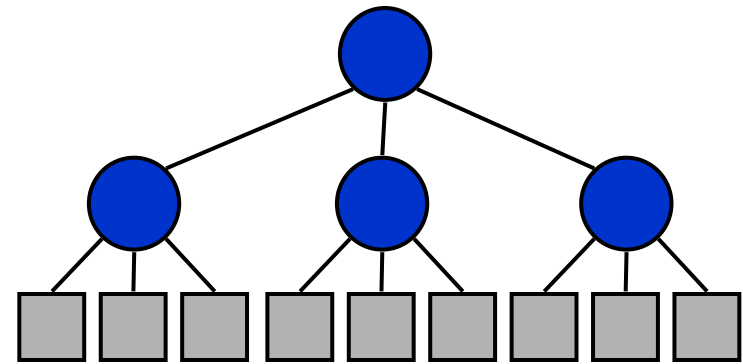




Divide-and-Conquer Technique: Merge Sort, Quick Sort

Divide-and-Conquer

- **Divide-and-Conquer** is a general algorithm design paradigm:
 - **Divide** the problem into a number of subproblems that are smaller instances of the same problem
 - **Conquer** the subproblems by solving them recursively
 - **Combine** the solutions to the subproblems into the solution for the original problem
- The base case for the recursion are subproblems of constant size
- Analysis can be done using **recurrence equations**



Merge Sort and Quick Sort

Two well-known sorting algorithms adopt this divide-and-conquer strategy

- Merge sort
 - Divide step is trivial – just split the list into two equal parts
 - Work is carried out in the conquer step by merging two sorted lists
- Quick sort
 - Work is carried out in the divide step using a pivot element
 - Conquer step is trivial

Merge Sort: Algorithm

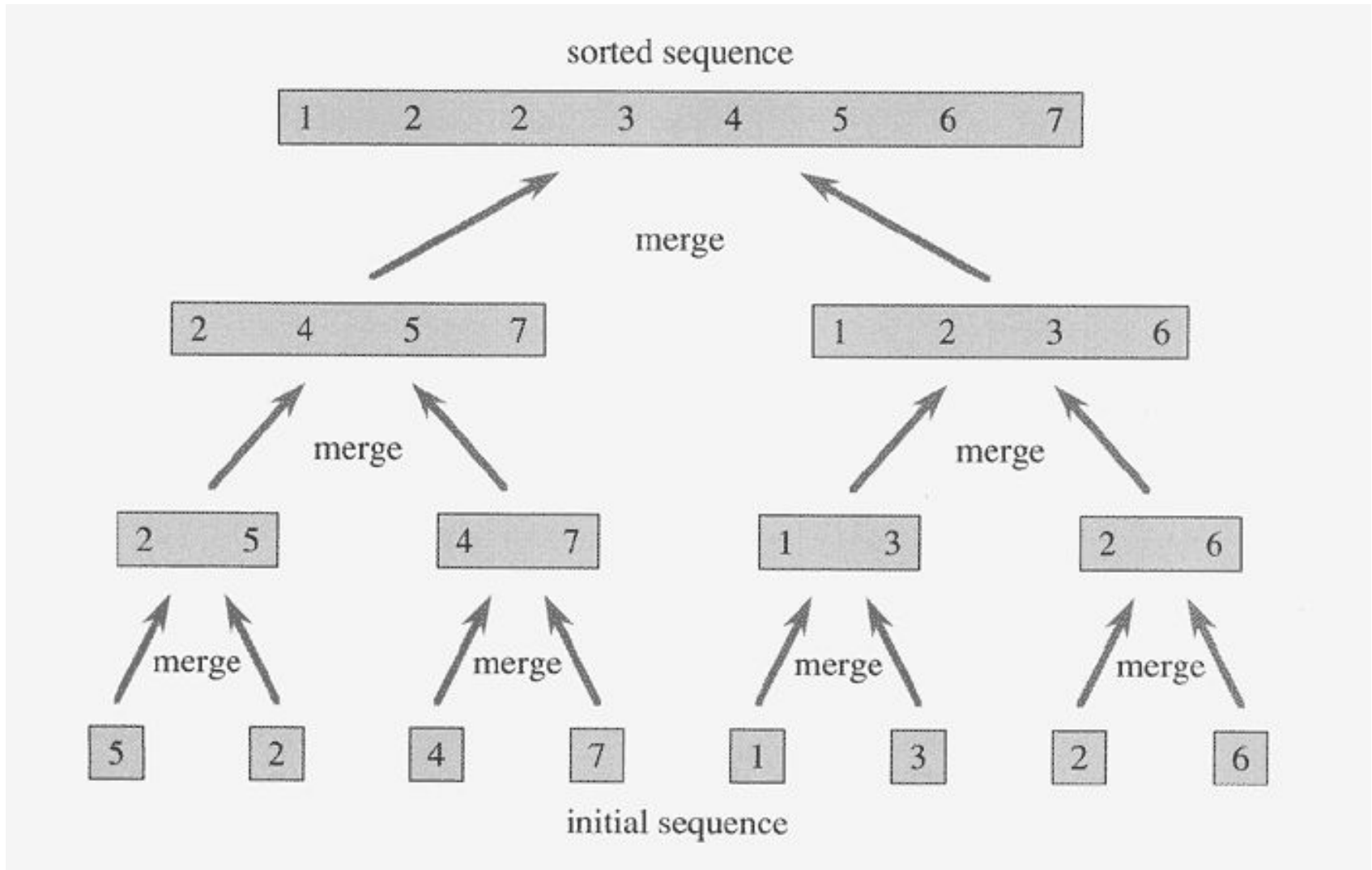
MERGE-SORT(A, p, r)

```
1  if  $p < r$ 
2      then  $q \leftarrow \lfloor (p + r)/2 \rfloor$ 
3          MERGE-SORT( $A, p, q$ )
4          MERGE-SORT( $A, q + 1, r$ )
5          MERGE( $A, p, q, r$ )
```

Merge Sort: Algorithm

```
MERGE( $A, p, q, r$ )
1   $n_1 \leftarrow q - p + 1$ 
2   $n_2 \leftarrow r - q$ 
3  create arrays  $L[1 \dots n_1 + 1]$  and  $R[1 \dots n_2 + 1]$ 
4  for  $i \leftarrow 1$  to  $n_1$ 
5      do  $L[i] \leftarrow A[p + i - 1]$ 
6  for  $j \leftarrow 1$  to  $n_2$ 
7      do  $R[j] \leftarrow A[q + j]$ 
8   $L[n_1 + 1] \leftarrow \infty$ 
9   $R[n_2 + 1] \leftarrow \infty$ 
10  $i \leftarrow 1$ 
11  $j \leftarrow 1$ 
12 for  $k \leftarrow p$  to  $r$ 
13     do if  $L[i] \leq R[j]$ 
14         then  $A[k] \leftarrow L[i]$ 
15              $i \leftarrow i + 1$ 
16         else  $A[k] \leftarrow R[j]$ 
17              $j \leftarrow j + 1$ 
```

