

Sorting & Master Theorem

CS16: Introduction to Data Structures & Algorithms
Spring 2019

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ Master Theorem
 - ▶ Quick Sort
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



The Problem

- ▶ Turn this

10	19	7	4	3	21	10	23	24	18	1	8	23	1	12
----	----	---	---	---	----	----	----	----	----	---	---	----	---	----

- ▶ Into this

1	1	3	4	7	8	10	10	12	18	19	21	23	23	24
---	---	---	---	---	---	----	----	----	----	----	----	----	----	----

- ▶ as efficiently as possible

Sorting is Serious!

Microsoft Research team shatters data sorting record, wrenches trophy from Yahoo



Alexis Santos
05.22.12

59
Shares



Before you read thi

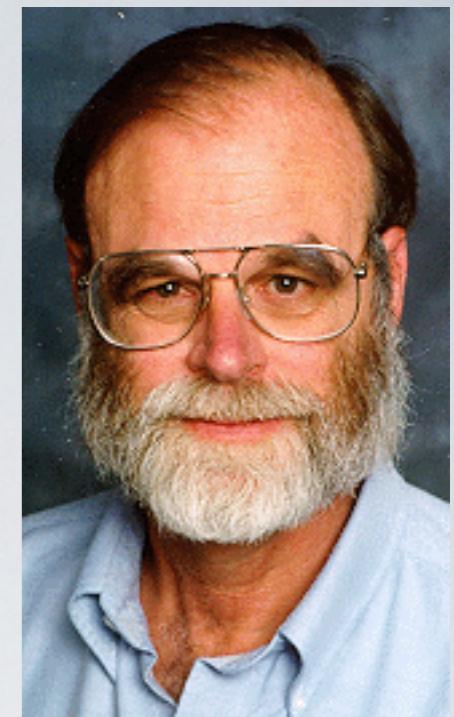


6 Ways Agents



Sorting Competition

- ▶ Sort Benchmark (sortbenchmark.org)
- ▶ Started by Jim Gray
 - ▶ Research scientist at Microsoft Research
 - ▶ Winner of 1998 Turing Award for contributions to databases
- ▶ Tencent Sort from Tencent Corp. (2016)
 - ▶ 100 TB in 134 seconds
 - ▶ 37 TB in 1 minute
- ▶



Why?

- ▶ Why do we care so much about sorting?
- ▶ Rule of thumb:
 - ▶ “good things happen when data is sorted”
 - ▶ we can find things faster (e.g., using binary search)

Sorting Algorithms

- ▶ There are many ways to sort arrays
 - ▶ Iterative vs. recursive
 - ▶ in-place vs. not-in-place
 - ▶ comparison-based vs. non-comparative
- ▶ In-place algorithms
 - ▶ transform data structure w/ small amount of extra storage (i.e., $O(1)$)
 - ▶ For sorting: array is overwritten by output instead of creating new array
- ▶ Most sorting algorithms in 16 are comparison-based
 - ▶ main operation is comparison
 - ▶ but not all (e.g., Radix sort)

“In-Placeness”

- ▶ Reversing an array

```
function reverse(A):  
    n = A.length  
    B = array of length n  
    for i = 0 to n - 1:  
        B[n-1-i] = A[i]  
    return B
```

Not in-place!

```
function reverse(A):  
    n = A.length  
    for i = 0 to n/2:  
        temp = A[i]  
        A[i] = A[n-1-i]  
        A[n-1-i] = temp
```

in-place



Return statement
not needed

Properties of In-Place Solutions

- ▶ Harder to write :-(
- ▶ Use less memory :-)
- ▶ Even harder to write for recursive algorithms :-(
- ▶ Tradeoff between simplicity an efficiency

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ **Selection sort**
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ Master Theorem
 - ▶ Quick Sort
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



Selection Sort

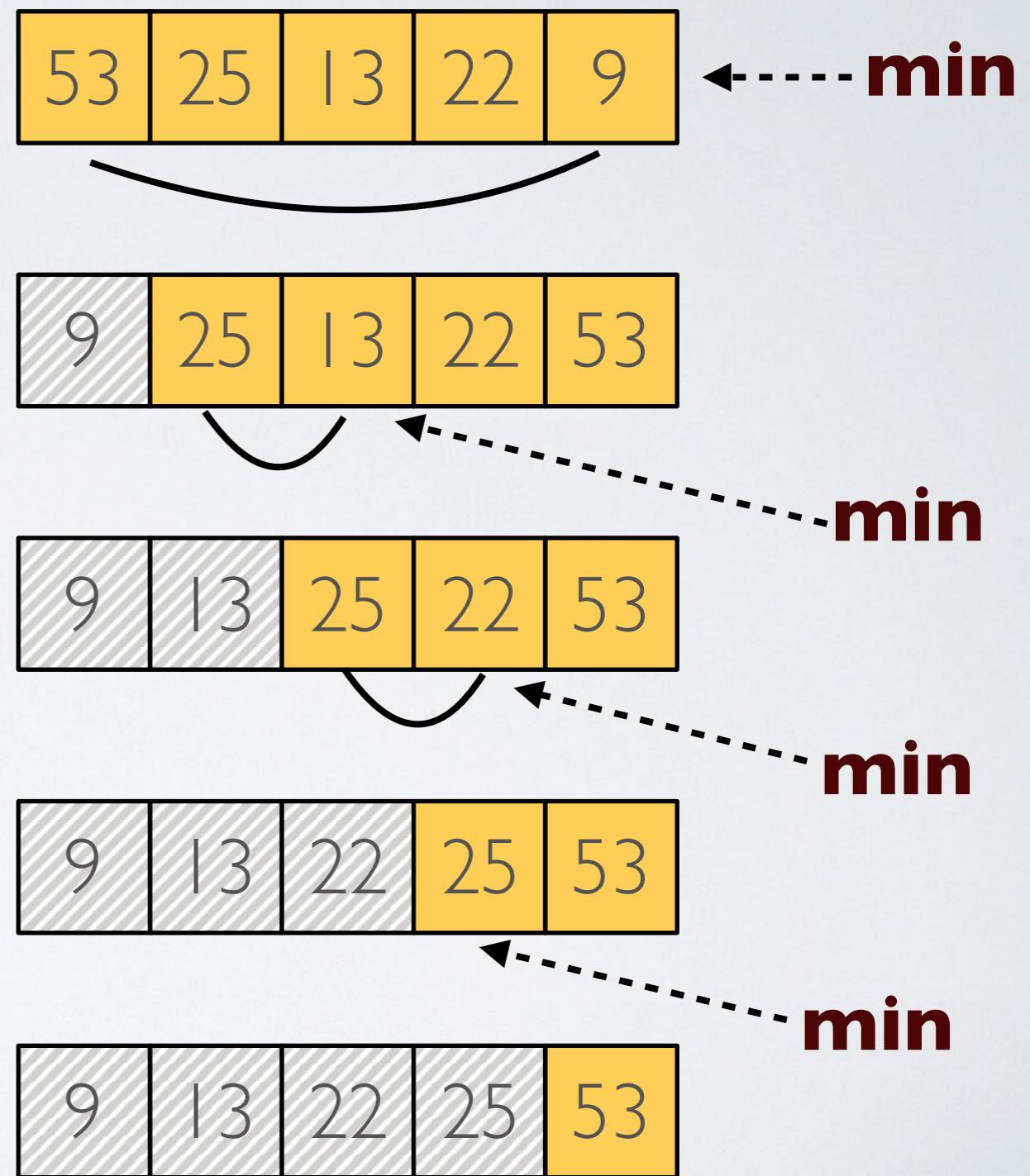
- ▶ Usually iterative and in-place
- ▶ Divides input array into two logical parts
 - ▶ elements already sorted
 - ▶ elements that still need to be sorted
- ▶ Selects smallest element & places it at index 0
 - ▶ then selects second smallest & places it in index 1
 - ▶ then the third smallest at index 2, etc..

Selection Sort

- ▶ Advantages
 - ▶ Very simple
 - ▶ Memory efficient if in-place (swaps elements in array)
- ▶ Disadvantages
 - ▶ Slow: $O(n^2)$

Selection Sort

- ▶ Iterate through positions
- ▶ At each position
 - ▶ store smallest element from remaining set



Selection Sort

```
function selection_sort(A):
    n = A.length
    for i = 0 to n-2:
        min = argmin(A[i:n-1])
        swap A[i] with A[min]
```

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ **Insertion sort**
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ Master Theorem
 - ▶ Quick Sort
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



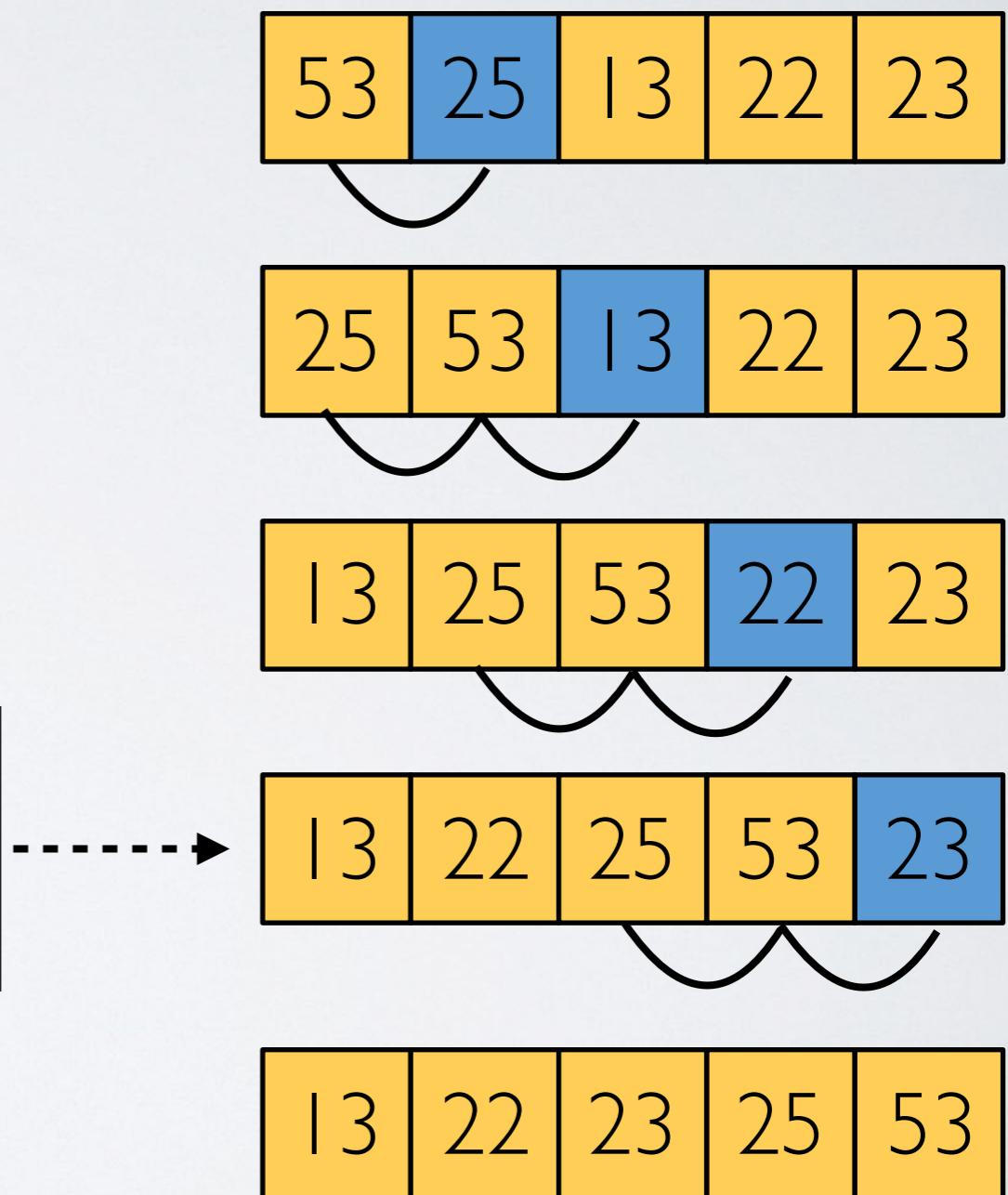
Insertion Sort

- ▶ Usually iterative and in-place
- ▶ Compares each item w/ all items before it...
 - ▶ ...and inserts it into correct position
- ▶ Advantages
 - ▶ Works really well if items partially sorted
 - ▶ Memory efficient if in-place (swaps elements in array)
- ▶ Disadvantages
 - ▶ Slow: $O(n^2)$

Insertion Sort

- ▶ Compares each item w/ all items before it...
 - ▶ ...and inserts it into correct position

Note: $23 > 22$ so don't need to keep checking since rest is already sorted



Insertion Sort

```
function insertion_sort(A):
    n = A.length
    for i = 1 to n-1:
        for j = i down to 1:
            if a[j] < a[j-1]:
                swap a[j] and a[j-1]
            else:
                break # out of the inner for loop
                    # this prevents double checking the
                    # already sorted portion
```

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ **Merge Sort**
 - ▶ Master Theorem
 - ▶ Quick Sort
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



Divide & Conquer

- ▶ Algorithmic design paradigm
 - ▶ divide: divide input S into disjoint subsets S_1, \dots, S_k
 - ▶ recur: solve sub-problems on S_1, \dots, S_k
 - ▶ conquer: combine solutions for S_1, \dots, S_k into solution for S
- ▶ Base case is usually sub-problem of size 1 or 0

