

# CH08-320201

# Algorithms and Data Structures

## **Lecture 11/12 — 13 Mar 2018**

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Spring 2018

# This & that

- Don't forget: midterm next week

# Last time

- Comparison sorting
  - The lower bound for comparison sorting  $\Omega(n \lg n)$ .
- Is it possible to avoid comparisons between elements?
- Yes, if we can make assumptions on the input data.
- E.g., trivial case
  - Input:  $A[1 \dots n]$ , where  $A[j] \in \{1, 2, \dots, n\}$ , and  $a_i \neq a_j$  for all  $i \neq j$
  - Output:  $B[1 \dots n]$

## 2.4 Counting Sort

# Problem statement

Input: **A[1..n]**, where  $A[j] \in \{1, 2, \dots, k\}$ .

Output: **B[1..n]**,  
which is a sorted version of A [1..n].

Auxiliary storage: **C[1..k]**.

# Counting Sort

```
for  $i := 1$  to  $k$ 
    do  $C[i] := 0$ 
for  $j := 1$  to  $n$ 
    do  $C[A[j]] := C[A[j]] + 1$             $// C[i] = |\{key = i\}|$ 
for  $i := 2$  to  $k$ 
    do  $C[i] := C[i] + C[i-1]$             $// C[i] = |\{key \leq i\}|$ 
for  $j := n$  downto  $1$ 
    do  $B[C[A[j]]] \leftarrow A[j]$ 
         $C[A[j]] \leftarrow C[A[j]] - 1$ 
```

# Example: loop 1

A: 

4	1	3	4	3
---	---	---	---	---

C: 

0	0	0	0
---	---	---	---

B: 

--	--	--	--	--

```
for i := 1 to k  
  do C[i] := 0
```

# Example: loop 2

A: 

4	1	3	4	3
---	---	---	---	---

C: 

0	0	0	0
---	---	---	---

B: 

--	--	--	--	--

```
for j := 1 to n  
  do C[A[j]] := C[A[j]] + 1           // C[i] = |{key=i}|
```



# Example: loop 2

A: 

4	1	3	4	3
---	---	---	---	---

C: 

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

B: 

--	--	--	--	--

```
for j := 1 to n  
  do C[A[j]] := C[A[j]] + 1           // C[i] = |{key=i}|
```

# Example: loop 2

A:

4	1	3	4	3
---	---	---	---	---

C:

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

B:

--	--	--	--	--

```
for j := 1 to n  
  do C[A[j]] := C[A[j]] + 1           // C[i] = |{key=i}|
```

# Example: loop 2

etc. etc.

```
for j := 1 to n  
  do C[A[j]] := C[A[j]] + 1           // C[i] = |{key=i}|
```

# Example: loop 2

A: 

4	1	3	4	3
---	---	---	---	---

C: 

1	0	2	2
---	---	---	---

B: 

--	--	--	--	--

```
for j := 1 to n  
  do C[A[j]] := C[A[j]] + 1           // C[i] = |{key=i}|
```

# Example: loop 3

A:

4	1	3	4	3
---	---	---	---	---

C:

1	0	2	2
---	---	---	---

B:

--	--	--	--	--

C':

1	1	3	5
---	---	---	---

```
for i := 2 to k  
  do C[i] := C[i] + C[i-1]           // C[i] = |{key ≤ i}|
```

# Example: loop 4

A: 

4	1	3	4	3
---	---	---	---	---

C: 

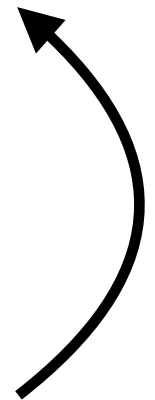
1	1	3	5
---	---	---	---

B: 

--	--	--	--	--

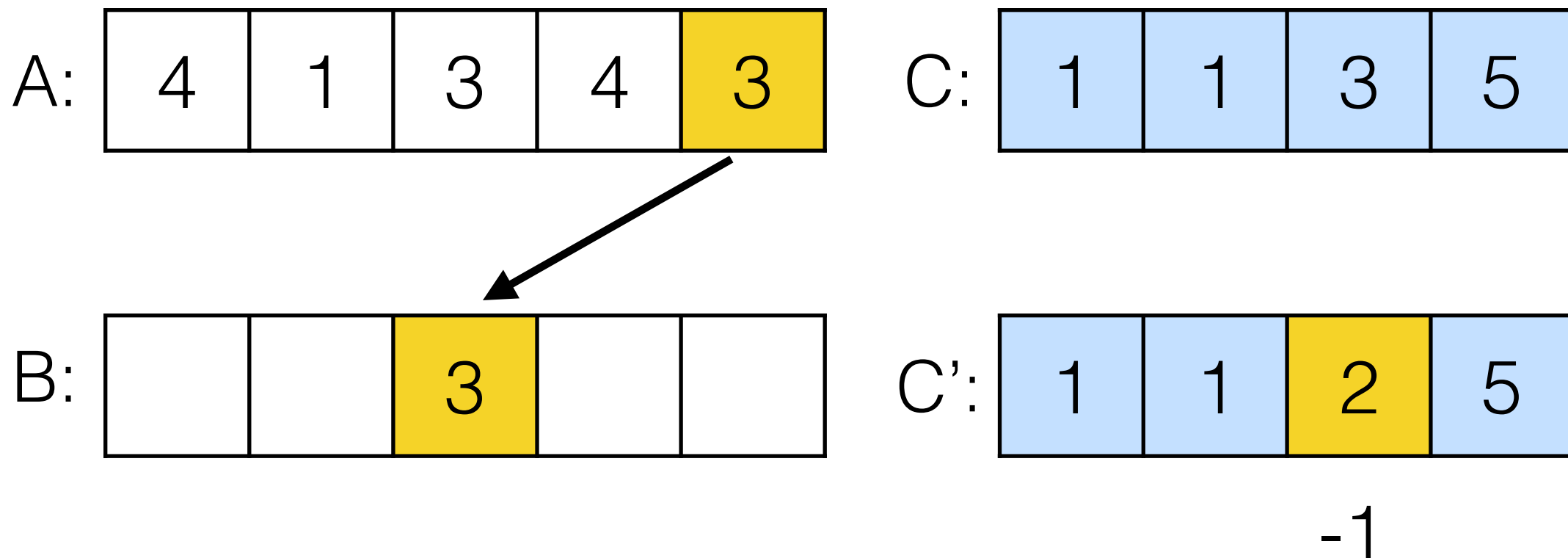
C': 

