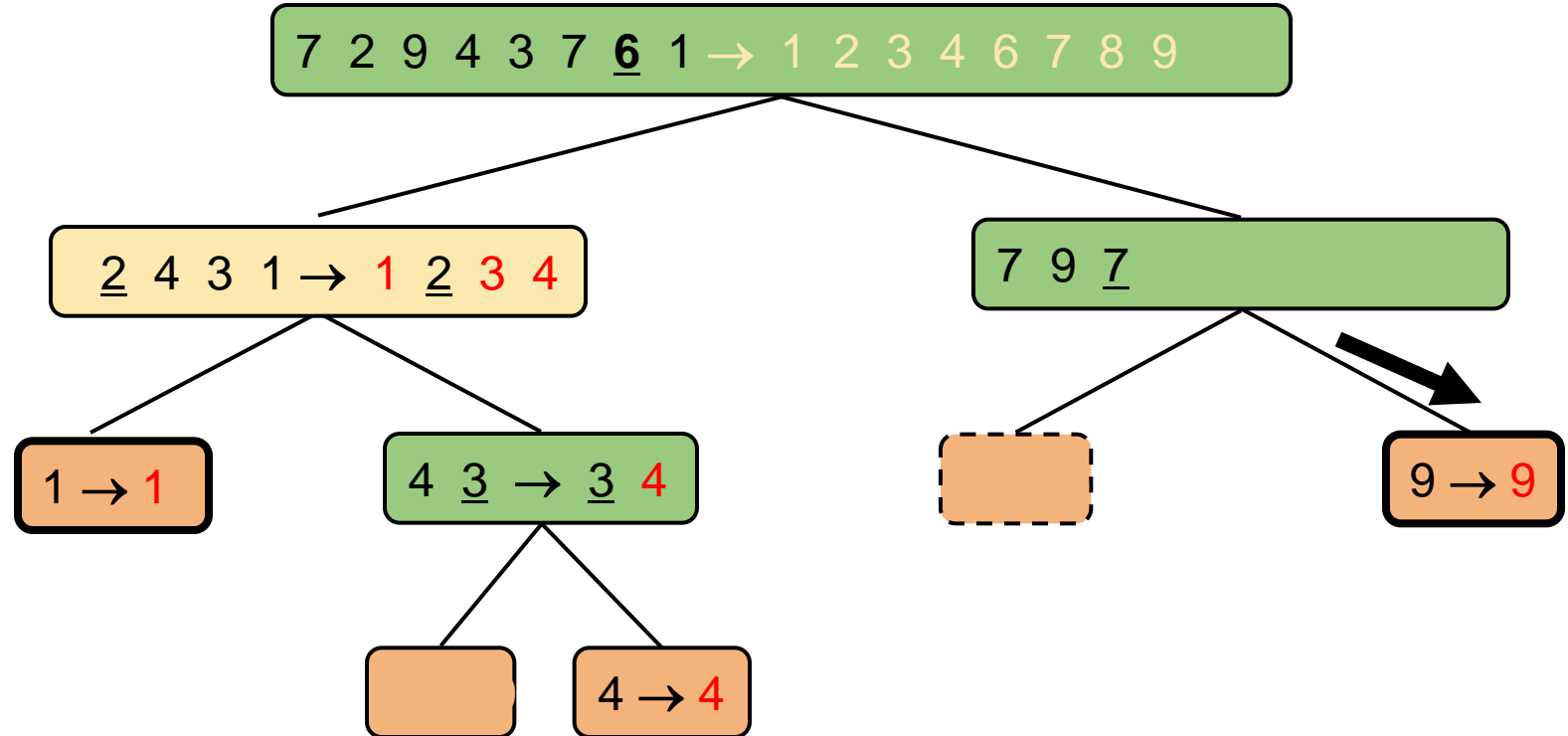
## Execution Example (cont.)

