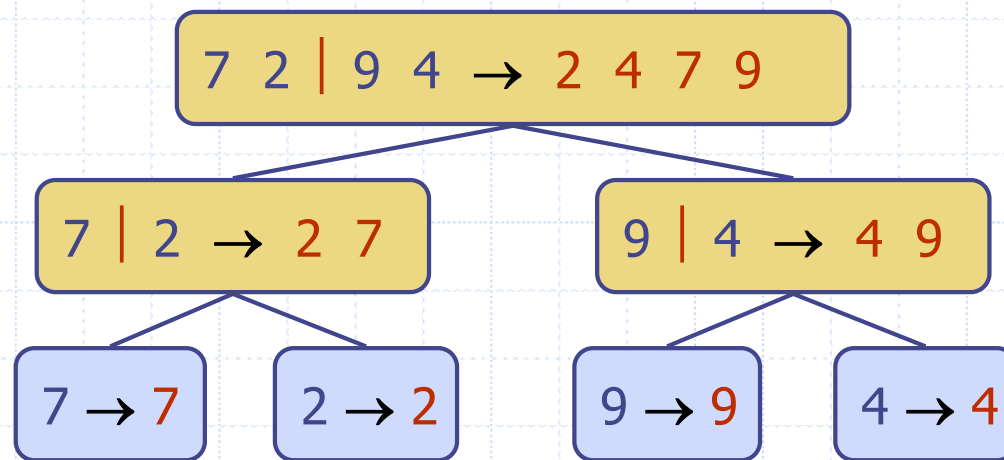


Merge Sort



We will look at this table later ...

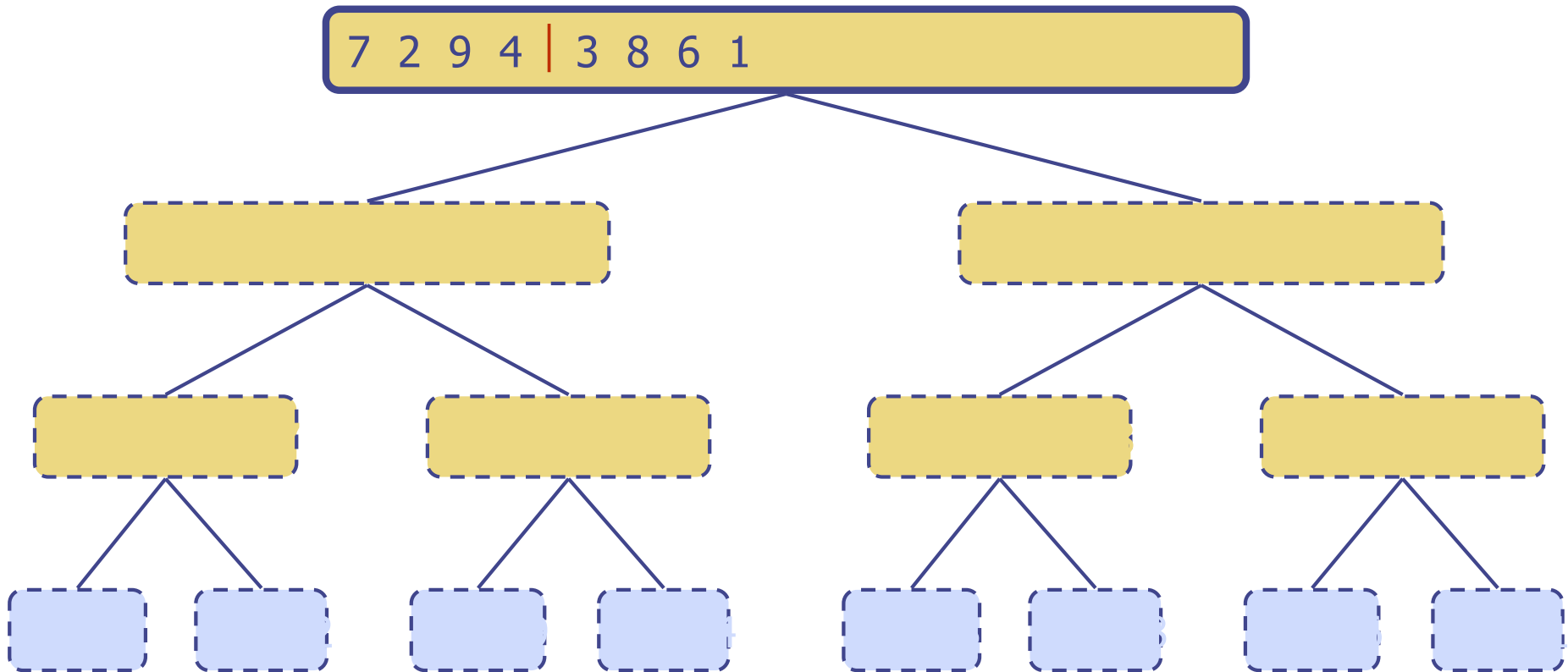
Algorithm	Time	Notes
selection-sort	$O(n^2)$	<ul style="list-style-type: none">▪ slow▪ in-place▪ for small data sets (< 1K)
insertion-sort	$O(n^2)$	<ul style="list-style-type: none">▪ slow▪ in-place▪ for small data sets (< 1K)
heap-sort	$O(n \log n)$	<ul style="list-style-type: none">▪ fast▪ in-place▪ for large data sets (1K — 1M)
merge-sort	$O(n \log n)$	<ul style="list-style-type: none">▪ fast▪ sequential data access▪ for huge data sets (> 1M)