1	1	3	5
---	---	---	---



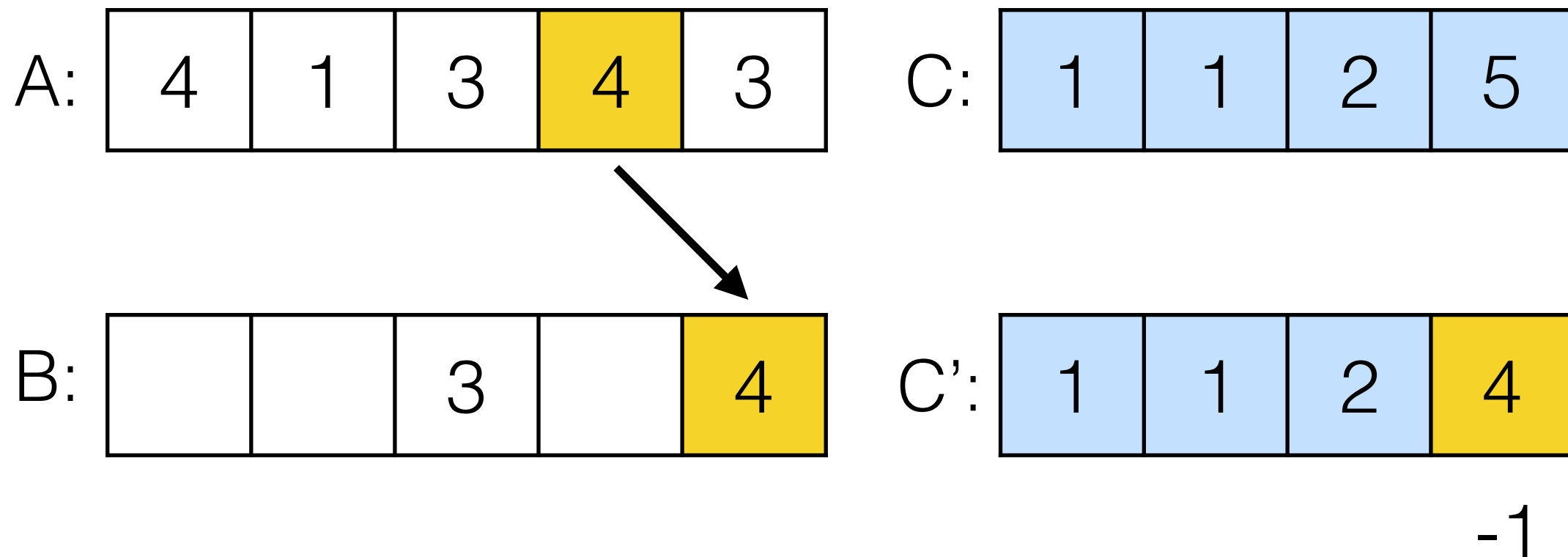
```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```

# Example: loop 4