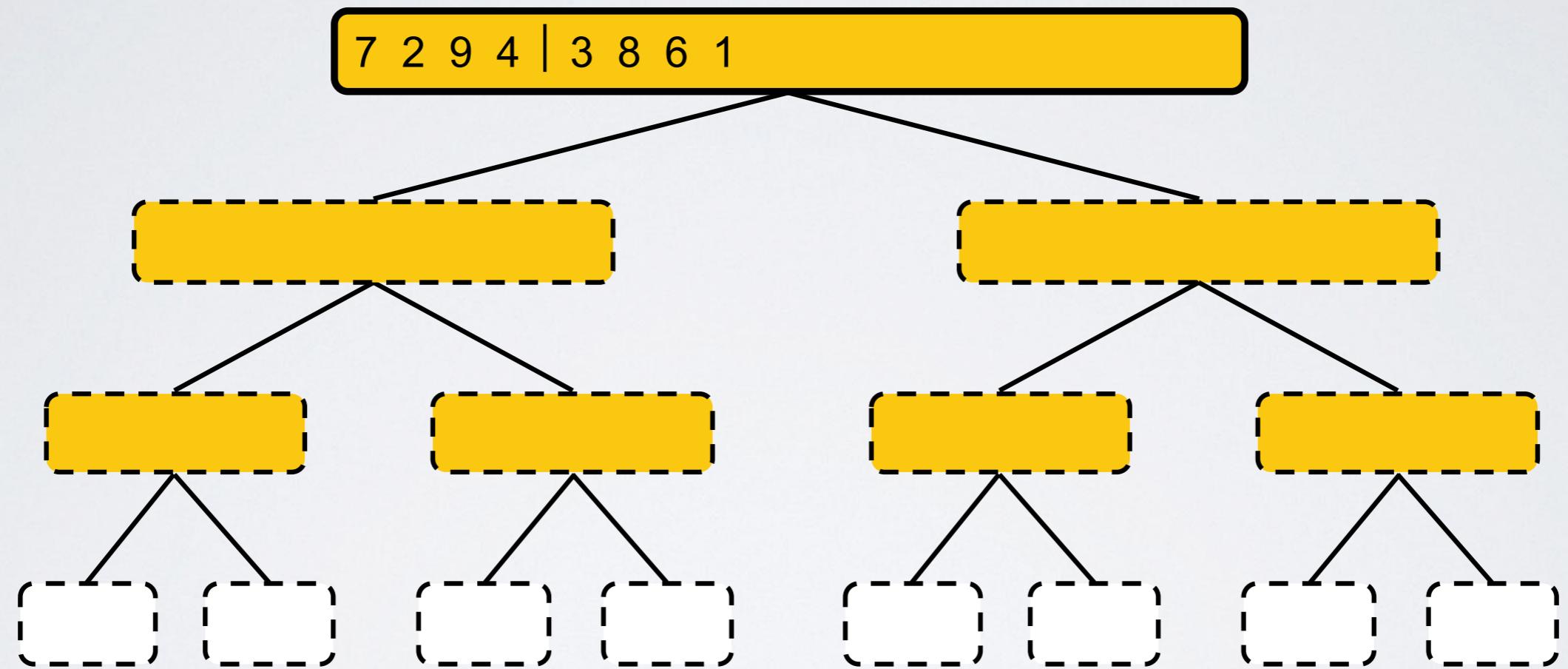
Merge Sort

- ▶ Sorting algorithm based on divide & conquer
- ▶ Like quadratic sorts
 - ▶ comparative
- ▶ Unlike quadratic sorts
 - ▶ recursive
 - ▶ linearithmic $O(n \log n)$

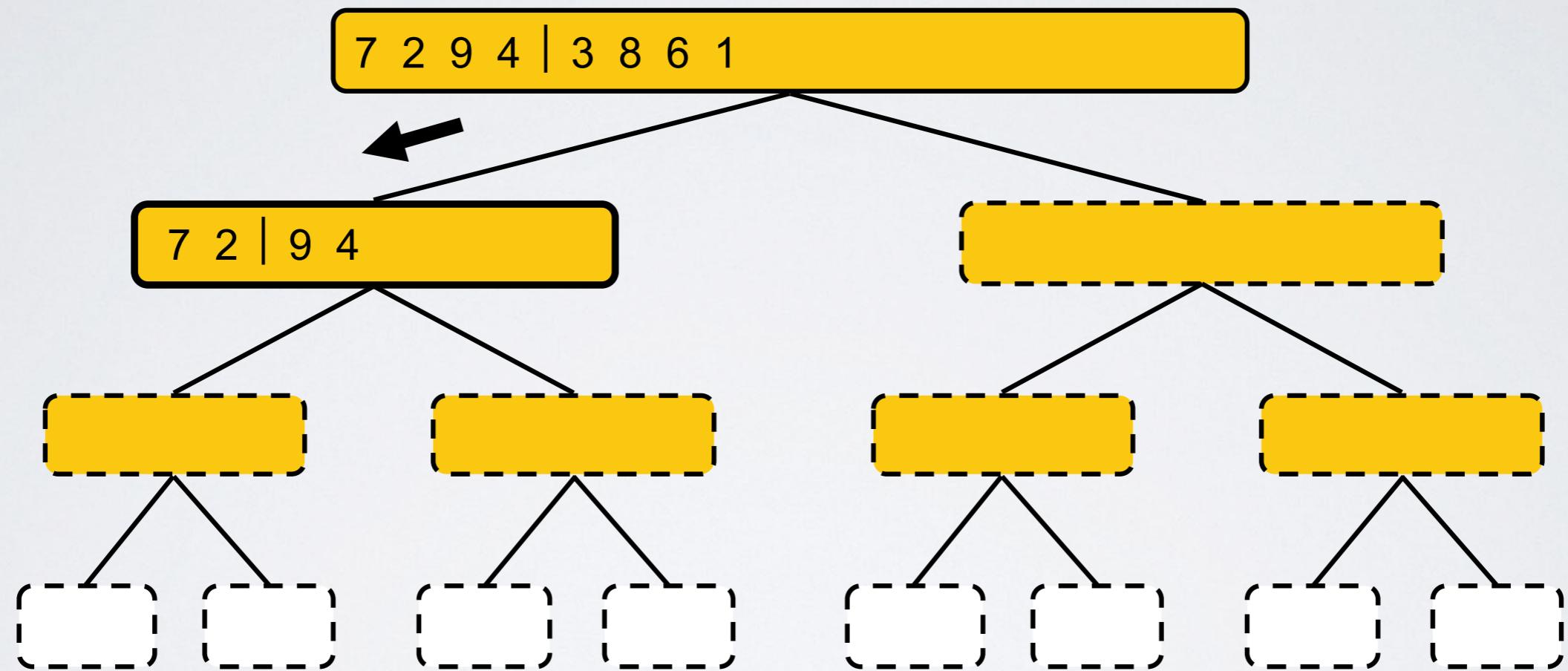
Merge Sort

- ▶ Merge sort on n-element sequence S
 - ▶ divide: divide S into disjoint subsets S_1 and S_2
 - ▶ recur: recursively merge sort S_1 and S_2
 - ▶ conquer: merge S_1 and S_2 into sorted sequence
- ▶ Suppose we want to sort
 - ▶ $7, 2, 9, 4, 3, 8, 6, 1$