New things that we will learn from this part

- ◆ Divide-and-Conquer rationale
- ◆ Complexity analysis based on recurrence relation

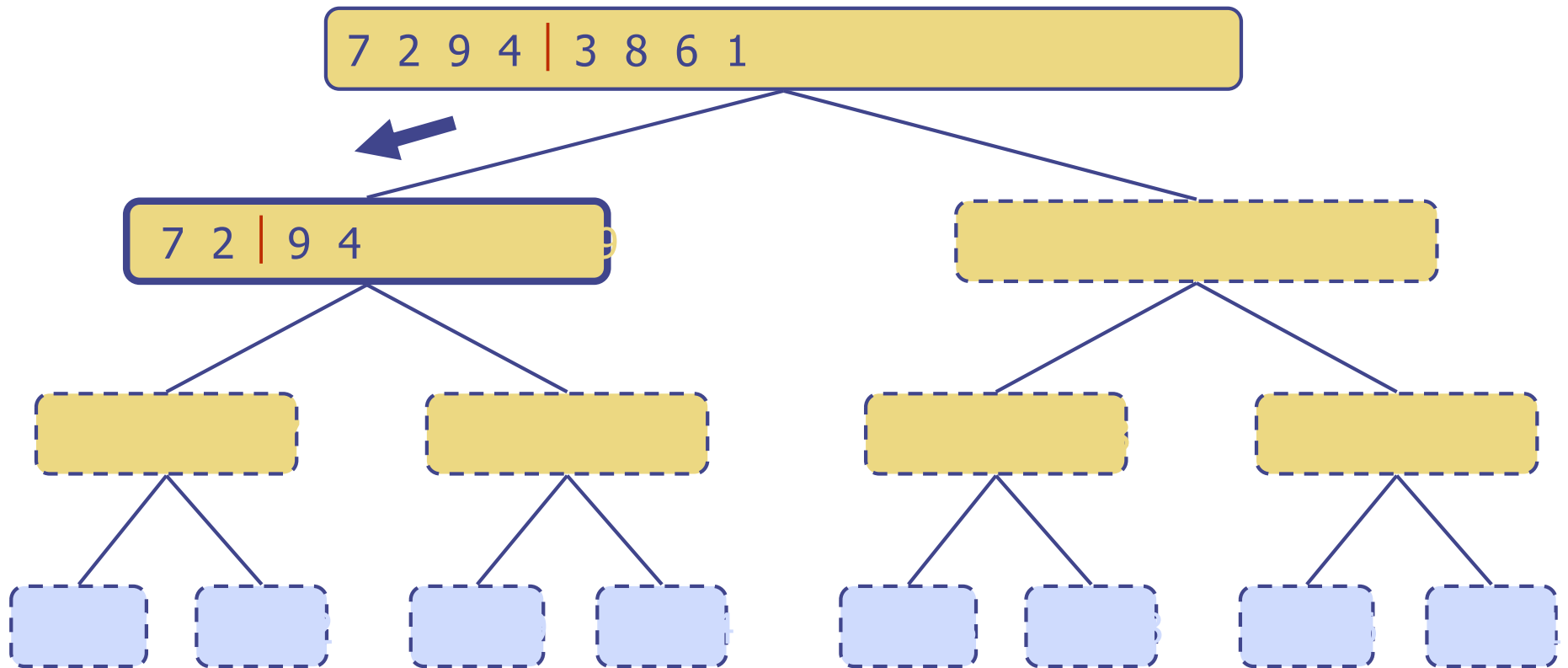
Execution Example