```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```

# Example: loop 4



```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```



# Example: loop 4

A:

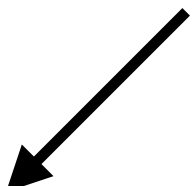
4	1	3	4	3
---	---	---	---	---

C:

1	1	2	4
---	---	---	---

B:

	3	3		4
--	---	---	--	---



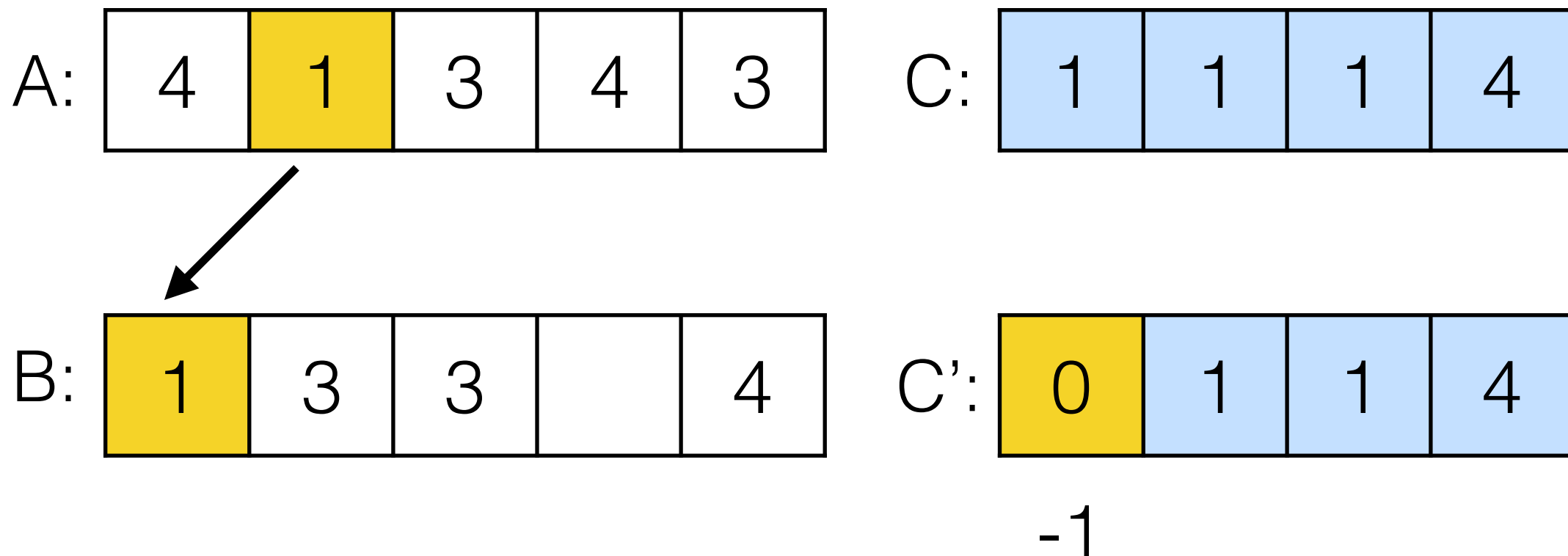
C':

1	1	1	4
---	---	---	---

-1

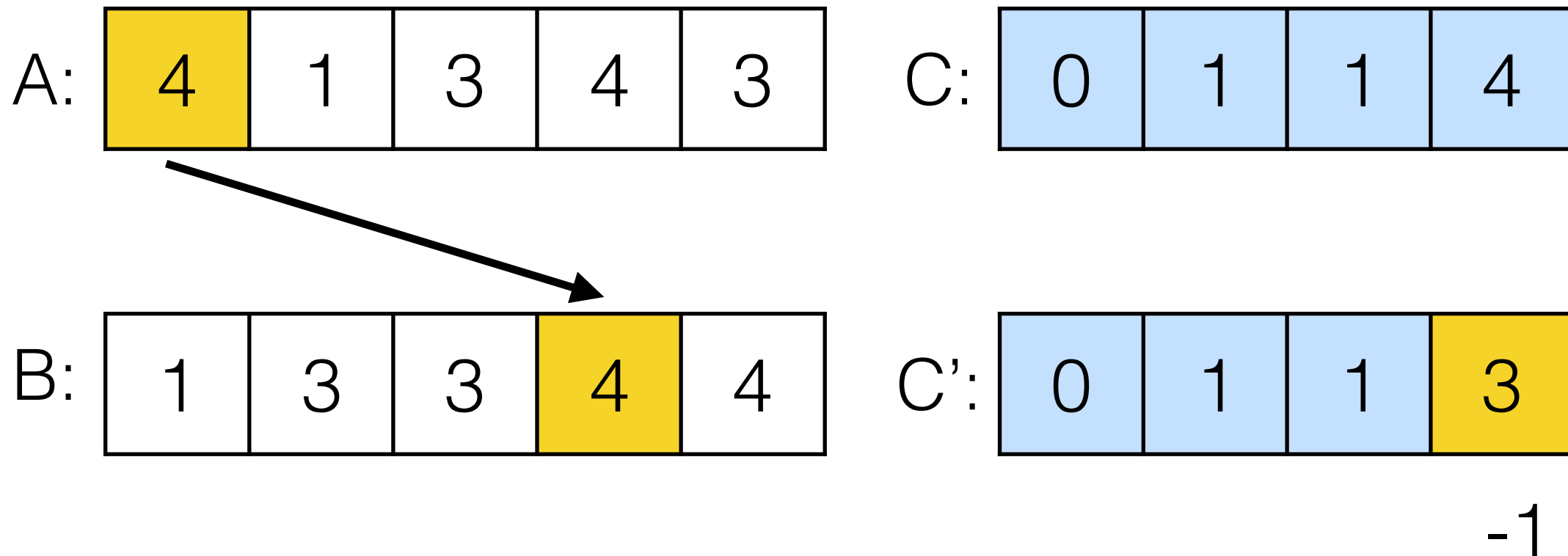
```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```