- Recursive call, ..., base case, join



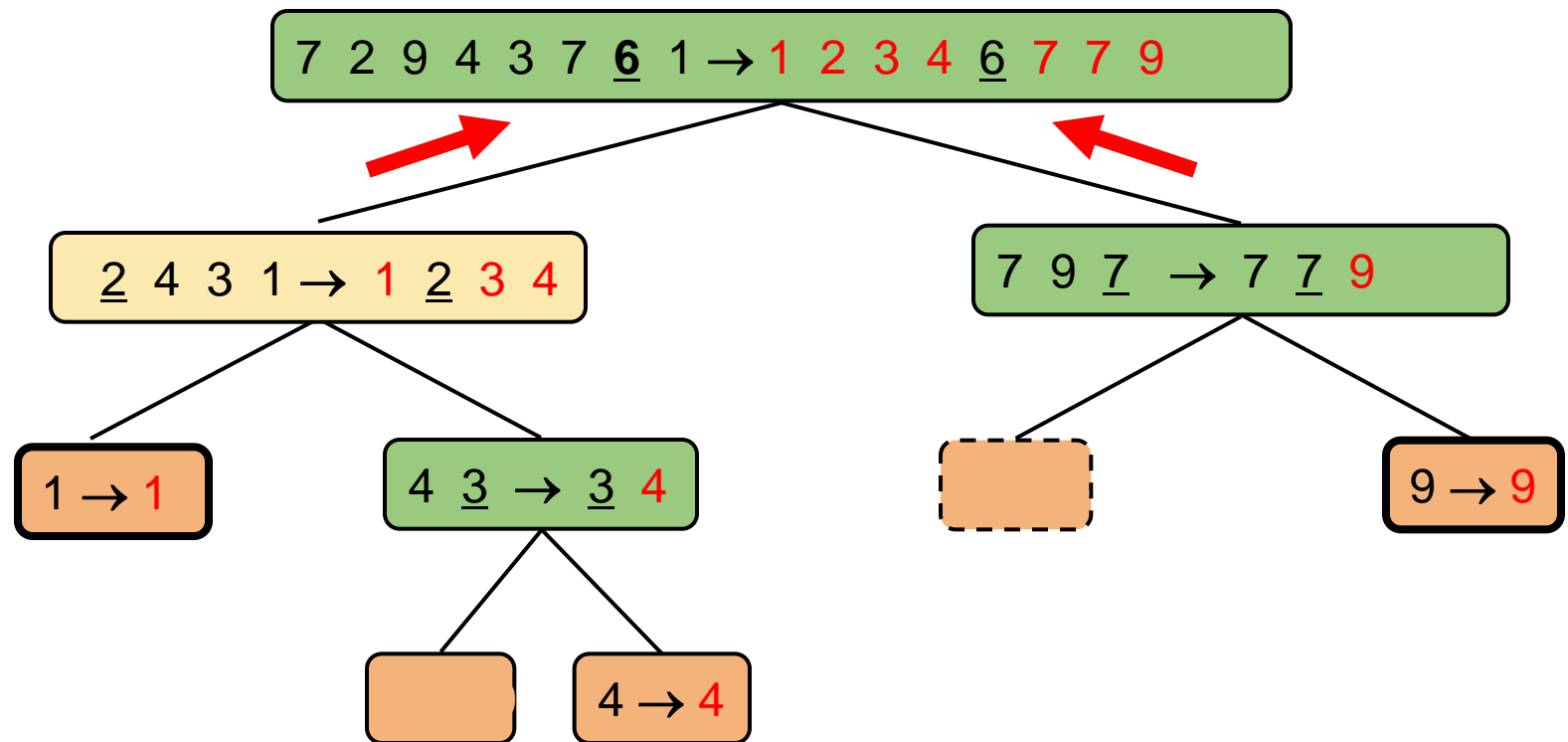
## Execution Example (cont.)