◆ Partition



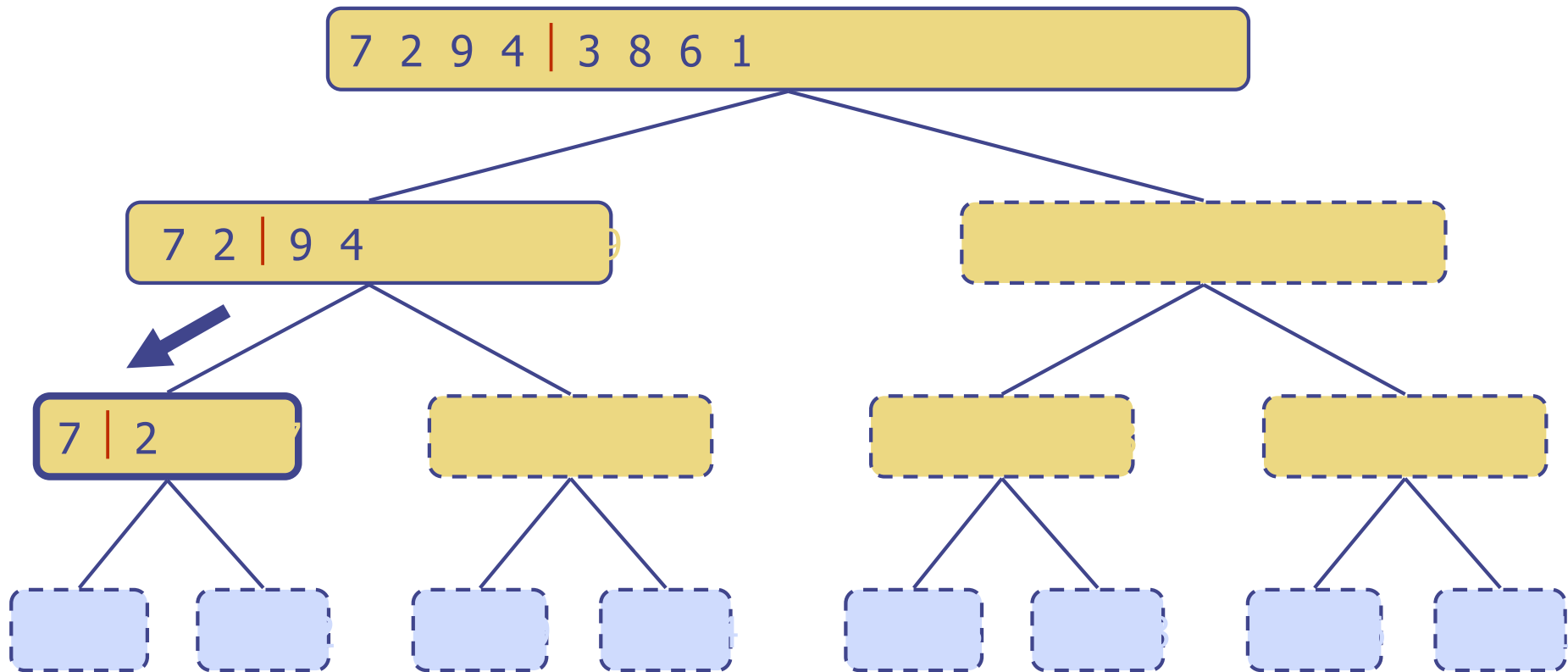
Execution Example (cont.)

- ◆ Recursive call, partition



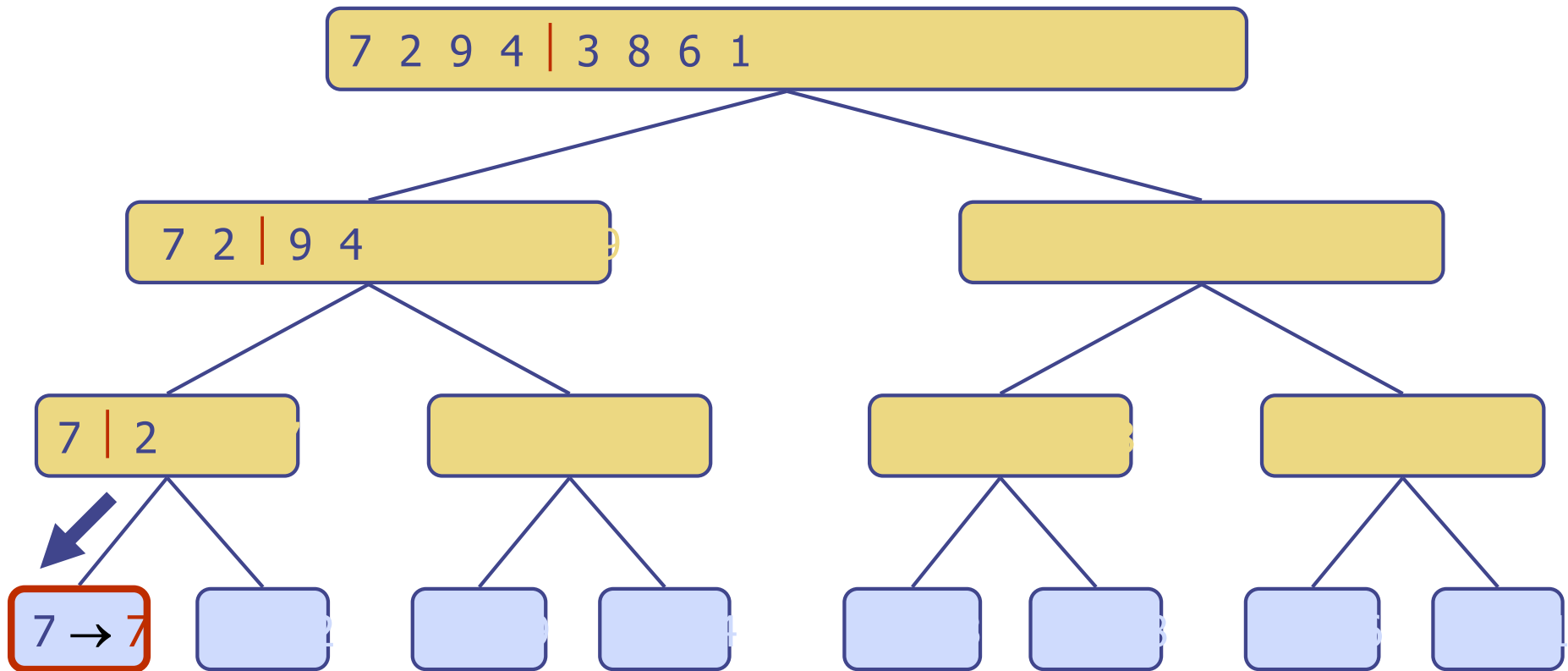
Execution Example (cont.)