# Example: loop 4



```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```

# Example: loop 4



```
for j := n downto 1
  do B[C[A[j]]] ← A[j]
      C[A[j]] ← C[A[j]] - 1
```

# Asymptotic Analysis

$\Theta(k)$       **for**  $i := 1$  **to**  $k$   
                  **do**  $C[i] := 0$

$\Theta(n)$       **for**  $j := 1$  **to**  $n$   
                  **do**  $C[A[j]] := C[A[j]] + 1$

$\Theta(k)$       **for**  $i := 2$  **to**  $k$   
                  **do**  $C[i] := C[i] + C[i-1]$

$\Theta(n)$       **for**  $j := n$  **downto**  $1$   
                  **do**  $B[C[A[j]]] \leftarrow A[j]$   
                       $C[A[j]] \leftarrow C[A[j]] - 1$

---

$\Theta(n + k)$

# Asymptotic Analysis

- If  $k=O(n)$ , then Counting Sort takes  $\Theta(n)$  time.
- Remark:
  - Comparison sorting takes  $\Omega(n \lg n)$  time.  
Counting Sort is not a comparison sort. In fact, not a single comparison between elements occurs!

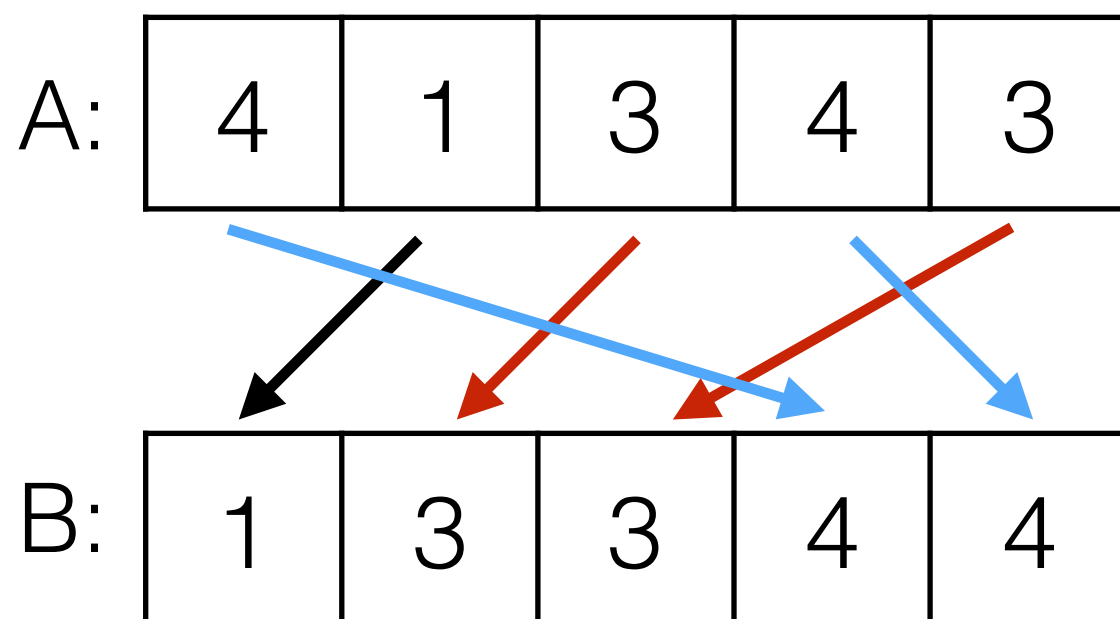
# Stable Sorting (hw 4)

Definition:

- Stable sorting: Stable sorting algorithms maintain the relative order of records with equal keys (i.e., values).
- Thus, a sorting algorithm is stable, if whenever there are two records R and S with the same key and with R appearing before S in the original list, R will appear before S in the sorted list.

Is Counting Sort stable?

Yes!



## 2.5 Radix Sort

# Motivation

- Counting Sort is less efficient when processing numbers from a large range, i.e.,  $k$  is large.
- Can we find an algorithm that efficiently sorts  $n$  numbers for large  $k$ ?



# Origin

- The 1880 U.S. Census took almost 10 years to process.
- Herman Hollerith (1860-1929) prototyped a punched-card technology.
- His machines, including a “card sorter”, allowed the 1890 census total to be reported in 6 weeks.