- Partition, ..., recursive call, base case



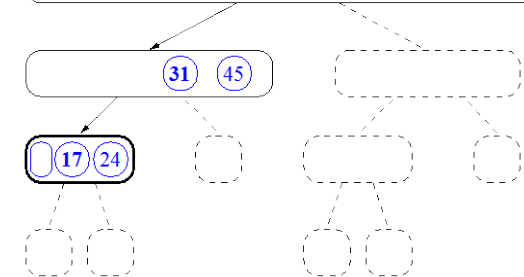
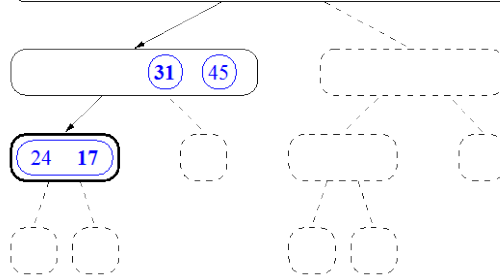
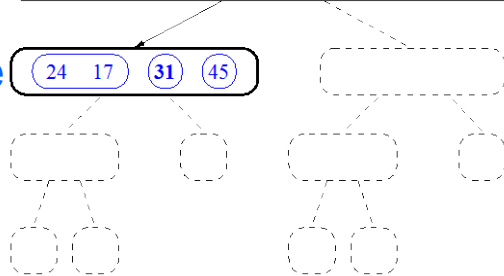
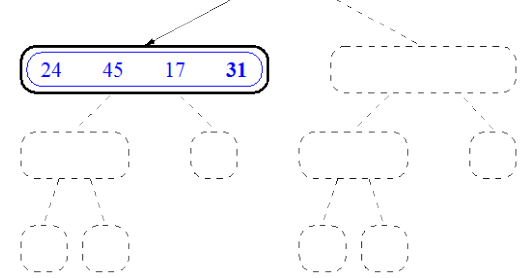
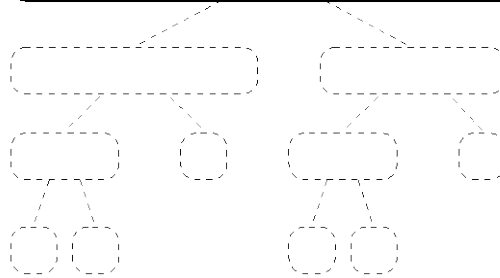
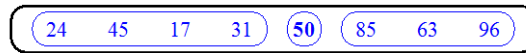
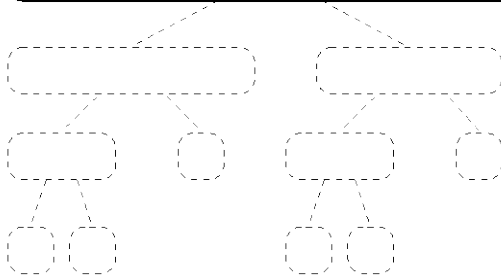
## Execution Example (cont.)

– Join, join



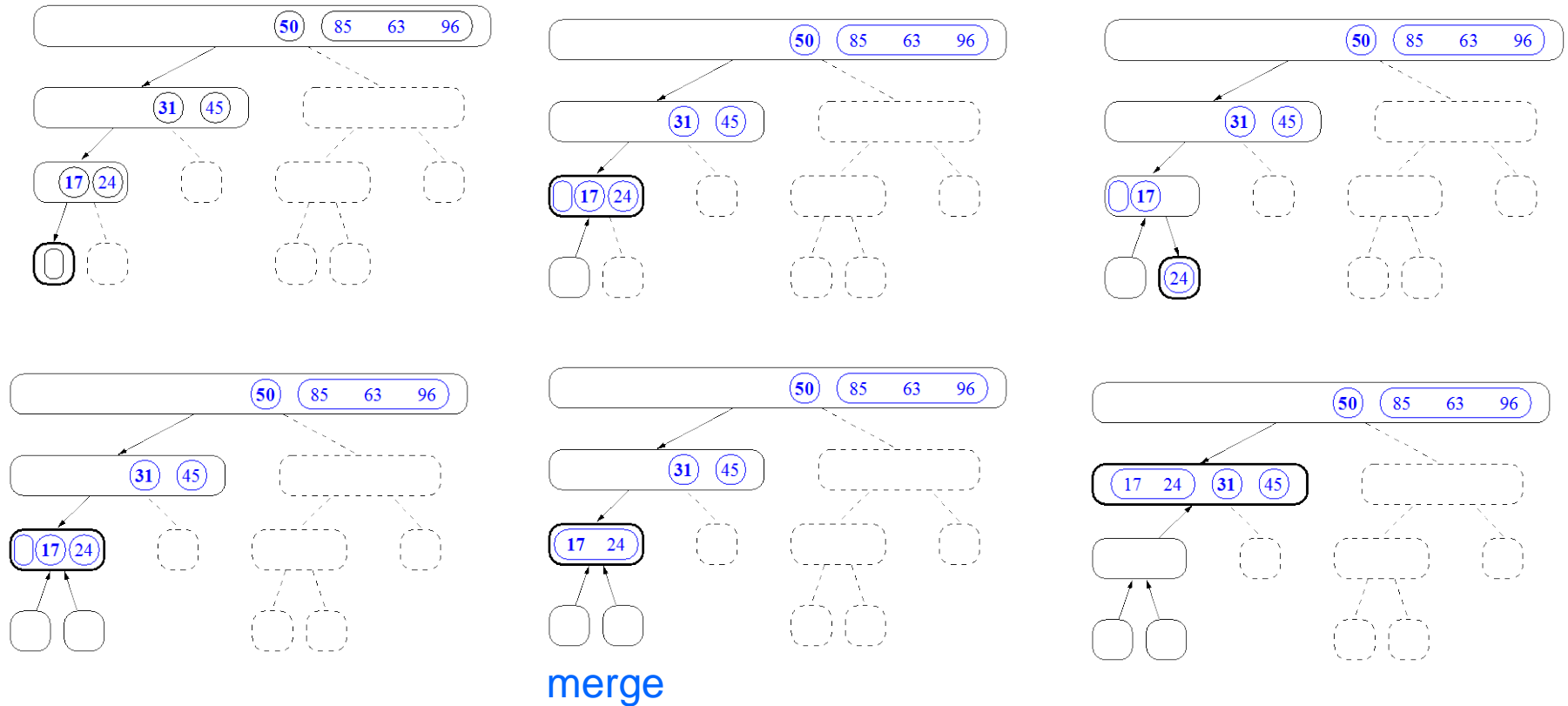
## Example 2 (pivot is the last element)