Merge Sort: Example



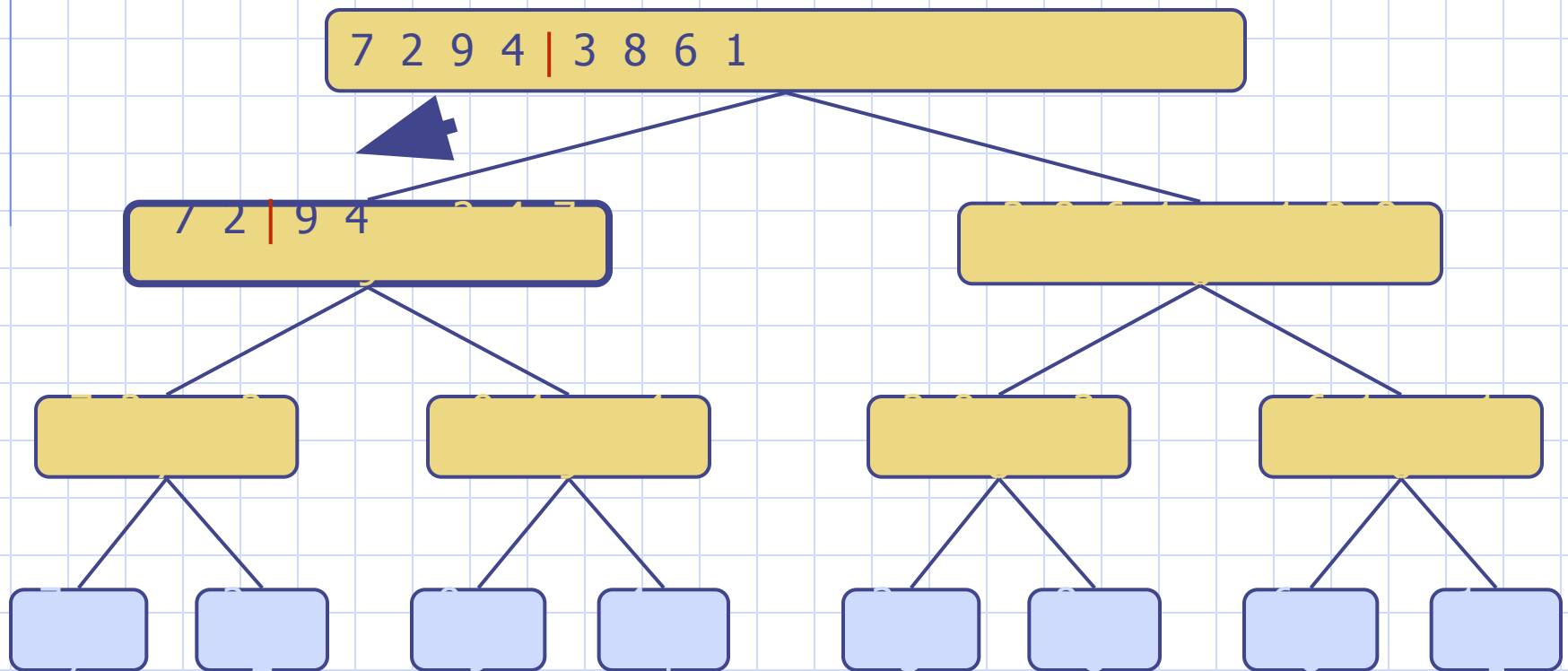
- Partition

- Partition



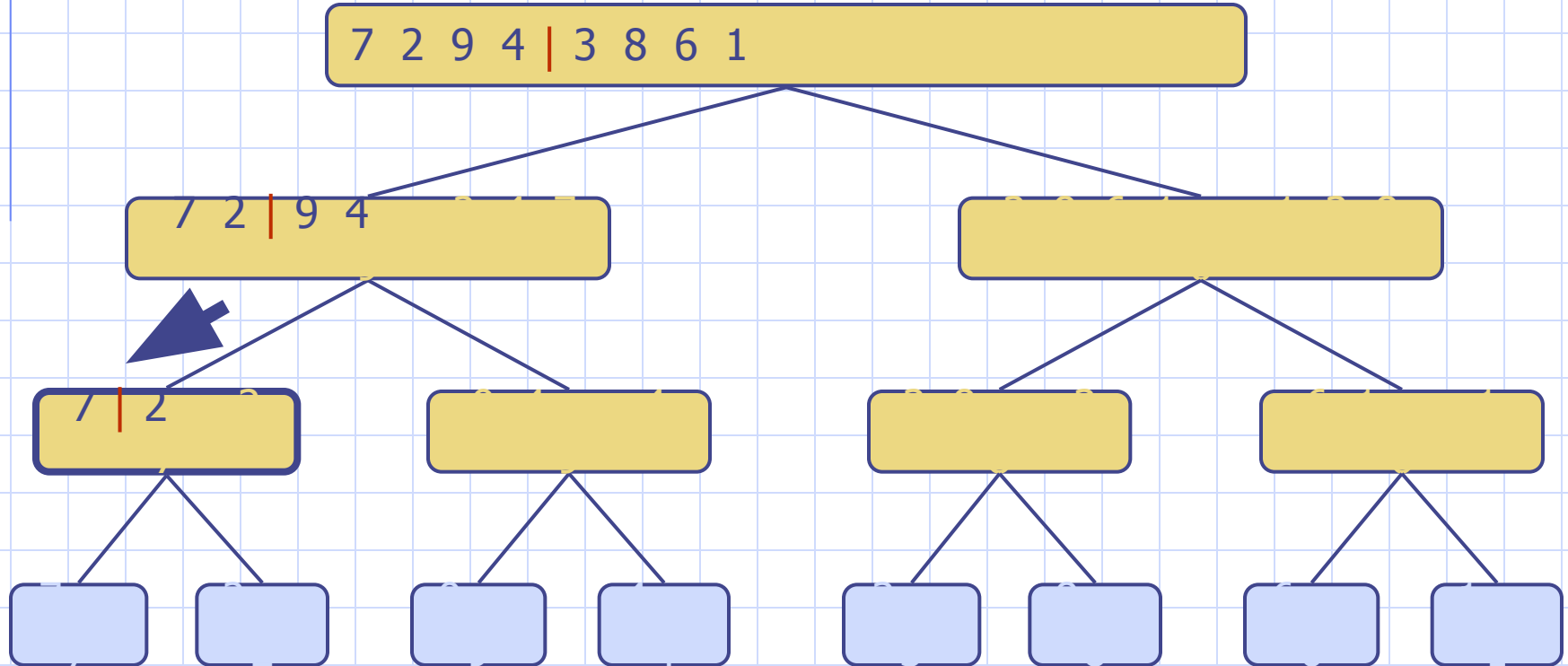
Execution Example

- Recursive call, partition



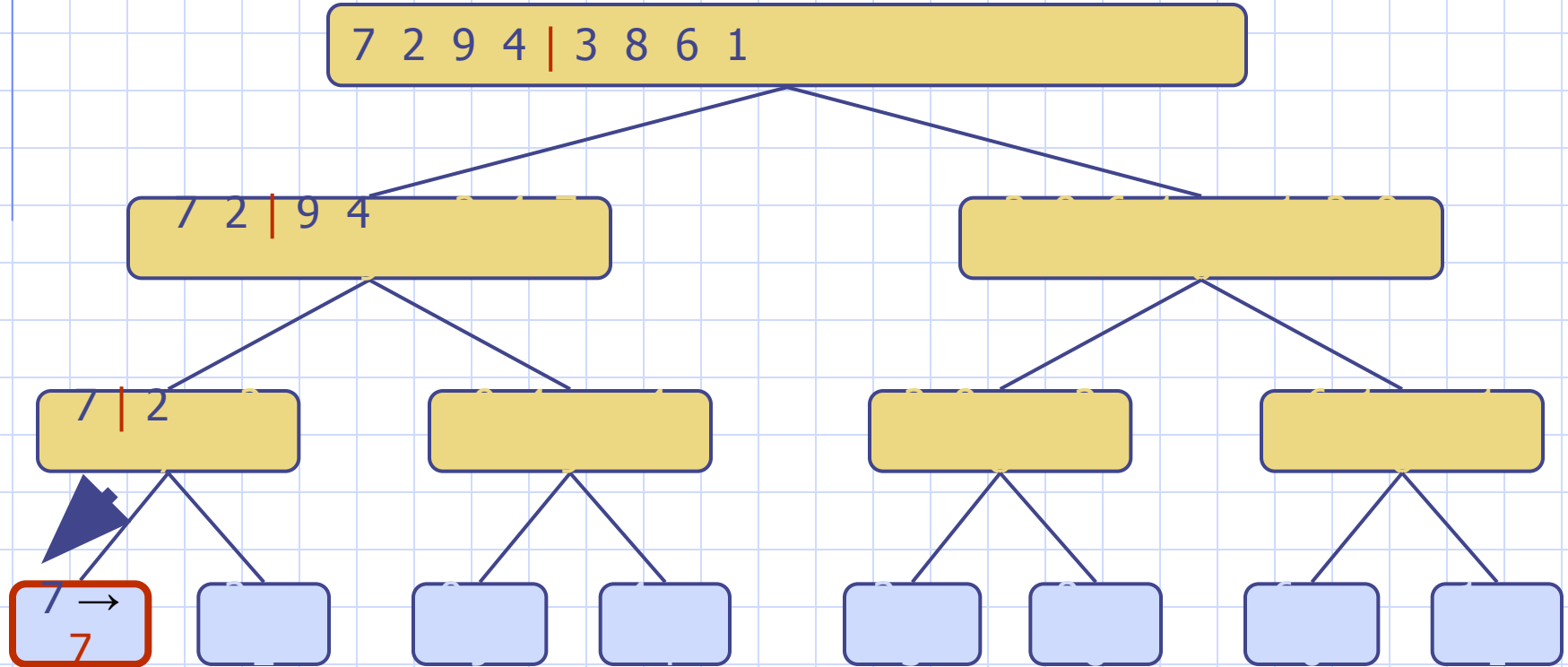
Execution Example

- Recursive call, partition



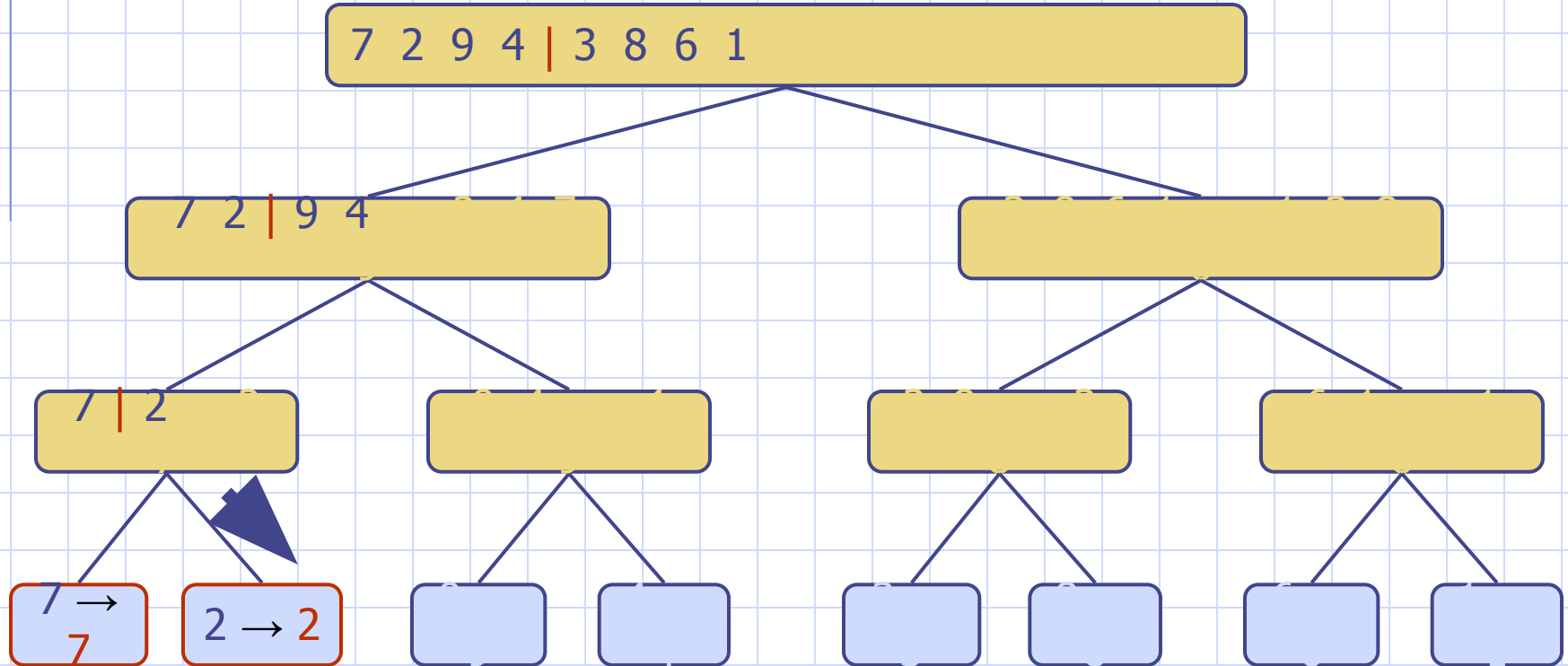
Execution Example