- He founded the Tabulating Machine Company in 1911, which merged with other companies in 1924 to form International Business Machines (IBM).

# Idea

- Hollerith's idea was to use a digit-by-digit sort.
- He sorted on most significant digit first.
- However, it requires us to keep one sequence for each digit, which then gets sorted recursively.
- It is more efficient to sort on least significant digit first.
- This idea requires a stable sorting algorithm.

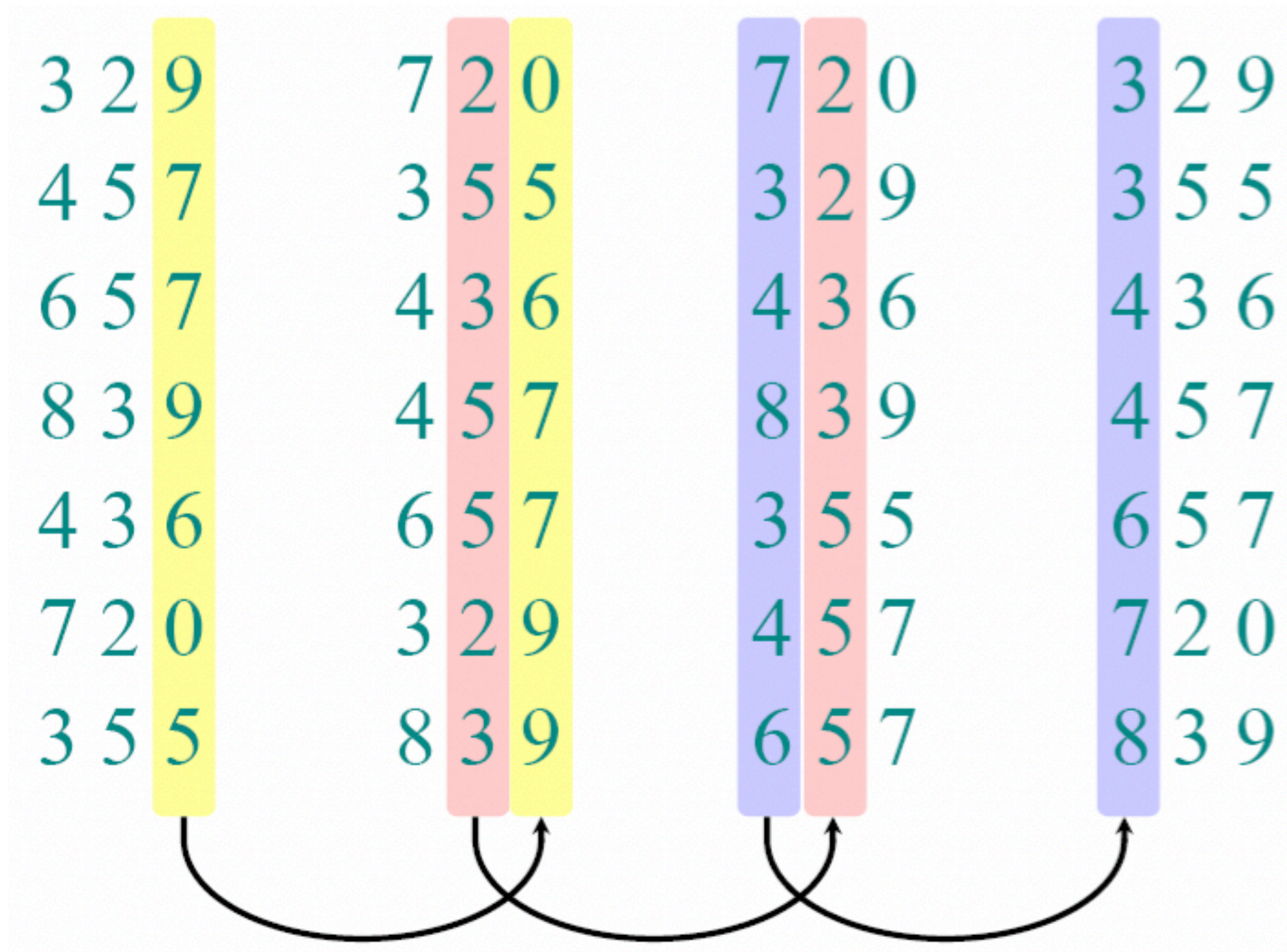
# Radix Sort

**RADIX-SORT( $A, d$ )**

**1   for  $i = 1$  to  $d$**

**2       use a stable sort to sort array  $A$  on digit  $i$**

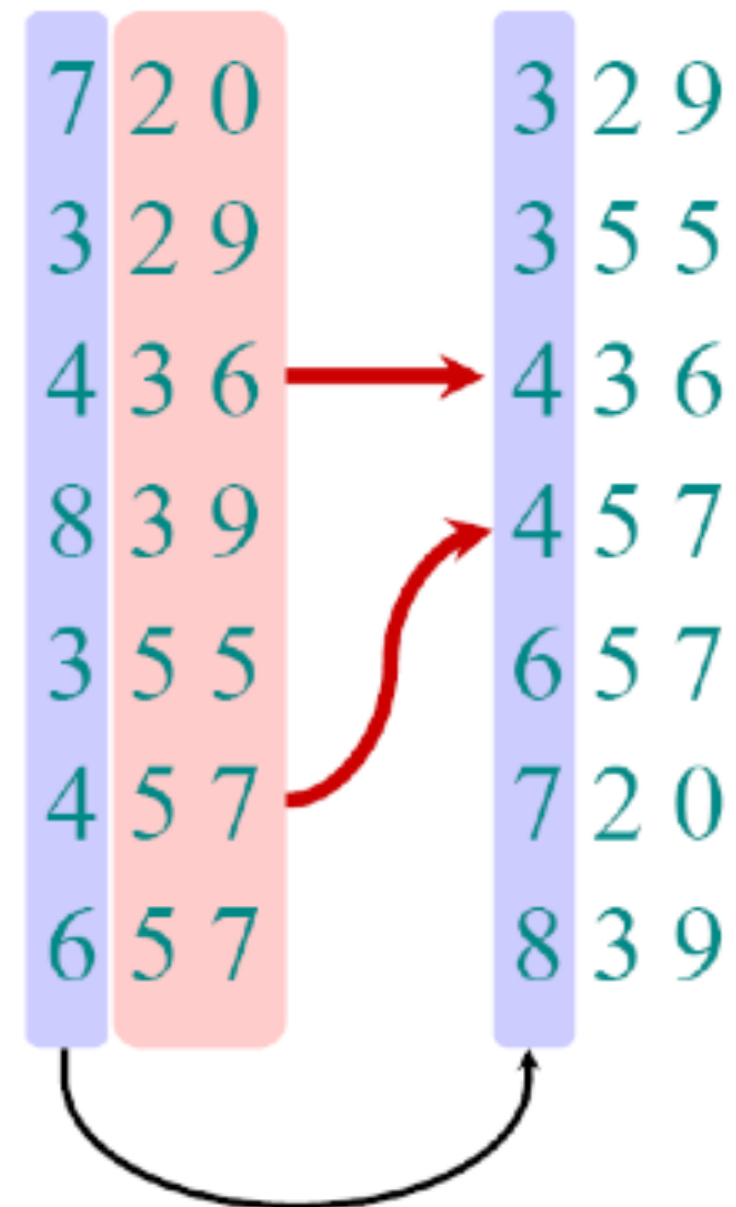
# Example



# Correctness of radix sort

Induction on digit position:

- Only one digit: trivial.
- Assume that the numbers are sorted by their low-order  $t-1$  digits.
- Sort on digit  $t$ :
  - Two numbers that differ in digit  $t$  are correctly sorted.
  - Two numbers equal in digit  $t$  are put in the same order as the input, i.e., correct order.



# Asymptotic Analysis

- Use Counting Sort as stable sorting algorithm.
- Sort  $n$  computer words of  $b$  bits each.
- Each word can be viewed as having  $b/r$  base- $2^r$  digits.
- Example: 32-bit word.
  - $r = 8$ :  
 $d = b/r = 4$  passes of counting sort on base- $2^8$  digits;
  - $r = 16$ :  
 $d = b/r = 2$  passes of counting sort on base- $2^{16}$  digits.
- **How many passes should we make?**



# Choosing $r$

- Counting Sort takes  $\Theta(n+k)$  time to sort  $n$  numbers in the range from 0 to  $k-1$ .
- If each  $b$ -bit word is broken into  $r$ -bit pieces, each pass of Counting Sort takes  $\Theta(n + 2^r)$  time.
- Since there are  $b/r$  passes, we have:

$$T(n, b) = \Theta\left(\frac{b}{r}(n + 2^r)\right).$$

- Choose  $r$  to minimize  $T(n, b)$ .