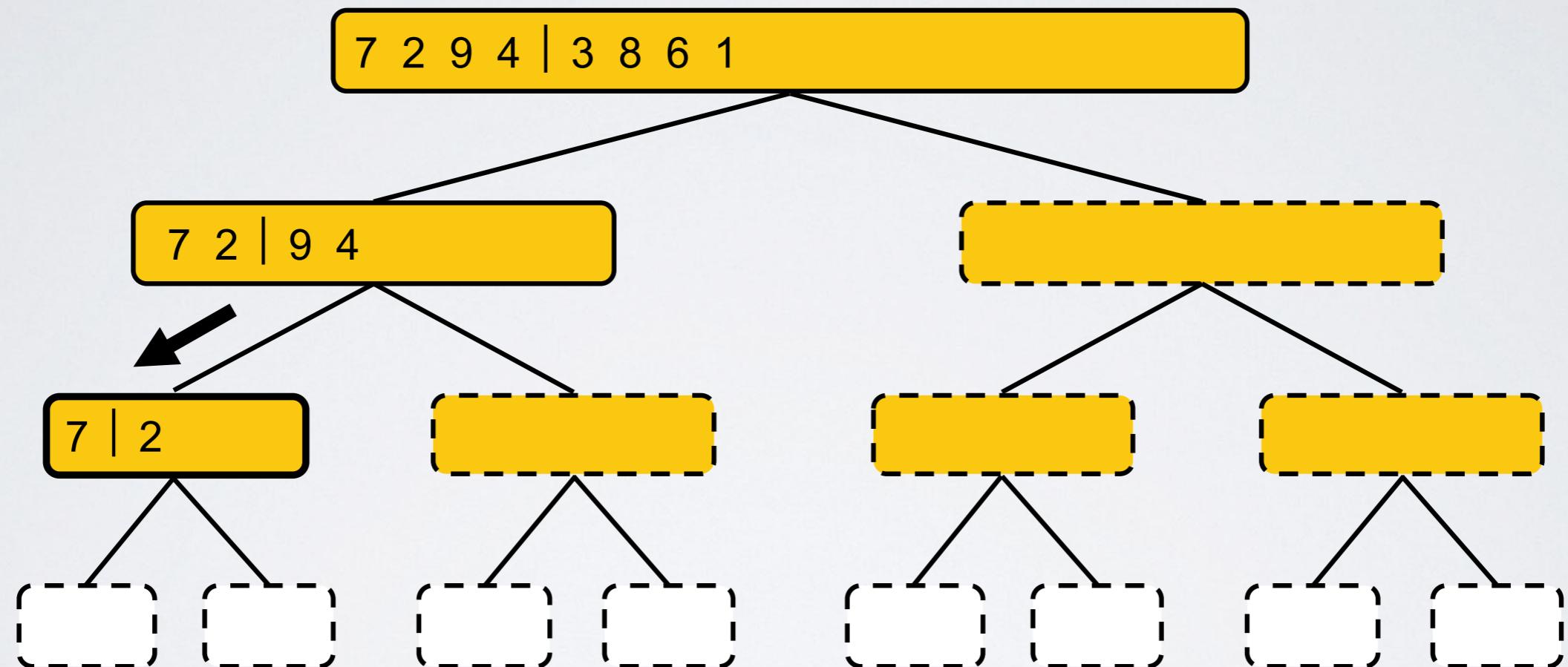
Merge Sort Recursion Tree



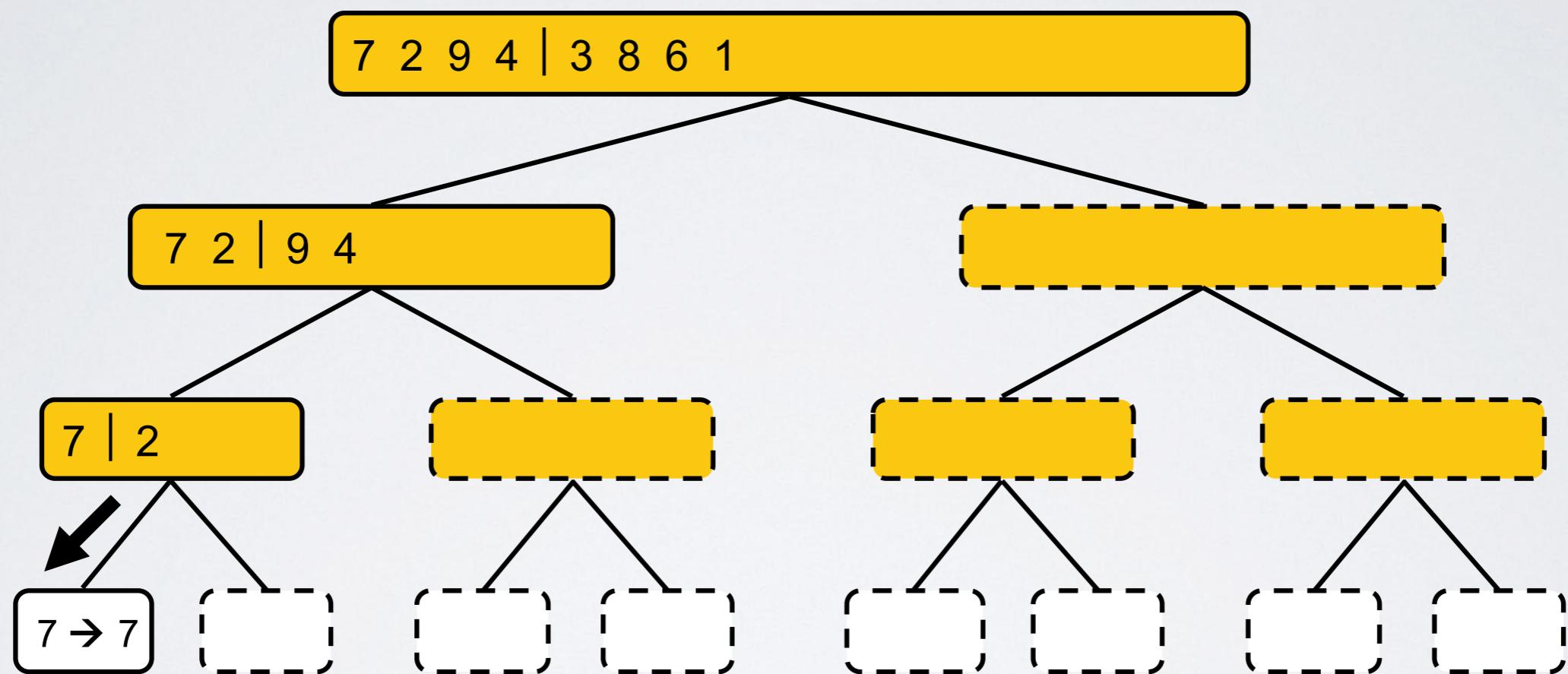
Merge Sort Recursion Tree



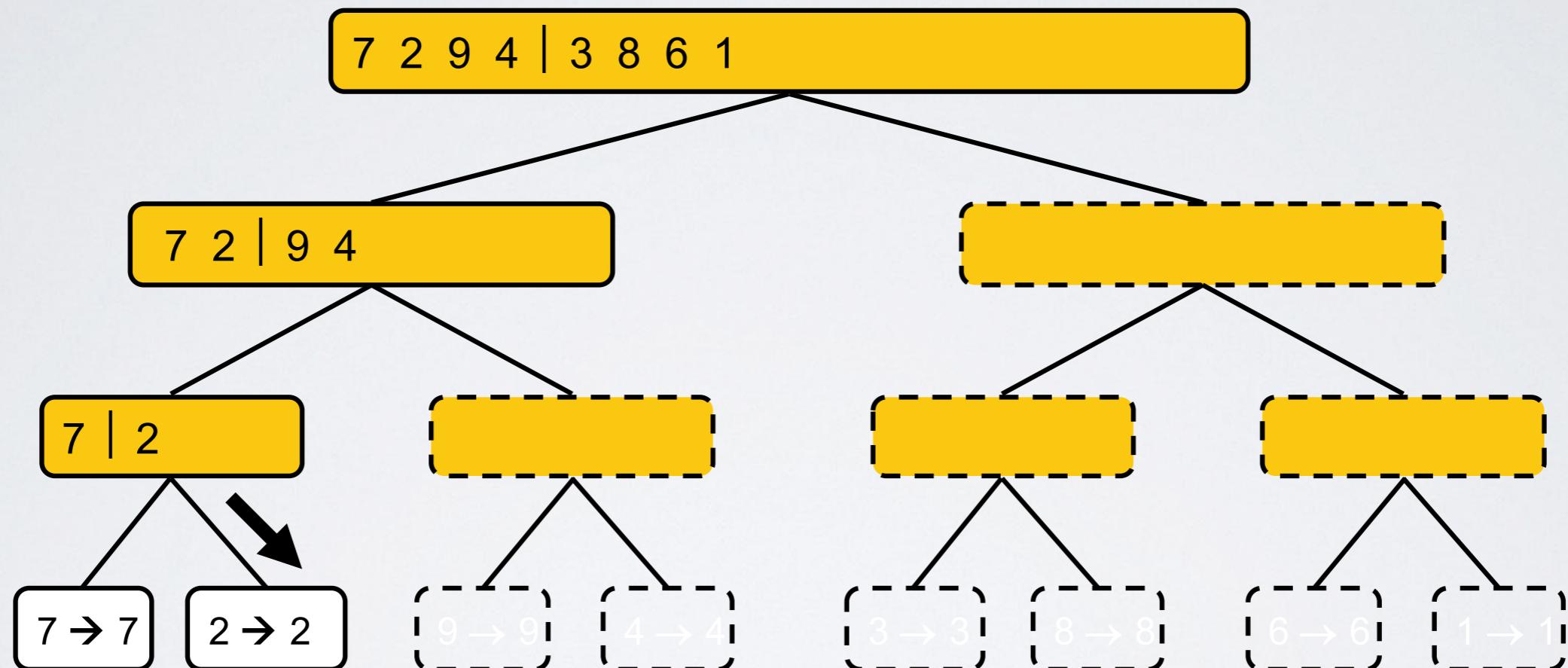
Merge Sort Recursion Tree



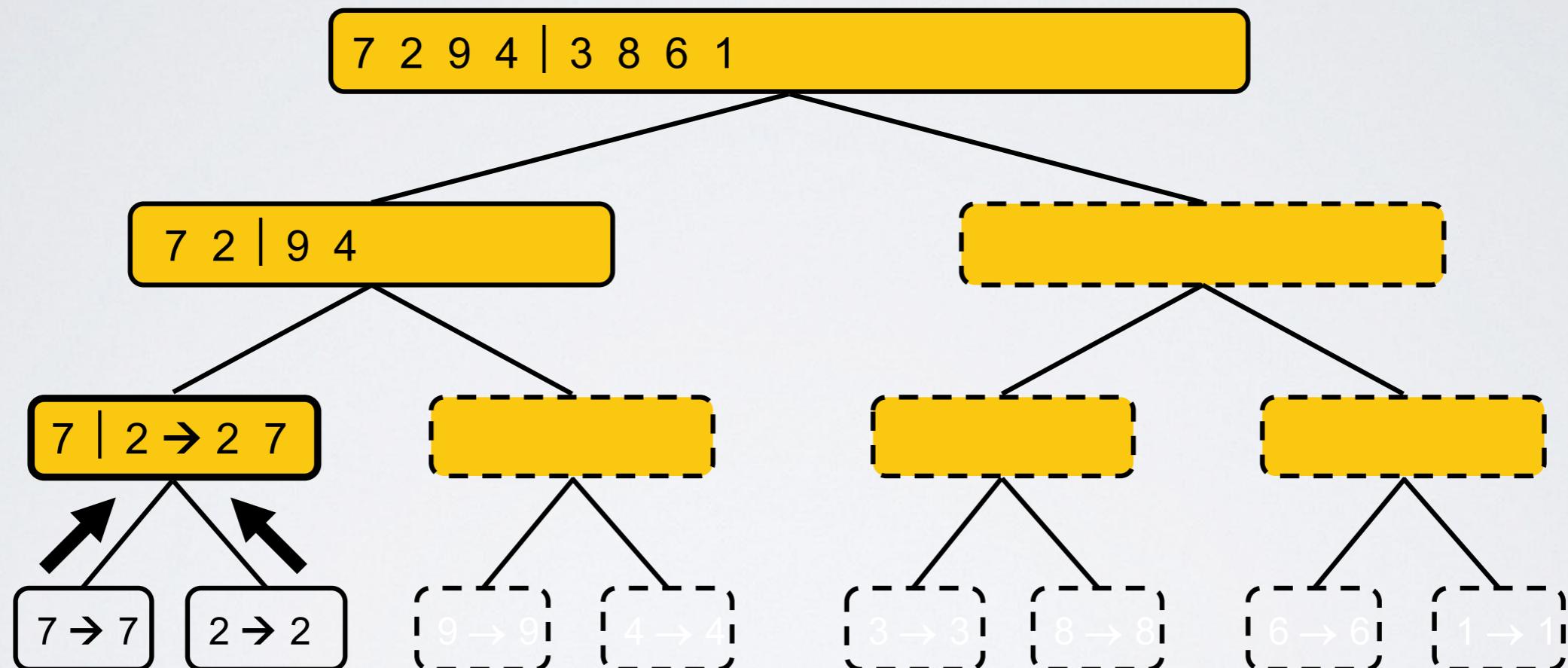
Merge Sort Recursion Tree



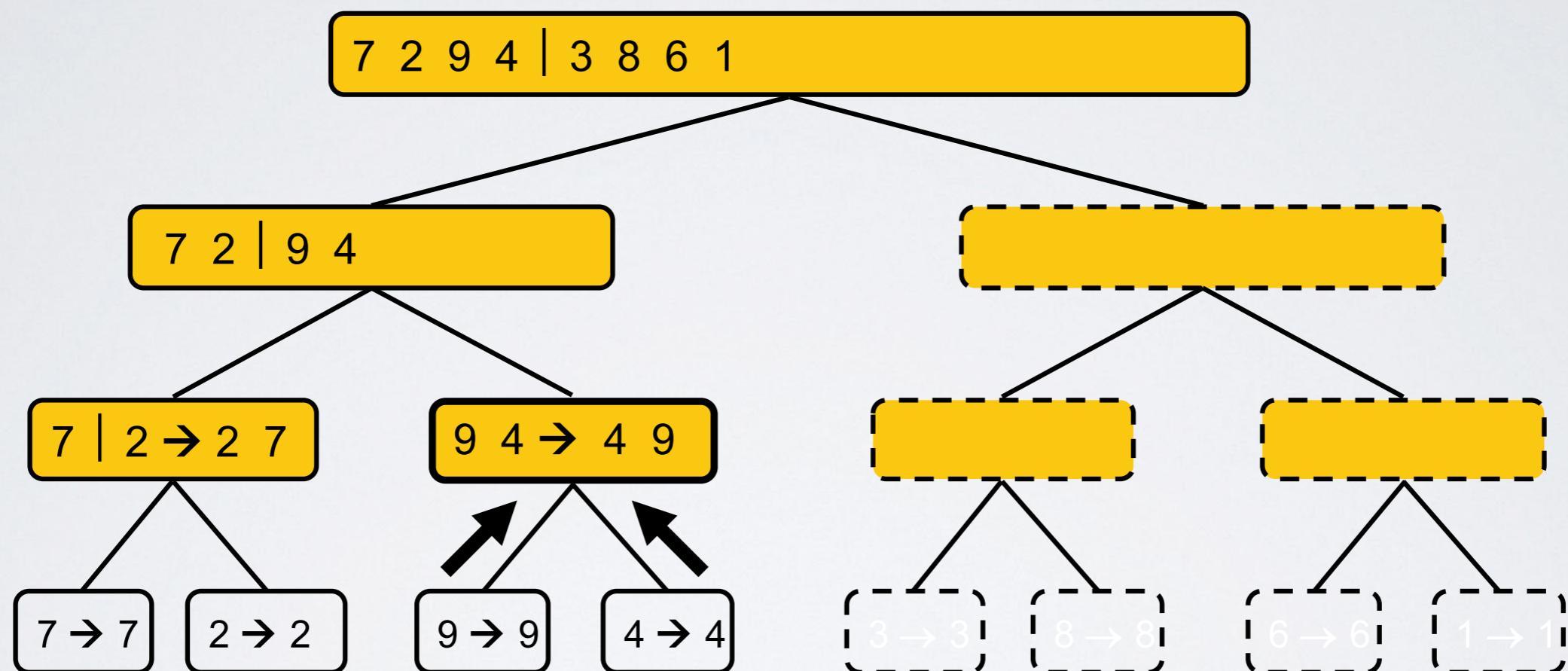
Merge Sort Recursion Tree



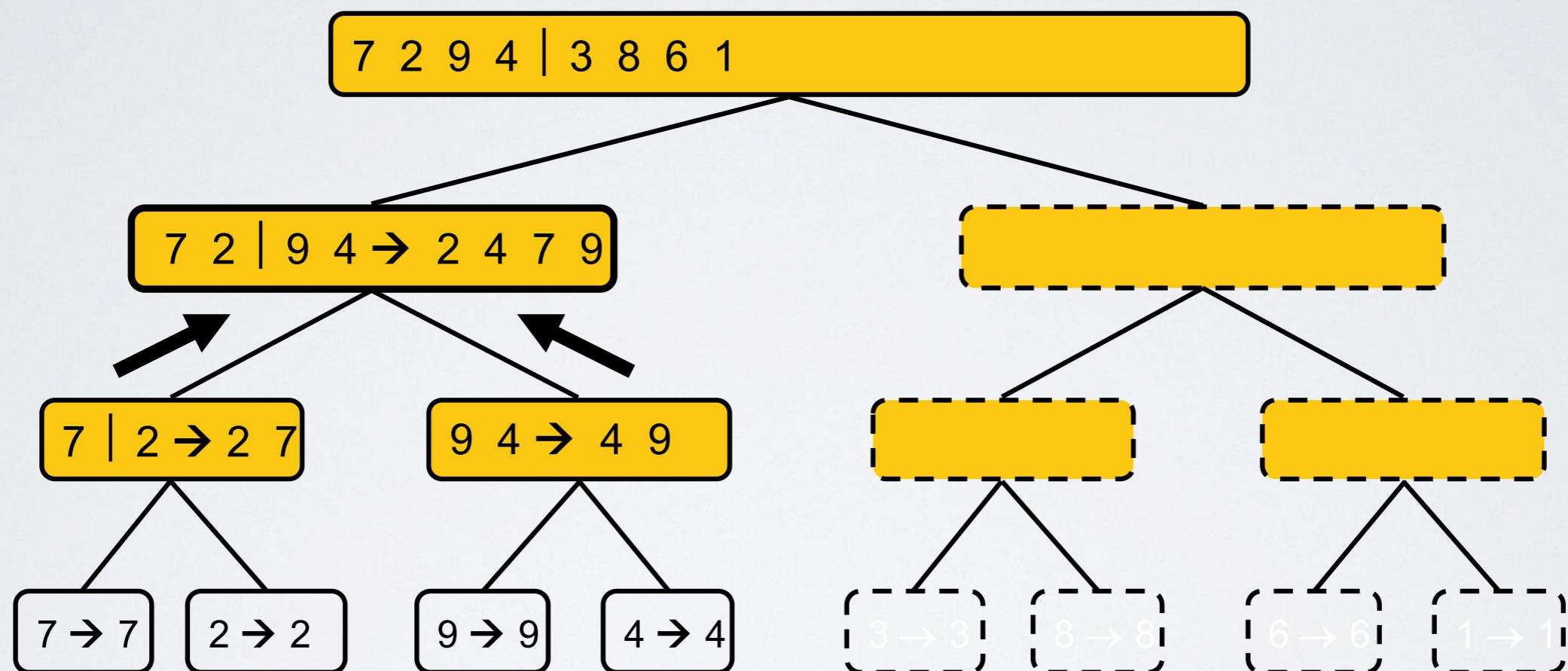
Merge Sort Recursion Tree



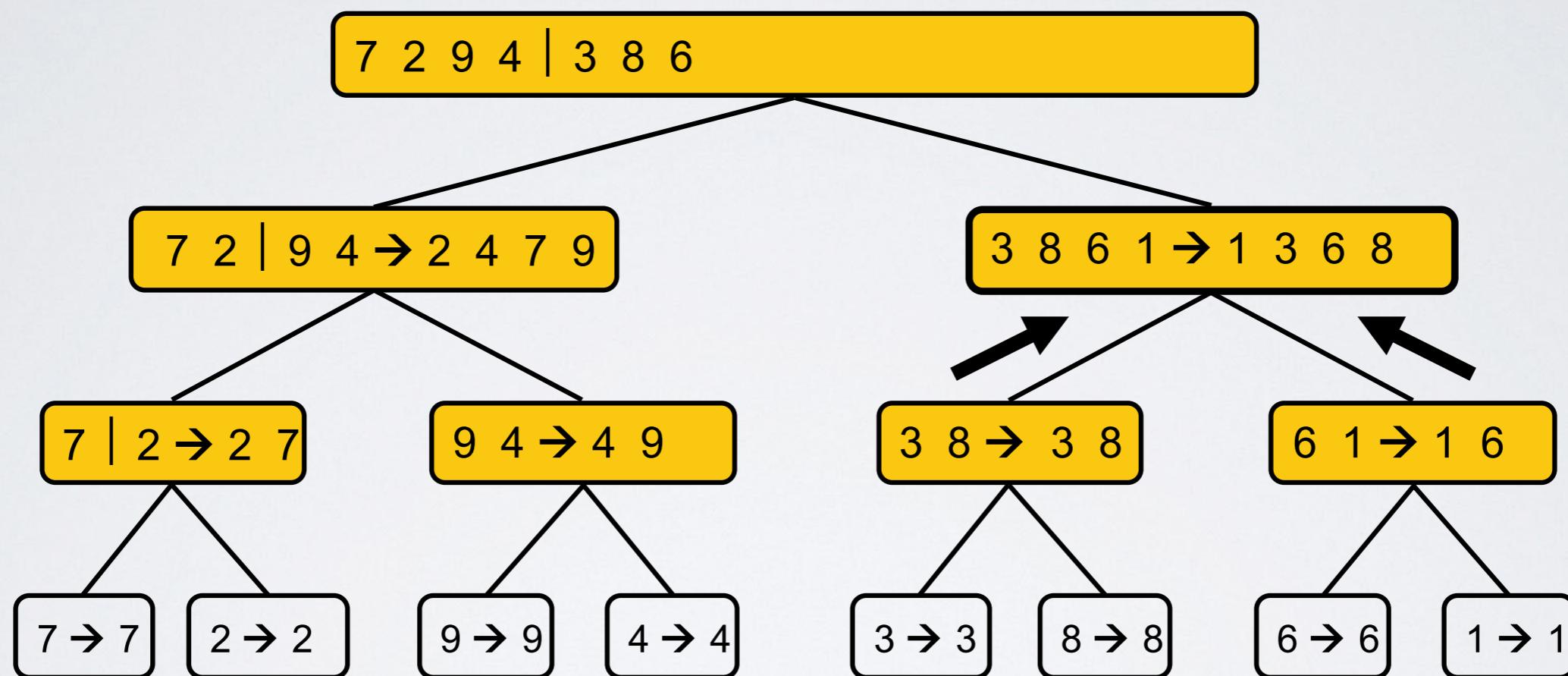
Merge Sort Recursion Tree



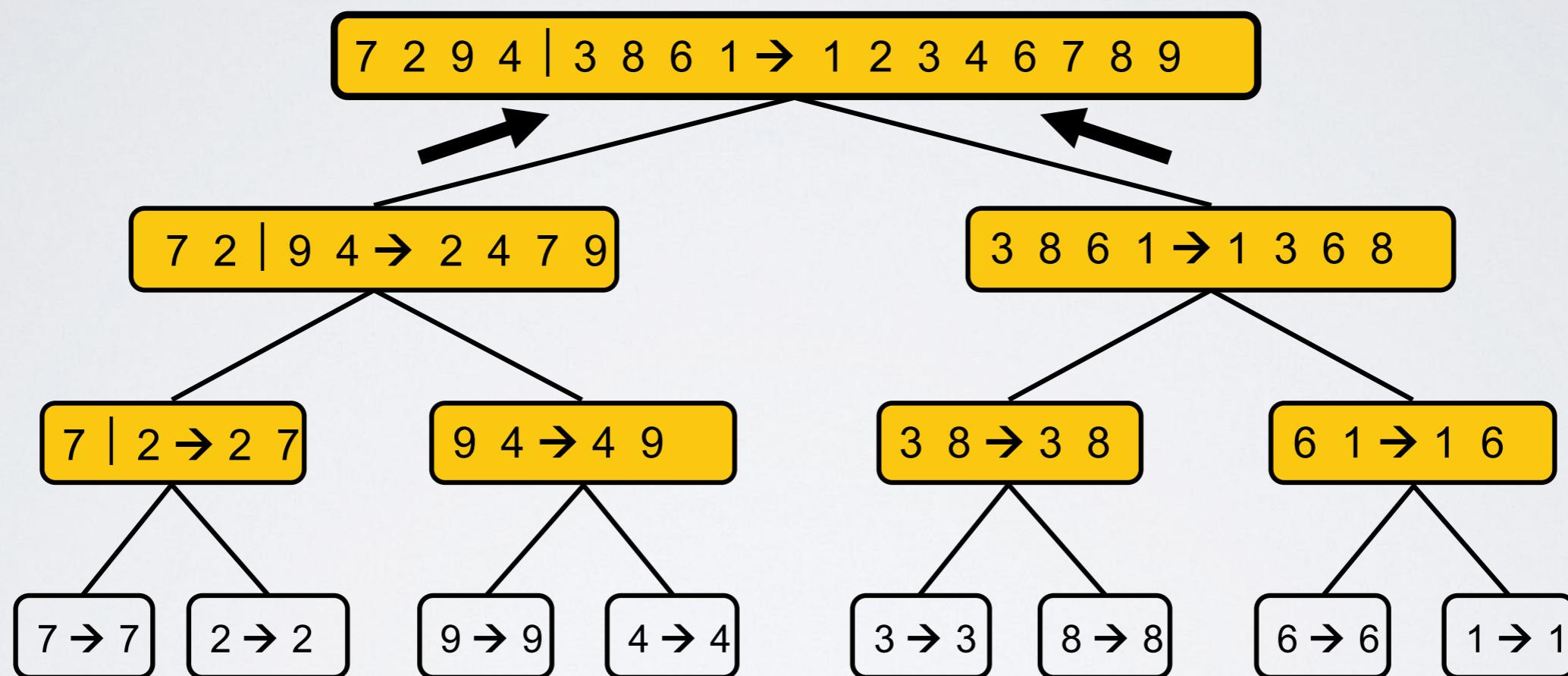
Merge Sort Recursion Tree



Merge Sort Recursion Tree



Merge Sort Recursion Tree



Merge Sort Pseudo-Code

```
function mergeSort(A):
    n = A.length
    if n <= 1:
        return A

    mid = n/2
    left = mergeSort(A[0...mid-1])
    right = mergeSort(A[mid...n-1])
    return merge(left, right)
```

Merge Sort Pseudo-Code

```
function merge(A, B):
    result = []
    aIndex = 0
    bIndex = 0
    while aIndex < A.length and bIndex < B.length:
        if A[aIndex] <= B[bIndex]:
            result.append(A[aIndex])
            aIndex++
        else:
            result.append(B[bIndex])
            bIndex++
        if aIndex < A.length:
            result = result + A[aIndex:end]
        if bIndex < B.length:
            result = result + B[bIndex:end]
    return result
```

Merge Sort

2 min

Activity #1

Merge Sort

2 min

Activity #1

Merge Sort

1 min

Activity #1

Merge Sort

Omin

Activity #1

Merge Sort Recurrence Relation

- ▶ Merge sort steps
 - ▶ Recursively merge sort left half
 - ▶ Recursively merge sort right half
 - ▶ Merge both halves
- ▶ $T(n)$: time to merge sort input of size n
 - ▶ $T(n) = \text{step 1} + \text{step 2} + \text{step 3}$
 - ▶ Steps 1 & 2 are merge sort on half input so $T(n/2)$
 - ▶ Step 3 is $O(n)$

Merge Sort Recurrence Relation

- ▶ General case

$$T(n) = T\left(\frac{n}{2}\right) + T\left(\frac{n}{2}\right) + O(n) = 2 \cdot T\left(\frac{n}{2}\right) + O(n)$$

- ▶ Base case

$$T(1) = c$$

Merge Sort Recurrence Relation

- ▶ Plug & chug

$$T(1) = c_1$$

$$T(2) = 2 \cdot T(1) + 2 = 2c_1 + 2$$

$$T(4) = 2 \cdot T(2) + 4 = 2(2c_1 + 2)4 = 4c_1 + 8$$

$$T(8) = 2 \cdot T(4) + 8 = 2(4c_1 + 8) + 8 = 8c_1 + 24$$

$$T(16) = 2 \cdot T(8) + 16 = 2(8c_1 + 24) + 16 = 16c_1 + 64$$

- ▶ Solution

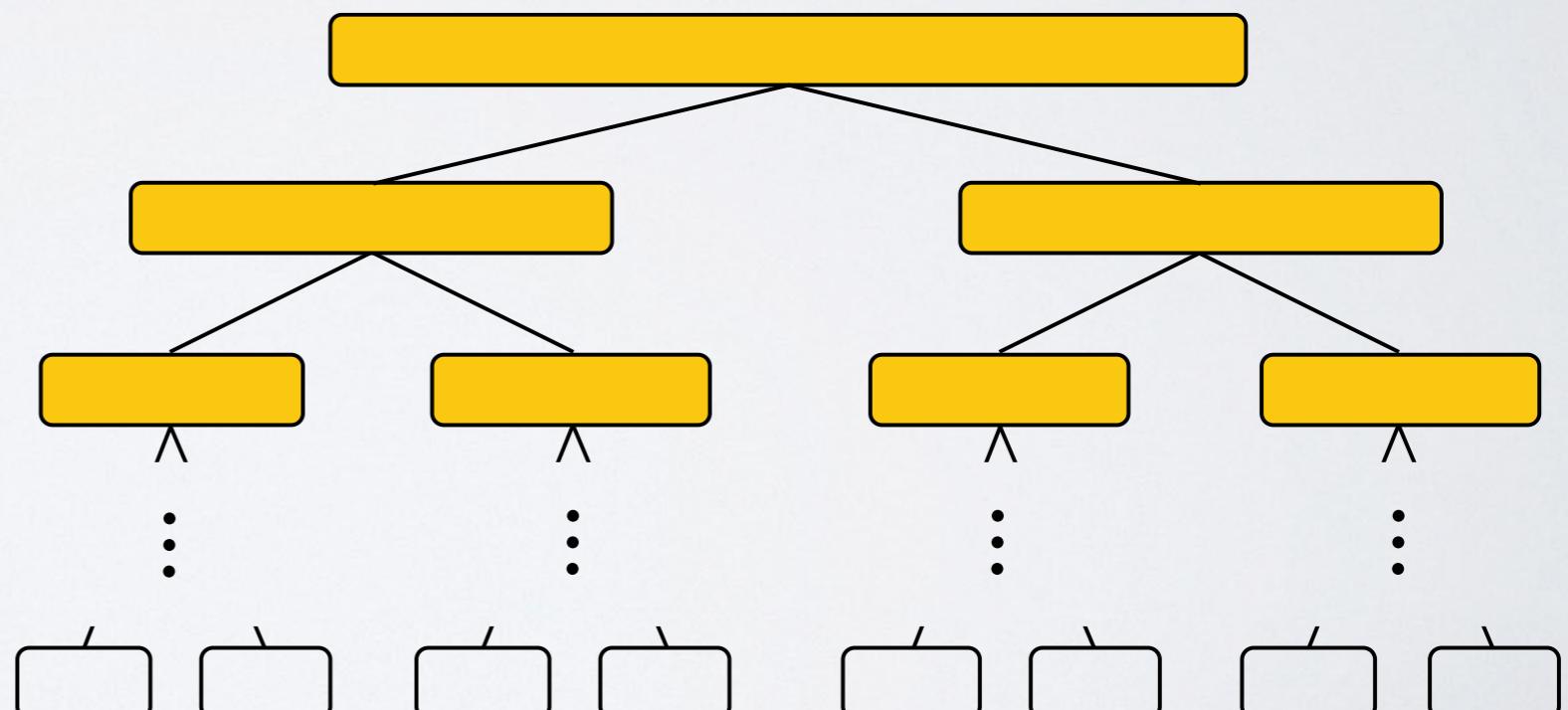
$$T(n) = nc_1 + n \log n = O(n \log n)$$

Analysis of Merge Sort

