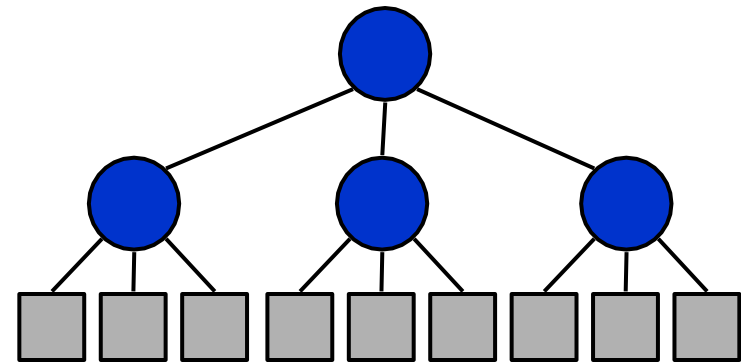




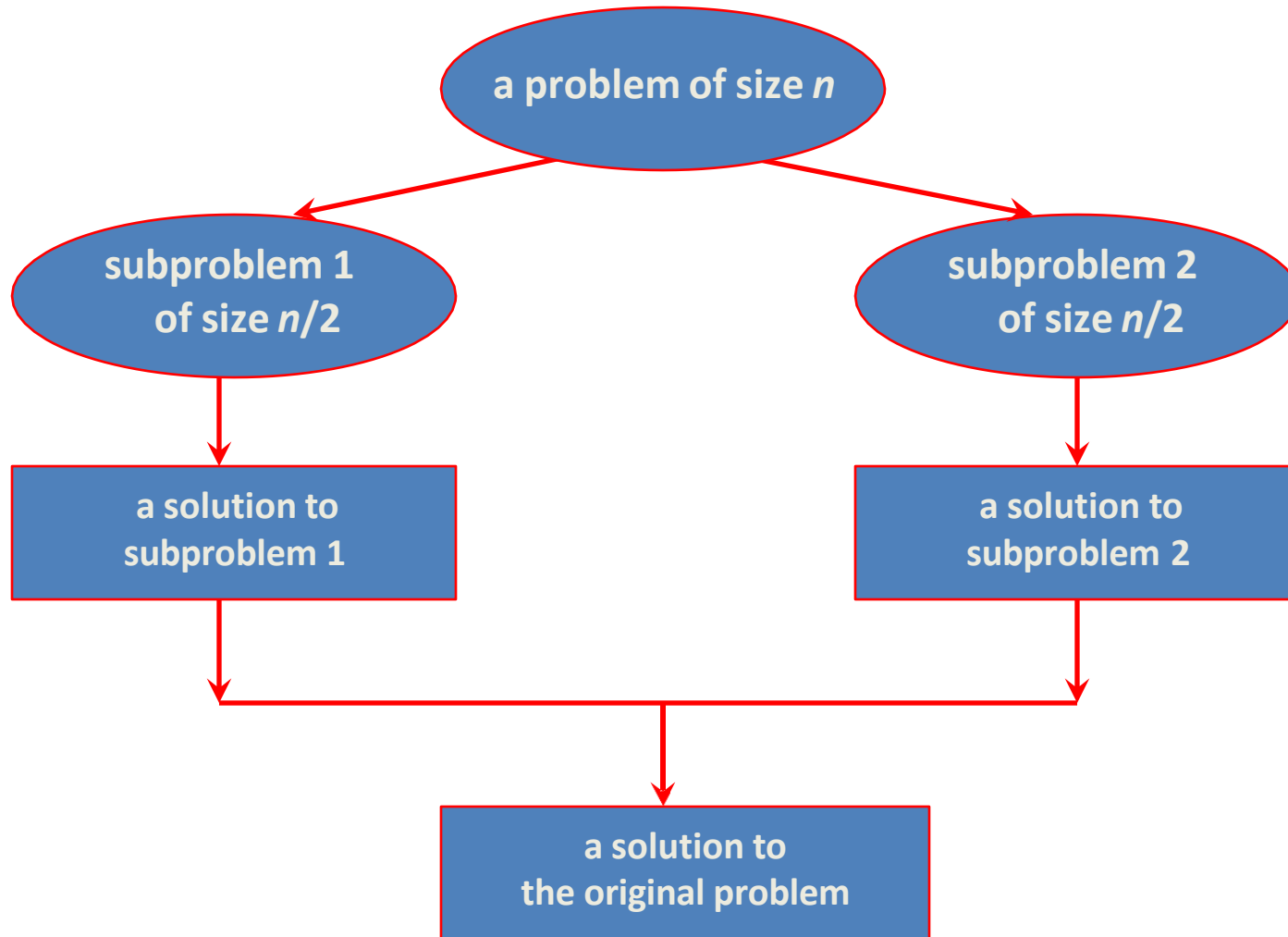
Divide-and-Conquer Technique: Merge Sort

Divide-and-Conquer

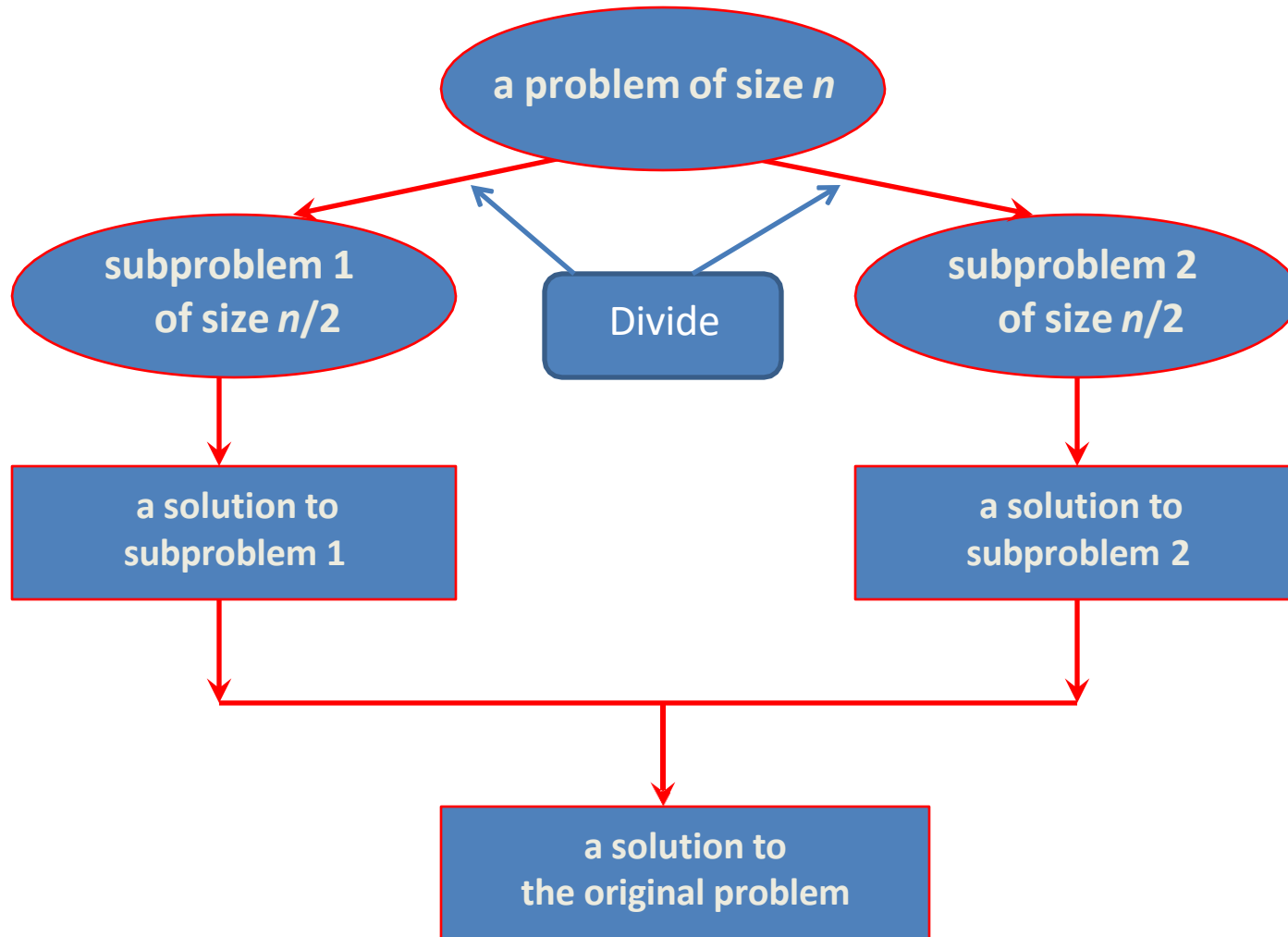
- **Divide-and-Conquer** is a general algorithm design paradigm:
 - **Divide** the problem into a number of