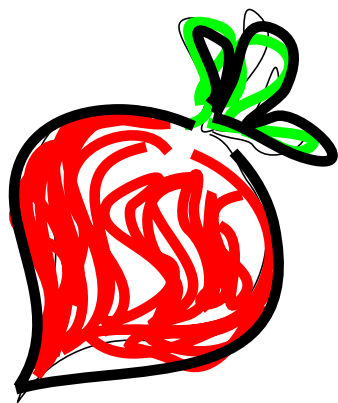


# More Sorting

- radix sort
- bucket sort
- in-place sorting
- how fast can we sort?



# Radix Sort

- Unlike other sorting methods, radix sort considers the structure of the keys
- Assume keys are represented in a base M number system ( $M = \text{radix}$ ), i.e., if  $M = 2$ , the keys are represented in binary

$$9 = \begin{array}{|c|c|c|c|} \hline 1 & 0 & 0 & 1 \\ \hline \end{array}$$

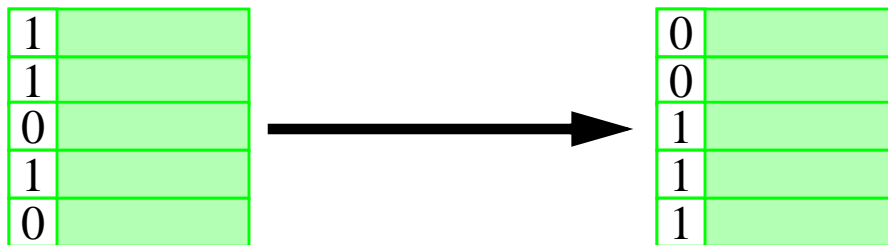
8    4    2    1                      weight  
( $b = 4$ )  
3    2    1    0                      bit #

- Sorting is done by comparing bits in the same position
- Extension to keys that are alphanumeric strings

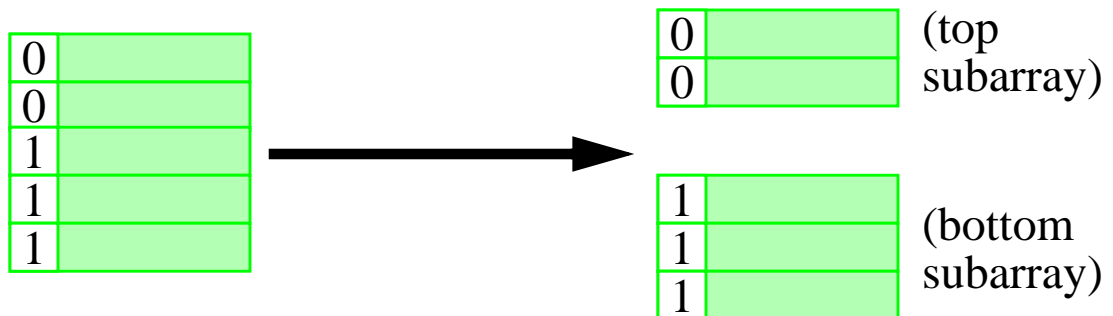
# Radix Exchange Sort

Examine bits from *left* to *right*:

## 1. Sort array with respect to leftmost bit



## 2. Partition array



## 3. Recursion

- recursively sort top subarray, ignoring leftmost bit
- recursively sort bottom subarray, ignoring leftmost bit

Time to sort  $n$   $b$ -bit numbers:  $O(b n)$

# Radix Exchange Sort

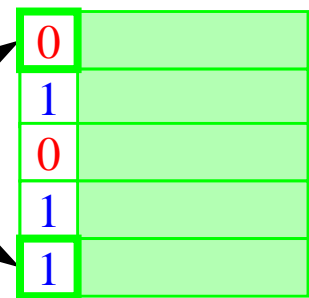
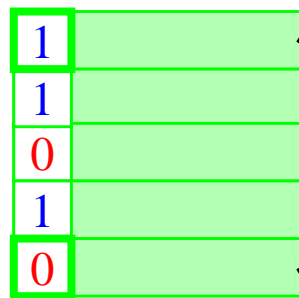
How do we do the sort from the previous page?  
Same idea as partition in [Quicksort](#).

