

COMP 2402

# Sorting

part 2

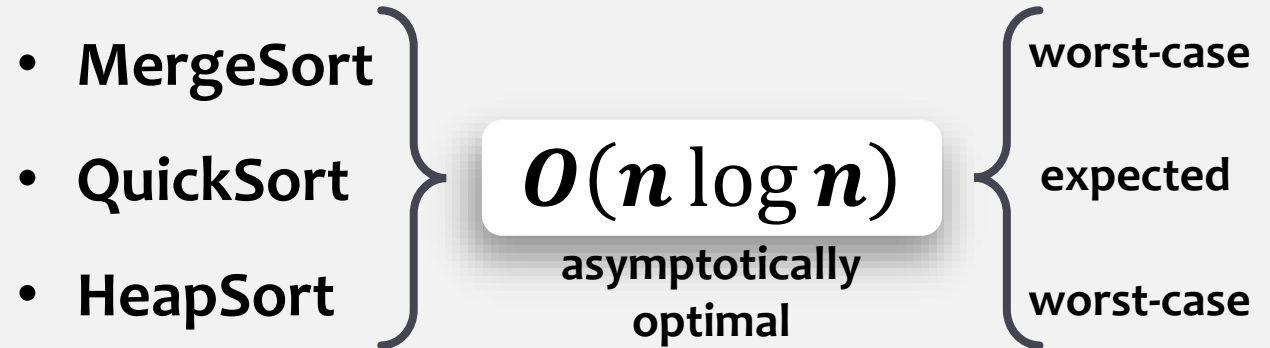


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# Sorting

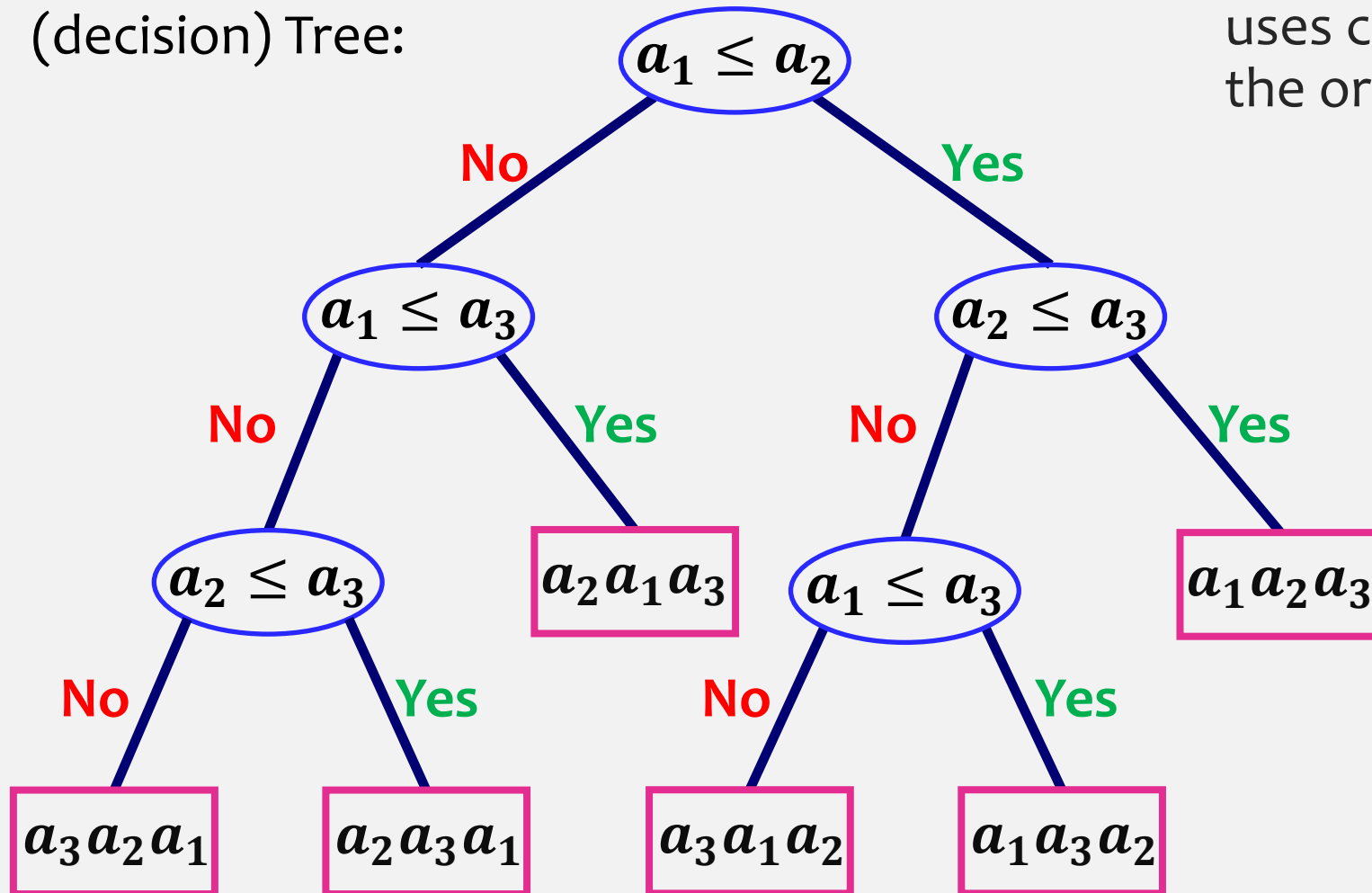
- Comparison-based Sorting:



- Comparison-based algorithms can be used to sort any **array** of **comparable** items.
- Is there a faster (maybe  $O(n)$  time) sorting algorithm (for general elements)?  
**No / Yes**
- All branching in the comparison-based algorithms is based on the results of comparisons of the form  $a[i] < a[j]$
- Every comparison-based sorting algorithm takes  $\Omega(n \log n)$  time for some input.