subproblems that are smaller instances of the same problem
 - **Conquer** the subproblems by solving them recursively
 - **Combine** the solutions to the subproblems into the solution for the original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



Divide-and-Conquer



Divide-and-Conquer



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

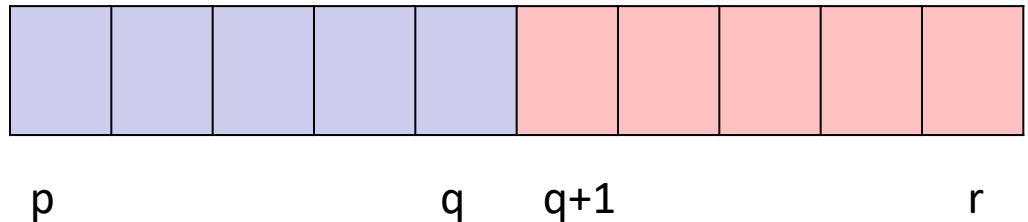
- Quick sort

- 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( $A, p, q, r$ )
```

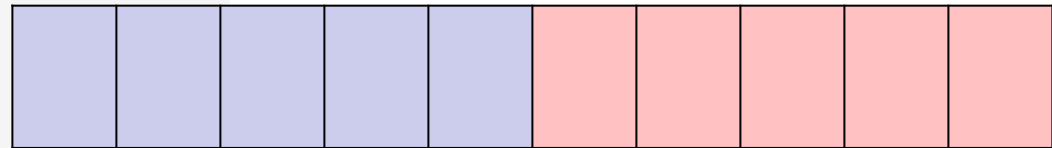


Merge Sort: Algorithm

MERGE(A, p, q, r)

```
1   $n_1 \leftarrow q - p + 1$ 