- ◆ Recursive call, partition



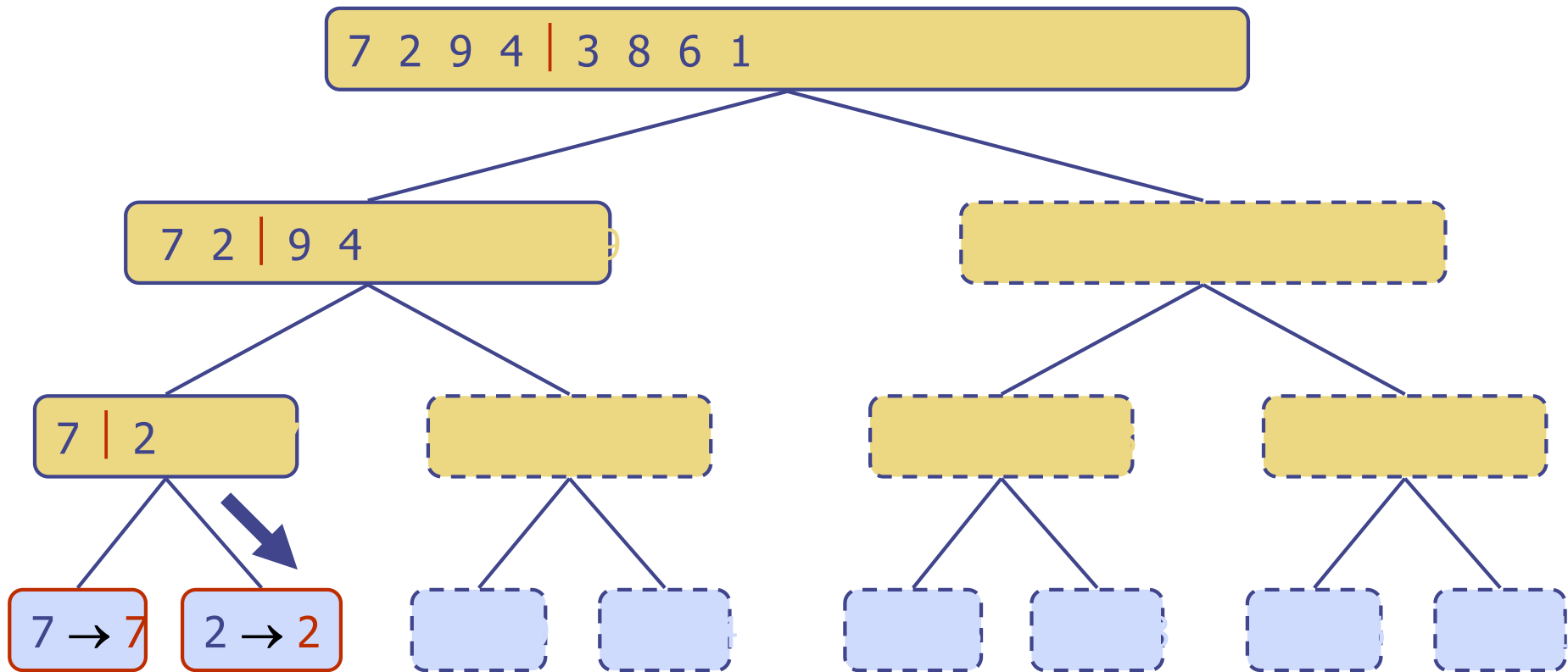
Execution Example (cont.)

- ◆ Recursive call, base case



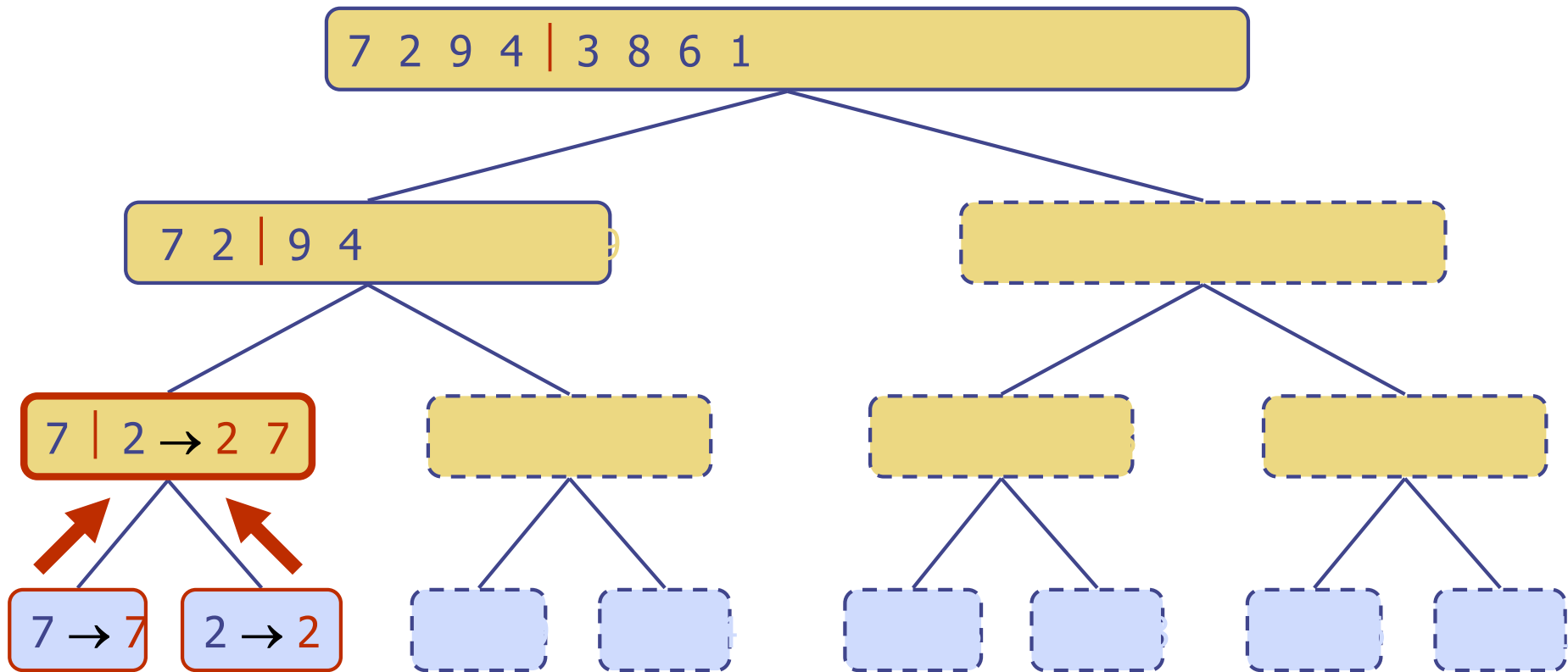
Execution Example (cont.)