# Comparison-based Sorting $\Omega(n \log n)$

Comparison  
(decision) Tree:



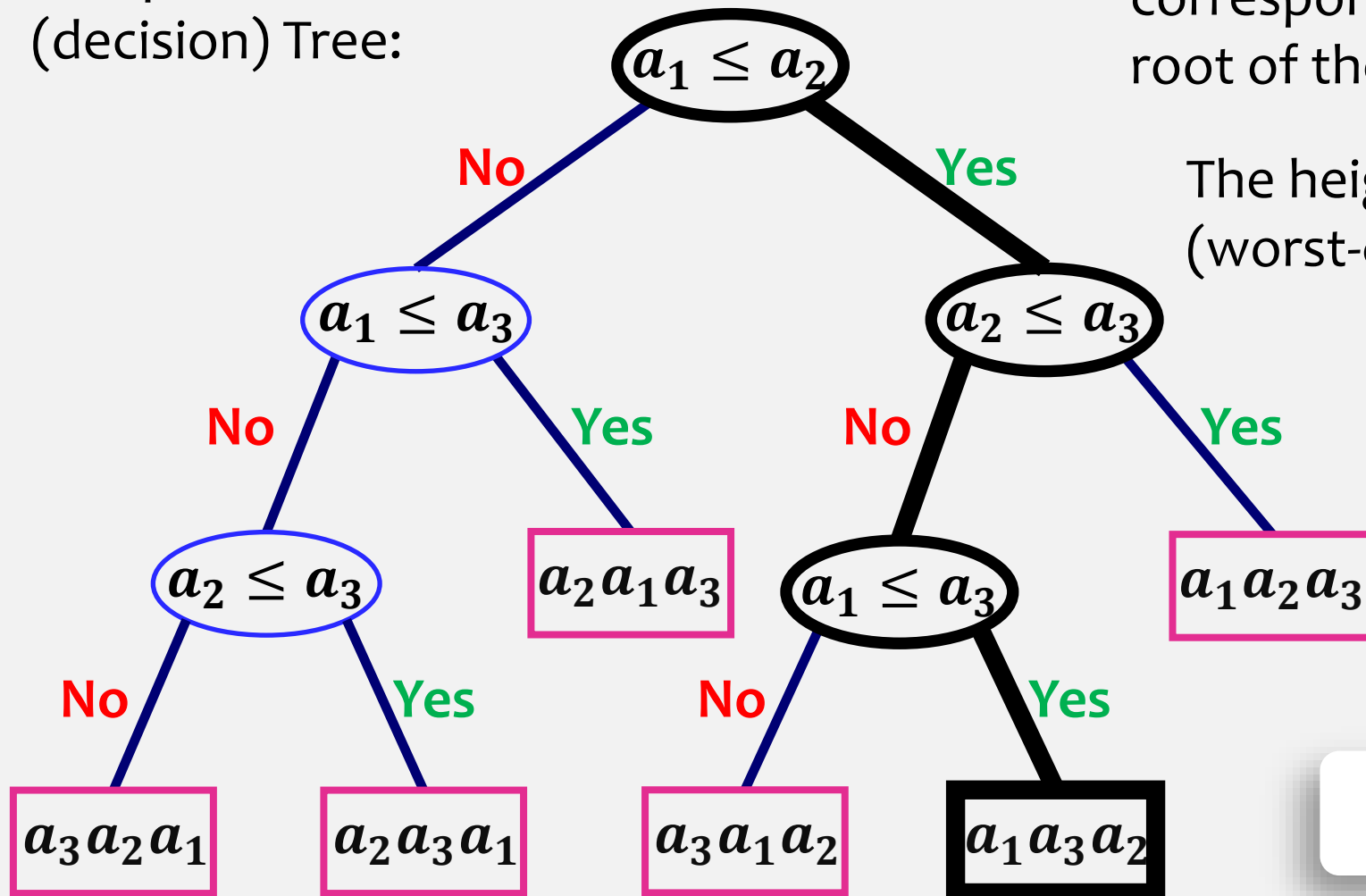
Comparison based sorting algorithm  
uses comparison operators to find  
the order between two numbers.

**Input:**  $a_1, a_2, a_3$

**Output:**  
reordering  $a'_1, a'_2, a'_3$   
of the input such that  
 $a'_1 \leq a'_2 \leq a'_3$

# Comparison-based Sorting $\Omega(n \log n)$

Comparison  
(decision) Tree:



The execution of the sorting algorithm corresponds to **tracing a path** from the root of the decision tree to a leaf.