- Recursive call, base case



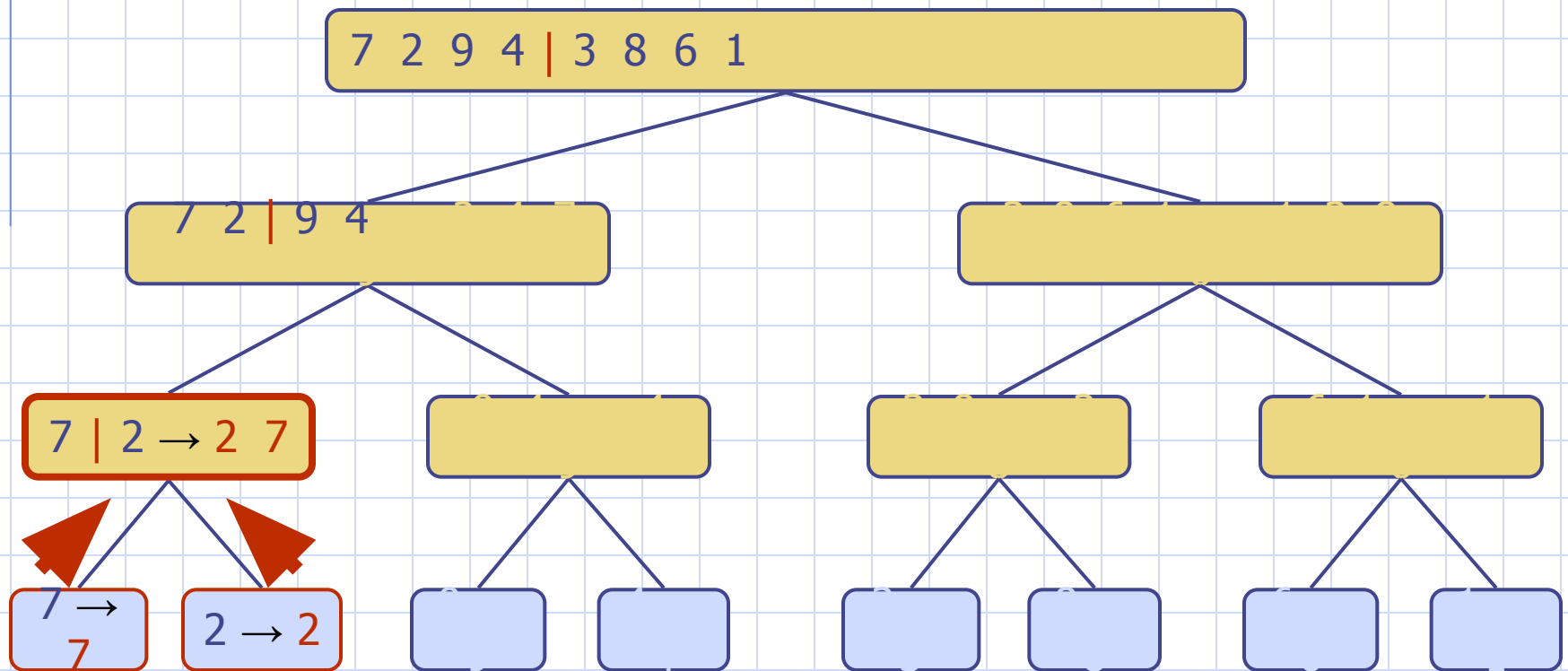
Execution Example

- Recursive call, base case



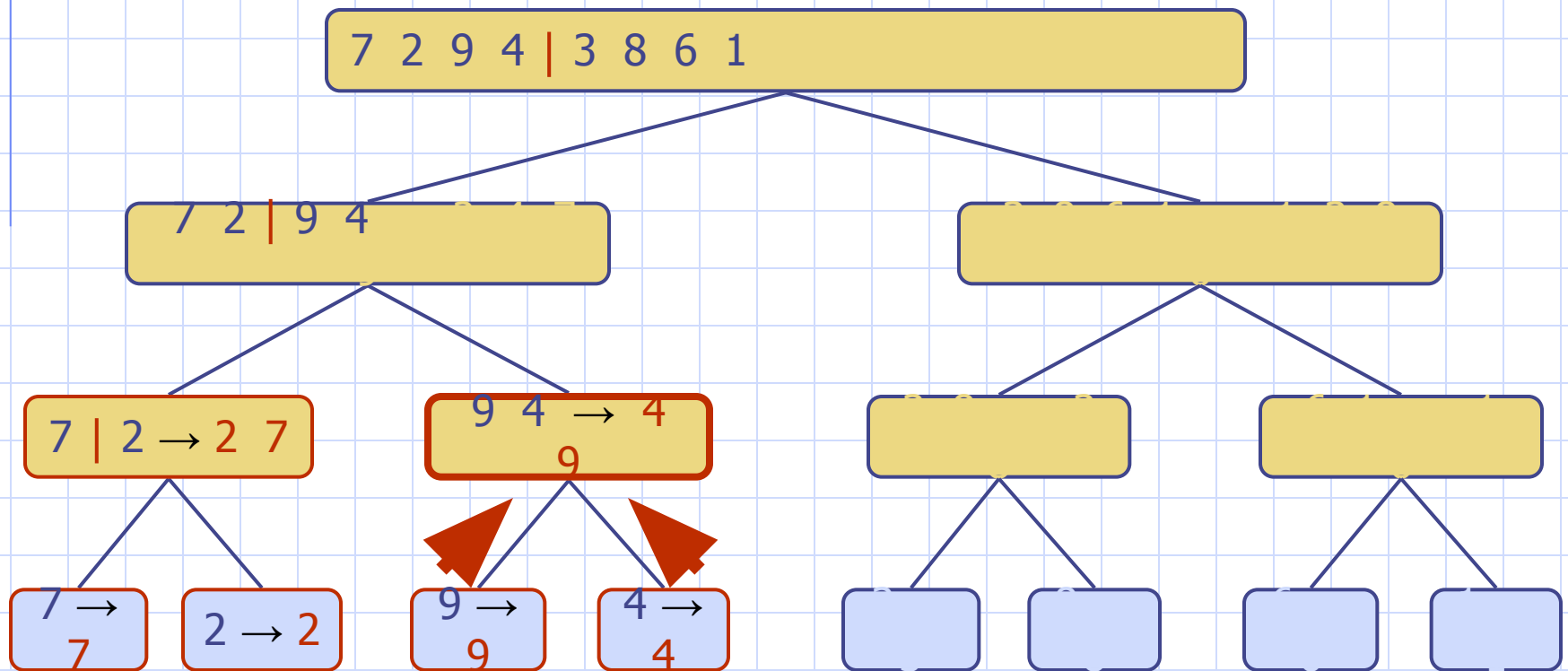
Execution Example

- Merge



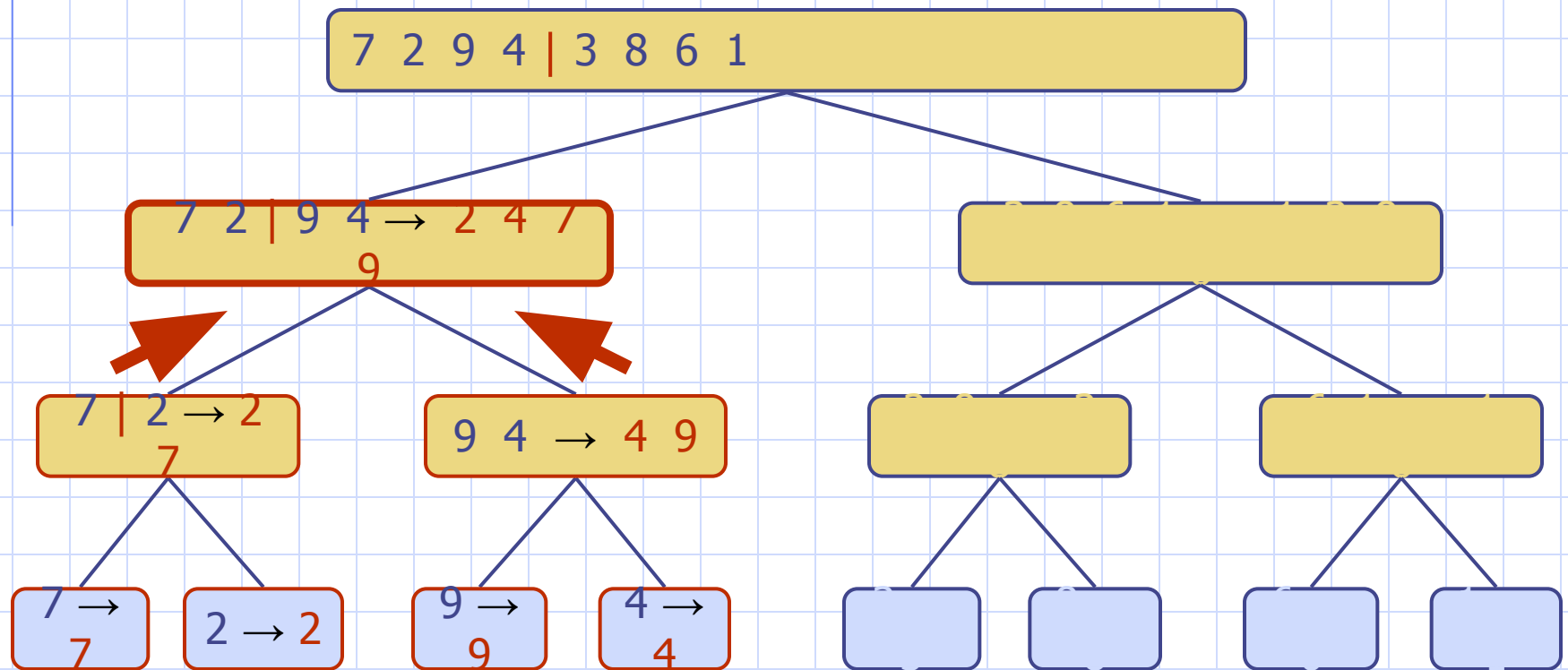
Execution Example

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