divide

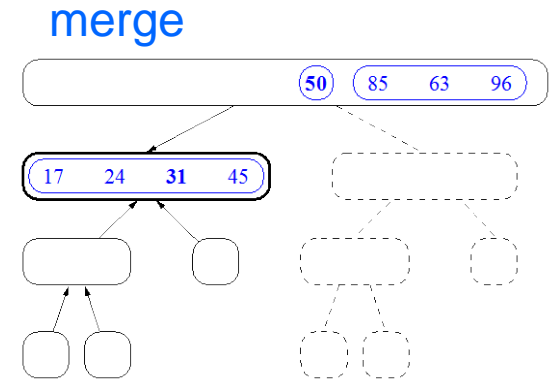
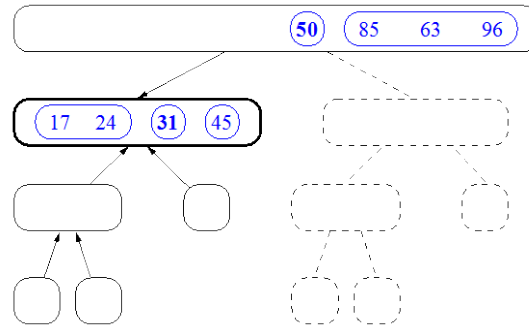
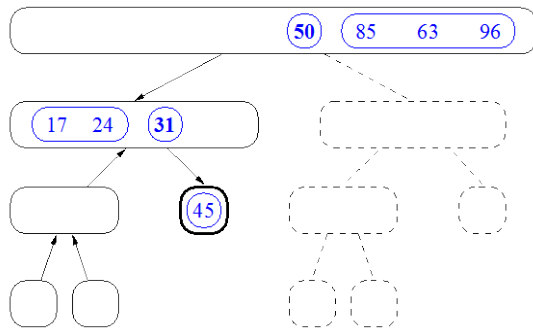