# Choosing $r$

$$T(n, b) = \Theta\left(\frac{b}{r} (n + 2^r)\right).$$

- Increasing  $r$  means fewer passes, but when  $r \gg \lg n$  the time grows exponentially.
- We do not want  $2^r > n$ , but there is no harm asymptotically in choosing  $r$  as large as possible subject to this constraint.
- Choosing  $r = \lg n$  implies  $T(n, b) = \Theta(bn/\lg n)$ .
- For numbers in the range from 0 to  $n^d - 1$ , we have  $b = dr = d \lg n$ , i.e., Radix Sort runs in  $\Theta(dn)$  time.



# Conclusions

- In practice, Radix Sort is fast for large inputs, as well as simple to code and maintain.
- Example (32-bit numbers, i.e.,  $b=32$ , and  $n=2000$ ):
  - **$dn$** : At most  $d = 3$  passes when sorting 2000 numbers.
  - **$n \lg n$** : Merge Sort and Quicksort do at least ceiling ( $\lg 2000$ ) = 11 passes.

## 2.6 Bucket Sort

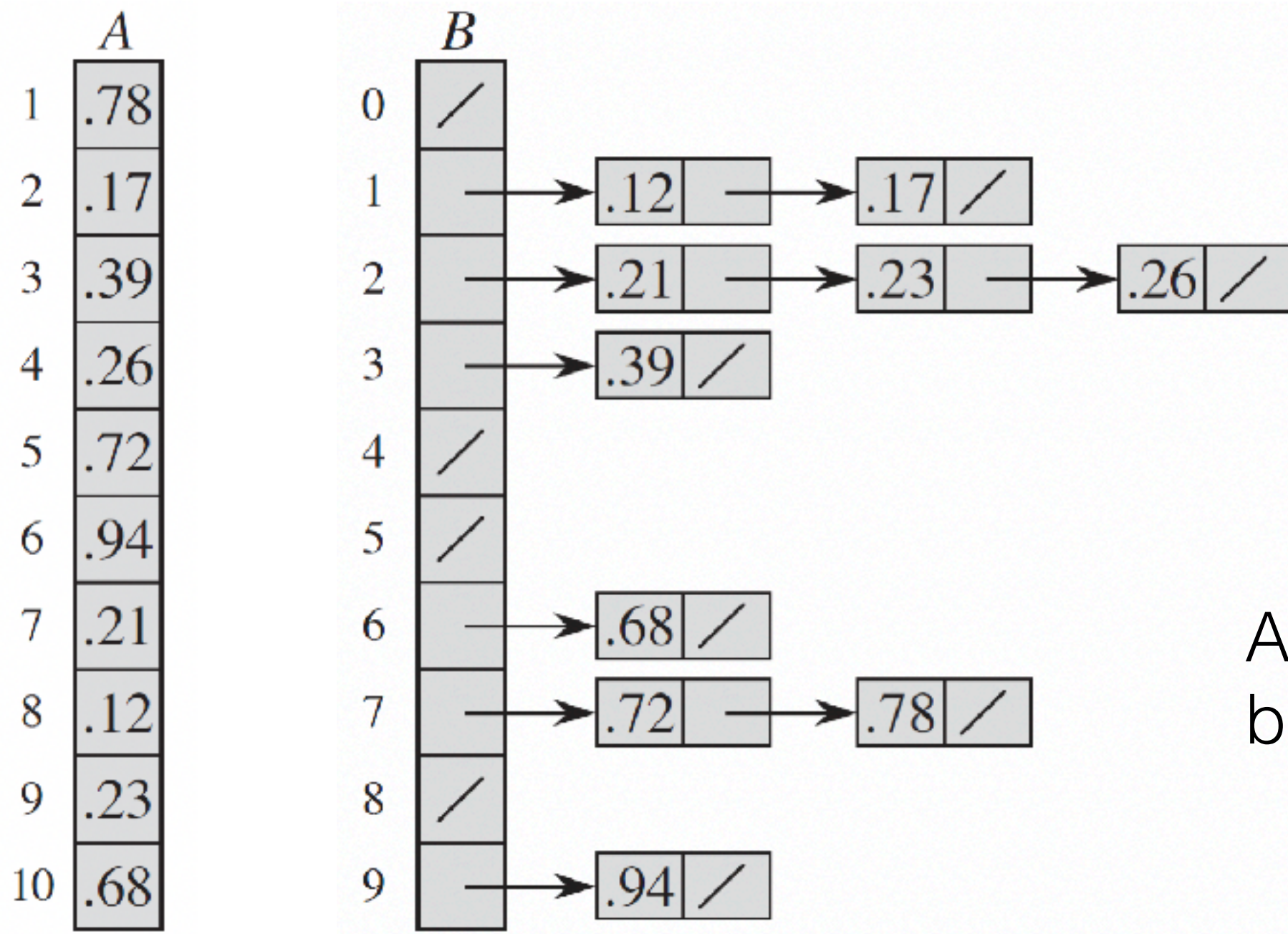
# Motivation

- Can we use the idea of Radix Sort to sort any numbers, i.e., without assuming them to be integers?
- In order to do this efficiently, we make a new assumption:
  - The to-be-sorted elements shall distribute uniformly and independently over the interval  $[0,1)$ .
- Remark:
  - Interval  $[0,1)$  is not a real restriction, as we can normalize the elements to this interval in linear time.
  - However, uniform distribution and independence are restrictions and we will see that we need this to assure good expected running time.

# Idea

- Assuming that we have to sort  $n$  numbers, we split the interval  $[0, 1)$  into  $n$  subintervals or *buckets*.
- Then, we can distribute the  $n$  numbers to the  $n$  buckets.
- Assuming uniform distribution, we can conclude that we have only few numbers falling into each bucket.

# Example ( $n=10$ )



$A[i]$  is put into  
bucket  $\lfloor nA[i] \rfloor$



# Bucket Sort

BUCKET-SORT( $A$ )