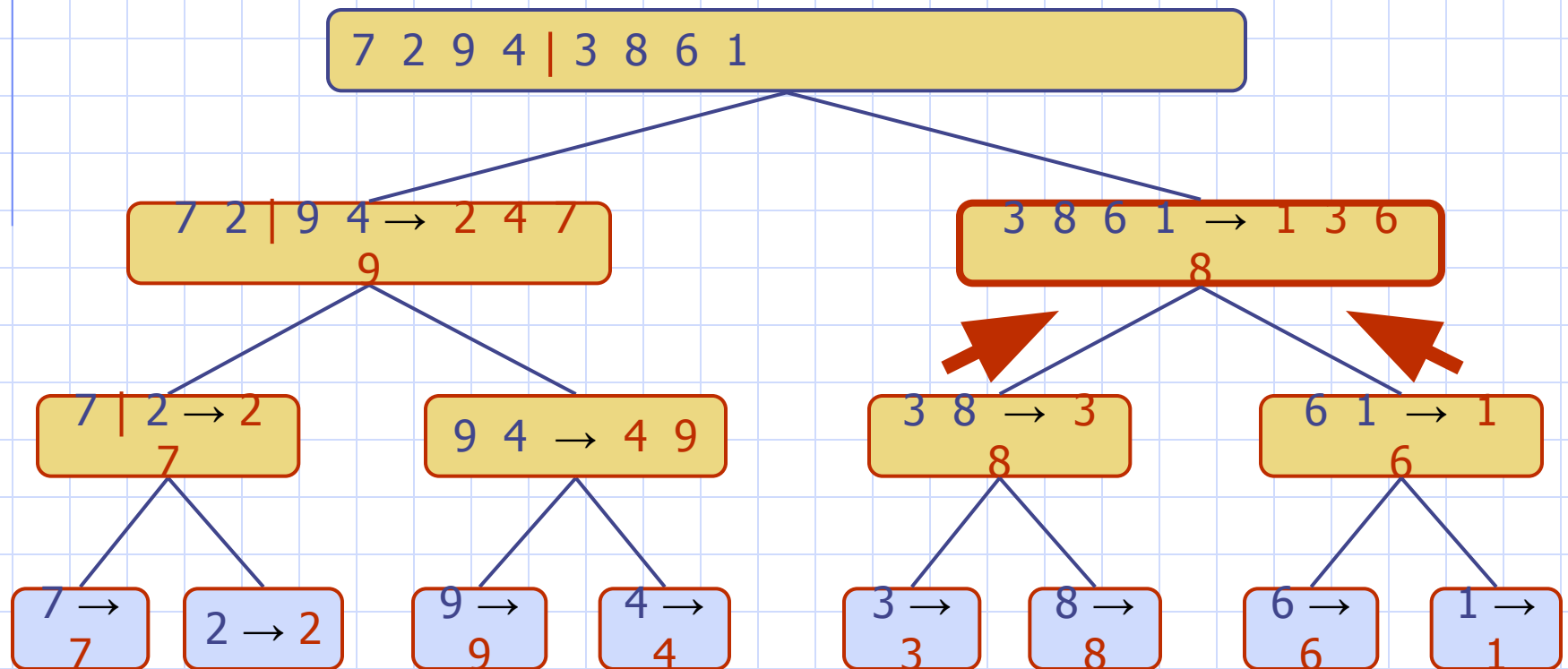
Execution Example

- Merge



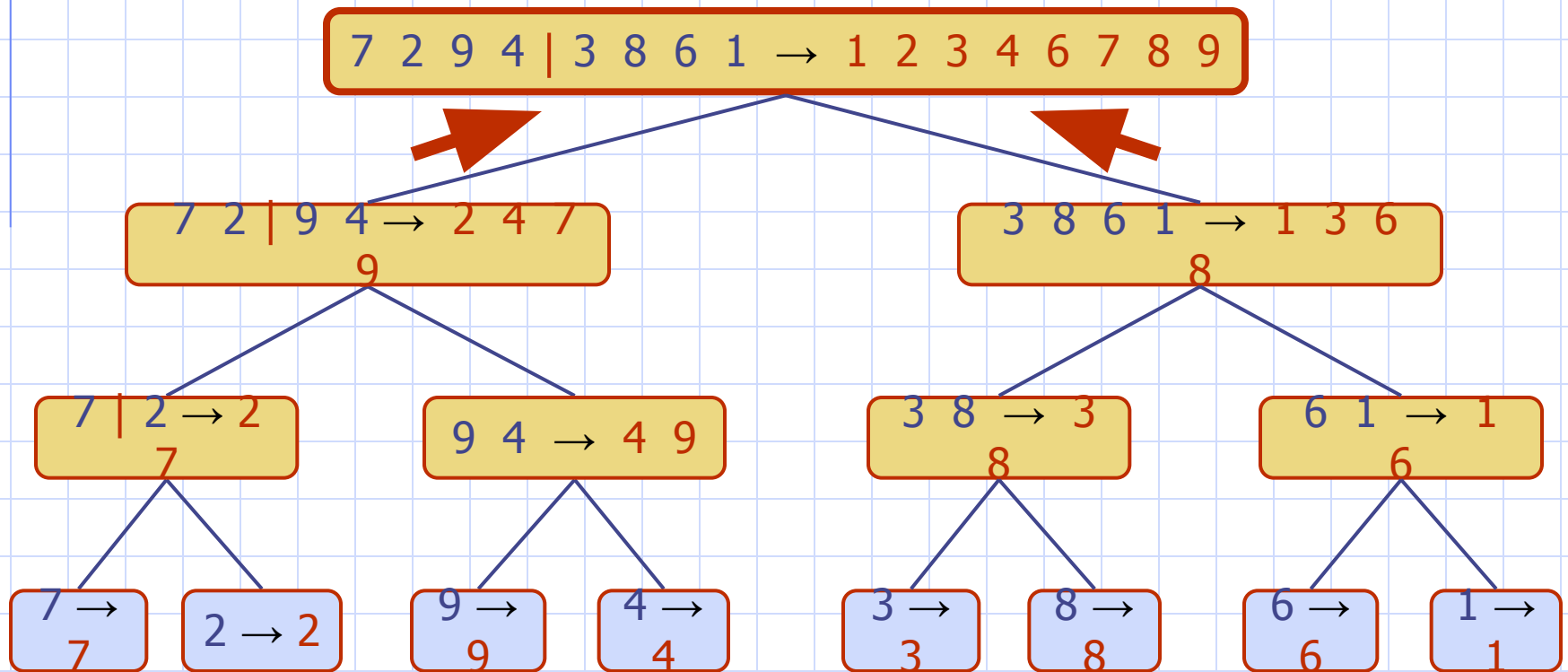
Execution Example

- Recursive call, ..., merge, merge



Execution Example

- Merge



Merge Sort: Running Time

The recurrence for the worst-case running time $T(n)$ is

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 2T(n/2) + \Theta(n) & \text{if } n > 1 \end{cases}$$