2   $n_2 \leftarrow r - q$ 
3  create arrays  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$ 
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          $j \leftarrow j + 1$ 
```

A



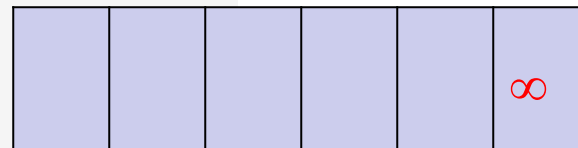
p

q

$q+1$

r

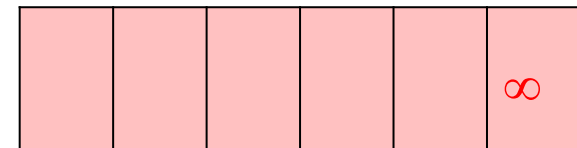
L



1

n_1+1

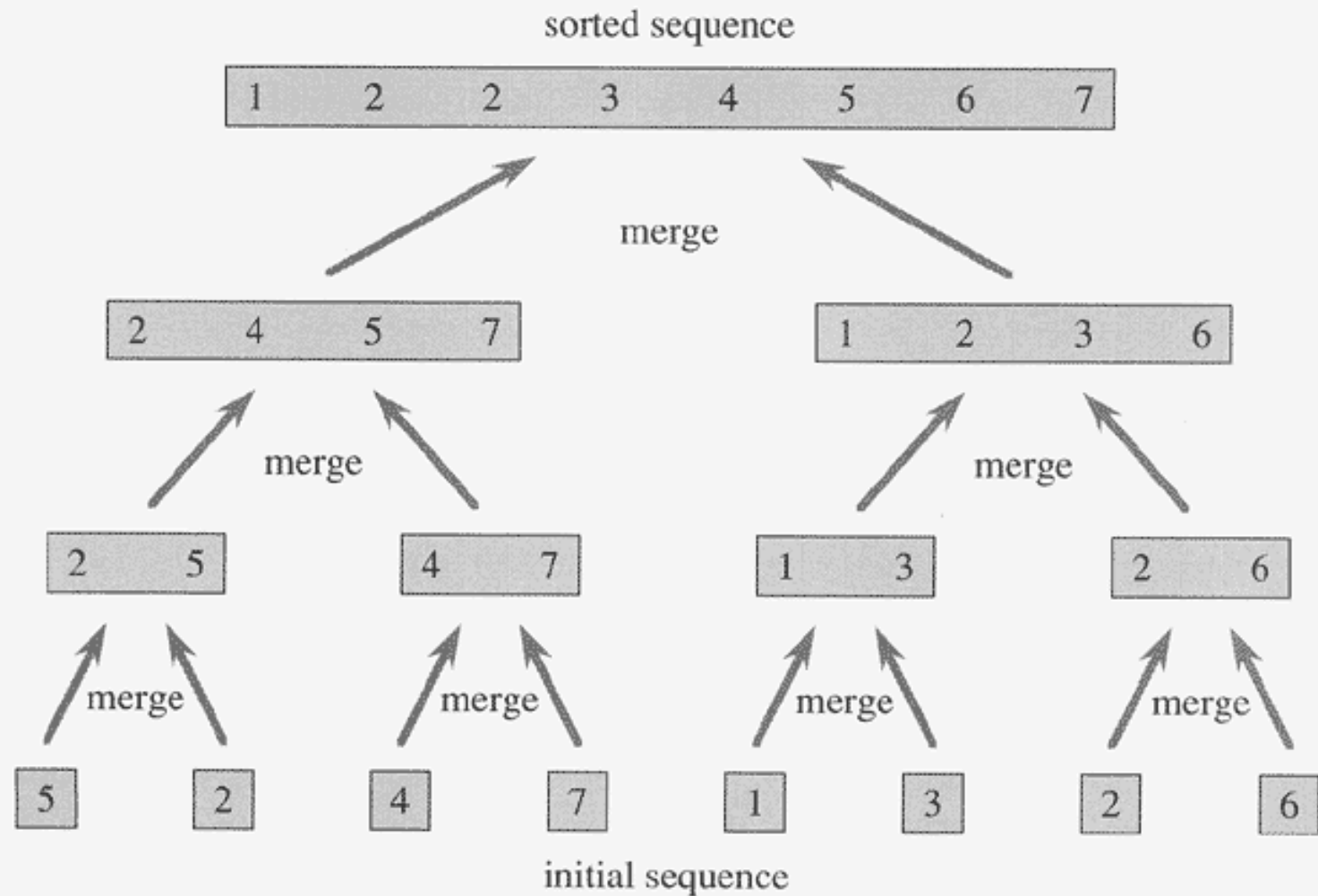
R



1

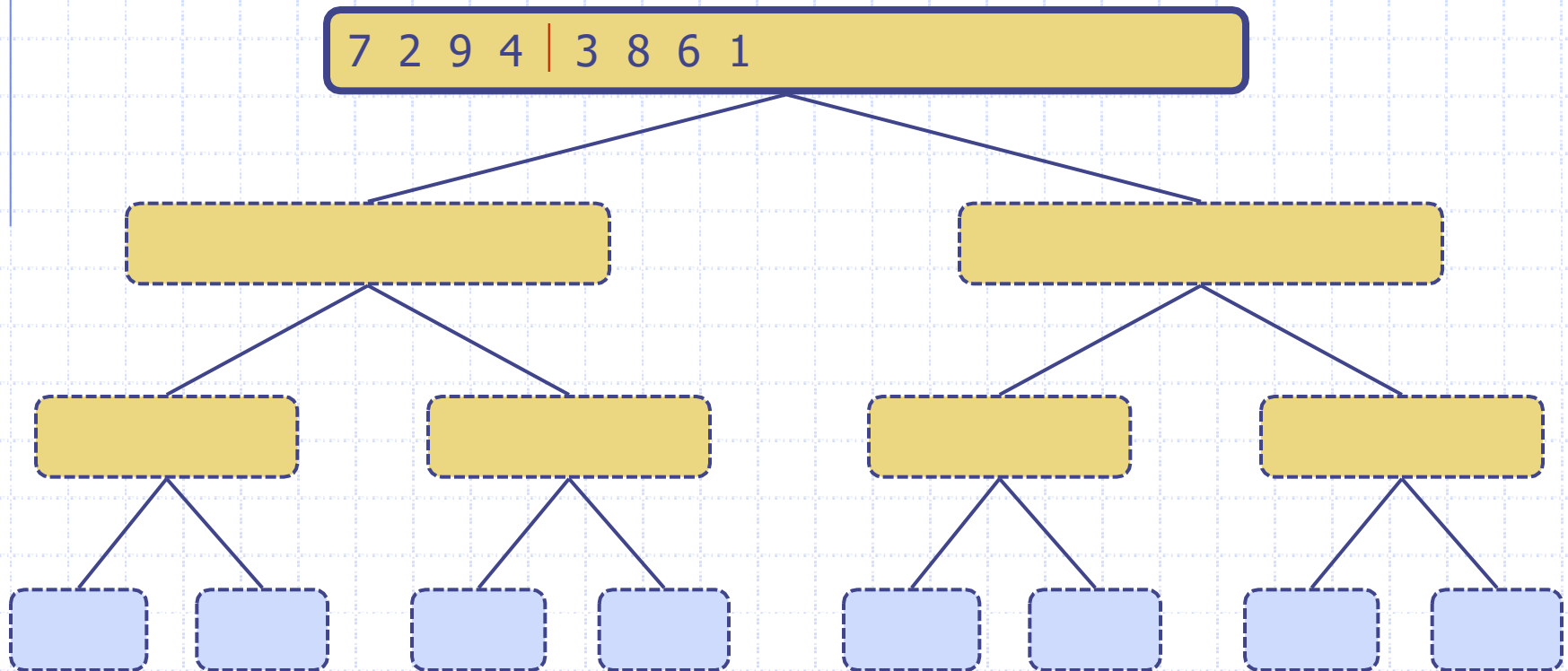
n_2+1

Merge Sort: Example



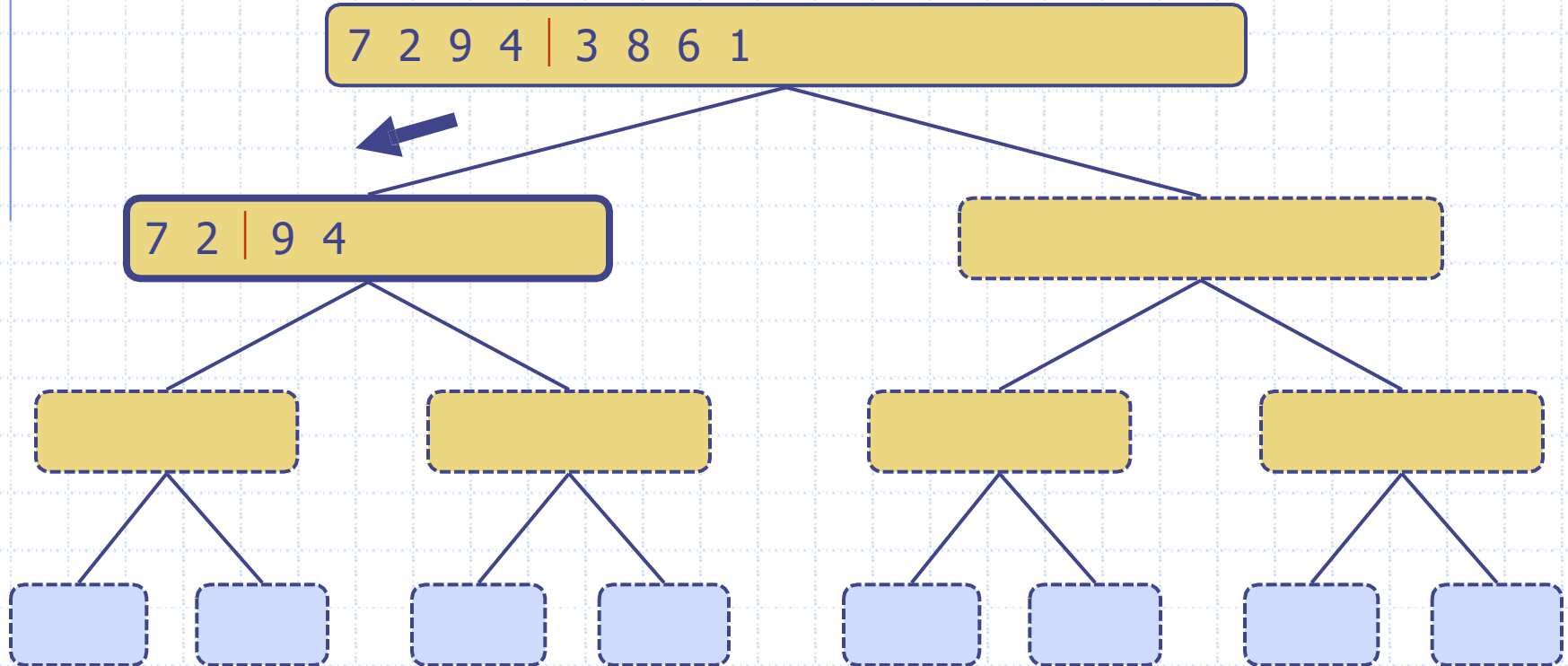
Execution Example

◆ Partition



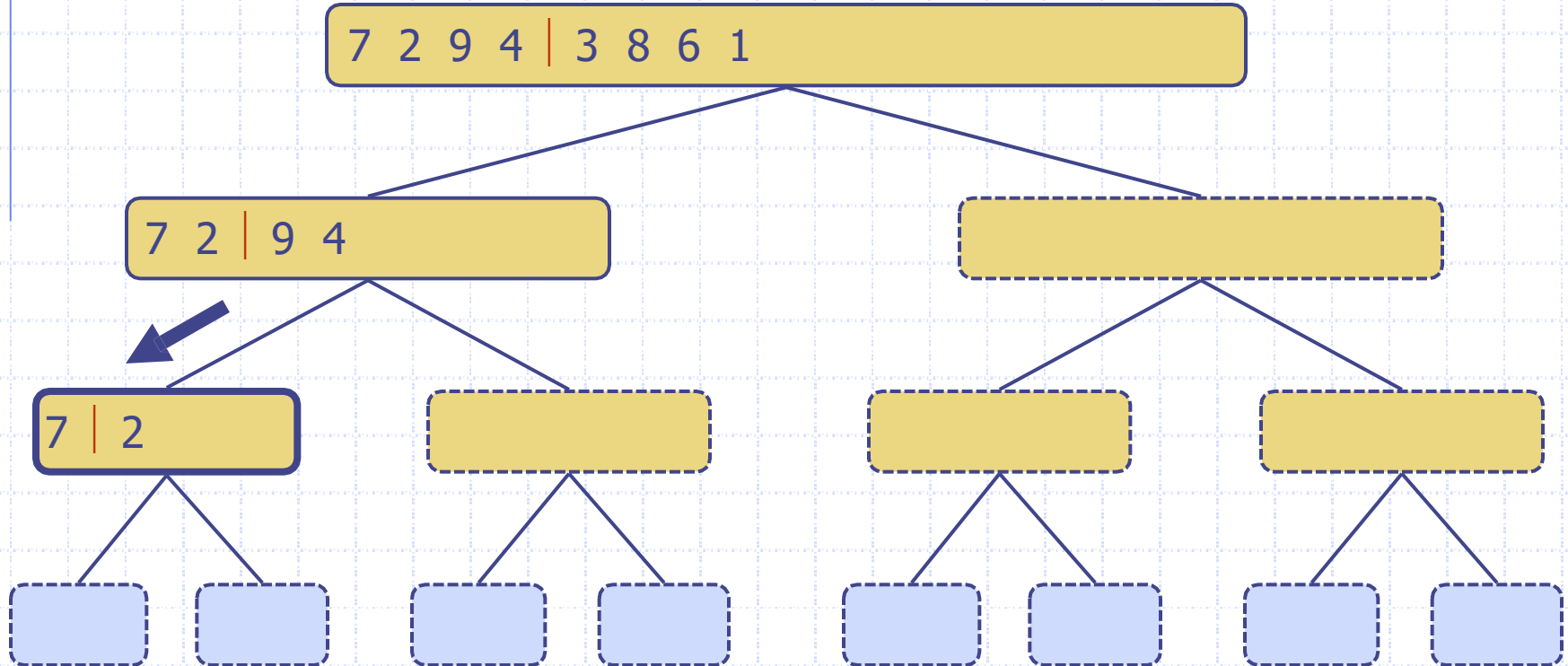
Execution Example

◆ Recursive call, partition