- ▶ Merge sort recursive tree is perfect binary tree so has height $O(\log n)$
- ▶ At each depth k : need to merge 2^{k+1} sequences of size $n/2^{k+1}$
 - ▶ work at each depth is $O(n)$

depth	sequence	size
-------	----------	------

0	2	$n/2$
1	4	$n/4$
2	8	$n/4$
:	:	:
k	2^{k+1}	$n/2^{k+1}$



Analysis of Merge Sort

- ▶ To determine that Merge sort was $O(n \log n)$
 - ▶ Use plug and chug to guess a solution
 - ▶ Prove that $O(n \log n)$ is correct (e.g., using induction)
- ▶ Can be a lot of work

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ **Master Theorem**
 - ▶ Quick Sort
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



The Master Theorem

- ▶ Solves large class of recurrence relations
 - ▶ we will learn how to use it but not its proof
 - ▶ See Dasgupta et al. p. 58-60 for proof
- ▶ Let $T(n)$ be a monotonically-increasing function of form
$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \Theta(n^d)$$
 - ▶ **a**: number of sub-problems
 - ▶ **n/b**: size of each sub-problem
 - ▶ **n^d** : work to prepare sub-problems & combine their solutions



The Master Theorem

- ▶ If $a \geq 1, b > 1, d \geq 0$, then
 - ▶ if $a < b^d$ then $T(n) = \Theta(n^d)$
 - ▶ if $a = b^d$ then $T(n) = \Theta(n^d \log n)$
 - ▶ if $a > b^d$ then $T(n) = \Theta(n^{\log_b a})$
- ▶ Applying Master Theorem to merge sort
 - ▶ Recurrence relation of merger sort: $T(n) = 2T(n/2) + O(n^1)$
 - ▶ $a=2, b=2$ and $d=1$ so $a=b^d$
 - ▶ and $T(n) = \Theta(n^1 \log n)$
 $= \Theta(n^1 \log n)$
 $= \Theta(n \log n)$

Master Theorem

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \Theta(n^d)$$

- ▶ $T(n) = \Theta(n^d)$ if $a < b^d$
- ▶ $T(n) = \Theta(n^d \log n)$ if $a = b^d$
- ▶ $T(n) = \Theta(n^{\log_b a})$ if $a > b^d$

2 min

• Activity #2+3

Master Theorem

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \Theta(n^d)$$

- ▶ $T(n) = \Theta(n^d)$ if $a < b^d$
- ▶ $T(n) = \Theta(n^d \log n)$ if $a = b^d$
- ▶ $T(n) = \Theta(n^{\log_b a})$ if $a > b^d$

2 min

• Activity #2+3

Master Theorem

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \Theta(n^d)$$

- ▶ $T(n) = \Theta(n^d)$ if $a < b^d$
- ▶ $T(n) = \Theta(n^d \log n)$ if $a = b^d$
- ▶ $T(n) = \Theta(n^{\log_b a})$ if $a > b^d$

1 min

Activity #2+3

Master Theorem

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \Theta(n^d)$$

- ▶ $T(n) = \Theta(n^d)$ if $a < b^d$
- ▶ $T(n) = \Theta(n^d \log n)$ if $a = b^d$
- ▶ $T(n) = \Theta(n^{\log_b a})$ if $a > b^d$