- ◆ Recursive call, base case



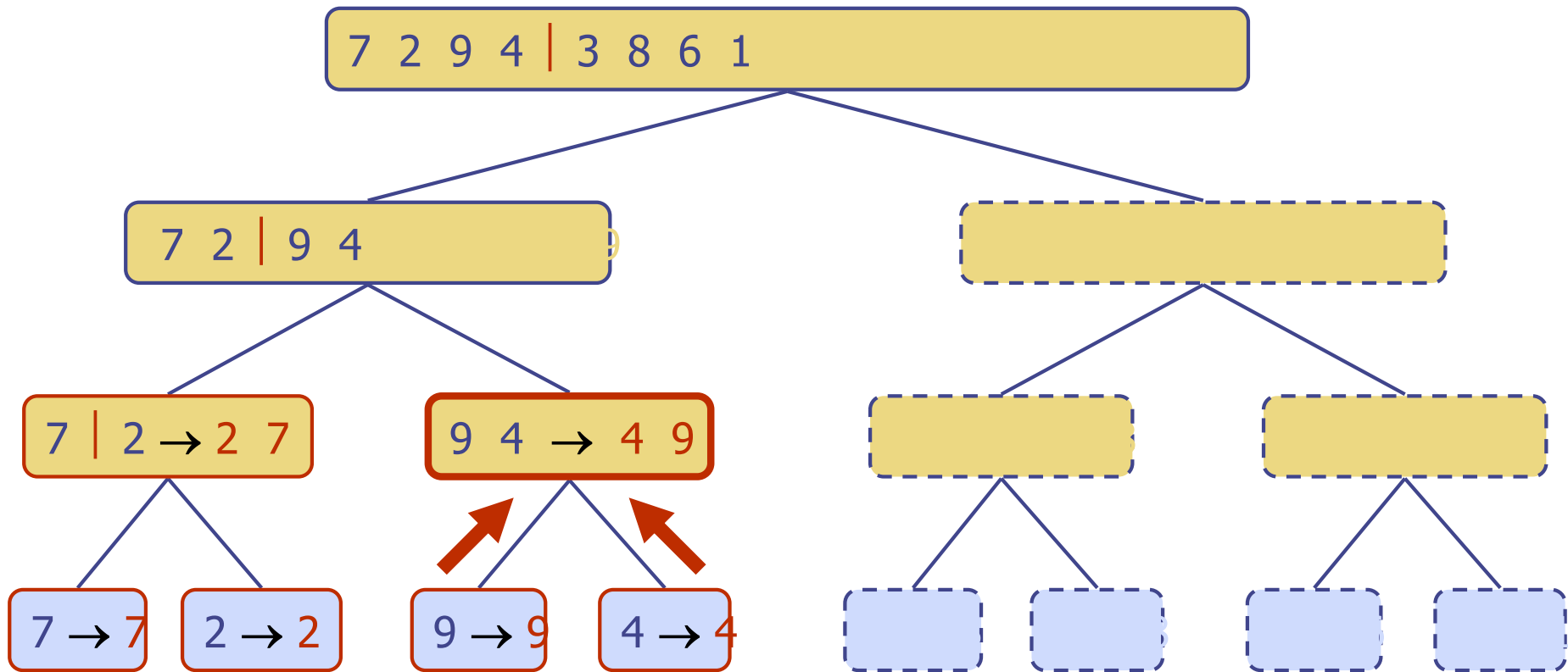
Execution Example (cont.)