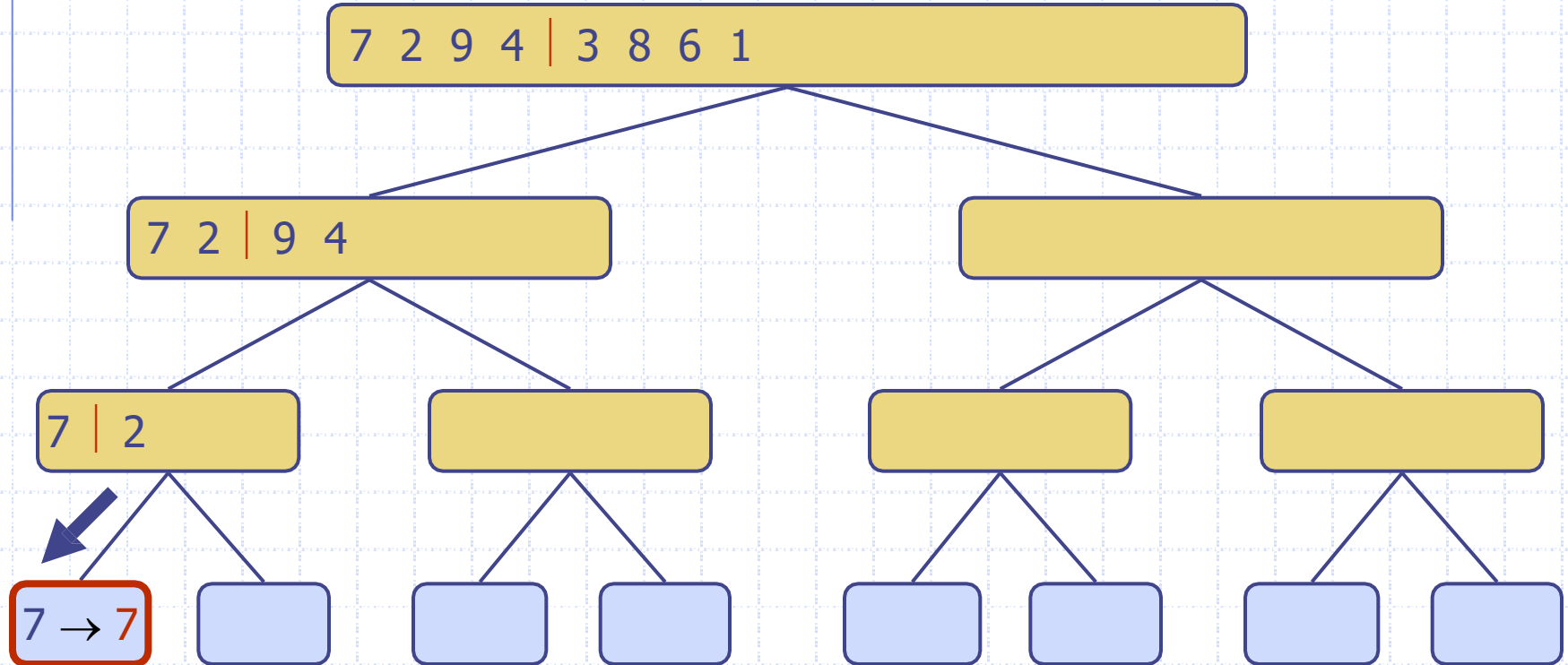
Execution Example

◆ Recursive call, partition



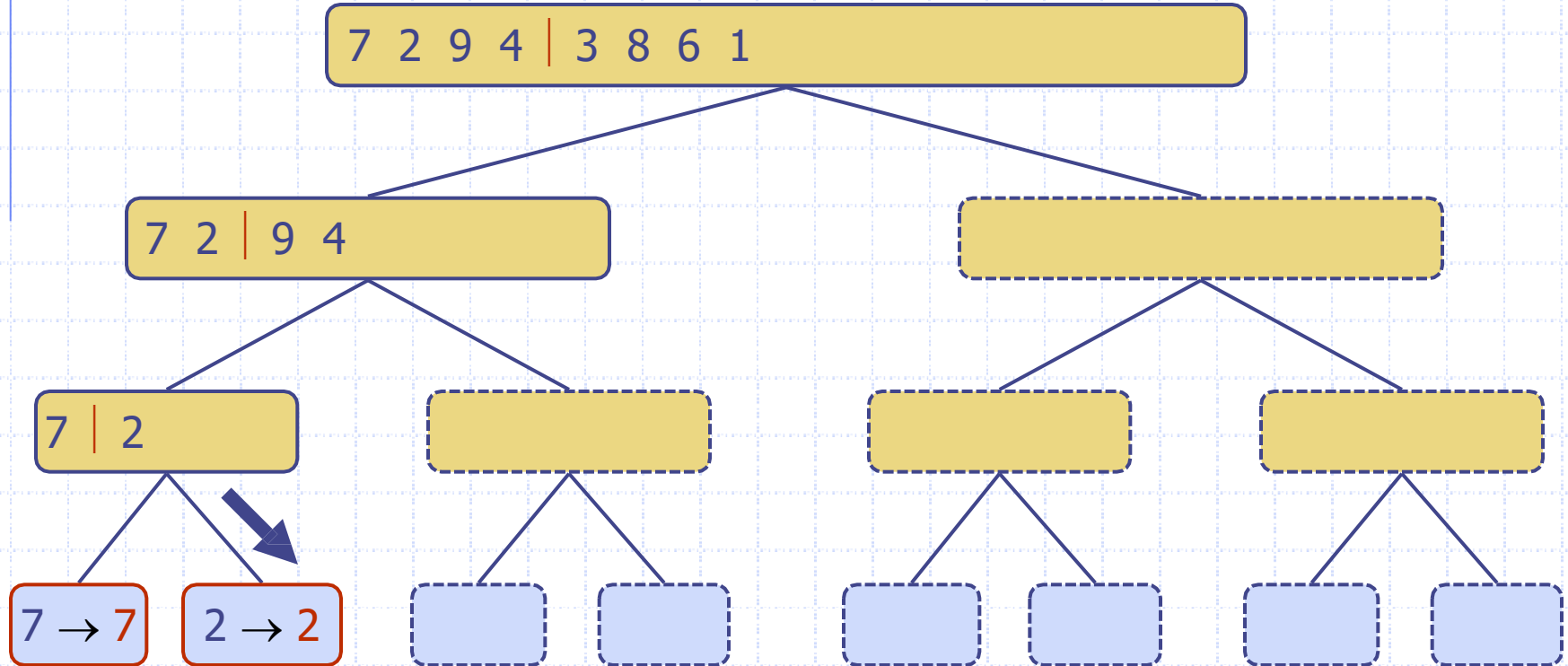
Execution Example

◆ Recursive call, base case



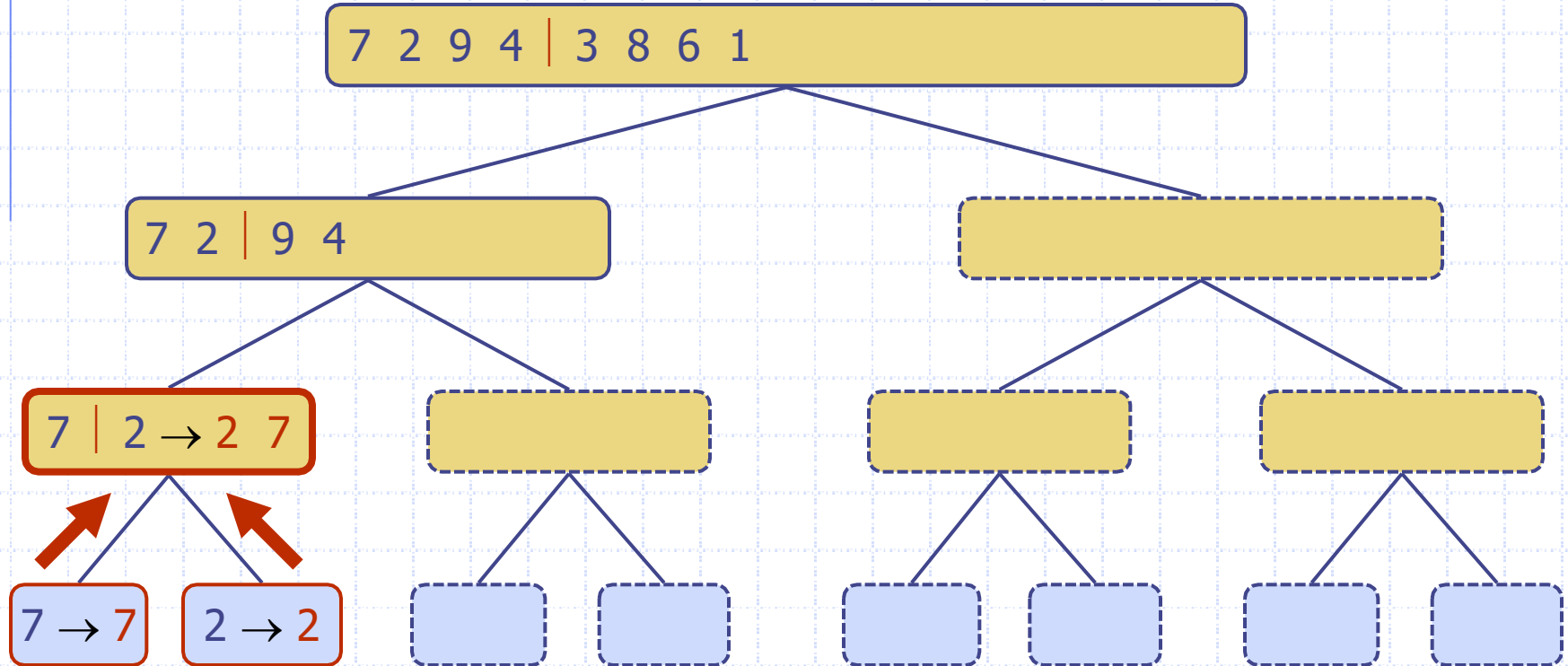
Execution Example

◆ Recursive call, base case



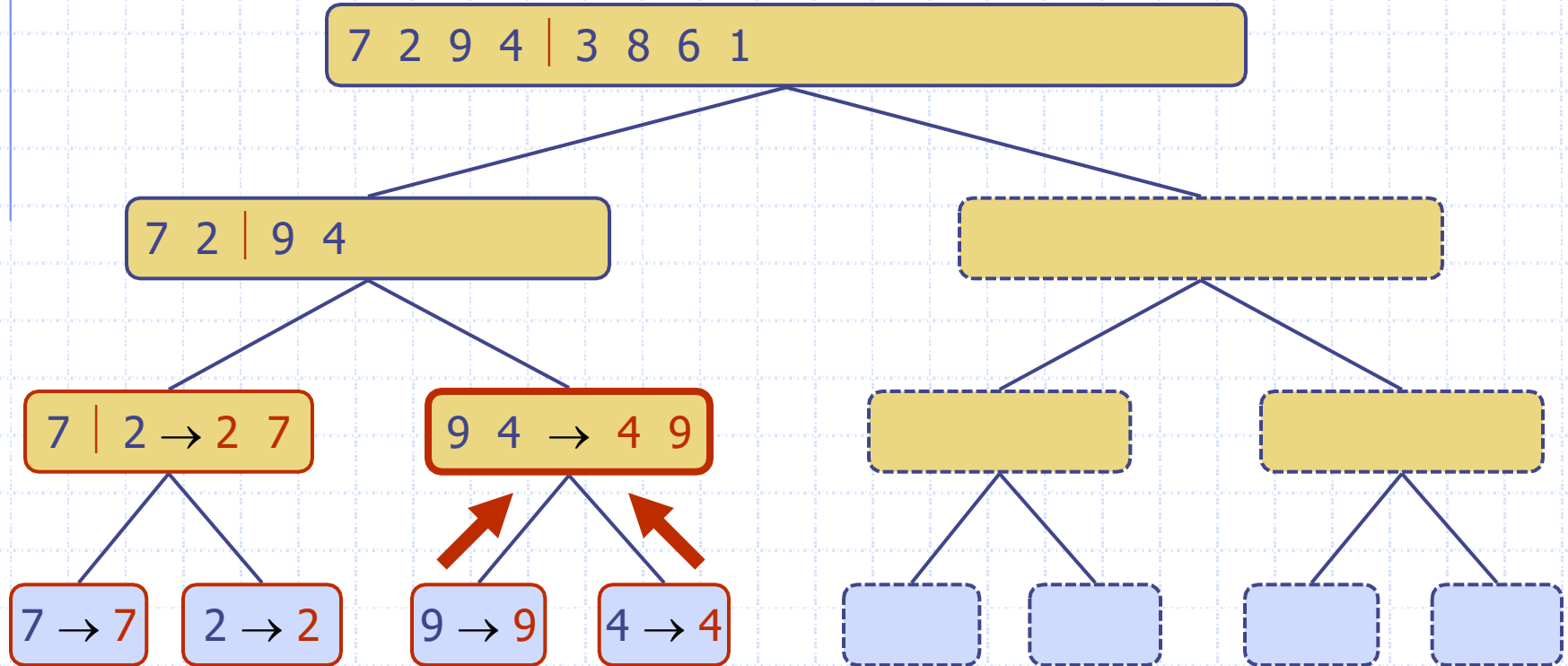