The height of the tree is equal to the (worst-case) number of comparisons

**Leaf:** sorted sequence.

**Input:**  $a_1, a_2, \dots, a_n$

**# of leaves:**  $n!$

**max height:**  $\log(n!)$

$$\log(n!) \geq cn \log n, c > 0$$

# Comparison-based Sorting $\Omega(n \log n)$

Every comparison tree that sorts any input of length  $n$  has **height** at least  $\frac{n}{2} \log_2 \frac{n}{2}$

- The **height** of a binary tree with  $m$  leaves is at least  $\log_2 m$
- The **height** of a binary tree with  $n!$  leaves is at least  $\log_2(n!)$

$$\begin{aligned}\log_2 n! &= \log_2(n) + \log_2(n-1) + \cdots + \log_2(1) \\ &\geq \log_2(n) + \cdots + \log_2(n/2) \\ &\geq \log_2(n/2) + \cdots + \log_2(n/2) \\ &= (n/2) \log_2(n/2)\end{aligned}$$

- Lower bound can be improved to  $n \ln n - O(n)$ .

$$\log(n!) \geq cn \log n, \quad c > 0$$

# Theorems 11.5 and 11.6

For every deterministic comparison-based sorting algorithm  $A$  and any integer  $n \geq 1$ , there exists an input array  $a$  of length  $n$  such that  $A$  requires  $\Omega(n \log n)$  comparisons to sort  $a$ .

For every comparison-based sorting algorithm  $A$ , the expected number of comparisons done by  $A$  when sorting a random permutation of  $\{1, \dots, n\}$  is  $\Omega(n \log n)$ .

# Summary

- **MergeSort** –  $n \log n$  comparisons
  - **QuickSort** –  $1.38n \log n$  comparisons
  - **HeapSort** –  $2n \log n$  comparisons
- 
- Any comparison-based sorting algorithm requires  $\Omega(n \log n)$  time.
  - **MergeSort**, **QuickSort**, and **HeapSort** are **optimal** comparison-based sorting algorithms.

# NOT comparison-based sorting

***a*:**

3	1	3	1	0	3	1	3	1	0	3
---	---	---	---	---	---	---	---	---	---	---

***c*:**

2	4	0	5
0	1	2	3

count the number of occurrences  
of ***i*** in ***a*** and store this in ***c*[*i*]**

counters

***a*:**

0	0	1	1	1	1	3	3	3	3	3
2		4				0	5			



# NOT comparison-based sorting

- **CountingSort**
- **RadixSort**

← **stable**

Suppose we have an input array  $a$  consisting of  $n$  integers, each in the range  $\{0, \dots, k - 1\}$ .

Specialized for sorting “small” integers.

They use (parts of) the elements in  $a$  as indices into an array.

They can sort faster than comparison-based algorithms – faster than  $\Omega(n \log n)$ .

**CountingSort** is very efficient for sorting an array of integers when  $n \not\ll k - 1$

not much less

**RadixSort** uses several passes of **CountingSort** to allow for a much greater range of maximum values.

length of the array

maximum  
value in the  
array

# CountingSort

```
int[] countingSort(int[] a, int k) {  
    int c[] = new int[k];  
    for (int i = 0; i < a.length; i++)  
        c[a[i]]++;  
    for (int i = 1; i < k; i++)  
        c[i] += c[i-1];  
    int b[] = new int[a.length];  
    for (int i = a.length-1; i >= 0; i--)  
        b[--c[a[i]]] = a[i];  
    return b;  
}
```

<i>a</i> :	3	1	3	1	0	3	1	3	1	0	3
<i>c</i> :	2	4	0	5							
	0	1	2	3							

- Suppose we have an input array  $a$  consisting of  $n$  integers, each in the range  $\{0, \dots, k-1\}$ .
- For each  $i \in \{0, \dots, k-1\}$ , count the number of occurrences of  $i$  in  $a$  and store this in  $c[i]$ .
- Compute a **running-sum** of the counters so that  $c[i]$  becomes the number of elements in  $a$  that are less than or equal to  $i$ .
- Scan  $a$  backwards to place its elements, in order, into an output array  $b$ . When scanning, the element  $a[i]$  is placed at location  $b[c[a[i]] - 1]$  and the value  $c[a[i]]$  is decremented.

$c'$ :

0	1	2	3	4	5
2	7	9	13	13	16

stable

 $c$ :
















0	1	2	3	4	5
2	5	2	4	0	3

there are 16  
items in  $a$

 $b$ :

0	0	1	1	1	1	1	2	2	3	3	3	3	5	5	5
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