**repeat**

scan top-down to find key starting with 1;  
scan bottom-up to find key starting with 0;  
exchange keys;

**until** scan indices cross;

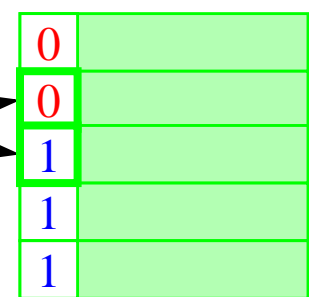
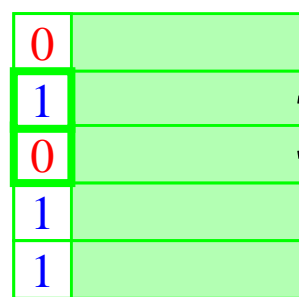
scan from top



first  
exchange

scan from bottom

scan from top

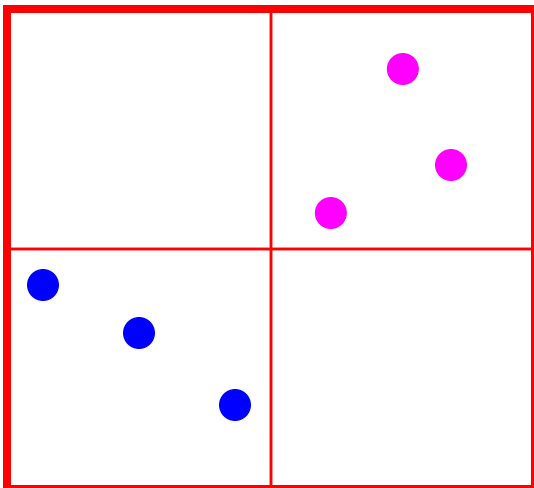
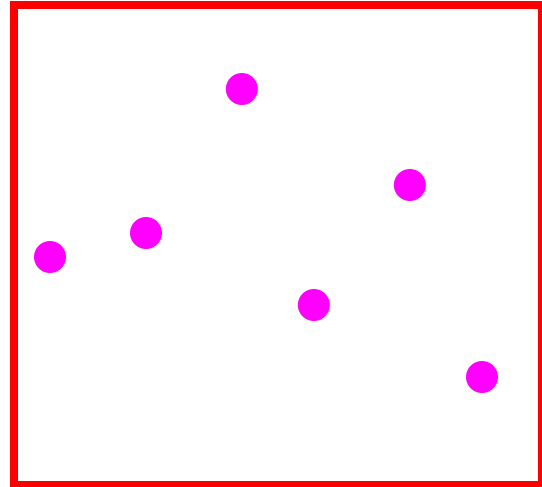


second  
exchange

scan from bottom

# Radix Exchange Sort

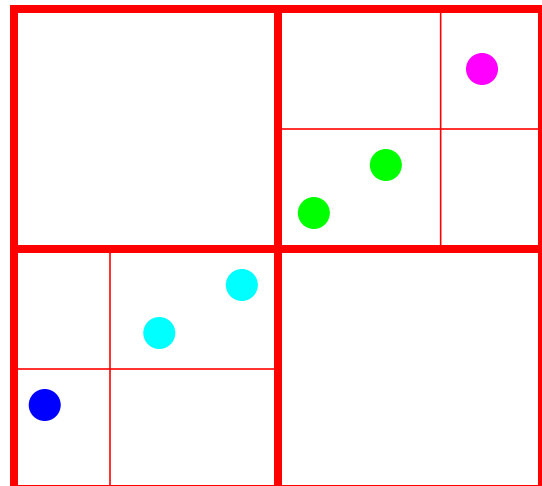
array before sort



$2^{b-1}$

array after sort  
on leftmost bit

array after recursive  
sort on second from  
leftmost bit



# Radix Exchange Sort vs. Quicksort

## Similarities

- both partition array
- both recursively sort sub-arrays

## Differences

- *Method of partitioning*
  - radix exchange divides array based on greater than or less than  $2^{b-1}$
  - quicksort partitions based on greater than or less than some element of the array
- *Time complexity*
  - Radix exchange  $O(bn)$
  - Quicksort average case  $O(n \log n)$

# Straight Radix Sort

Examines bits from *right* to *left*

**for**  $k := 0$  **to**  $b-1$

sort the array in a *stable* way,

looking only at bit  $k$

First,  
sort  
these

Next, sort  
these digits

Last, sort  
these.