Omin

• Activity #2+3

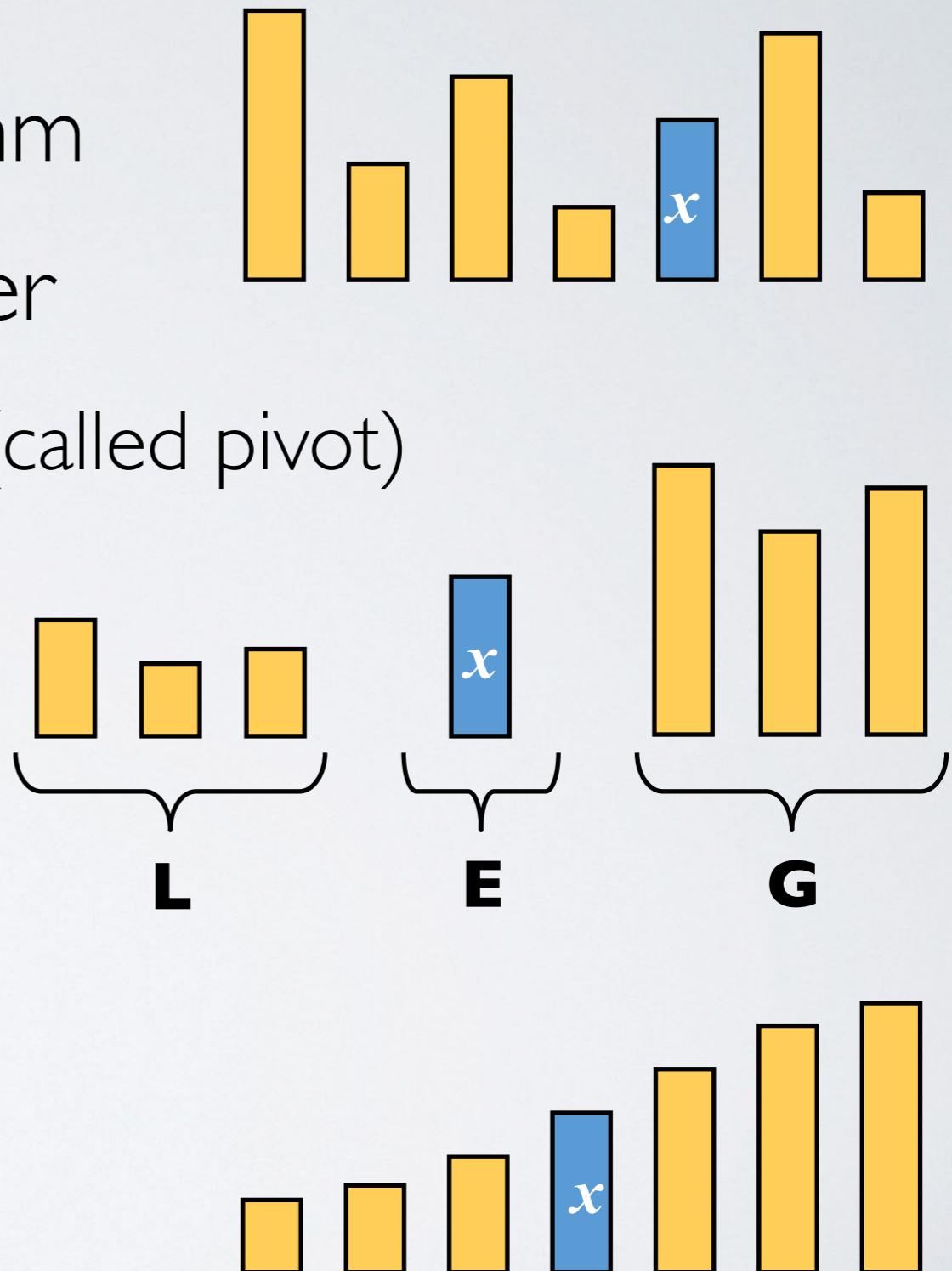
Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ Master Theorem
 - ▶ **Quick Sort**
- ▶ Comparative sorting lower bound
- ▶ Linear Sorting
 - ▶ Radix Sort



Quicksort

- ▶ Randomized sorting algorithm
- ▶ Based on divide-and-conquer
 - ▶ divide: pick random element (called pivot) and partition set into
 - ▶ **L:** elements less than x
 - ▶ **E:** elements equal to x
 - ▶ **G:** elements larger than x
 - ▶ recur: quicksort L and G
 - ▶ conquer: join L, E and G



Quicksort

2 min

Activity #4

Quicksort

2 min

Activity #4

Quicksort

1 min

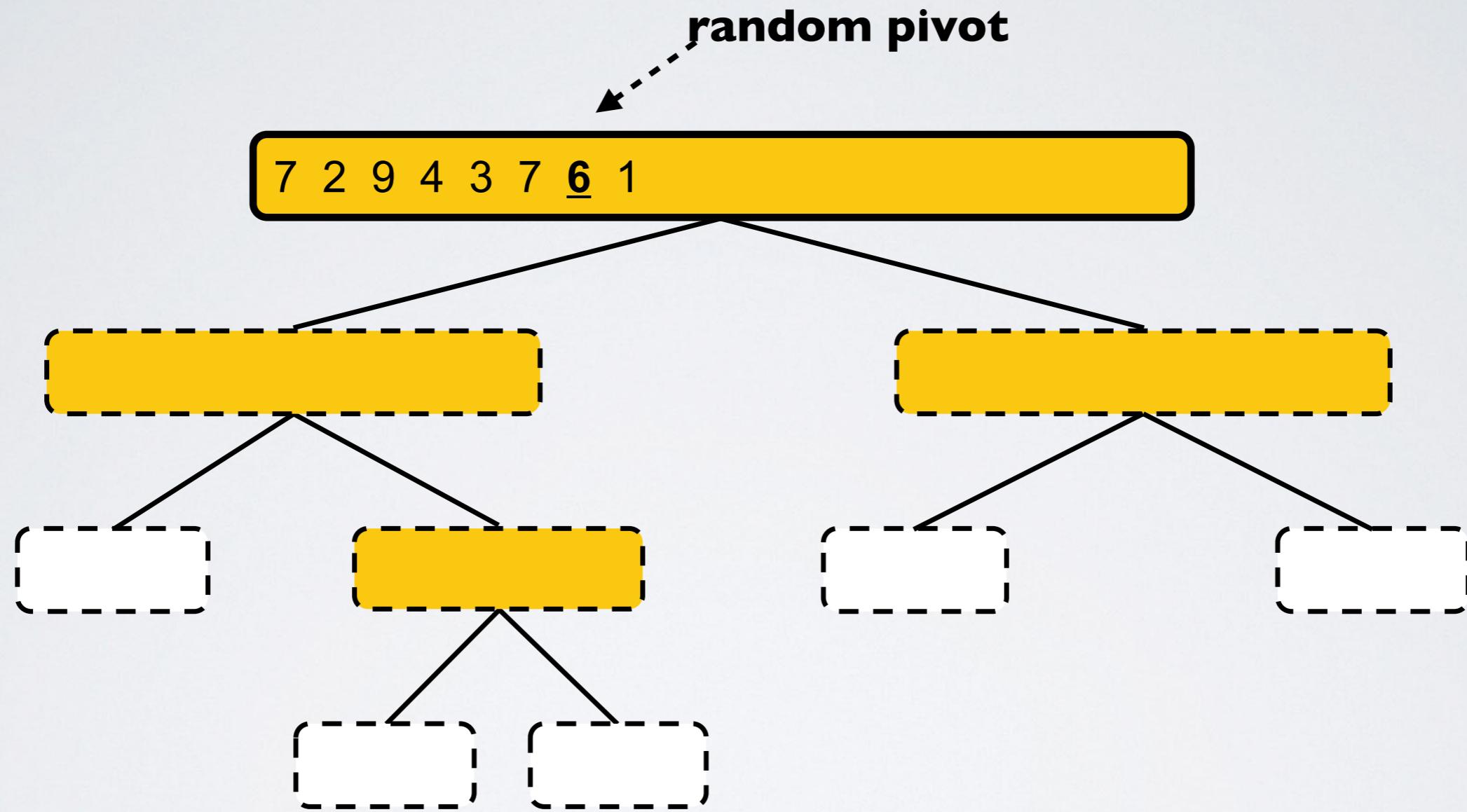
Activity #4

Quicksort

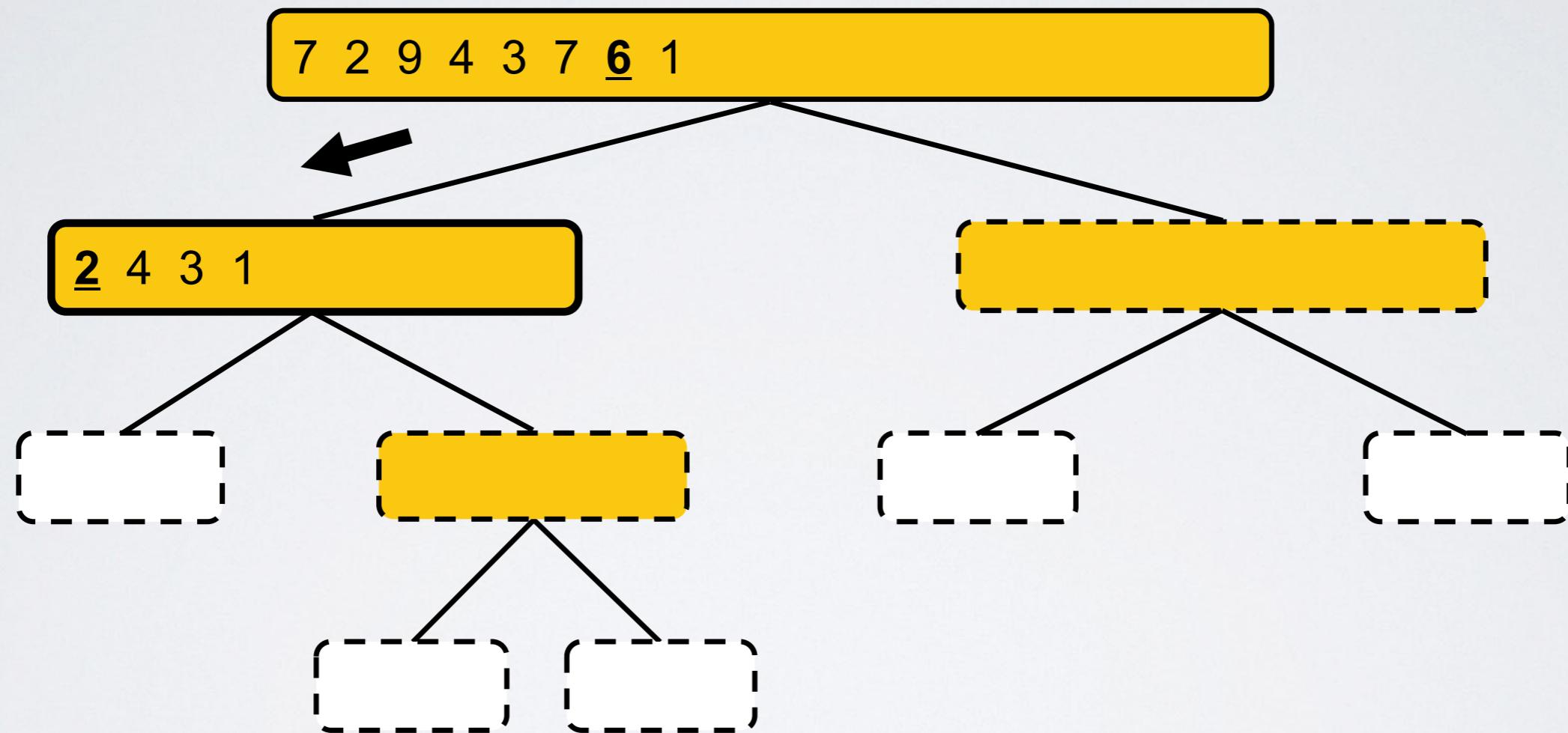
O min.