divide

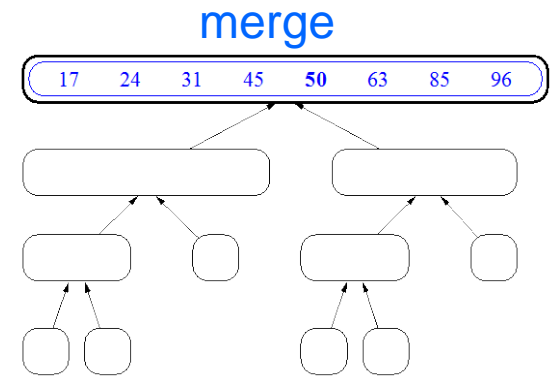
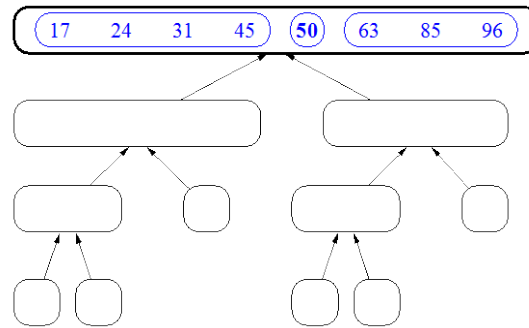
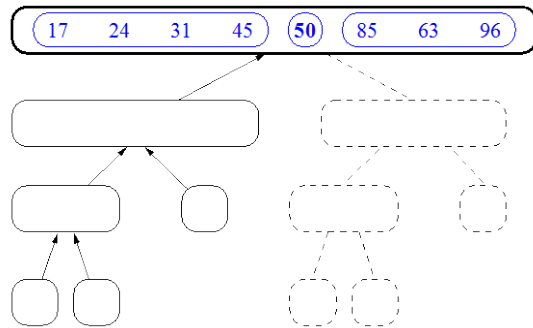
## Example 2 (pivot is the last element) cont..



## Example 2 (pivot is the last element) cont..



omitted





# Quick-sort Worst-case Running Time

- The worst case for quick-sort occurs when the **pivot** is the unique **minimum** or **maximum** element
- One of  $L$  and  $G$  has size  $n - 1$  and the other has size 0
- The running time is proportional to the sum
$$n + (n - 1) + \dots + 2 + 1$$
- Thus, the worst-case running time of quick-sort is  $O(n^2)$

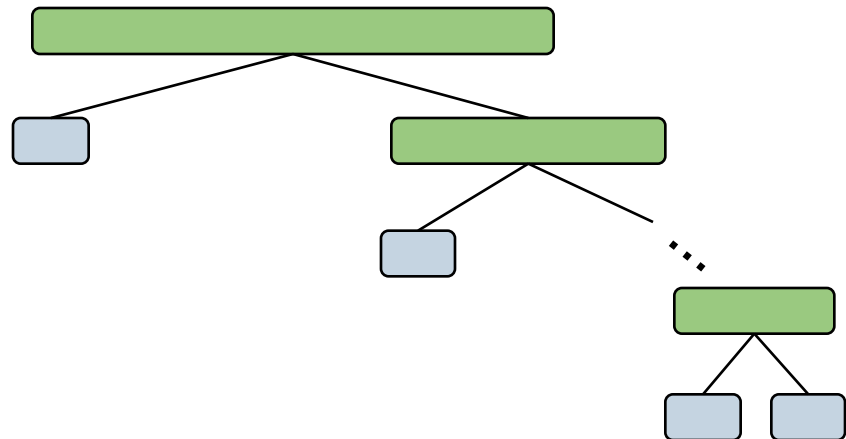
depth    time

0         $n$

1         $n - 1$

...        ...

$n - 1$     1



# Quick-sort Average-case running time

- While the **worst-case for quick-sort is  $O(n^2)$** , many (most) executions run a lot faster
  - when the pivot is closer to the **middle value** of the input for that step, then the two recurrence steps are close to half the size
    - similar to what happens in merge-sort ( $n/2$  each)
    - $n/4$  and  $3n/4$ : height  $O(\log n)$
- Fact: the **average running time is  $O(n \log n)$** 
  - take the running time of many different executions, with **pivots chosen randomly**, and average these running times
  - Expected running time: over all possible random choices (independent from input distribution)

# In-Place Quick-Sort

- Quick-sort can be implemented to run in-place
- In the **partition step**, we use replace operations to rearrange the elements of the input sequence such that
  - the elements less than the pivot have rank less than  $h$
  - the elements equal to the pivot have rank between  $h$  and  $k$
  - the elements greater than the pivot have rank greater than  $k$
- The recursive calls consider
  - elements with rank less than  $h$
  - elements with rank greater than  $k$