0	1	0
0	0	0
1	0	1
0	0	1
1	1	1
0	1	1
1	0	0
1	1	0

0	1	0
0	0	0
1	0	0
1	1	0
1	0	1
0	0	1
1	1	1
0	1	1

0	0	0
1	0	0
1	0	1
0	0	1
0	1	0
1	1	0
1	1	1
0	1	1

0	0	0
0	0	1
0	1	0
0	1	1
1	0	0
1	0	1
1	1	0
1	1	1


Note order of these bits after sort.

# I forgot what it means to “sort in a stable way”!!!

In a stable sort, the initial relative order of equal keys is unchanged.

For example, observe the first step of the sort from the previous page:

0	1	0
0	0	0
1	0	1
0	0	1
1	1	1
0	1	1
1	0	0
1	1	0



0	1	0
0	0	0
1	0	0
1	1	0
1	0	1
0	0	1
1	1	1
0	1	1

Note that the relative order of those keys ending with 0 is unchanged, and the same is true for elements ending in 1



# The Algorithm is Correct (right?)

- We show that any two keys are in the correct relative order at the end of the algorithm
- Given two keys, let  $k$  be the leftmost bit-position where they differ

0	1	0	1	1
---	---	---	---	---

0	1	1	0	1
---	---	---	---	---

$k$

- At step  $k$  the two keys are put in the correct relative order
- Because of *stability*, the successive steps do not change the relative order of the two keys

# For Instance,

Consider a sort on an array with these two keys:

0	1	0	1	1
---	---	---	---	---

0	1	1	0	1
---	---	---	---	---

$k$

0	1	1	0	1
0	1	0	1	1

It makes no difference what order they are in when the sort begins.

0	1	0	1	1
0	1	1	0	1

When the sort visits bit  $k$ , the keys are put in the correct relative order.

0	1	0	1	1
0	1	1	0	1

Because the sort is stable, the order of the two keys will not be changed when bits  $> k$  are compared.

# Radix sorting can be applied to decimal numbers

First, sort these digits

Next, sort these digits

Last, sort these.

0	3	2
2	2	4
0	1	6
0	1	5
0	3	1
1	6	9
1	2	3
2	5	2

0	3	1
0	3	2
2	5	2
1	2	3
2	2	4
0	1	5
0	1	6
1	6	9

0	1	5
0	1	6
1	2	3
2	2	4
0	3	1
0	3	2
2	5	2
1	6	9

0	1	5
0	1	6
0	3	1
0	3	2
1	2	3
1	6	9
2	2	4
2	5	2

Note order of these bits after sort.

Voila!

# Straight Radix Sort

## Time Complexity

**for**  $k = 0$  **to**  $b - 1$

sort the array in a *stable* way,  
looking only at bit  $k$

Suppose we can perform the stable sort above in  $O(n)$  time. The total time complexity would be

$$O(bn)$$

As you might have guessed, we can perform a stable sort based on the keys'  $k^{\text{th}}$  digit in  $O(n)$  time.

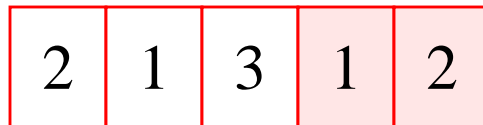
The method, you ask? Why it's **Bucket Sort**, of course.



# Bucket Sort

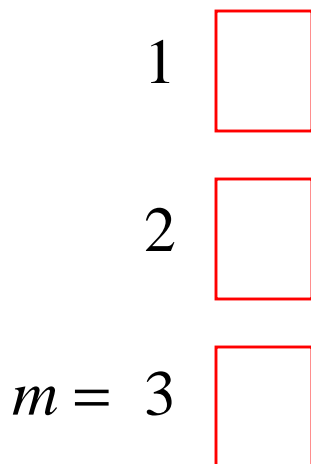
- $n$  numbers
- Each number  $\in \{1, 2, 3, \dots, m\}$
- Stable
- Time:  $O(n + m)$

For example,  $m = 3$  and our array is:



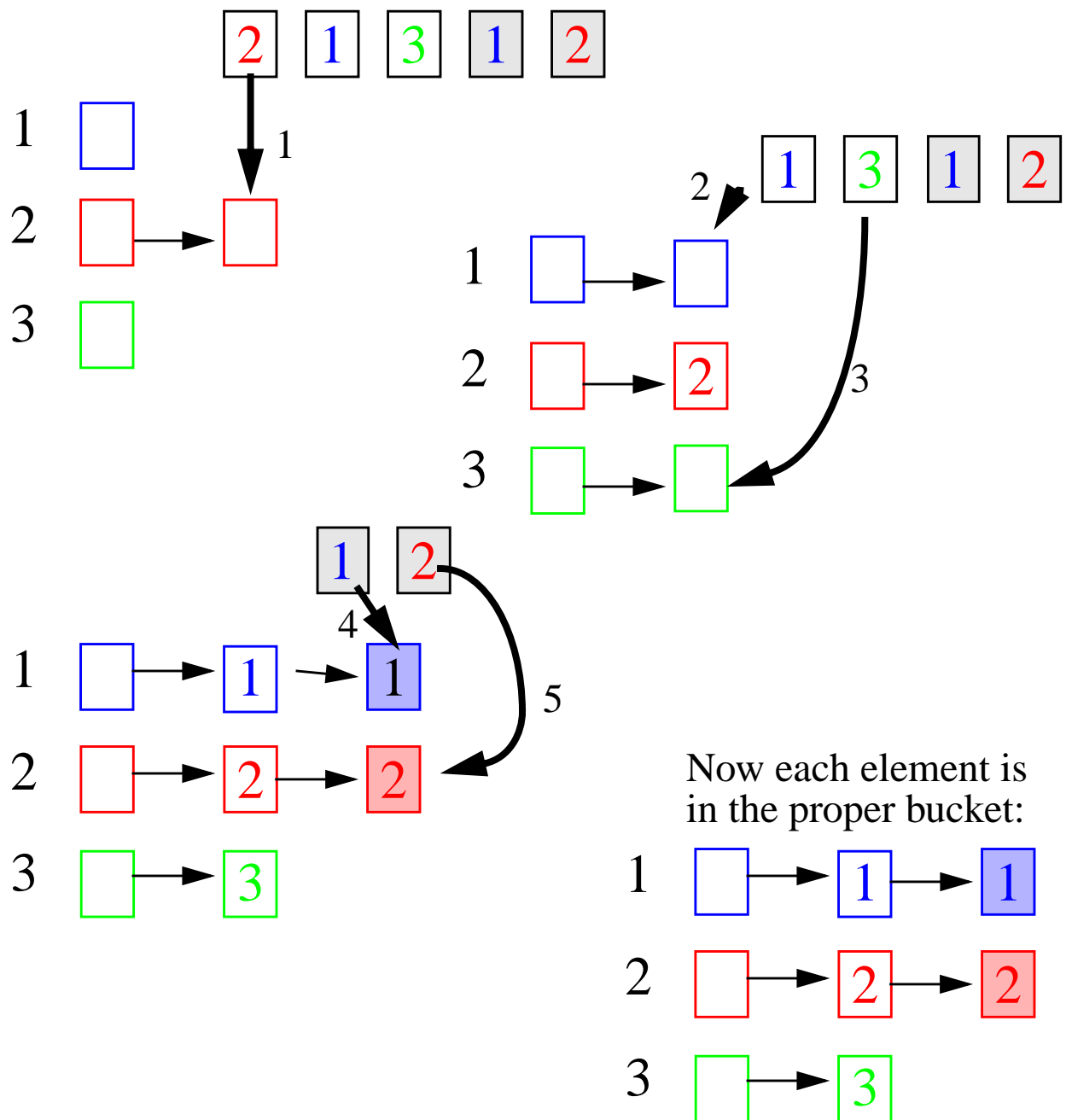
(note that there are two “2”s and two “1”s)

First, we create  $M$  “buckets”



