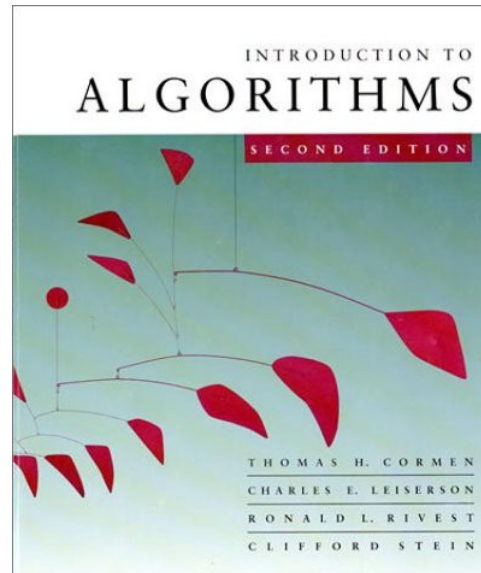


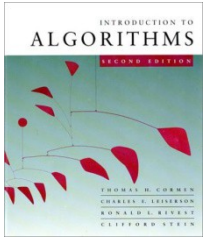
# *Introduction to Algorithms*

## 6.046J/18.401J/SMA5503



## *Lecture 2*

**Based on slides by Prof. Erik Demaine**



# How fast can we sort?

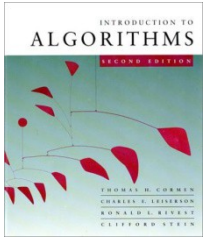
All the sorting algorithms we have seen so far are *comparison sorts*: only use comparisons to determine the relative order of elements.

- *E.g.*, insertion sort, merge sort, quicksort.

The best worst-case running time that we've seen for comparison sorting is  $O(n \lg n)$ .

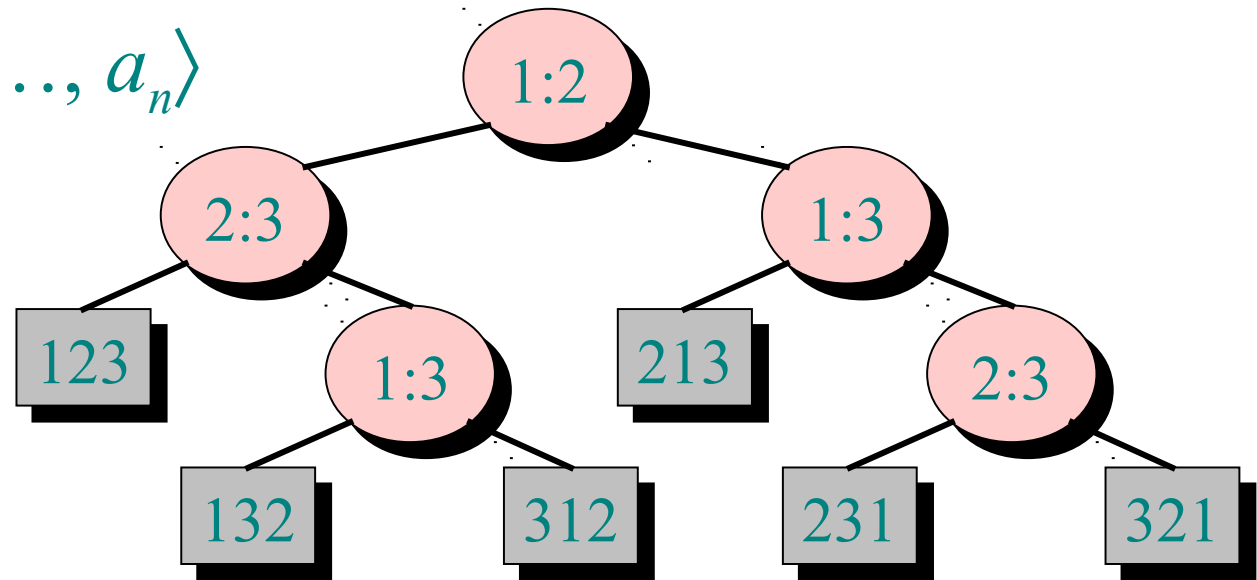
*Is  $O(n \lg n)$  the best we can do?*

*Decision trees* can help us answer this question.