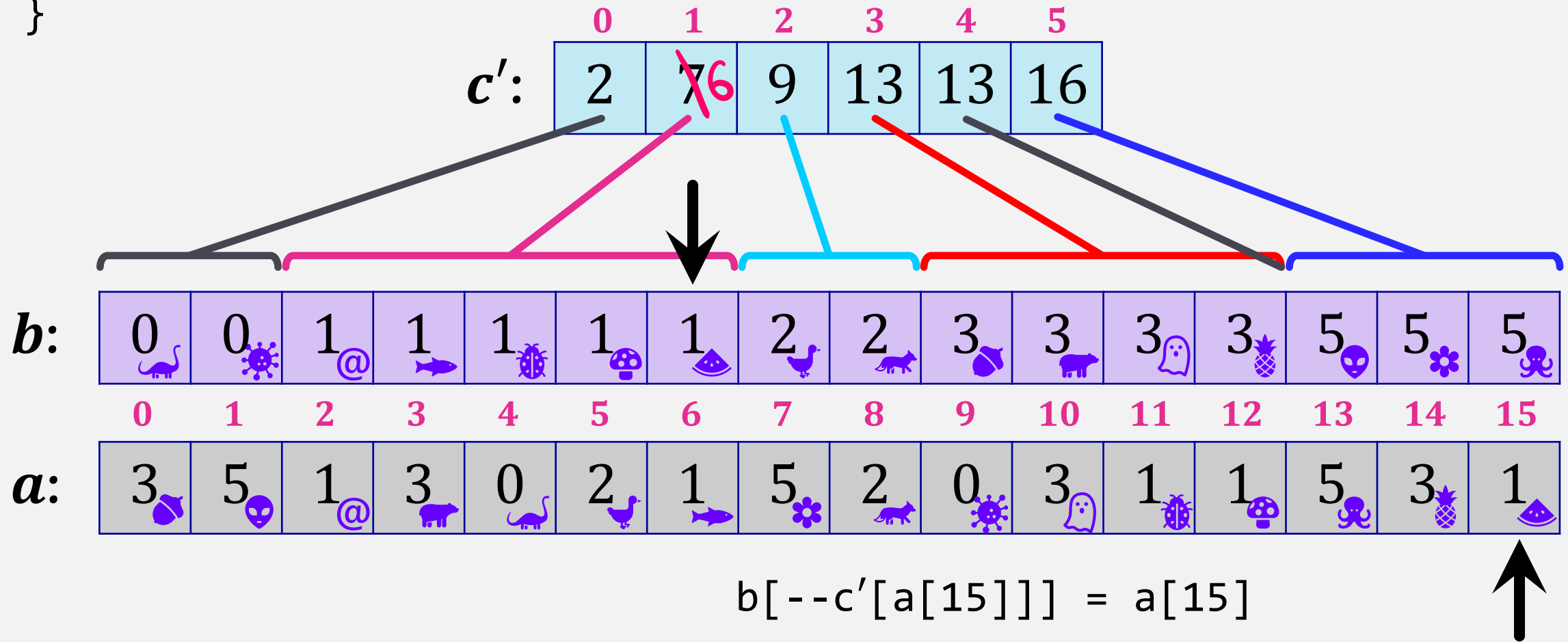
 $a$ :

3	5	1	3	0	2	1	5	2	0	3	1	1	5	3	1
		@													

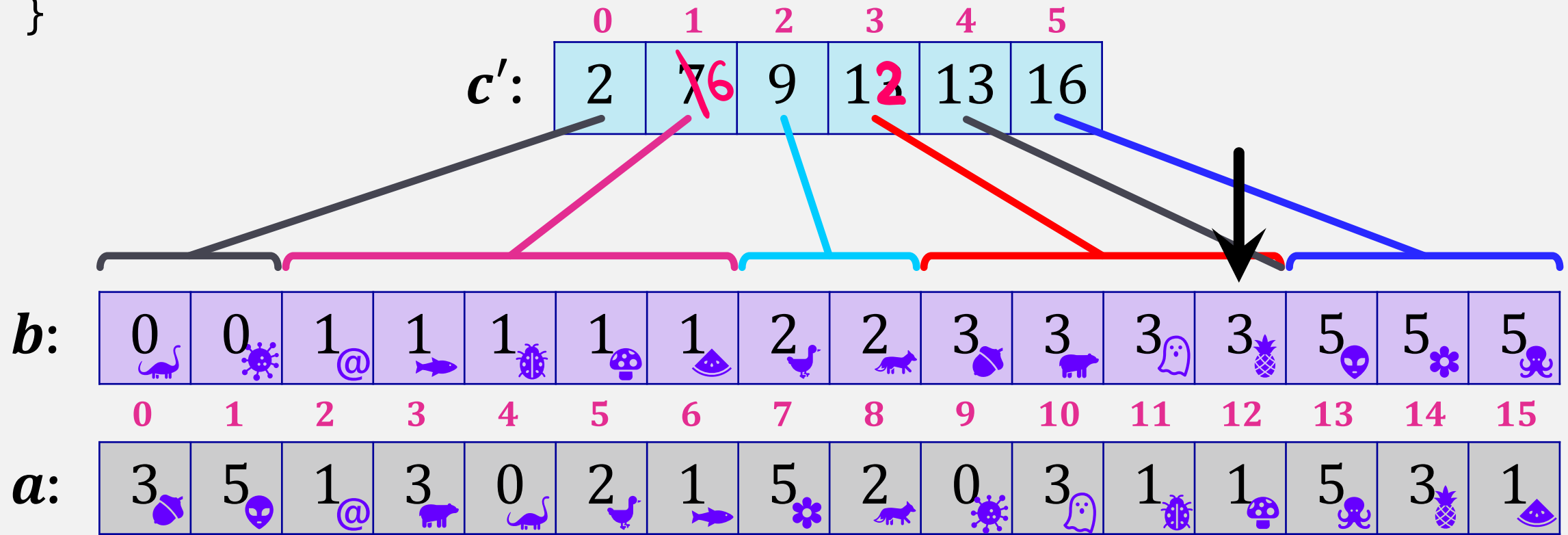
```

for (int i = a.length - 1; i >= 0; i--){
    b[--c'[a[i]]] = a[i];
}

```



```
for (int i = a.length - 1; i >= 0; i--){
    b[--c'[a[i]]] = a[i];
}
```