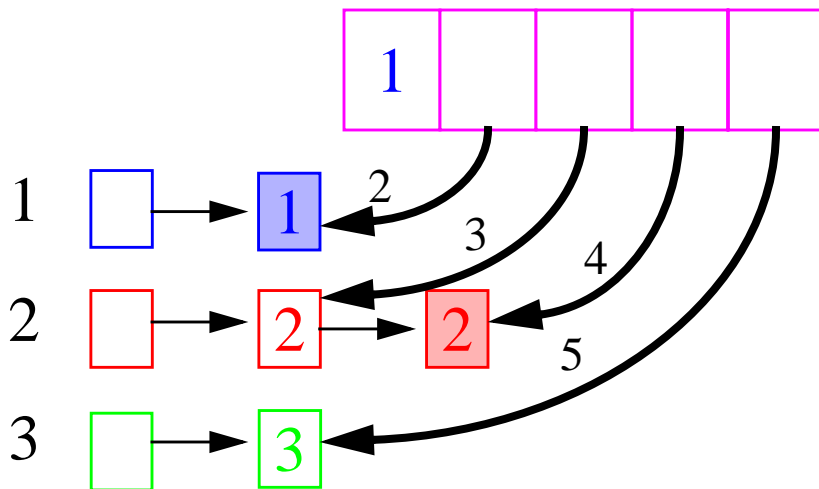
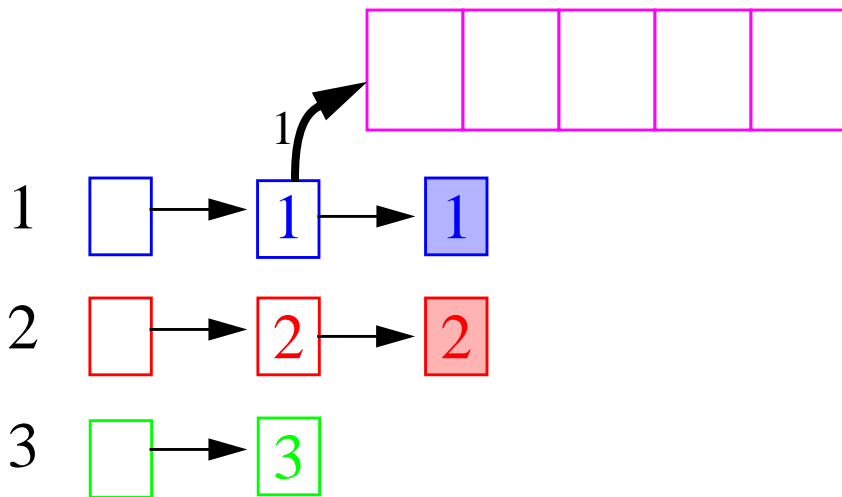
# Bucket Sort

Each element of the array is put in one of the  $m$  “buckets”

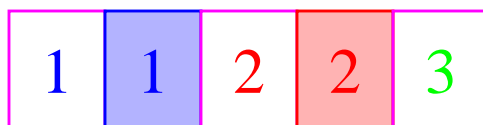


# Bucket Sort

Now, pull the elements from the buckets into the array



At last, the sorted array (sorted in a *stable* way):