**Algorithm** *inPlaceQuickSort*( $S, l, r$ )

**Input** sequence  $S$ , ranks  $l$  and  $r$

**Output** sequence  $S$  with the elements of rank between  $l$  and  $r$  rearranged in increasing order

**if**  $l \geq r$

**return**

$i \leftarrow$  a random integer between  $l$  and  $r$

$x \leftarrow S.\text{elemAtRank}(i)$

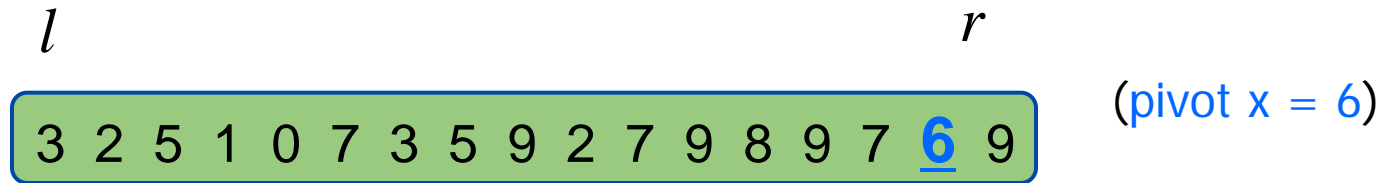
$(h, k) \leftarrow \text{inPlacePartition}(x)$

*inPlaceQuickSort*( $S, l, h - 1$ )

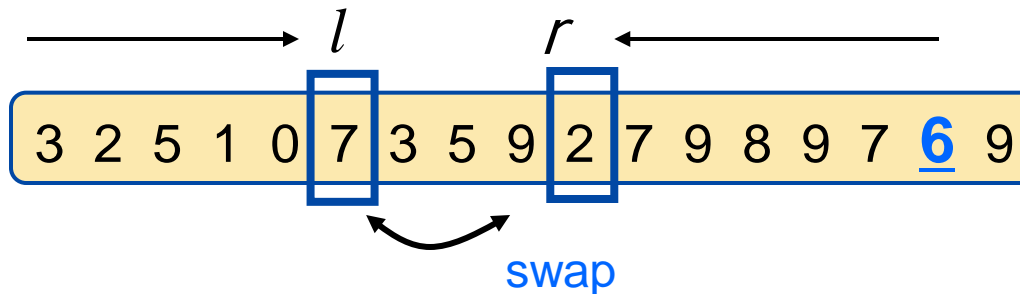
*inPlaceQuickSort*( $S, k + 1, r$ )

# In-Place Partitioning

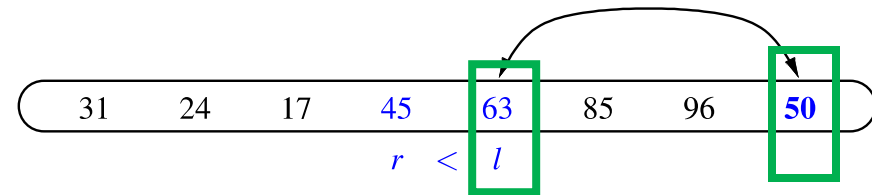
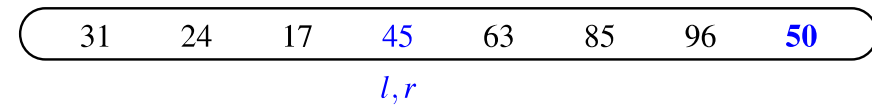
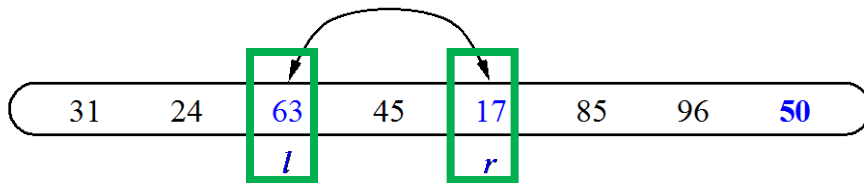
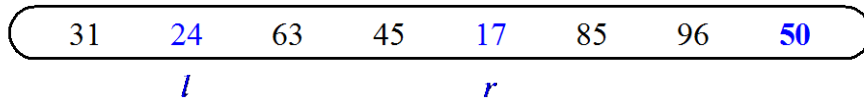
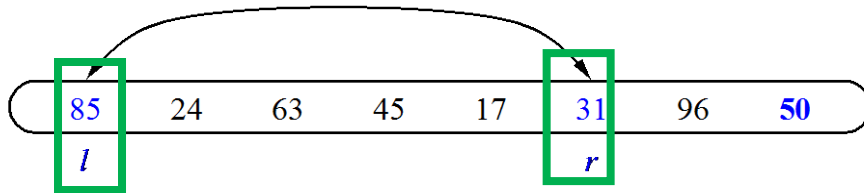
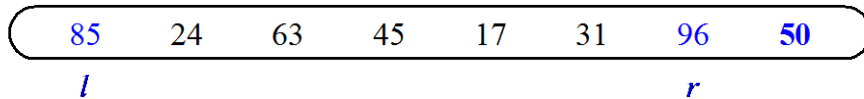
- Perform the **partition using two indices** to split  $S$  into  $L$  and  $E \cup G$  (a similar method can split  $E \cup G$  into  $E$  and  $G$ ).