equivalently

$$T(n) = \begin{cases} b & \text{if } n = 1 \\ 2T(n/2) + bn & \text{if } n > 1 \end{cases}$$

Solve this recurrence by

- (1) iteratively expansion
- (2) using the recursion tree

Merge Sort: Running Time (Iterative Expansion)

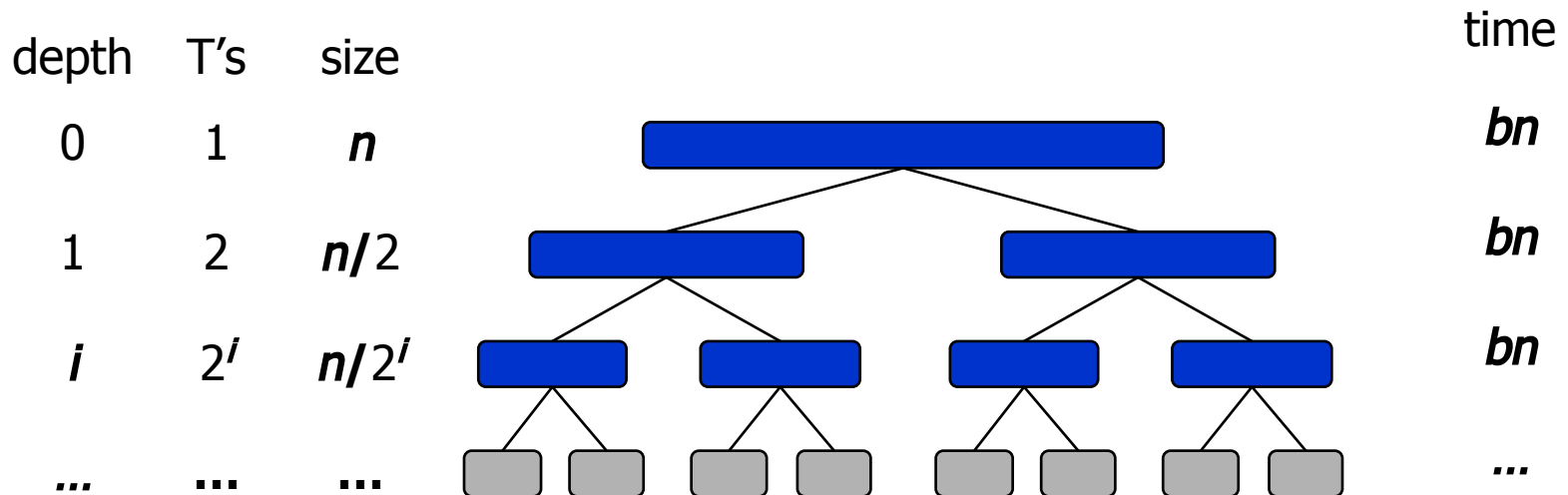
$$\begin{aligned}T(n) &= 2T(n/2) + bn \\&= 2(2T(n/2^2)) + b(n/2) + bn \\&= 2^2T(n/2^2) + 2bn \\&= 2^3T(n/2^3) + 3bn \\&= 2^4T(n/2^4) + 4bn \\&= \dots \\&= 2^i T(n/2^i) + ibn\end{aligned}$$

- Note that base, $T(n) = b$, case occurs when $2^i = n$.
That is, $i = \log n$.
- So, $T(n) = bn + bn \log n$
- Thus, $T(n)$ is $O(n \log n)$.