$b[--c'[a[14]]] = a[14]$

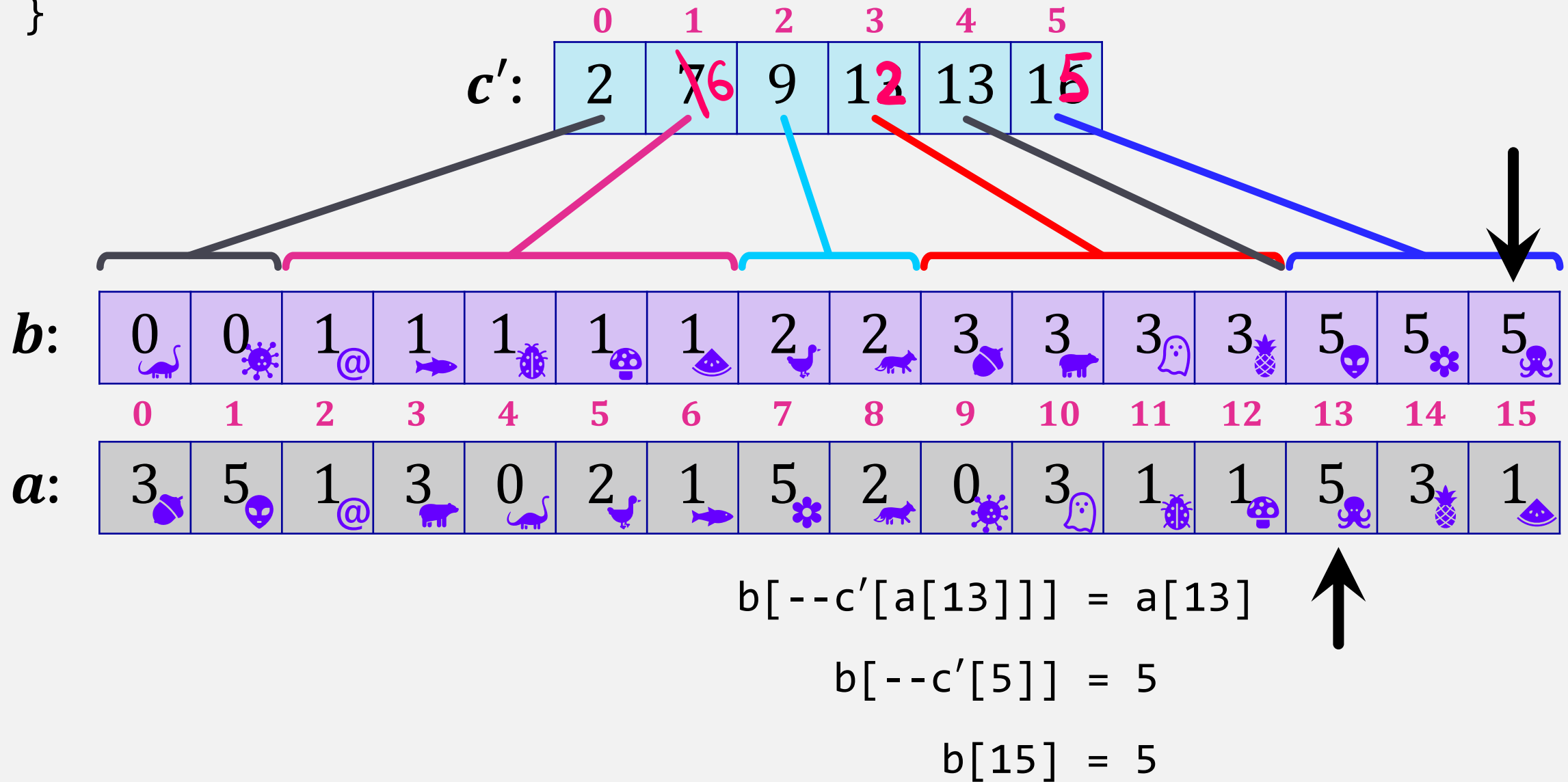
$b[--c'[3]] = 3$

$b[12] = 3$

```

for (int i = a.length - 1; i >= 0; i--){
    b[--c'[a[i]]] = a[i];
}