◆ Merge



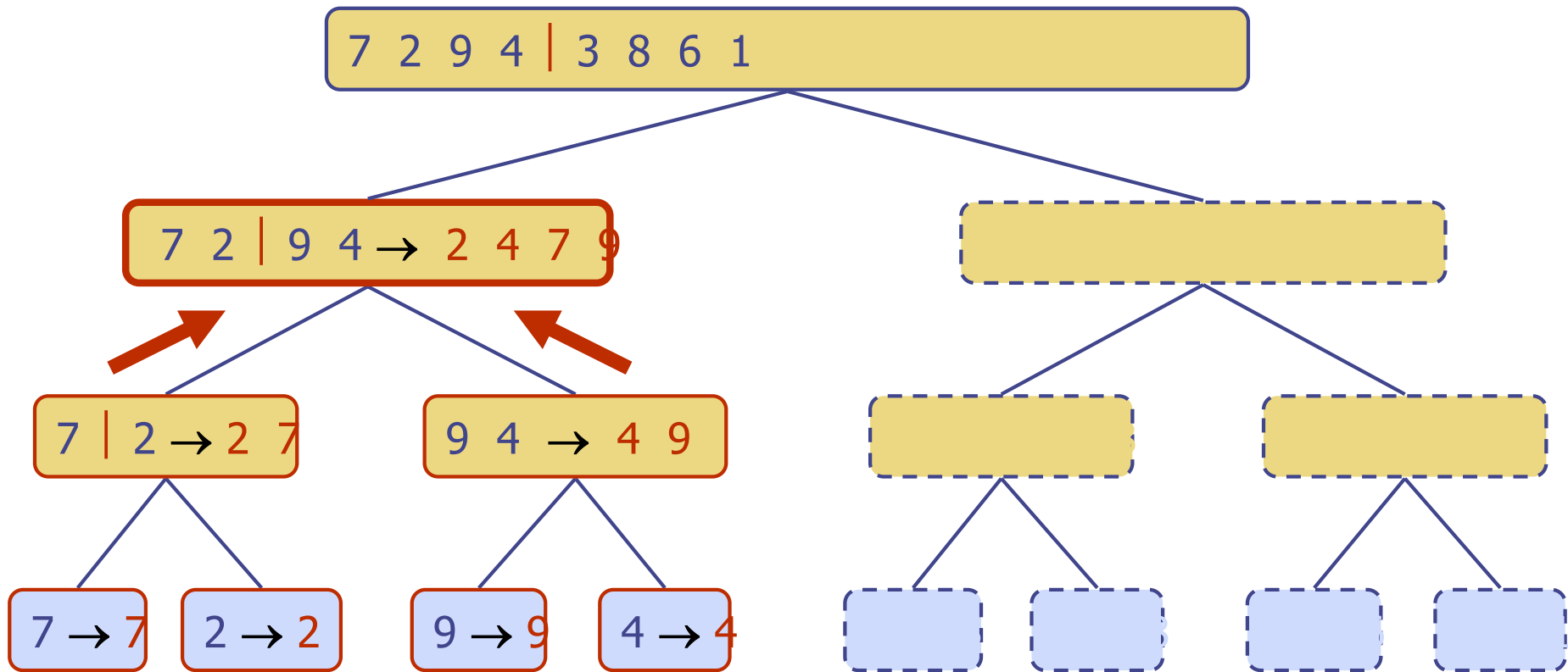
Execution Example (cont.)

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



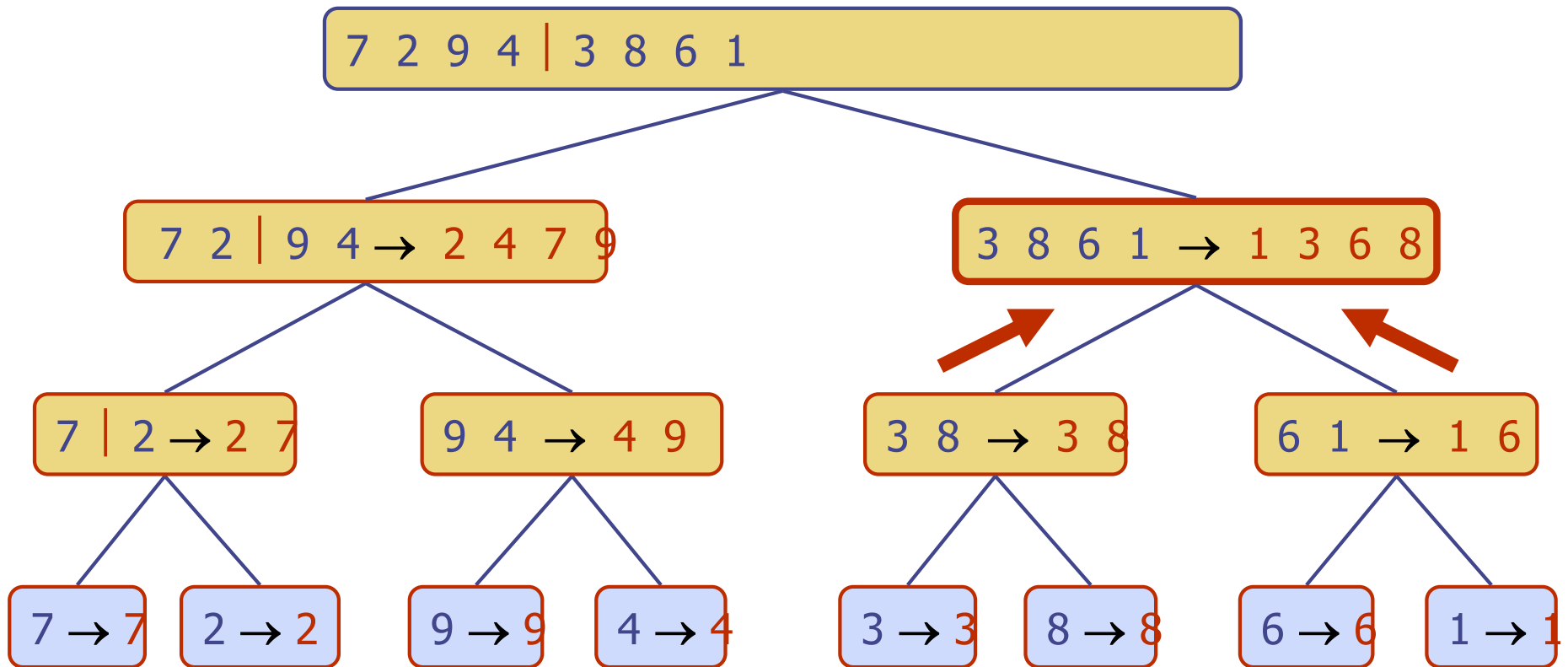
Execution Example (cont.)