Activity #4

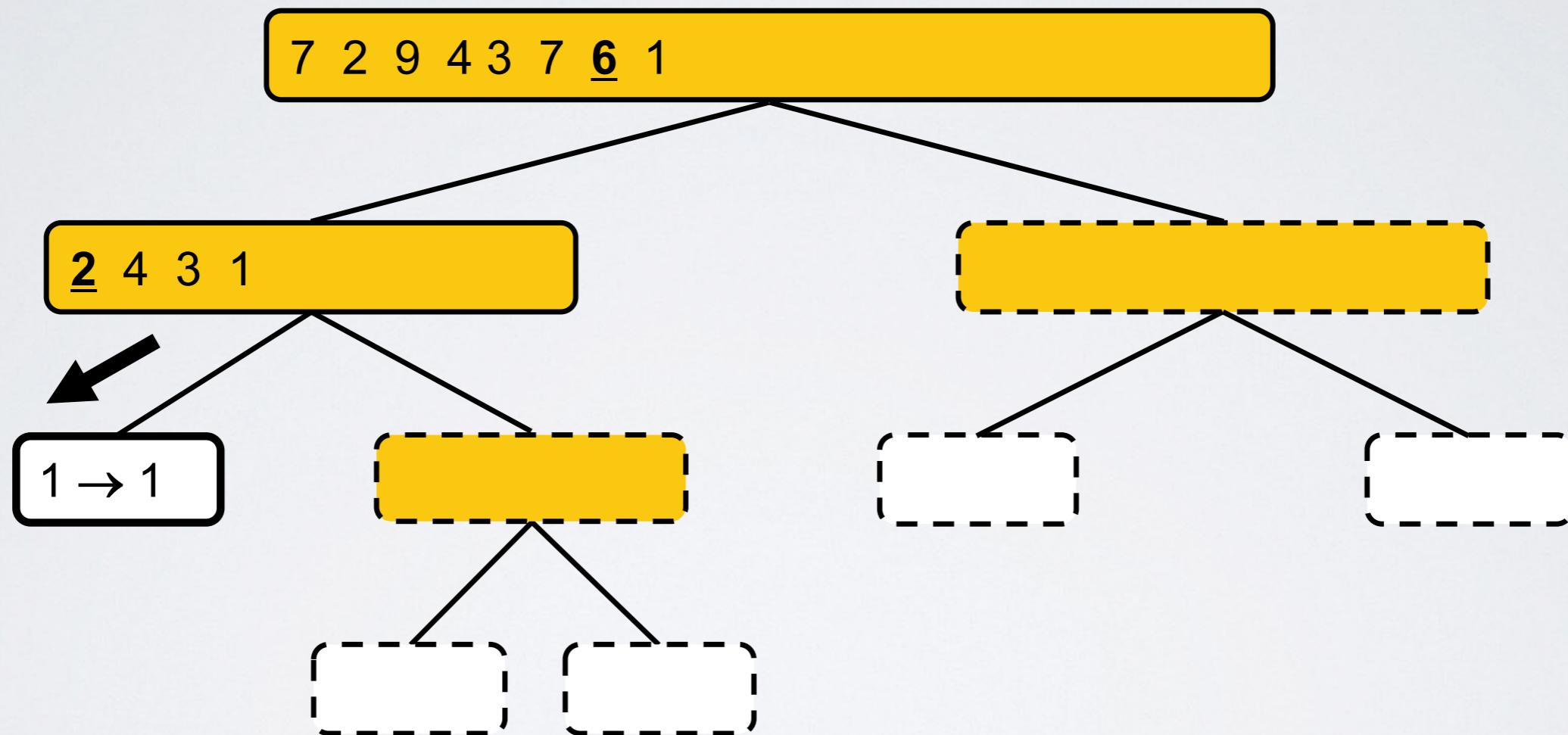
Quicksort Example



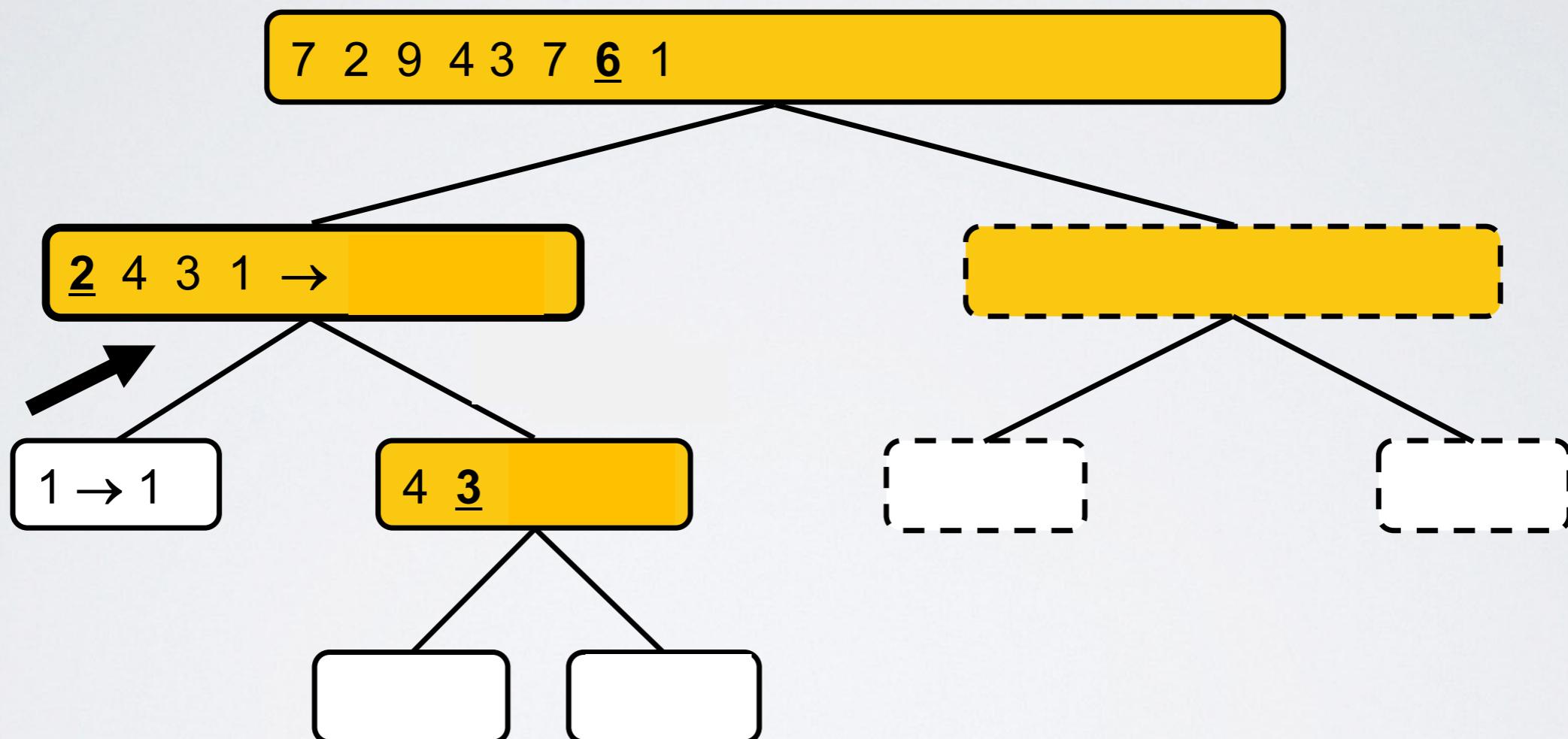
Quicksort Example



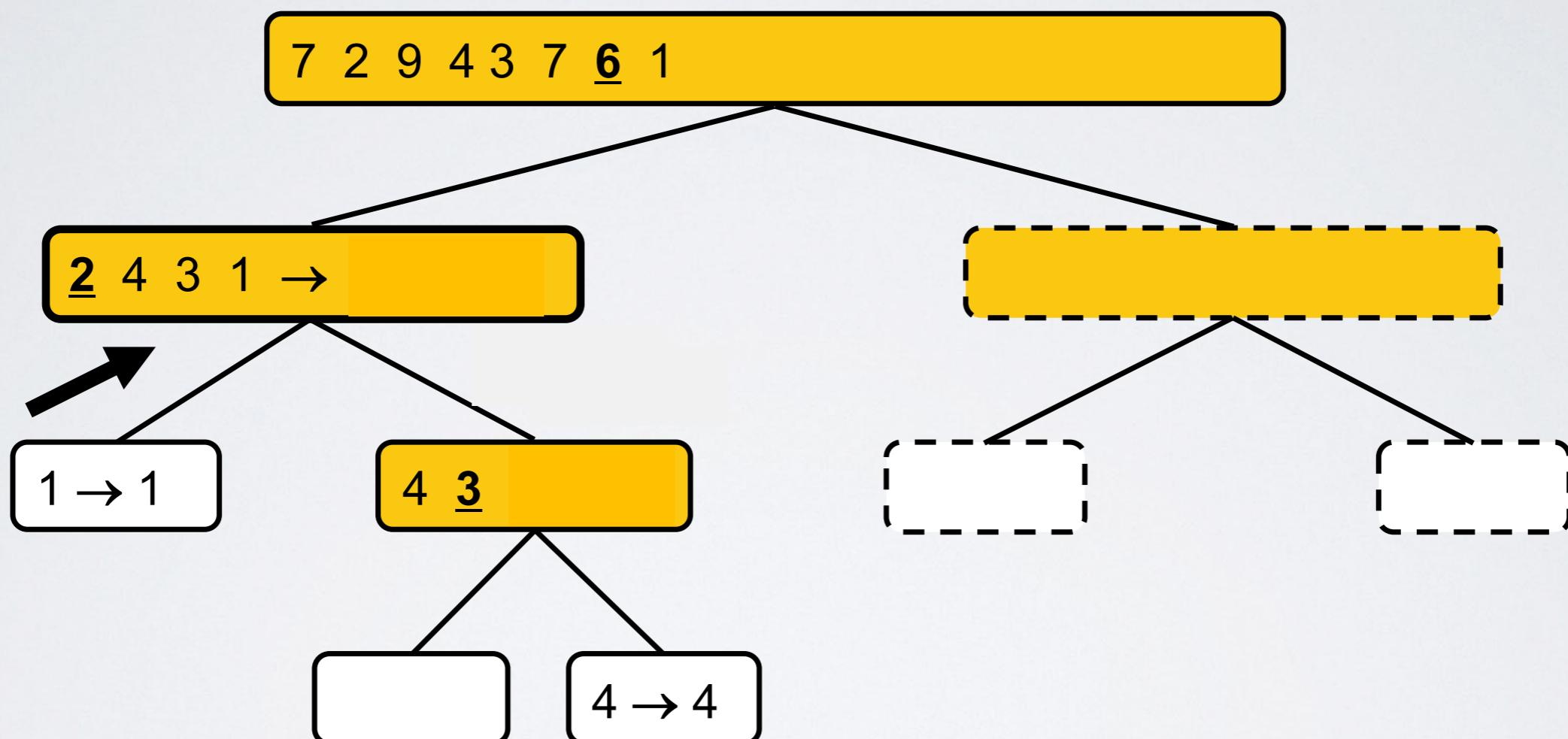
Quicksort Example



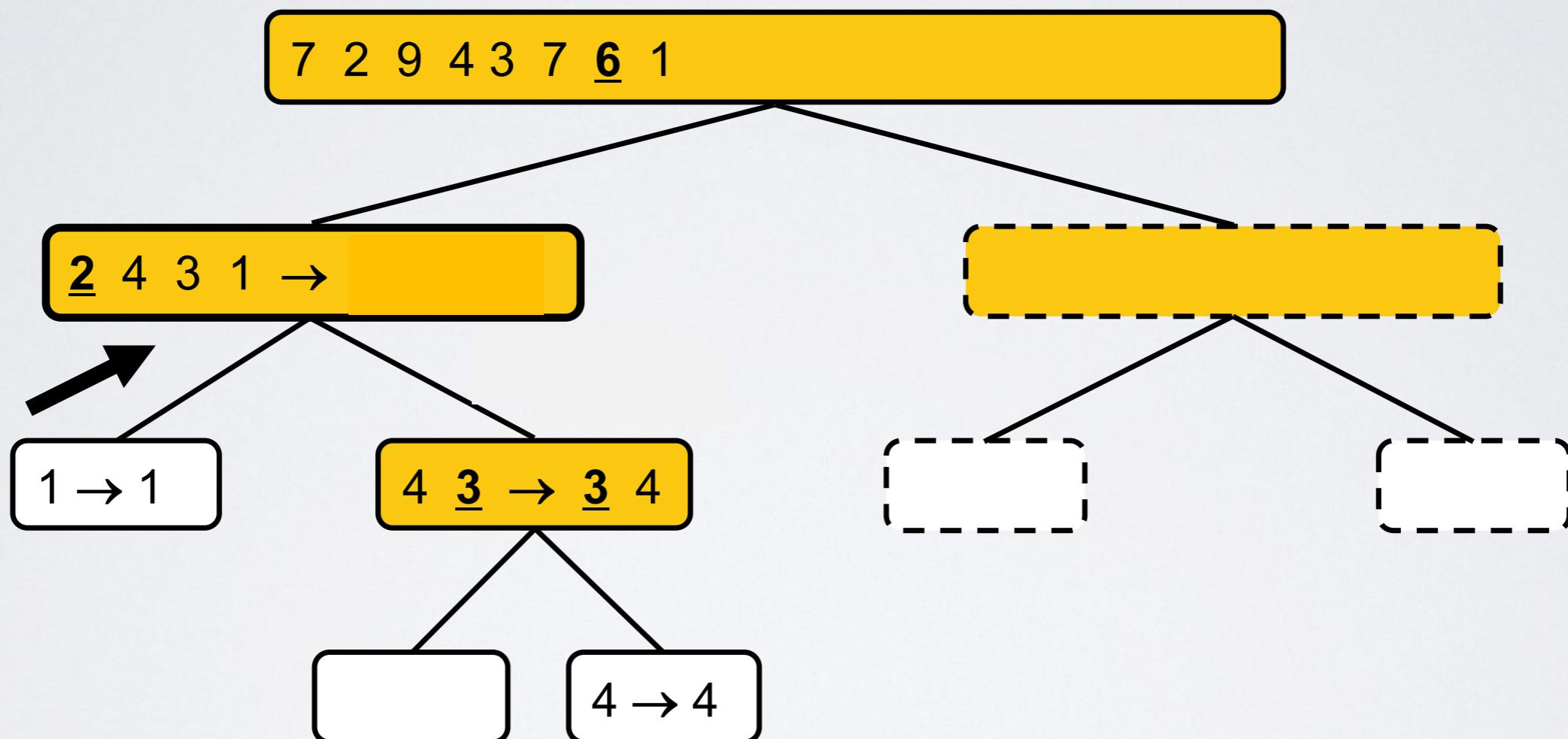
Quicksort Example



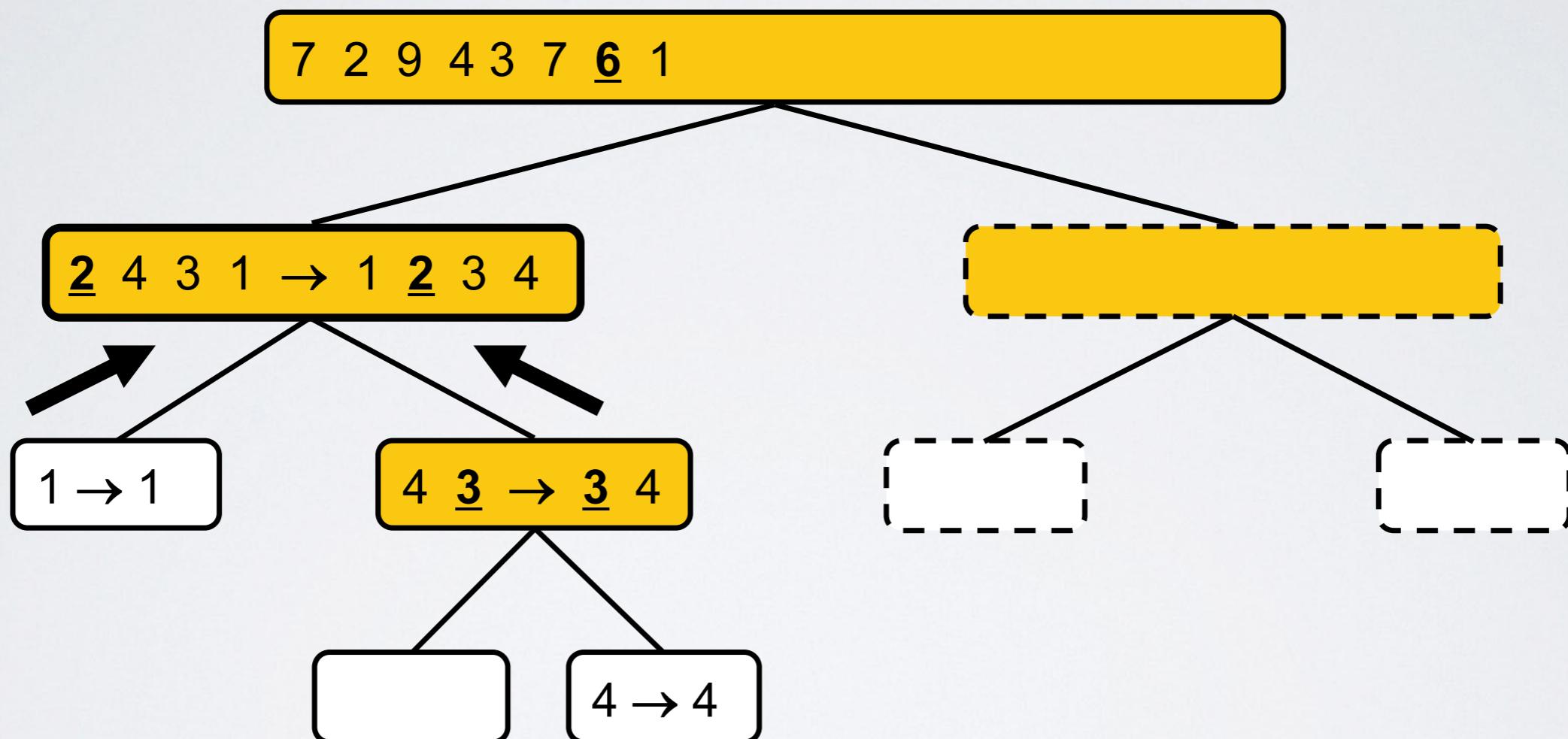
Quicksort Example



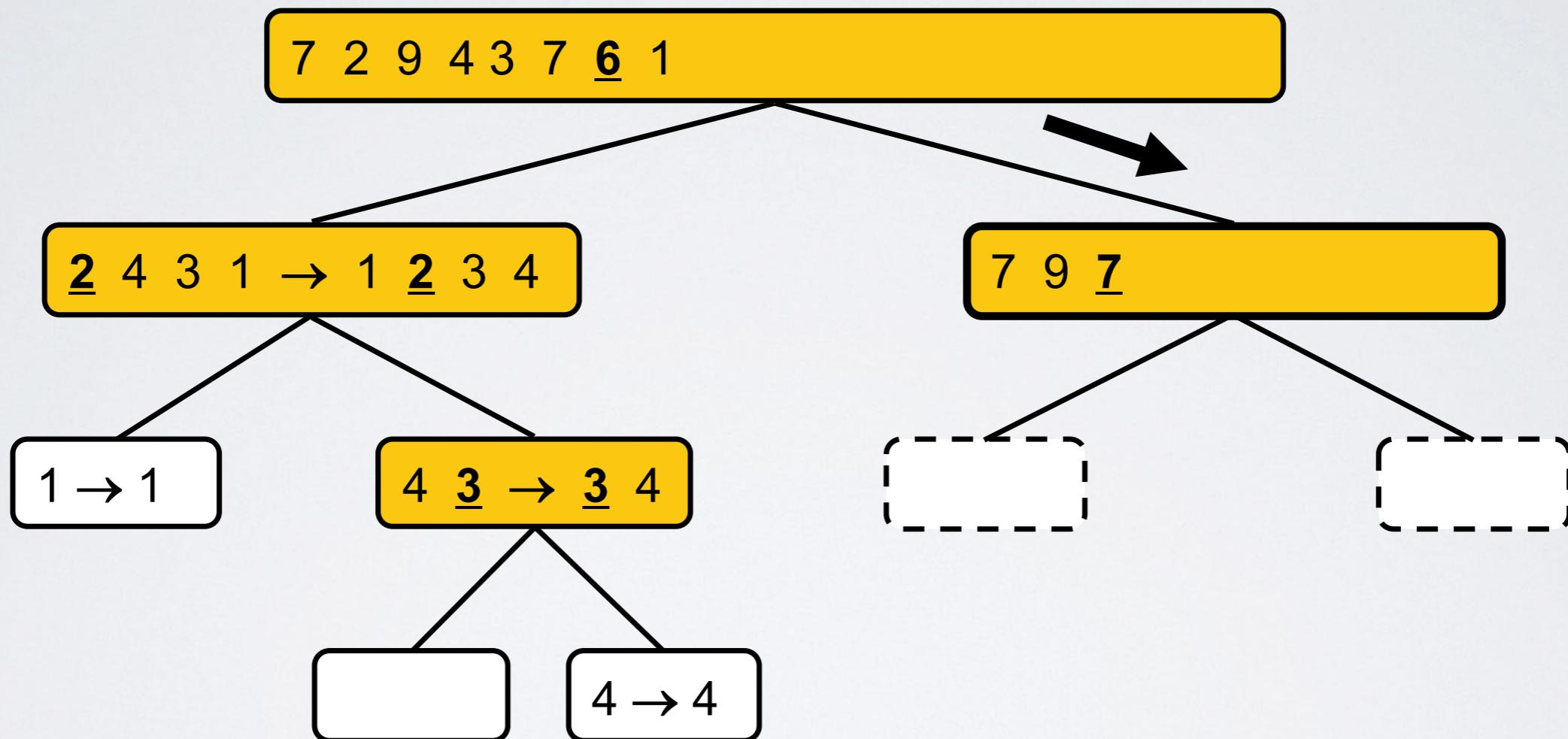
Quicksort Example



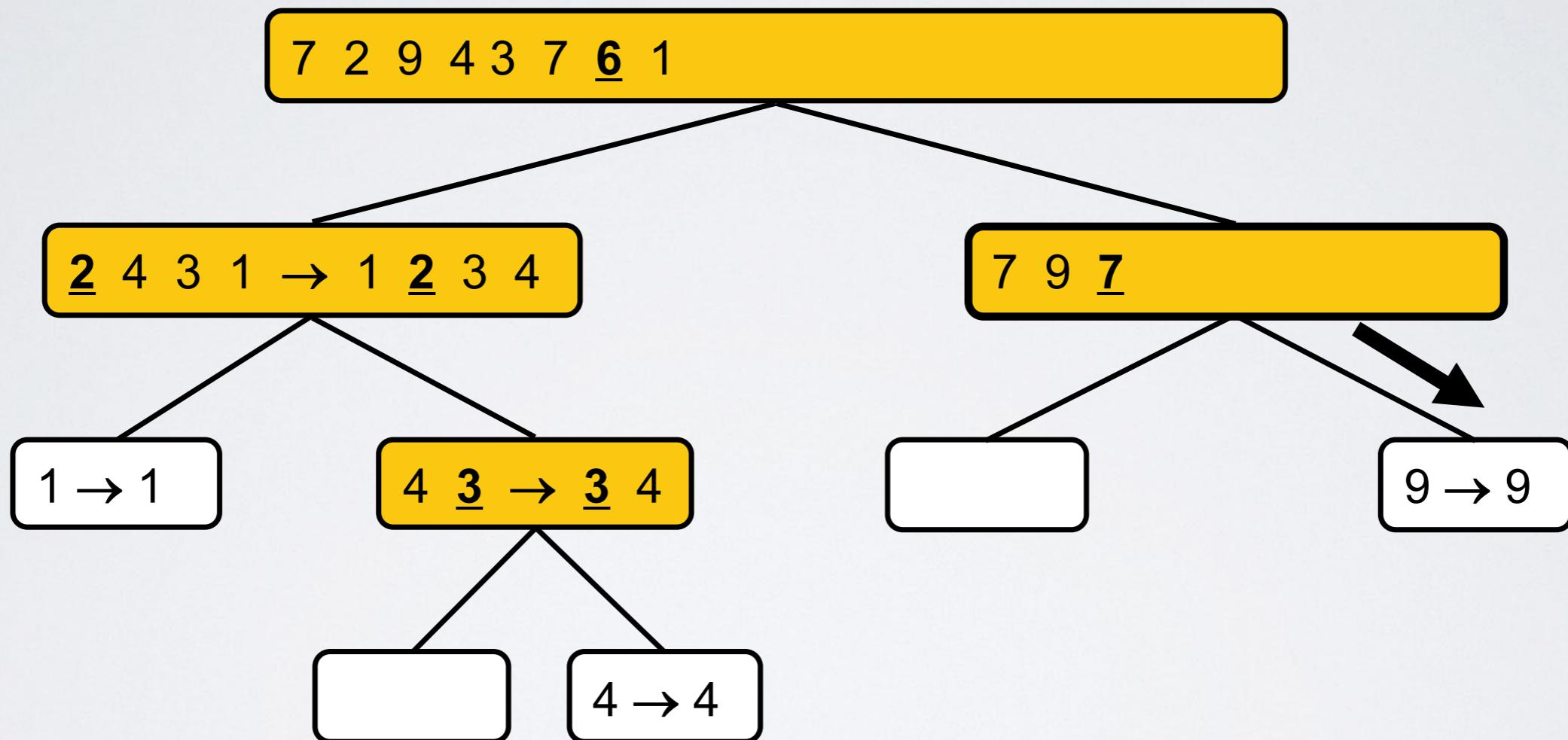
Quicksort Example



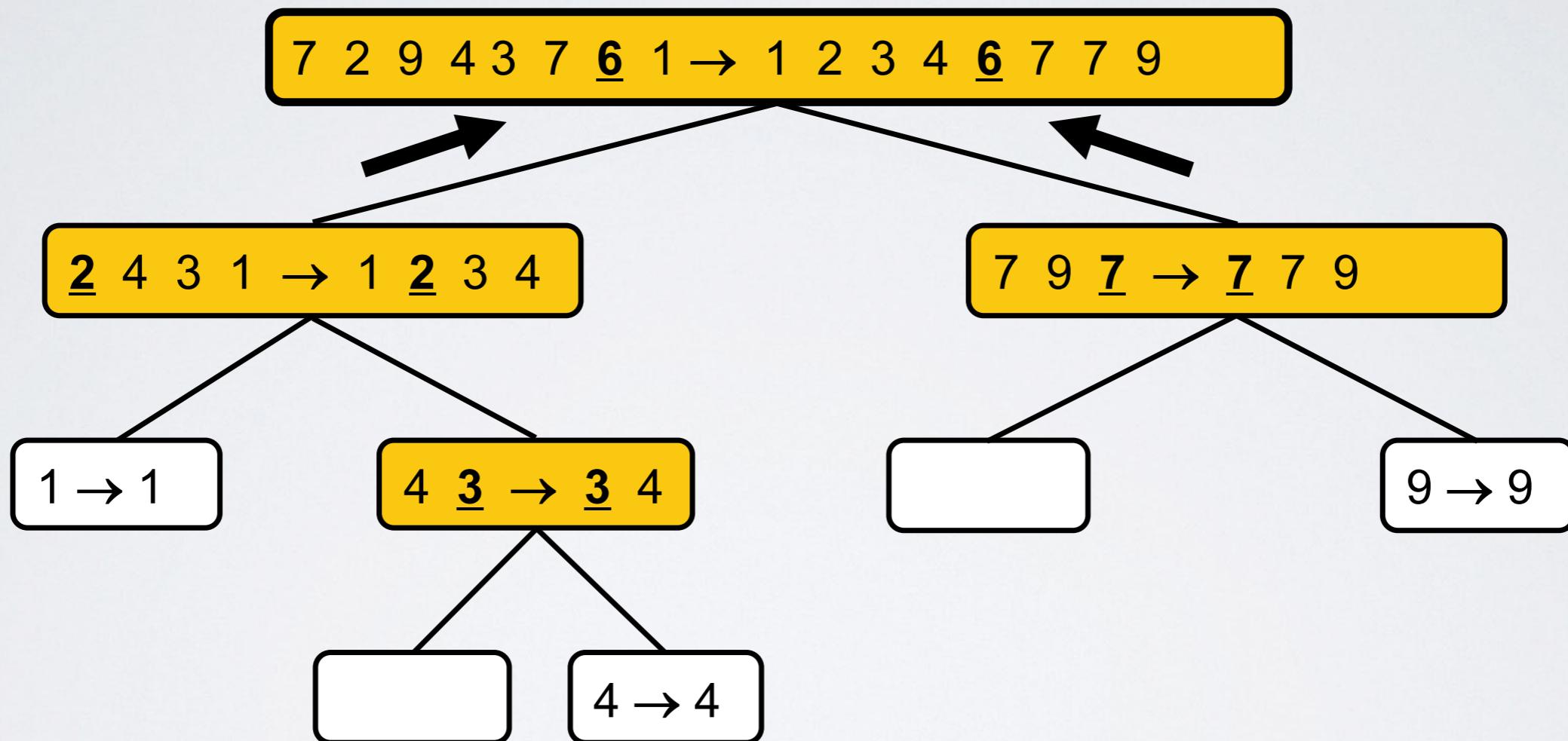
Quicksort Example



Quicksort Example



Quicksort Example



Quicksort Pseudo-Code

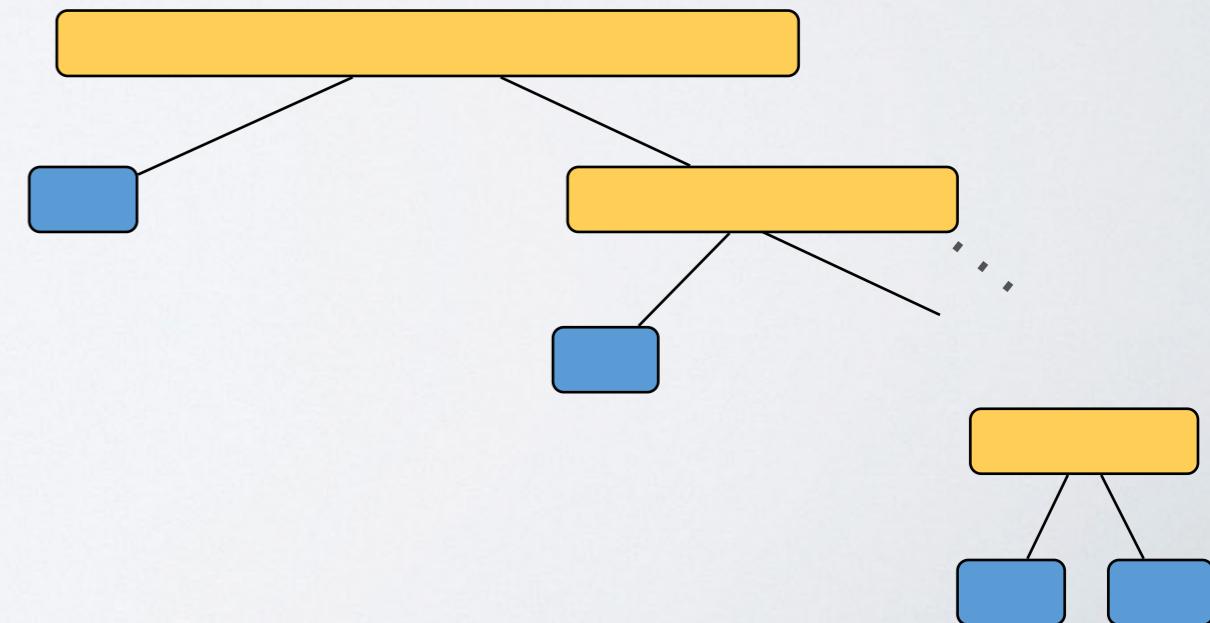
```
function quick_sort(A):
    if A.length ≤ 1
        return A

    pivot = random element from A
    L = [ ], E = [ ], G = [ ]
    for each x in A:
        if x < pivot:
            L.append(x)
        else if x > pivot:
            G.append(x)
        else E.append(x)
    return quick_sort(L) + E + quick_sort(G)
```

Worst-Case Running Time

- ▶ Worst-case for Quicksort
 - ▶ when pivot is (unique) min or max element
 - ▶ Either **L** or **G** has size **n-1** and other has size 0
 - ▶ Runtime is proportional to
 - ▶ $n + (n-1) + (n-2) + \dots + 2 + 1$
 - ▶ Which is $O(n^2)$

depth	time
0	n
1	$n-1$
2	$n-2$
\vdots	\vdots
$n-1$	1



Expected Runtime of Quicksort