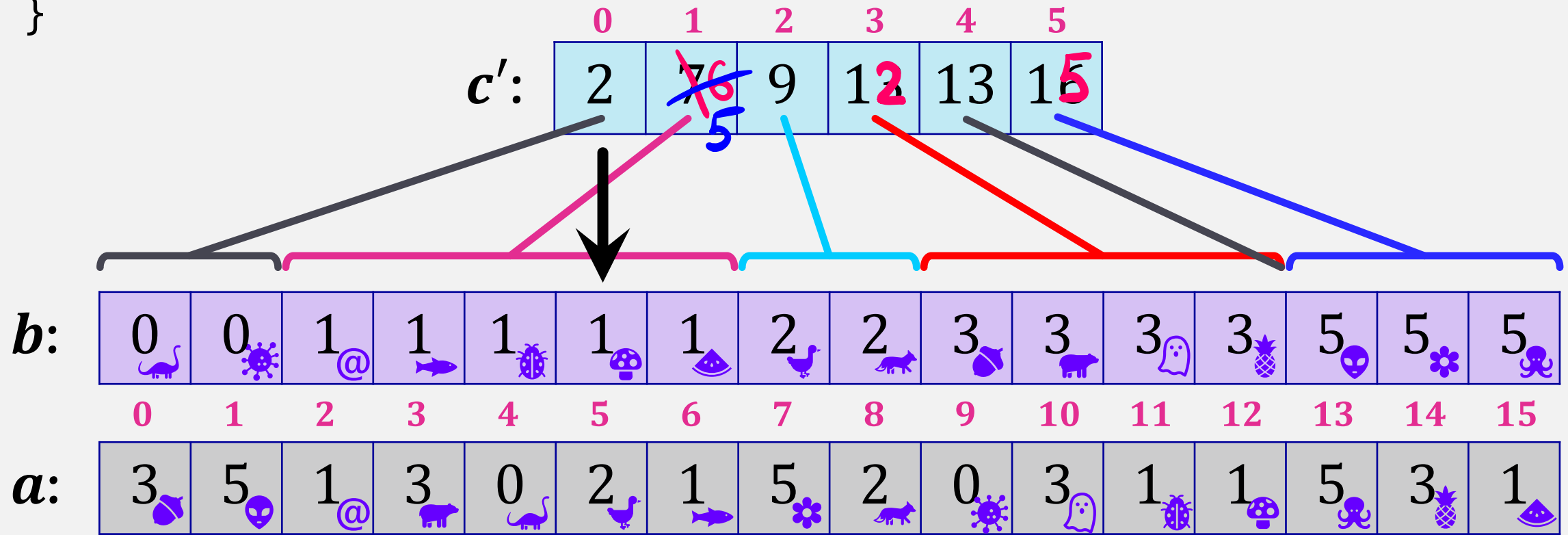
```



```

for (int i = a.length - 1; i >= 0; i--){
    b[--c'[a[i]]] = a[i];
}

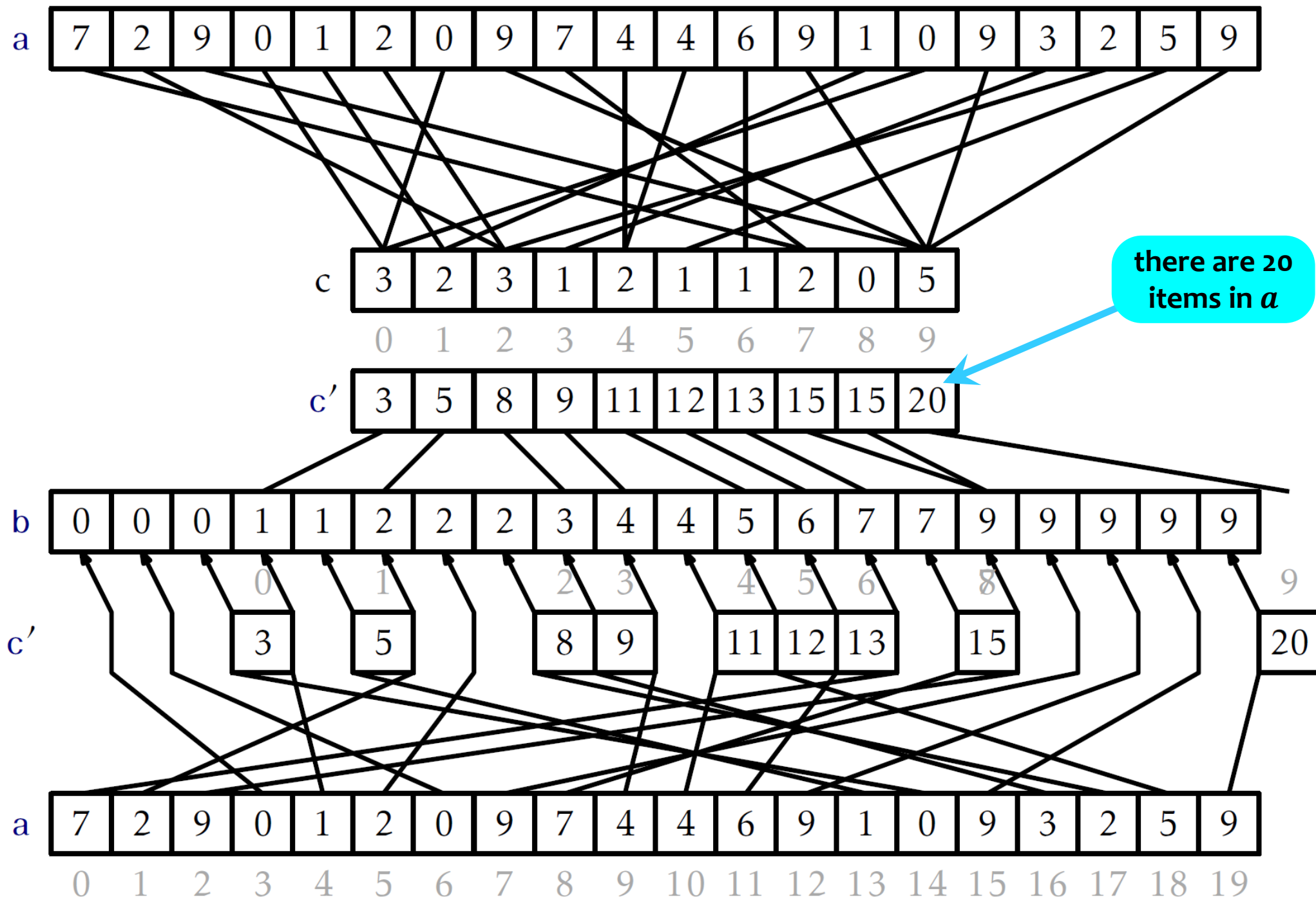
```