Merge Sort: Running Time (Recursion Tree)

- Draw the recursion tree for the recurrence relation and look for a pattern:

$$T(n) = \begin{cases} b & \text{if } n = 1 \\ 2T(n/2) + bn & \text{if } n \geq 2 \end{cases}$$



Total time = $bn + bn \log n$
(last level plus all previous levels)

Quick Sort: Algorithm

- Another divide-and-conquer algorithm
 - The array $A[p..r]$ is *partitioned* into two non-empty subarrays $A[p..q]$ and $A[q+1..r]$
 - Invariant: All elements in $A[p..q]$ are less than all elements in $A[q+1..r]$
 - The subarrays are recursively sorted by calls to quicksort
 - Unlike merge sort, no combining step: two subarrays form an already-sorted array

Quick Sort: Algorithm

QUICKSORT(A, p, r)

```
1  if  $p < r$ 
2      then  $q \leftarrow \text{PARTITION}(A, p, r)$ 
3          QUICKSORT( $A, p, q - 1$ )
4          QUICKSORT( $A, q + 1, r$ )
```

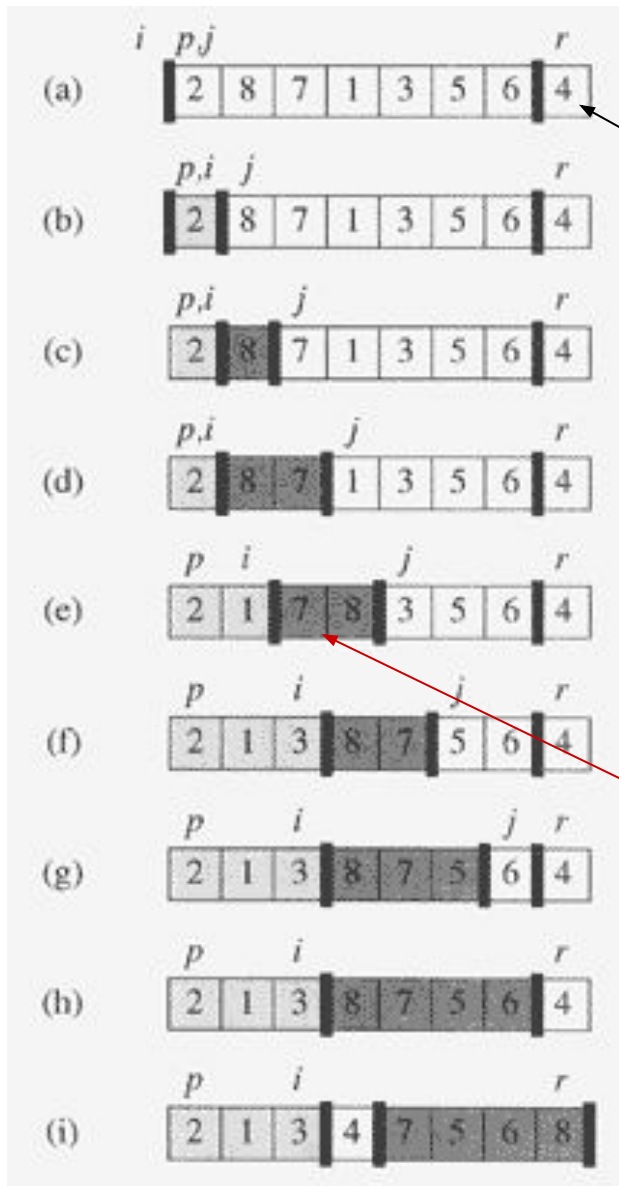
PARTITION(A, p, r)

```
1   $x \leftarrow A[r]$ 
2   $i \leftarrow p - 1$ 
3  for  $j \leftarrow p$  to  $r - 1$ 
4      do if  $A[j] \leq x$ 
5          then  $i \leftarrow i + 1$ 
6              exchange  $A[i] \leftrightarrow A[j]$ 
7  exchange  $A[i + 1] \leftrightarrow A[r]$ 
8  return  $i + 1$ 
```

Quick Sort: Algorithm (Partition)

- Clearly, all the actions take place in the **partition()** function
 - Rearranges the subarrays in place
 - End result:
 - Two subarrays
 - All values in first subarray \leq all values in the second
 - Returns the index of the “pivot” element separating the two subarrays

Quick Sort: Algorithm



pivot

From $i + 1$ to j is a window of elements $> A[r]$. The cursor j moves right one step at a time.

If the cursor j "discovers" an element $\leq A[r]$, then this element is swapped with the front element of the window, effectively moving the window right one step; if it discovers an element $> A[r]$, then the window simply becomes longer one unit.

Quick Sort: Algorithm

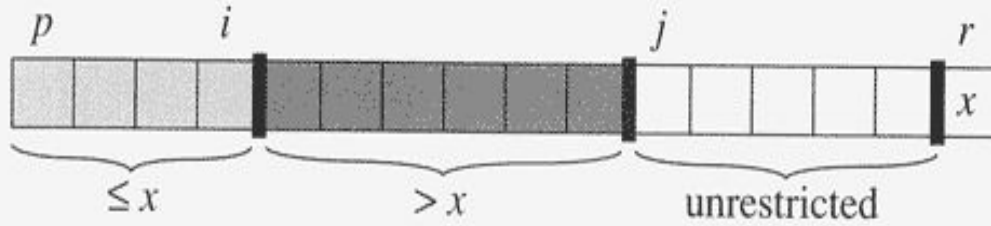


Figure 7.2 The four regions maintained by the procedure PARTITION on a subarray $A[p..r]$. The values in $A[p..i]$ are all less than or equal to x , the values in $A[i+1..j-1]$ are all greater than x , and $A[r] = x$. The values in $A[j..r-1]$ can take on any values.