- ▶ Assume there are no duplicates
 - ▶ if there are then we have even less recursive calls
- ▶ At each level of recursion, Quicksort can make n different & unique recursive calls depending on the chosen split/pivot
 - ▶ $|L|=0$ and $|G|=n-1$
 - ▶ $|L|=1$ and $|G|=n-2$
 - ▶ ...
 - ▶ $|L|=n$ and $|G|=0$
- ▶ Since there are n possible splits...
- ▶ ...and since the split is chosen uniformly at random...
- ▶ ...each split is chosen with probability $1/n$

Expected Runtime of Quicksort

- ▶ Each split is chosen with probability $1/n$
- ▶ So expected running time is

prob of first split cost of first split prob of last split cost of last split

$$\begin{aligned} T(n) &= n + \frac{1}{n} \cdot \left(T(0) + T(n-1) \right) + \cdots + \frac{1}{n} \cdot \left(T(n-1) + T(n-1-(n-1)) \right) \\ &= n + \frac{1}{n} \cdot \sum_{i=0}^{n-1} \left(T(i) + T(n-1-i) \right) \end{aligned}$$

- ▶ Solution is $T(n) = 2n \ln n = 1.39 \cdot n \log_2 n = O(n \log n)$

Quicksort Pseudo-Code

```
function quick_sort(A):
    if A.length ≤ 1
        return A

    pivot = random element from A
    L = [ ], E = [ ], G = [ ]
    for each x in A:
        if x < pivot:
            L.append(x)
        else if x > pivot:
            G.append(x)
        else E.append(x)
    return quick_sort(L) + E + quick_sort(G)
```

Not in place!