```
1  let  $B[0 \dots n - 1]$  be a new array
2   $n = A.length$ 
3  for  $i = 0$  to  $n - 1$ 
4      make  $B[i]$  an empty list
5  for  $i = 1$  to  $n$ 
6      insert  $A[i]$  into list  $B[\lfloor n A[i] \rfloor]$ 
7  for  $i = 0$  to  $n - 1$ 
8      sort list  $B[i]$  with insertion sort
9  concatenate the lists  $B[0], B[1], \dots, B[n - 1]$  together in order
```

# Time complexity

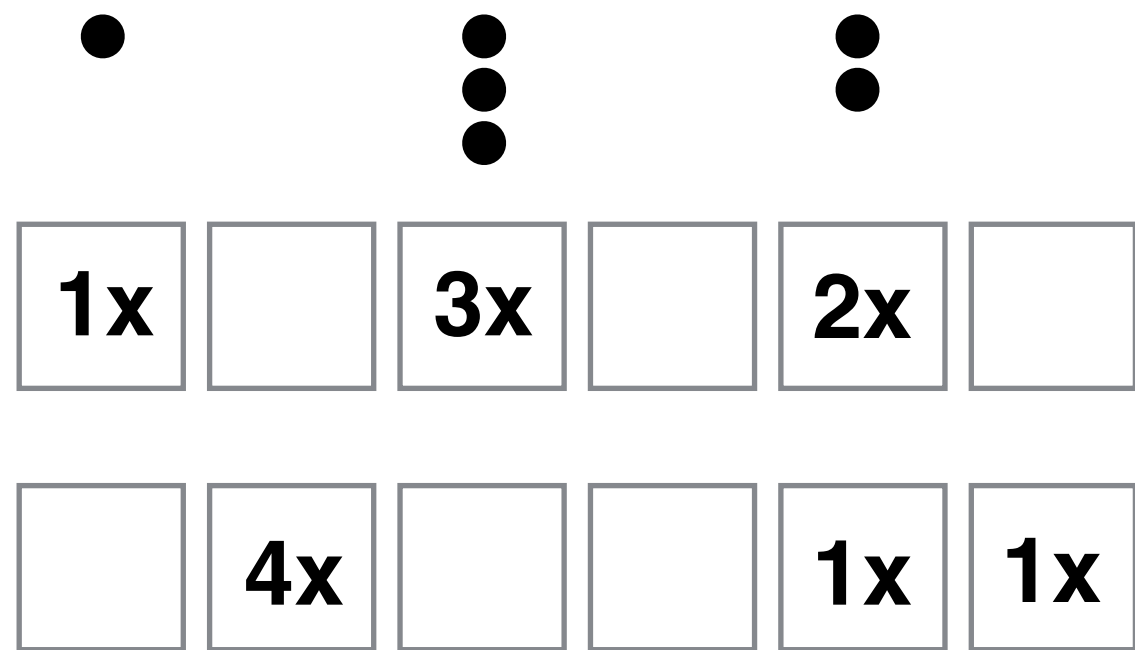
BUCKET-SORT( $A$ )