Quick Sort: Analysis

- *What will be the worst case for the algorithm?*
 - Partition is always unbalanced
- *What will be the best case for the algorithm?*
 - Partition is perfectly balanced
- *Which is more likely?*
 - The latter, by far, except...
- *Will any particular input elicit the worst case?*
 - Yes: Already-sorted input

Quick Sort: Analysis

- **In the worst case:**

$$T(1) = \Theta(1)$$

$$T(n) = T(n - 1) + \Theta(n)$$

Works out to

$$T(n) = \Theta(n^2)$$

- **In the best case:**

$$T(1) = \Theta(1)$$

$$T(n) = 2T(n/2) + \Theta(n)$$

Works out to

$$T(n) = \Theta(n \lg n)$$

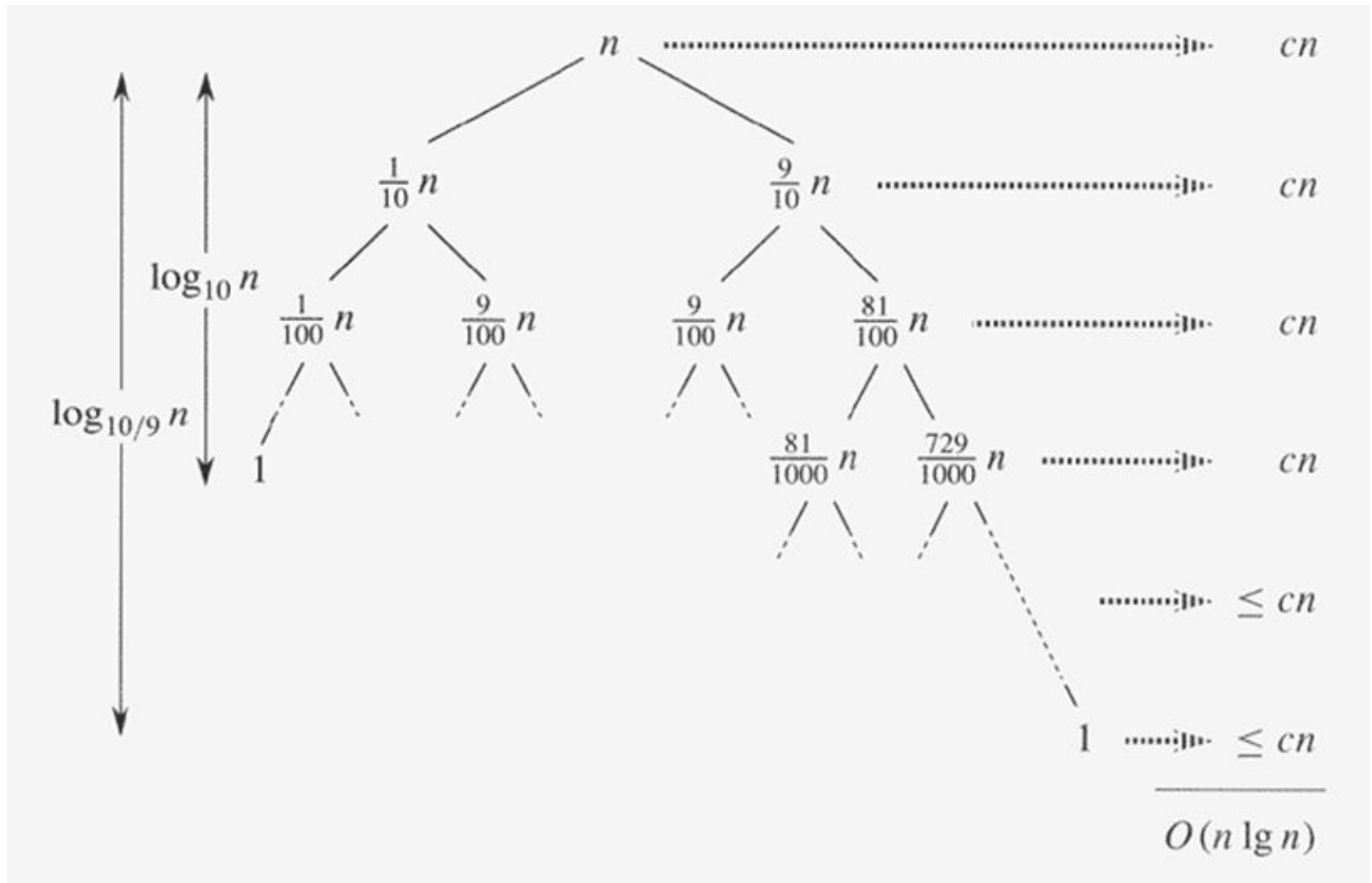
Quick Sort: Analysis

- The real liability of quicksort is that it runs in $O(n^2)$ on already-sorted input
- Book discusses two solutions:
 - Randomize the input array, OR
 - *Pick a random pivot element*
- *How will these solve the problem?*
 - By ensuring that no particular input can be chosen to make quicksort run in $O(n^2)$ time

Analyzing Quicksort: Average Case

- Assuming random input, average-case running time is much closer to $O(n \lg n)$ than $O(n^2)$
- First, a more intuitive explanation/example:
 - Suppose that partition() always produces a 9-to-1 split. This looks quite unbalanced!
 - The recurrence is thus:
$$T(n) = T(9n/10) + T(n/10) + n$$
 - *How deep will the recursion go?* (draw it)

Analyzing Quicksort: Average Case



Analyzing Quicksort: Average Case

- Intuitively, a real-life run of quicksort will produce a mix of “bad” and “good” splits
 - Randomly distributed among the recursion tree
 - Pretend for intuition that they alternate between best-case ($n/2 : n/2$) and worst-case ($n-1 : 1$)
 - *What happens if we bad-split root node, then good-split the resulting size $(n-1)$ node?*

Analyzing Quicksort: Average Case

- Intuitively, a real-life run of quicksort will produce a mix of “bad” and “good” splits
 - Randomly distributed among the recursion tree
 - Pretend for intuition that they alternate between best-case ($n/2 : n/2$) and worst-case ($n-1 : 1$)
 - *What happens if we bad-split root node, then good-split the resulting size $(n-1)$ node?*
 - We end up with three subarrays, size 1, $(n-1)/2$, $(n-1)/2$
 - Combined cost of splits = $n + n - 1 = 2n - 1 = O(n)$
 - No worse than if we had good-split the root node!

Analyzing Quicksort: Average Case

- Intuitively, the $O(n)$ cost of a bad split (or 2 or 3 bad splits) can be absorbed into the $O(n)$ cost of each good split
- Thus running time of alternating bad and good splits is still $O(n \lg n)$, with slightly higher constants
- How can we be more rigorous?