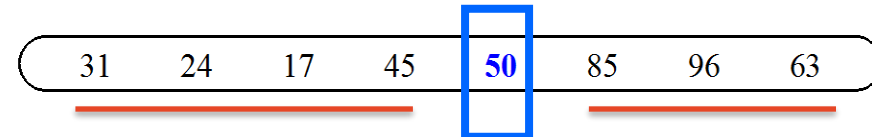
- Repeat until  $l$  and  $r$  cross:
  - Scan  $l$  to the right until finding an element  $\geq x$ .
  - Scan  $r$  to the left until finding an element  $< x$ .
  - Swap elements at indices  $l$  and  $r$



# In-Place : divide step



Put pivot in final place



Make recursive calls...

# Java Implementation

```
1  /** Sort the subarray S[a..b] inclusive. */
2  private static <K> void quickSortInPlace(K[] S, Comparator<K> comp,
3                                          int a, int b) {
4      if (a >= b) return;          // subarray is trivially sorted
5      int left = a;
6      int right = b-1;
7      K pivot = S[b];
8      K temp;                      // temp object used for swapping
9      while (left <= right) {
10         // scan until reaching value equal or larger than pivot (or right marker)
11         while (left <= right && comp.compare(S[left], pivot) < 0) left++;
12         // scan until reaching value equal or smaller than pivot (or left marker)
13         while (left <= right && comp.compare(S[right], pivot) > 0) right--;
14         if (left <= right) {      // indices did not strictly cross
15             // so swap values and shrink range
16             temp = S[left]; S[left] = S[right]; S[right] = temp;
17             left++; right--;
18         }
19     }
20     // put pivot into its final place (currently marked by left index)
21     temp = S[left]; S[left] = S[b]; S[b] = temp;
22     // make recursive calls
23     quickSortInPlace(S, comp, a, left - 1);
24     quickSortInPlace(S, comp, left + 1, b);
25 }
```

# Summary of sorting algorithms

Sort algorithm	Time cost	Comments
Selection-sort	$O(n^2)$	can be done in-place
Insertion-sort	$O(n^2)$	can be done in-place
Heap-sort	$O(n \log n)$	can be done in-place
Bubble-sort	$O(n^2)$	can be done in-place sequential access, so works well with data on disk
Merge-sort	$O(n \log n)$	can be done in-place sequential access, so works well with data on disk
Quick-sort	worst: $O(n^2)$ Average/expected: $O(n \log n)$	can be done in-place requires randomization

Lower bound of comparison-based sorting algorithm:  $\Omega(n \log n)$  necessary  
Linear-time sorting: Bucket sort, Radix sort (special assumption on input)

# Summary

- Sorting algorithms and their costs
  - Review of pq-sorting algorithms: insertion, selection sort, heap-sort
  - In-place sorting
  - **Bubble sort**
  - **Merge-sort (section 13.1)**
  - **Quick-sort (section 13.2)**