◆ Merge



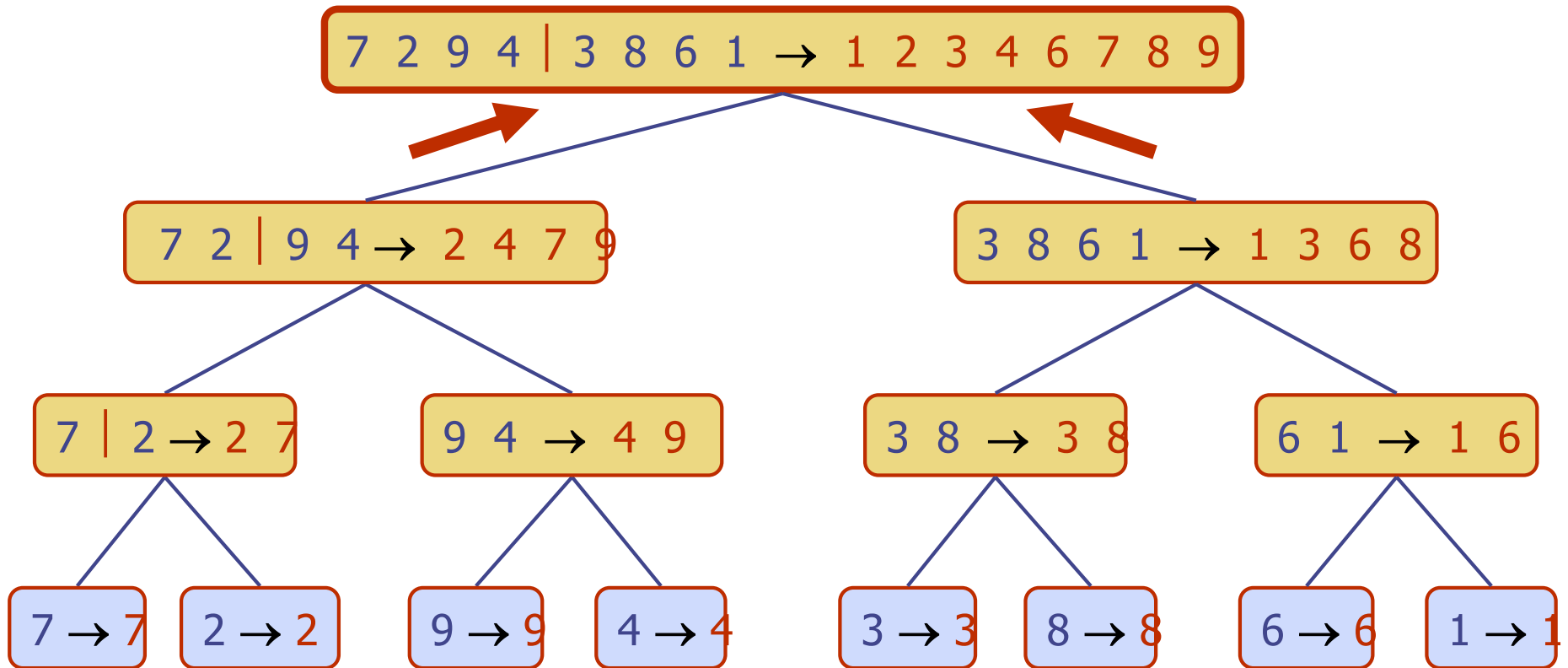
Execution Example (cont.)

- ◆ Recursive call, ..., merge, merge



Execution Example (cont.)

◆ Merge



Divide-and-Conquer (§ 10.1.1)

- ◆ **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
- ◆ Like heap-sort
 - It uses a comparator
 - It has $O(n \log n)$ running time
- ◆ Unlike heap-sort
 - It does not use an auxiliary priority queue
 - It accesses data in a sequential manner (suitable to sort data on a disk)
 - Disk
 - ◆ Fast when accessing data sequentially

Merge-Sort (§ 10.1)

◆ 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, C)

Input sequence S with n elements, comparator C

Output sequence S sorted according to C

if $S.size() > 1$

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

mergeSort(S_1, C)

mergeSort(S_2, C)

$S \leftarrow merge(S_1, S_2)$

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 sequences A and B with $n/2$ elements each