# In-Place Sorting

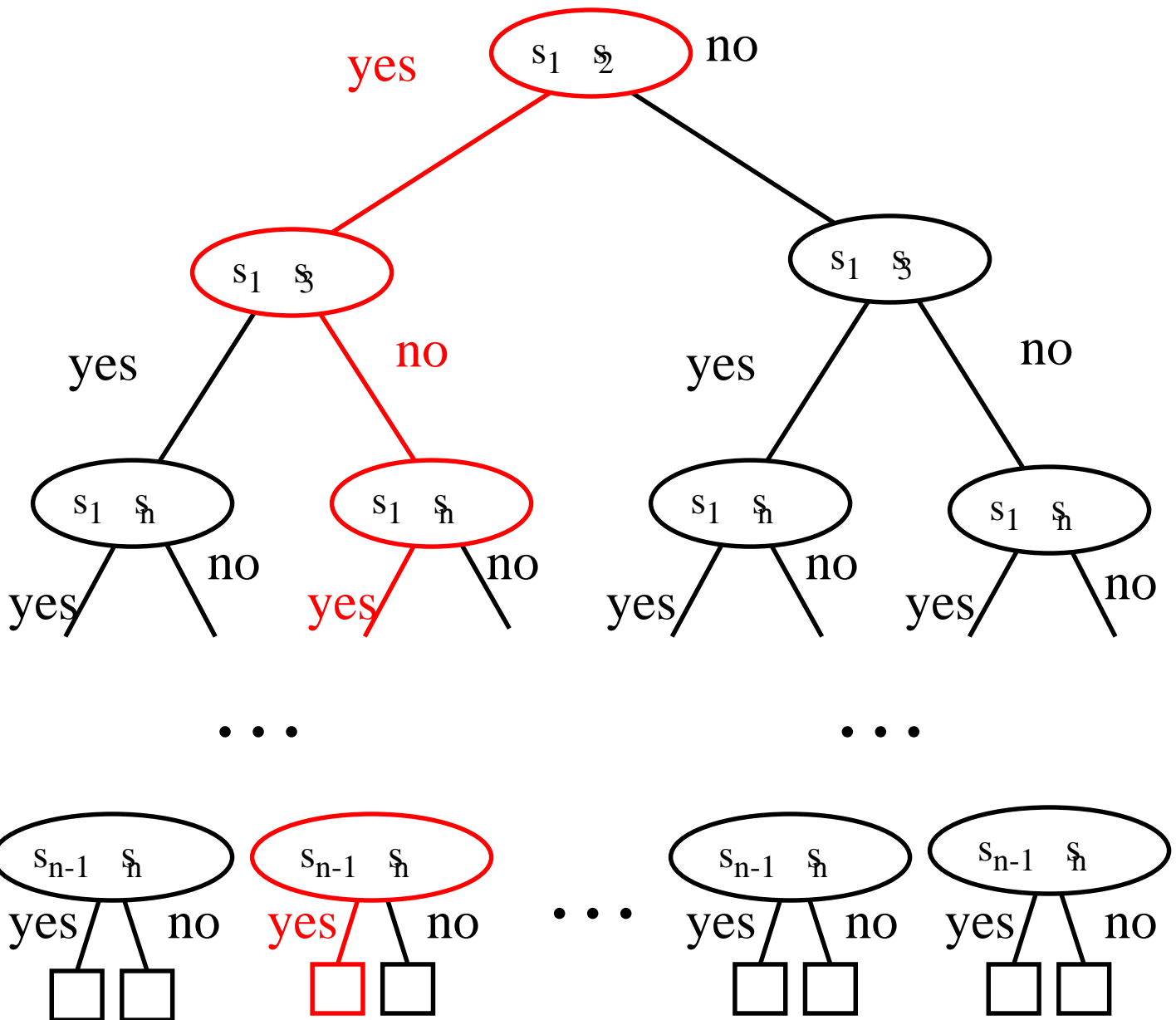
- A sorting algorithm is said to be *in-place* if
  - it uses **no auxiliary data structures** (however,  $O(1)$  auxiliary variables are allowed)
  - it updates the input sequence only by means of operations **replaceElement** and **swapElements**
- Which sorting algorithms seen so far can be made to work in place?

bubble-sort	Y
selection-sort	
insertion-sort	
heap-sort	
merge-sort	
quick-sort	
radix-sort	
bucket-sort	



# Decision Tree for Comparison Based Sorting

- internal node: comparison
- external node: permutation
- algorithm execution: **root-to-leaf path**



# How Fast Can We Sort?

- **Proposition:** The running time of any comparison-based algorithm for sorting an  $n$ -element sequence  $S$  is  $\Omega(n \log n)$ .
- **Justification:**
  - The running time of a comparison-based sorting algorithm must be equal to or greater than the depth of the decision tree  $T$  associated with this algorithm.
  - Each internal node of  $T$  is associated with a comparison that establishes the ordering of two elements of  $S$ .
  - Each external node of  $T$  represents a distinct permutation of the elements of  $S$ .
  - Hence  $T$  must have at least  $n!$  external nodes which again implies  $T$  has a height of at least  $\log(n!)$
  - Since  $n!$  has at least  $n/2$  terms that are greater than or equal to  $n/2$ , we have:
$$\log(n!) \geq (n/2) \log(n/2)$$
- **Total Time Complexity:**  $\Omega(n \log n)$ .