Execution Example

◆ Merge



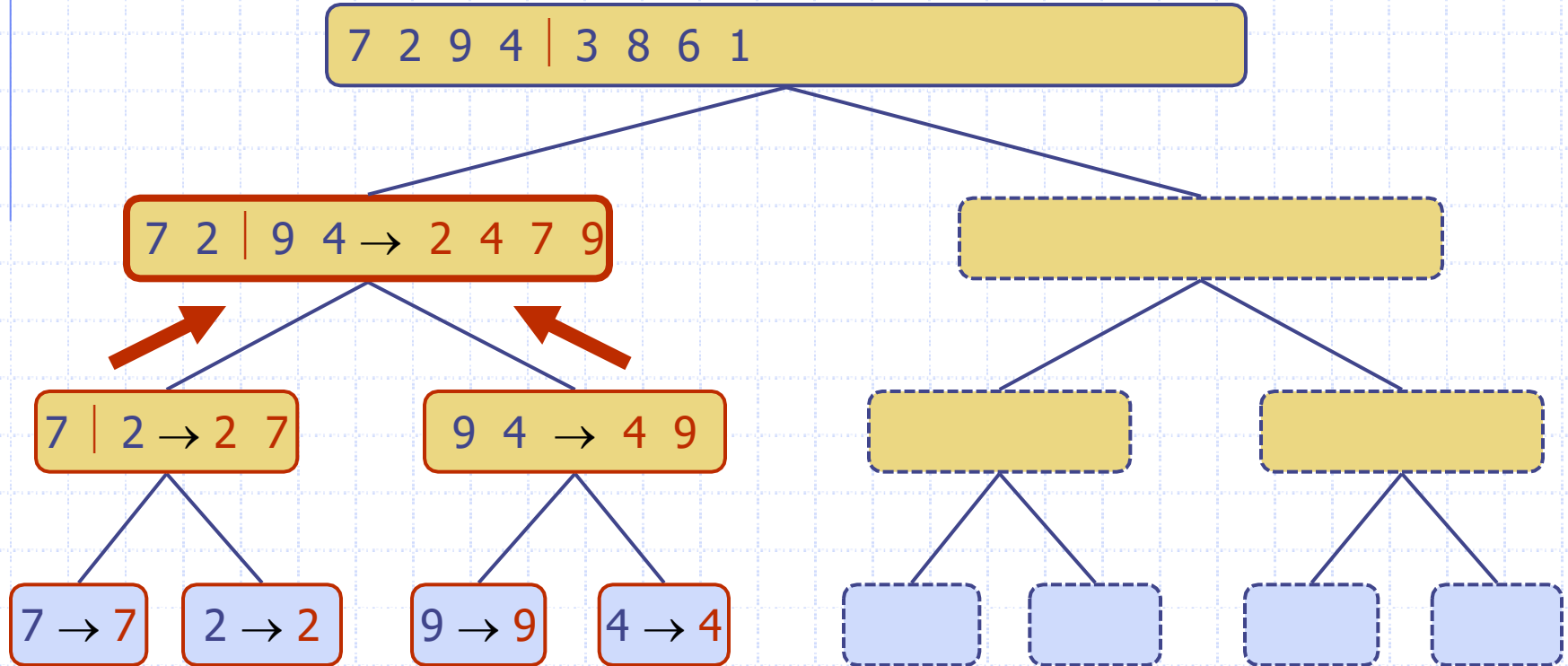
Execution Example

◆ Recursive call, ..., base case, merge



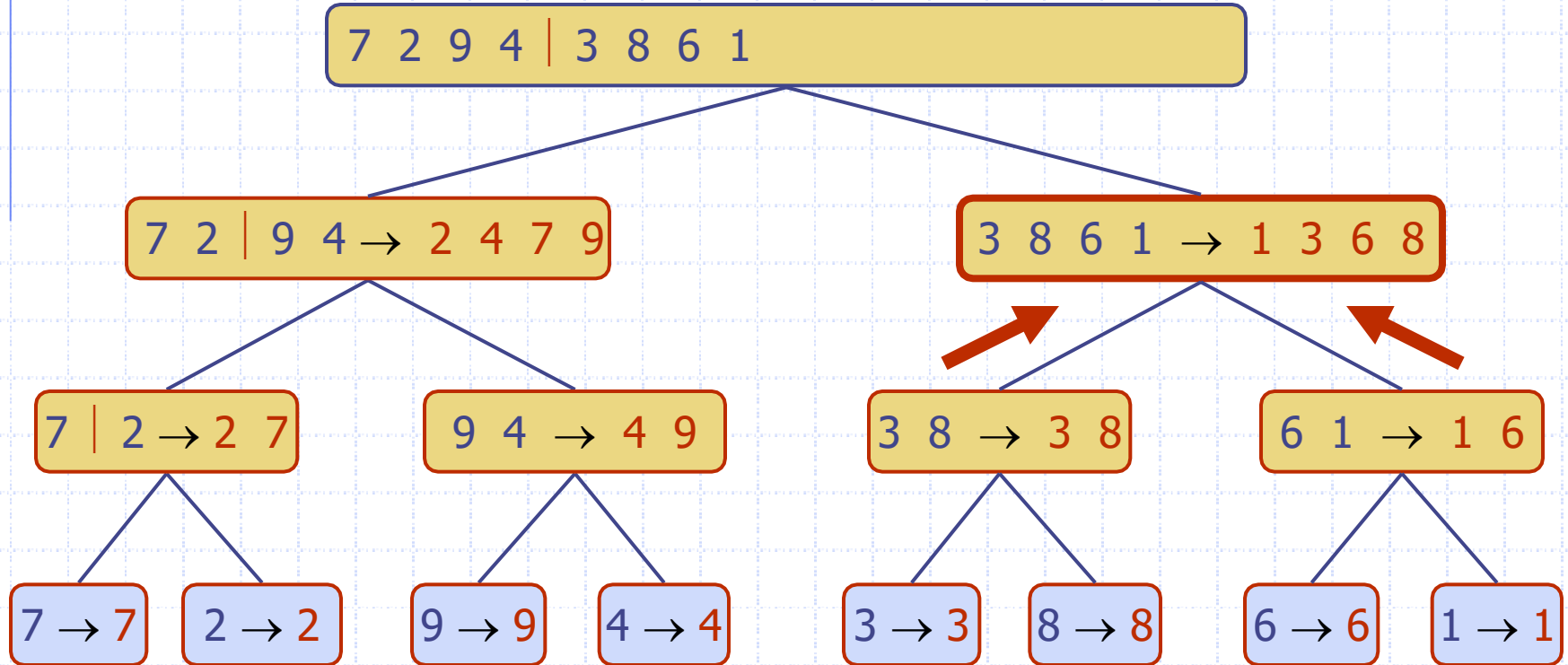
Execution Example

◆ Merge



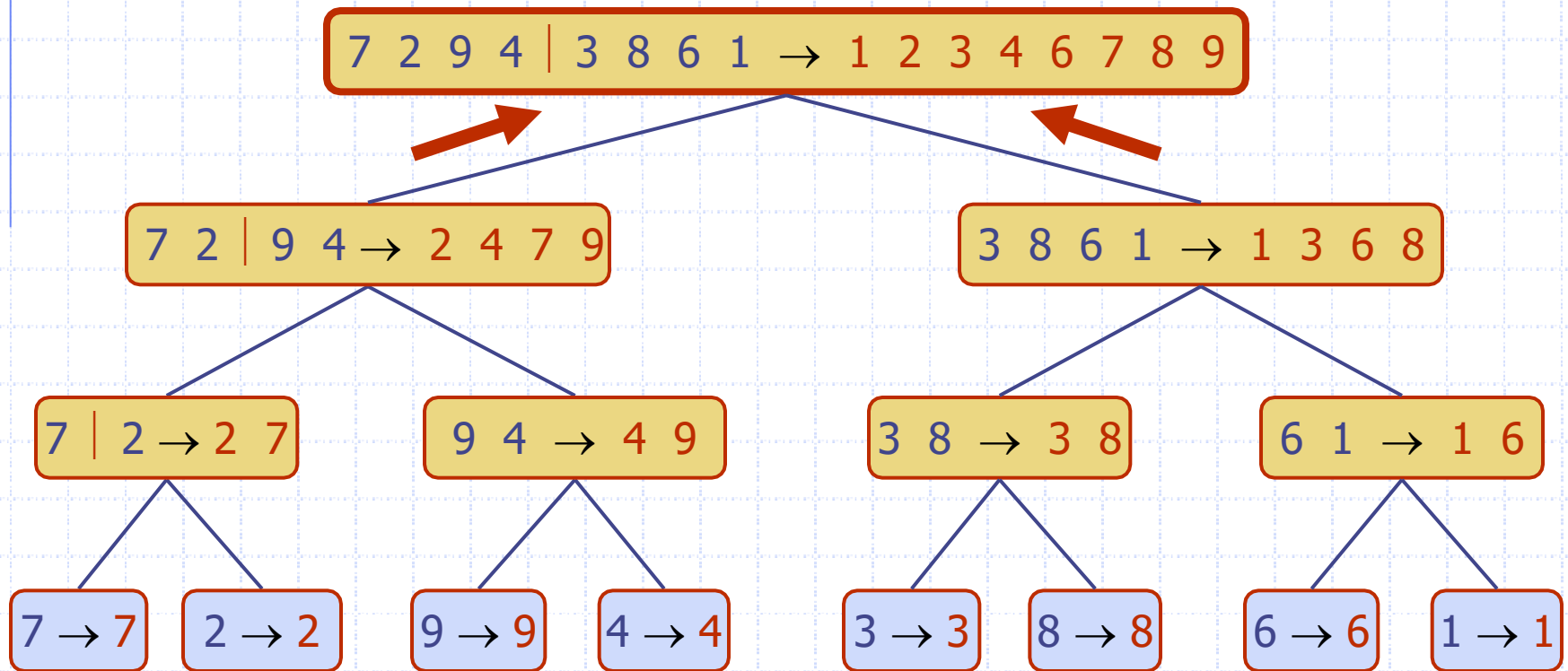
Execution Example