$b[--c'[a[12]]] = a[12]$

$b[--c'[1]] = 1$

$b[5] = 1$





# CountingSort

executes  $n$  times

executes  $k$  times

executes  $n$  times

```
int[] countingSort(int[] a, int k) {  
    int c[] = new int[k];  
    for (int i = 0; i < a.length; i++)  
        c[a[i]]++;  
    for (int i = 1; i < k; i++)  
        c[i] += c[i-1];  
    int b[] = new int[a.length];  
    for (int i = a.length-1; i >= 0; i--)  
        b[--c[a[i]]] = a[i];  
    return b;  
}
```

$O(n + k)$

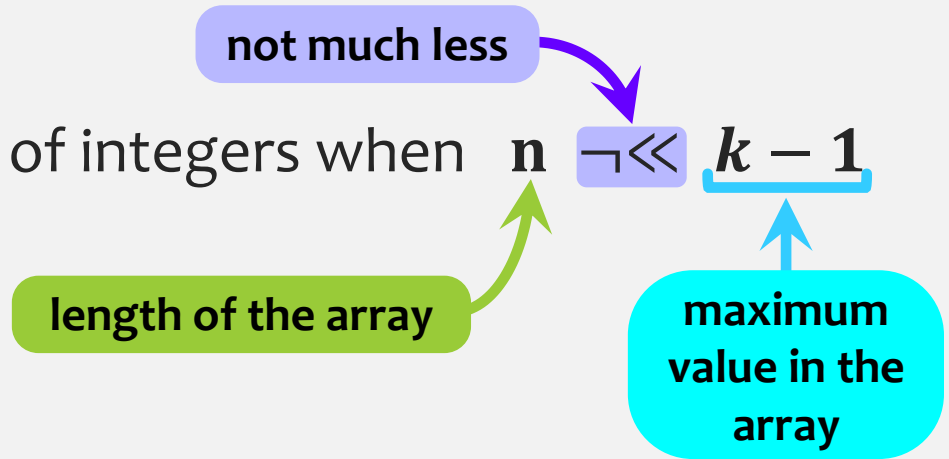
# Theorem 11.7

The **countingSort**( $a, k$ ) method can sort an array  $a$  containing  $n$  integers in the set  $\{0, \dots, k - 1\}$  in  $O(n + k)$  time.

The **CountingSort** algorithm is **stable**; it preserves the relative order of equal elements: If two elements  $a[i]$  and  $a[j]$  have the same value, and  $i < j$  then  $a[i]$  will appear before  $a[j]$  in  $b$ .

# RadixSort

**CountingSort** is very efficient for sorting an array of integers when  $n \ll k$



**RadixSort** uses several passes of **CountingSort** to allow for a much greater range of maximum values.

**RadixSort** sorts integers **one digit at a time**:

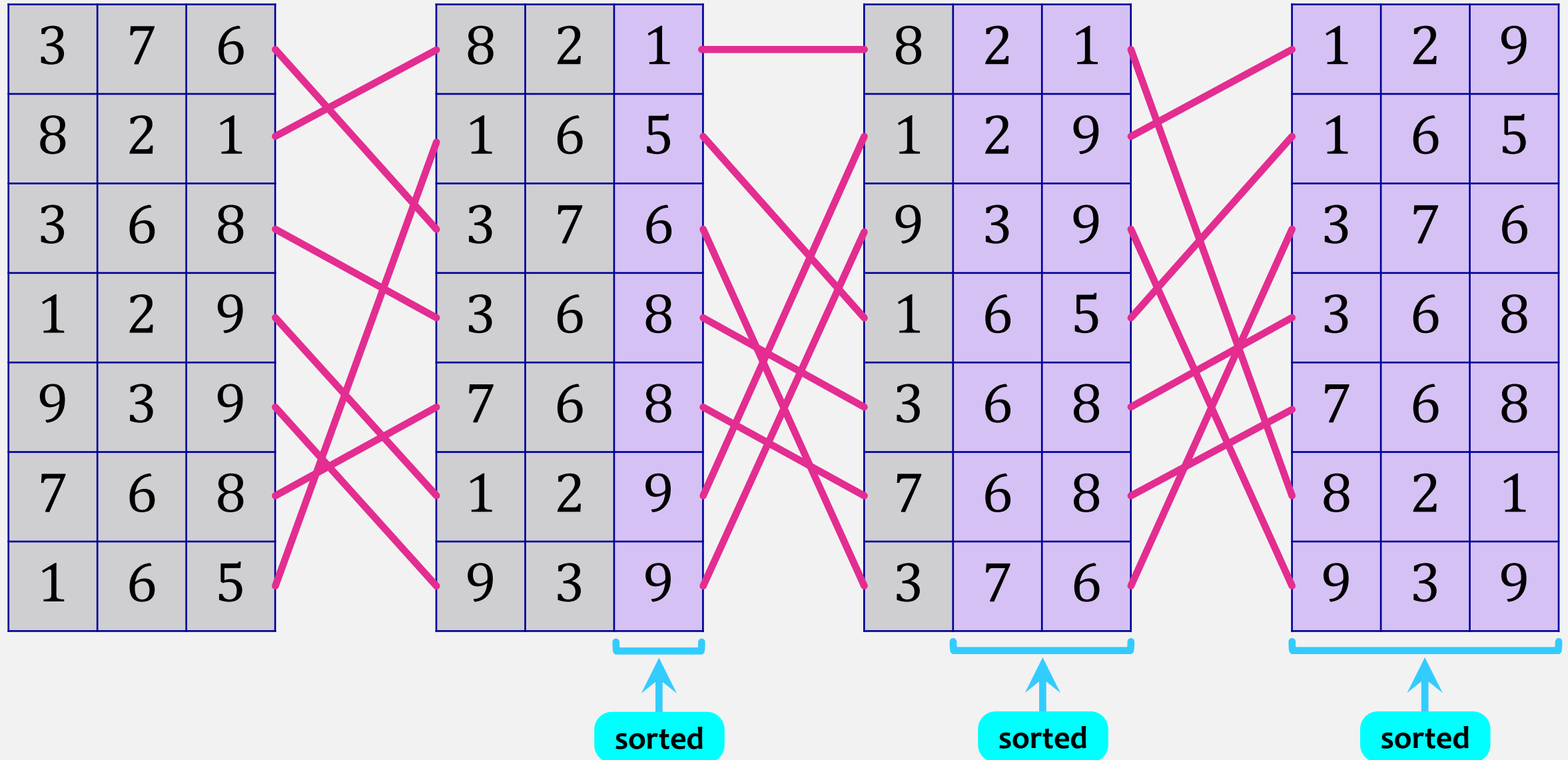
- integers have  $w$  bits
- **digit** has  $d$  bits
- uses  $w/d$  passes of **CountingSort**