In-Place Quicksort

```
function partition(A, low, high):  
    pivotIndex = random index between low and high  
    pivotValue = A[pivotIndex]  
    swap A[pivotIndex] and A[high] # move pivot to end  
    currIndex = low  
    for i from low to high - 1:  
        if A[i] <= pivotValue :  
            swap A[i] and A[currIndex]  
            currIndex++  
    swap A[currIndex] and A[high] # move the pivot back  
    return currIndex
```

In-Place Quicksort

```
function quicksort(A, low, high):  
    if low < high:  
        pivotIndex = partition(A, low, high)  
        quicksort(A, low, pivotIndex - 1)  
        quicksort(A, pivotIndex + 1, high)
```

Merge Sort vs. Quicksort

- ▶ Merge sort is worst-case $O(n \log n)$
- ▶ Quicksort is expected $O(n \log n)$
- ▶ Which is better?
- ▶ In practice quicksort is faster!
 - ▶ it also uses less space
 - ▶ constants are better

Outline

- ▶ Motivation
- ▶ Quadratic Sorting
 - ▶ Selection sort
 - ▶ Insertion sort
- ▶ Linearithmic Sorting
 - ▶ Merge Sort
 - ▶ Master Theorem
 - ▶ Quick Sort
- ▶ **Comparative sorting lower bound**
- ▶ Linear Sorting
 - ▶ Radix Sort



How Fast Can We Sort?

- ▶ Merge sort and Quicksort are $O(n \log n)$
- ▶ Can we do better?
 - ▶ No!
 - ▶ Well kind of...

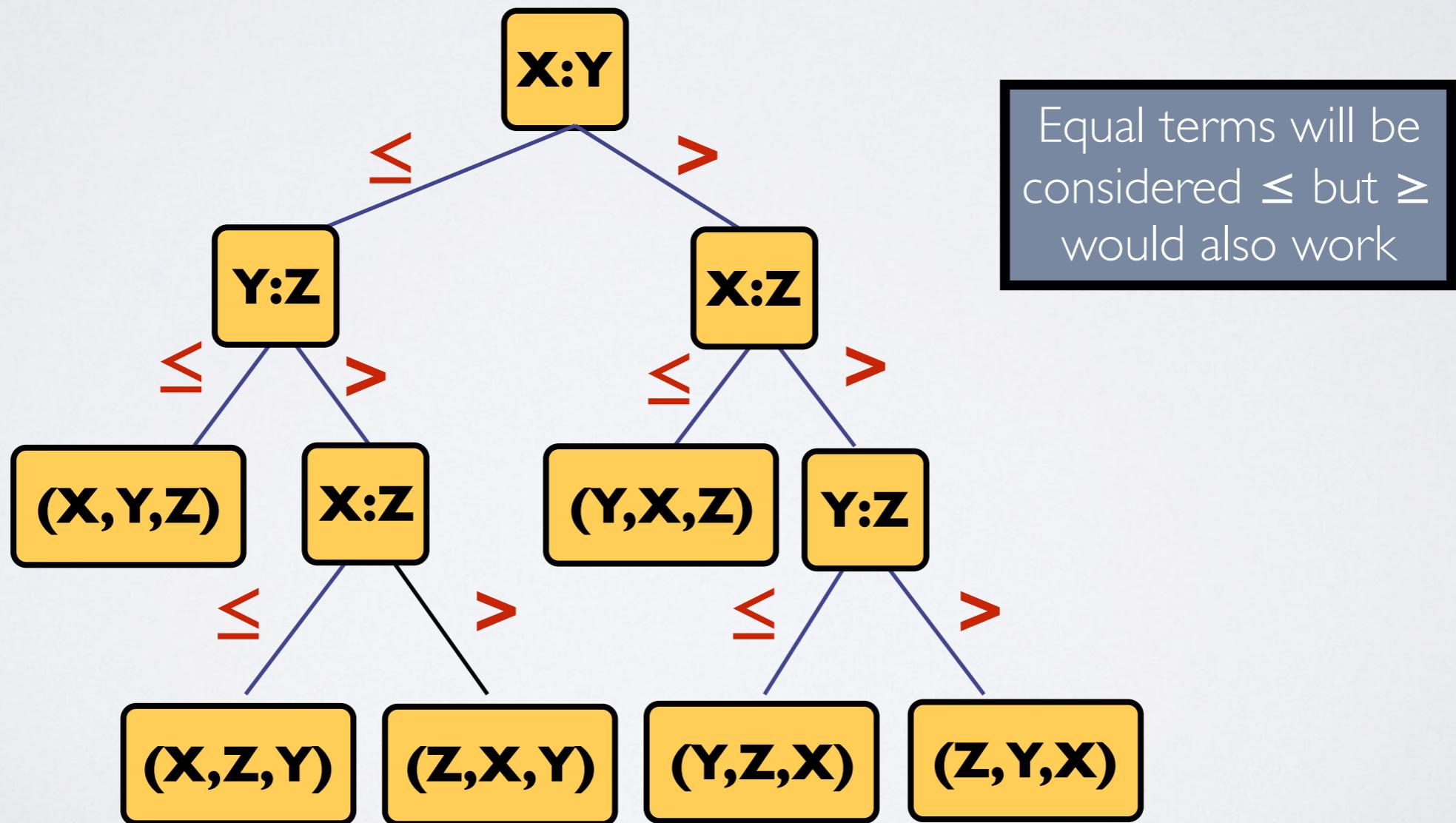
Any comparison-based sorting algorithm has to make at least $\Omega(n \log n)$ comparisons in the worst-case to sort n keys

Lower Bound on Comparative Sorting

- ▶ Viewed abstractly, a sorting algorithm
 - ▶ takes a sequence of keys k_1, \dots, k_n
 - ▶ outputs a permutation of the keys that has them in order
- ▶ We can view the optimal (i.e., best possible) algorithm as a perfect binary **decision** tree
 - ▶ internal nodes do comparisons of keys
 - ▶ leaves are the correct permutation
- ▶ To sort a sequence, we traverse tree
- ▶ Worst-case number of comparisons is height of tree

Lower Bound on Comparative Sorting

- ▶ Suppose our input is X, Y, Z...
- ▶ ...and the proper order is Z, X, Y



Lower Bound on Comparative Sorting

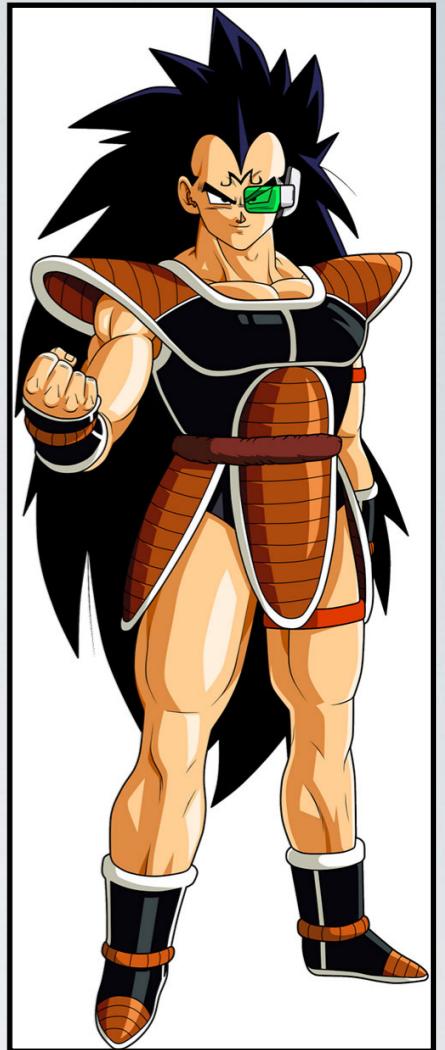
- ▶ How many leaves does tree have?
 - ▶ $n!$ because there are $n!$ permutations of a sequence of n elements
 - ▶ A perfect binary tree with L leaves has height $\log L$
 - ▶ So tree with $n!$ leaves has height $\log(n!)$
 - ▶ Based on Stirling's formula: $n! > \left(\frac{n}{e}\right)^n$
$$\log(n!) > \log\left(\left(\frac{n}{e}\right)^n\right)$$
$$\log(n!) > n \log n - n \log e$$
- ▶ So height of tree (and # of comparisons) is $\Omega(n \log n)$

Non-Comparative Sorting

- ▶ Sorting functions are used on different types of inputs
 - ▶ Integers, floats, strings, arrays, other objects...
 - ▶ As long as we can compare the inputs we can use comparative sorting algorithms
- ▶ But for certain kinds of inputs, we can sometimes do better
 - ▶ example: for positive integers we can use Radix sort

Radix Sort

- ▶ How would you sort 258391 and 258492?
 - ▶ digit by digit
 - ▶ the 3 high order digits are same...
 - ▶ ...so you keep going until you see $3 < 4$ so 258391 must less than 258492



Radix Sort

- ▶ How would you sort an array of numbers between 0 and 9?
- ▶ example: [5 , 1 , 6 , 2 , 3 , 1] → [1 , 1 , 2 , 3 , 5 , 6]
- ▶ Create array of 10 buckets
- ▶ for each number x , add it to bucket at index x
- ▶ Return concatenation of all buckets (in order)
 - ▶ print out [1 , 1]+[2]+[3]+[5]+[6]
- ▶ Runtime is $O(n)$

Radix Sort

- ▶ Radix sort combines both approaches
 - ▶ iterate from least significant to most significant digit
 - ▶ sort number by digit
- ▶ Takes advantage of
 - ▶ the “**digit-iness**” of integers
 - ▶ for every digit there are $O(1)$ number of options



Radix Sort

- ▶ Sort [273, 279, 8271, 7891, 8736, 8735]
- ▶ Start with lowest-order digit (the 1's place)
 - ▶ add number to bucket corresponding to that digit

0	1	2	3	4	5	6	7	8	9
	8271		273		8735	8736			279
	7891								

- ▶ Concatenate all buckets
 - ▶ [8271, 7891, 273, 8735, 8736, 279]
 - ▶ Now sorted by lowest-order digit

Radix Sort

- ▶ Sort [8271, 7891, 273, 8735, 8736, 279]
- ▶ Start with second lowest-order digit (the 10's place)
 - ▶ add number to bucket corresponding to that digit

0	1	2	3	4	5	6	7	8	9
			8735				8271		7891
			8736				273		
							279		

- ▶ Concatenate all buckets
 - ▶ [8735, 8736, 8271, 273, 279, 7891]
- ▶ Now sorted by second and lowest-order digit

Radix Sort

- ▶ Sort [8735, 8736, 8271, 273, 279, 7891]
- ▶ Start with third lowest-order digit (the 100's place)
 - ▶ add number to bucket corresponding to that digit

0	1	2	3	4	5	6	7	8	9
		8271					8735	7891	
		273					8736		
		279							

- ▶ Concatenate all buckets
 - ▶ [8271, 273, 279, 8735, 8736, 7891]
- ▶ Now sorted by third, second and lowest-order digit

Radix Sort

- ▶ Sort [8271, 273, 279, 8735, 8736, 7891]
- ▶ Start with third lowest-order digit (the 1000's place)
 - ▶ add number to bucket corresponding to that digit

0	1	2	3	4	5	6	7	8	9
273							7891	8271	
279								8735	
								8736	

- ▶ Concatenate all buckets
 - ▶ [273, 279, 7891, 8271, 8735, 8736]
- ▶ Now sorted by third, second and lowest-order digit

Radix Sort

```
function radix_sort(A):
    buckets = array of 10 lists
    for place = least to most significant
        for number in A
            d = digit in number at place
            buckets[d].append(number)
    A = concatenate all buckets in order
    empty all buckets
return A
```

- ▶ Very efficient!
 - ▶ $O(nd)$
 - ▶ d is number of digits in the largest number

More on Radix Sort

- ▶ Can be applied to
 - ▶ positive integers in base 10 (we just saw this)
 - ▶ Octals (base 8)
 - ▶ Hexadecimals (base 16)
 - ▶ Strings (one bucket for every valid character)
- ▶ Number of buckets can be different at each round
- ▶ Can represent almost anything as a bit string and radix sort with two buckets
 - ▶ number of digits will dominate runtime
 - ▶ for long sequences will be very slow

Radix Sort

2 min

Activity #5

Radix Sort

2 min

Activity #5

Radix Sort

1 min

Activity #5

Radix Sort

On min

Activity #5

Summary of Sorting Algorithms

Algorithm	Time	Notes
Selection sort	$O(n^2)$	in-place slow (good for small inputs)
Insertion sort	$O(n^2)$	in-place slow (good for small inputs)
Merge sort	$O(n \log n)$	fast (good for large inputs)
Quick sort	$O(n \log n)$ expected	randomized fastest (good for large inputs)
Radix sort	$O(nd)$	d is number of digits in largest number basically linear when d is small

Readings

- ▶ Dasgupta et al.
- ▶ **Section 2.1:** good intro to divide & conquer
- ▶ **Section 2.2:** review of recurrence rels. & master theorem
- ▶ **Section 2.3:** analysis of merge sort & lower bound on comparative sorting

References

- ▶ Slide #62
 - ▶ The character depicted is Raditz (sometimes called Radix) from the Anime *Dragon Ball Z*. He is the biological brother of Goku and one of the four remaining Universe 7 Saiyans.
- ▶ Slide #64
 - ▶ The RZA is the main producer and leader of the Wu-Tang Clan. He also released albums as his alter ego Bobby Digital