◆ 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)

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

◆ Note that base, $T(n) = b$, case occurs when $2^i = n$.

That is, $i = \log n$.

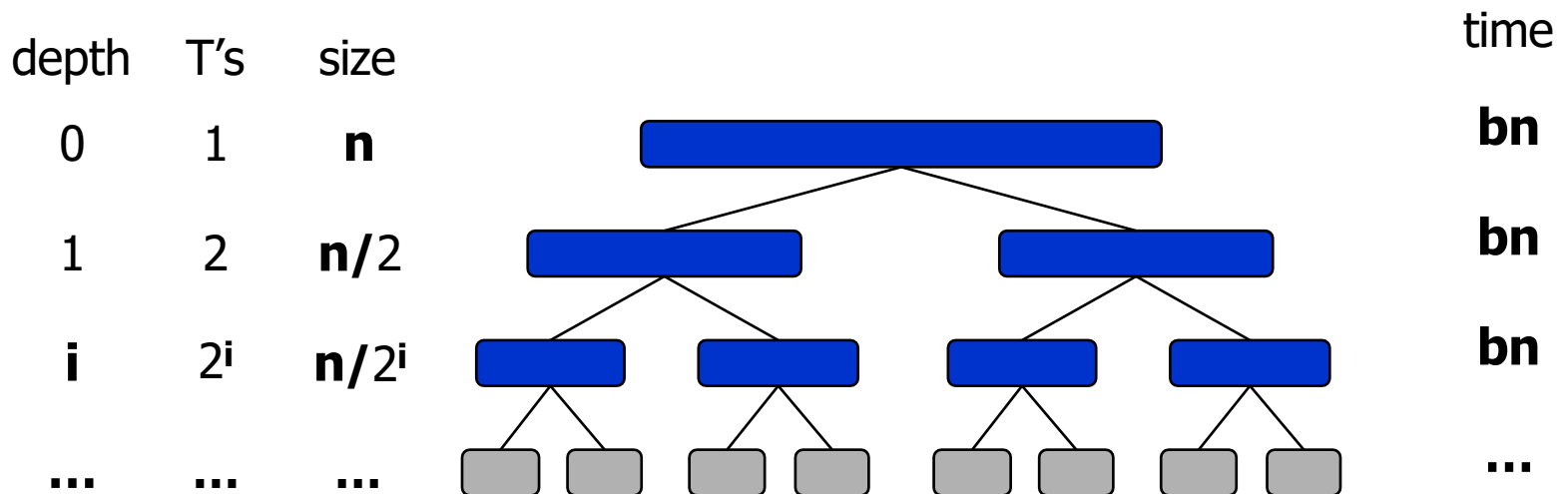
◆ So, $T(n) = bn + bn \log n$

◆ 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)