Output sorted sequence of $A \cup B$

$S \leftarrow$ empty sequence

while $\neg A.empty() \wedge \neg B.empty()$

if $A.front() < B.front()$

$S.addBack(A.front()); A.eraseFront();$

else

$S.addBack(B.front()); B.eraseFront();$

while $\neg A.empty()$

$S.addBack(A.front()); A.eraseFront();$

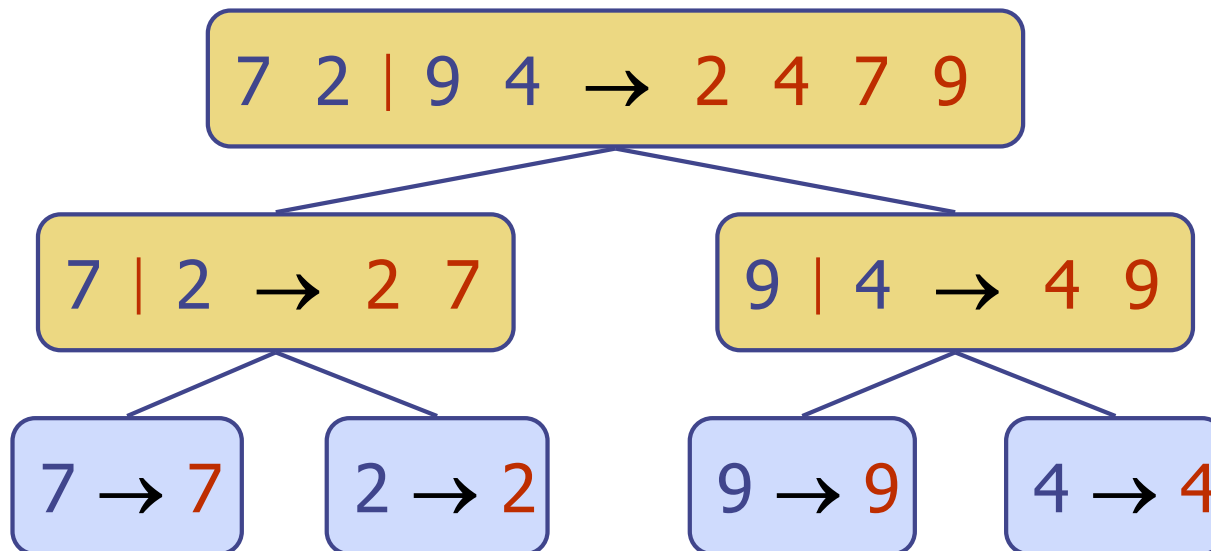
while $\neg B.empty()$

$S.addBack(B.front()); B.eraseFront();$

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



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 $O(n \log n)$

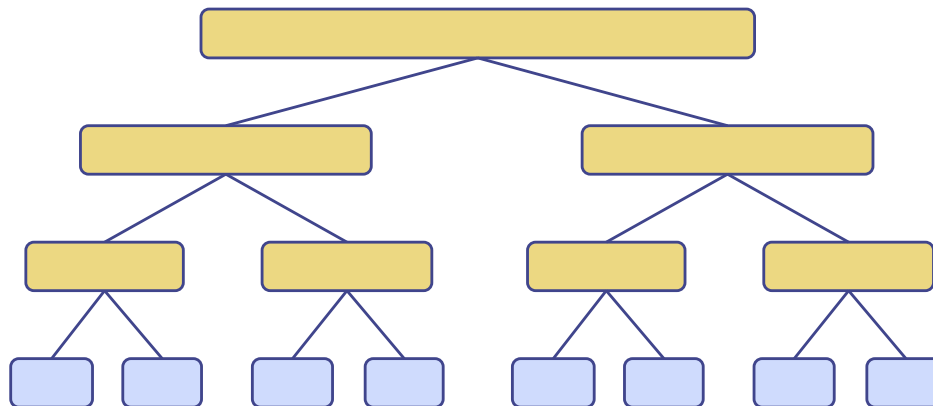
depth #seqs size

0 1 n

1 2 $n/2$

i 2^i $n/2^i$

...



Another Analysis: Recurrence Equation (1)

- ◆ $t(n)$: the worst-case running time of merge-sort
- ◆ For simplicity, n is a power of 2. Then, we have the following:

$$t(n) = \begin{cases} b & \text{if } n \leq 1 \\ 2t(n/2) + cn & \text{otherwise.} \end{cases}$$

- ◆ How to compute the order of $t(n)$?
- ◆ Applying the equation recursively,

$$\begin{aligned} t(n) &= 2(2t(n/2^2) + (cn/2)) + cn \\ &= 2^2t(n/2^2) + 2(cn/2) + cn = 2^2t(n/2^2) + 2cn. \end{aligned}$$

- ◆ We get the following general equation:

$$t(n) = 2^i t(n/2^i) + icn.$$

- ◆ We stop this when $n/2^i = 1$, i.e., $i = \log n$

Another Analysis: Recurrence Equation (2)

◆ Then, we have the following, and thus we are done.

$$\begin{aligned}t(n) &= 2^{\log n} t(n/2^{\log n}) + (\log n)cn \\ &= nt(1) + cn \log n \\ &= nb + cn \log n.\end{aligned}$$

Summary of Sorting Algorithms

Algorithm	Time	Notes
selection-sort	$O(n^2)$	<ul style="list-style-type: none">▪ slow▪ in-place▪ for small data sets (< 1K)
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Questions?