```
1  let  $B[0..n-1]$  be a new array
2   $n = A.length$ 
3  for  $i = 0$  to  $n - 1$ 
4      make  $B[i]$  an empty list
5  for  $i = 1$  to  $n$ 
6      insert  $A[i]$  into list  $B[\lfloor n A[i] \rfloor]$ 
7  for  $i = 0$  to  $n - 1$ 
8      sort list  $B[i]$  with insertion sort
9  concatenate the lists  $B[0], B[1], \dots, B[n-1]$  together in order
```

Time complexity:  $T(n) = \Theta(n) + \sum_{i=0}^{n-1} O(n_i^2)$

where  $n_i$  denotes the number of elements in bucket  $i$ .

# Average case



might fall into boxes like that

or that

...



# Expected time complexity

$$E[T(n)] = \Theta(n) + \sum_{i=0}^{n-1} O(E[n_i^2])$$

What is  $E[n_i^2]$ ?

Let  $X_{ij}$  be the event that  $A[j]$  falls into bucket  $i$ .

Then,

$$n_i = \sum_{j=1}^n X_{ij}$$

Use assumptions of uniform distribution and independence.

# Estimate $E[n_i^2]$

$$E[n_i^2] = E \left[ \left( \sum_{j=1}^n X_{ij} \right)^2 \right] = E \left[ \sum_{j=1}^n \sum_{k=1}^n X_{ij} X_{ik} \right]$$

$$\begin{aligned} &= E \left[ \sum_{j=1}^n X_{ij}^2 + \sum_{j=1}^n \sum_{\substack{k=1 \\ k \neq j}}^n X_{ij} X_{ik} \right] \\ &= \sum_{j=1}^n E[X_{ij}^2] + \sum_{j=1}^n \sum_{\substack{k=1 \\ k \neq j}}^n E[X_{ij} X_{ik}] \\ &= \sum_{j=1}^n E[X_{ij}^2] + \sum_{j=1}^n \sum_{\substack{k=1 \\ k \neq j}}^n E[X_{ij}] E[X_{ik}] \end{aligned}$$

# Estimate $E[n_i^2]$

$$E[X_{ij}] E[X_{ik}] = \frac{1}{n} \cdot \frac{1}{n} = \frac{1}{n^2}.$$

$$E[X_{ij}^2] = 1^2 \cdot \frac{1}{n} + 0^2 \cdot \left(1 - \frac{1}{n}\right) = \frac{1}{n}.$$

$$\begin{aligned} E[n_i^2] &= \sum_{j=1}^n E[X_{ij}^2] + \sum_{j=1}^n \sum_{\substack{k=1 \\ k \neq j}}^n E[X_{ij}] E[X_{ik}] \\ &= \sum_{j=1}^n \frac{1}{n} + \sum_{j=1}^n \sum_{\substack{k=1 \\ k \neq j}}^n \frac{1}{n^2} \\ &= \frac{n}{n} + n(n-1) \frac{1}{n^2} \\ &= 2 - \frac{1}{n}. \end{aligned}$$

# Expected time complexity

$$E[T(n)] = \Theta(n) + \sum_{i=0}^{n-1} O(E[n_i^2])$$

$$\begin{aligned} E[T(n)] &= \Theta(n) + n \cdot O(2 - 1/n) \\ &= \Theta(n) \end{aligned}$$

## 2.7 Searching

# Searching problem

- Given a sorted sequence.
- Find an element in that sequence.
- Example: Find element 9.

Sequence	3	5	7	8	9	12	15
----------	---	---	---	---	---	----	----

- Brute-force approach (going through the sequence from start until we find the 9) runs in  $O(n)$ .

# Binary Search

Idea: Use a divide & conquer strategy.

1. Divide:  
Check middle element.
2. Conquer:  
Recursively search 1 subarray.
3. Combine:  
Nothing to be done.

# Example (find 9)

3	5	7	8	9	12	15
---	---	---	---	---	----	----

3	5	7	8	9	12	15
---	---	---	---	---	----	----

3	5	7	8	9	12	15
---	---	---	---	---	----	----

3	5	7	8	9	12	15
---	---	---	---	---	----	----

3	5	7	8	9	12	15
---	---	---	---	---	----	----

3	5	7	8	9	12	15
---	---	---	---	---	----	----



# Time Complexity

$$T(n) = 1T(n/2) + \Theta(1)$$

$$a = 1, b = 2$$

$$n \log_b a = n^{\log_2 1} = 1$$

$$f(n) = \Theta(1)$$

$$\text{Case 2: } T(n) = \Theta(\lg n).$$

## 2.8 Summary

# Summary

## Sorting problem:

- Comparison sorts:
  - InsertionSort:  $\Theta(n)$  [best],  $\Theta(n^2)$  [average&worst].
  - MergeSort:  $\Theta(n \lg n)$ .
  - HeapSort:  $\Theta(n \lg n)$  / Heap as a data structure
  - Quicksort:  $\Theta(n \lg n)$  [best&average],  $\Theta(n^2)$  [worst].
  - Decision trees: Worst case does not get better than  $\Theta(n \lg n)$ .
- Sorting in linear time:
  - Counting Sort: small integers
  - Radix Sort: large integers
  - Bucket Sort: any numbers, but uniform distribution.

## Searching Problem:

- Linear Search:  $\Theta(1)$  [best],  $\Theta(n)$  [average&worst]
- Binary Search:  $\Theta(1)$  [best],  $\Theta(\lg n)$  [average&worst]