We assume that  $d$  divides  $w$ , otherwise we can always increase  $w$ .

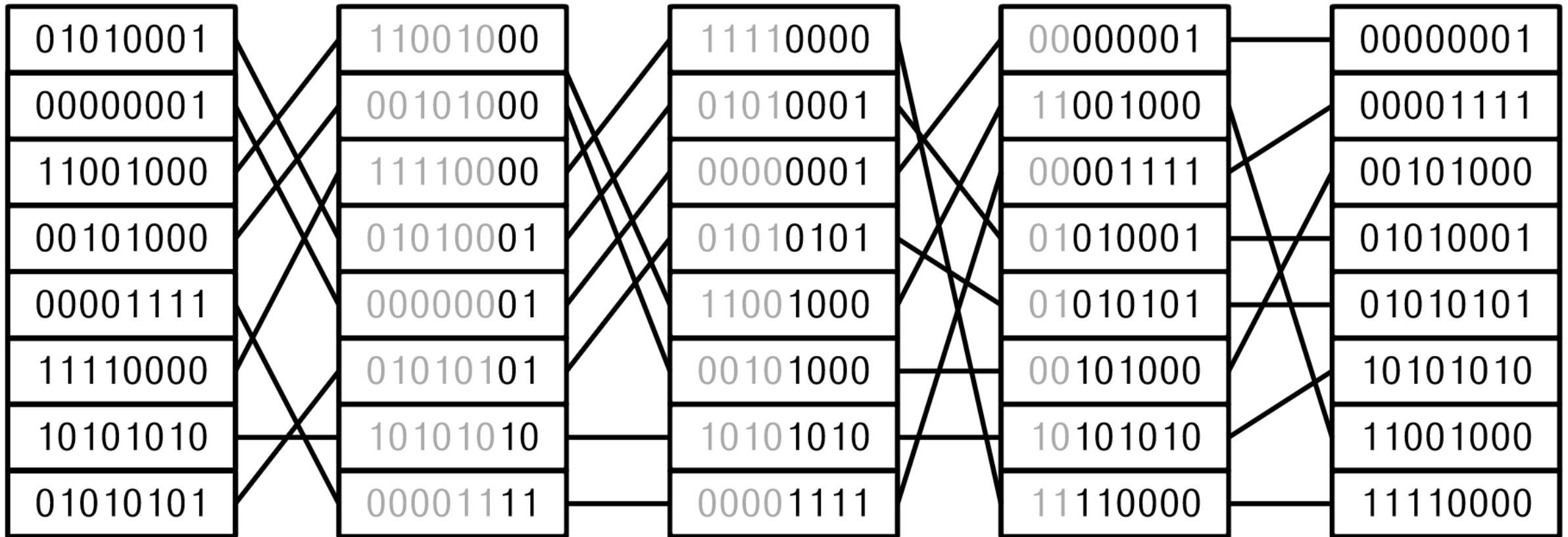
**RadixSort** starts by sorting the integers by their least-significant digit ( $d$  bits), then their next significant digit, and so on until, in the last pass, the integers are sorted by their most significant digit ( $d$  bits).

# RadixSort

The algorithm sorts correctly because  
CountingSort is a **stable** sorting algorithm



# RadixSort



# Theorem 11.8

**RadixSort** performs  $w/d$  passes of **CountingSort**.  
Each pass requires  $O(n + 2^d)$  time.

For any integer  $d > 0$ , the **RadixSort** algorithm can sort an array  $a$  containing  $n$   $w$ -bit integers in  $O\left(\left(\frac{w}{d}\right)(n + 2^d)\right)$  time.

Take  $d = \lceil \log_2 n \rceil$

The **RadixSort** algorithm can sort an array  $a$  containing  $n$  integers in the range  $\{0, \dots, n^c - 1\}$  in  $O(cn)$  time.

# Summary

## MergeSort

## QuickSort

## HeapSort

- can each sort an array of length  $n$  in  $O(n \log n)$  time.
- they work for any comparable data type.
- **QuickSort** and **HeapSort** are in-place but do more comparisons.
- **MergeSort** requires an auxiliary array.

## RadixSort

- can sort an array  $a$  of  $n$  integers in the range  $\{0, \dots, n^c - 1\}$  in  $O(cn)$  time (and does no comparisons).