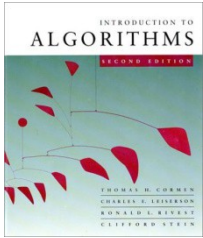
# Decision-tree example

Sort  $\langle a_1, a_2, \dots, a_n \rangle$



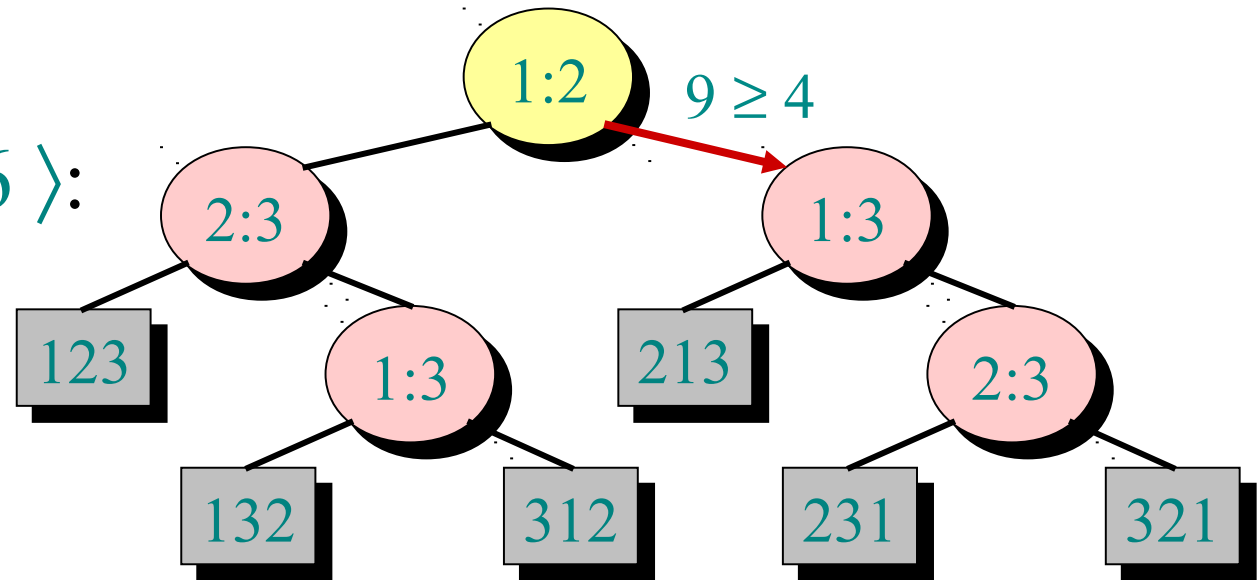
Each internal node is labeled  $i:j$  for  $i, j \in \{1, 2, \dots, n\}$ .

- The left subtree shows subsequent comparisons if  $a_i \leq a_j$ .
- The right subtree shows subsequent comparisons if  $a_i \geq a_j$ .



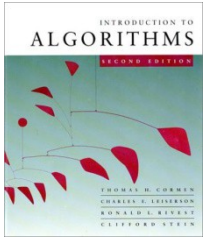
# Decision-tree example

Sort  $\langle a_1, a_2, a_3 \rangle = \langle 9, 4, 6 \rangle$ :



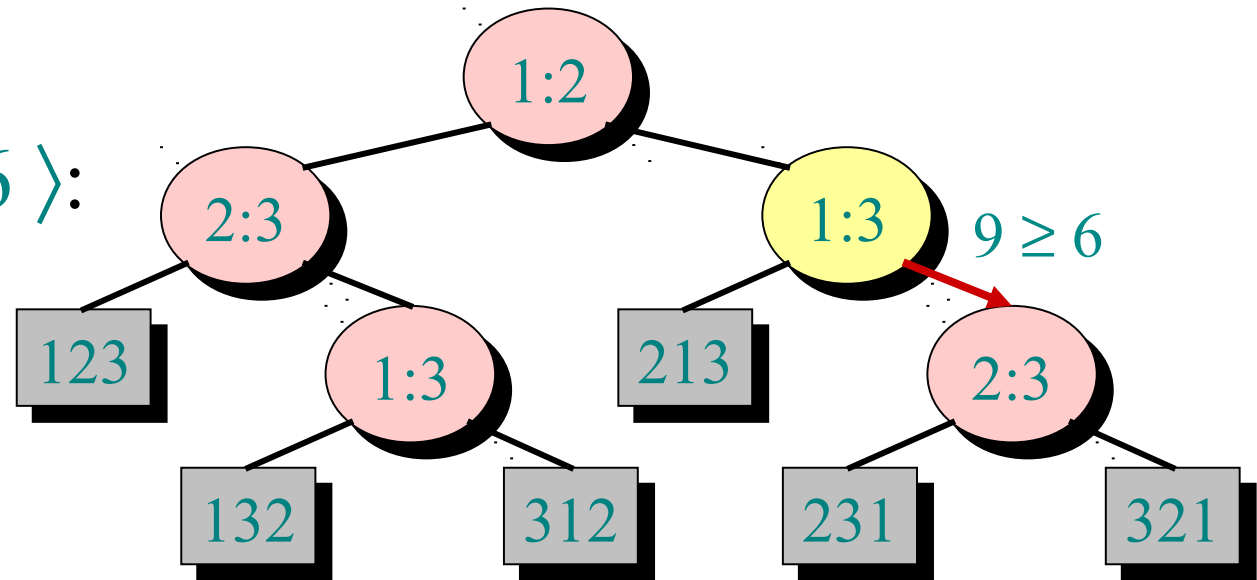
Each internal node is labeled  $i:j$  for  $i, j \in \{1, 2, \dots, n\}$ .

- The left subtree shows subsequent comparisons if  $a_i \leq a_j$ .
- The right subtree shows subsequent comparisons if  $a_i \geq a_j$ .



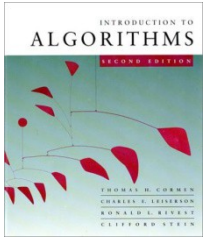
# Decision-tree example

Sort  $\langle a_1, a_2, a_3 \rangle = \langle 9, 4, 6 \rangle$ :



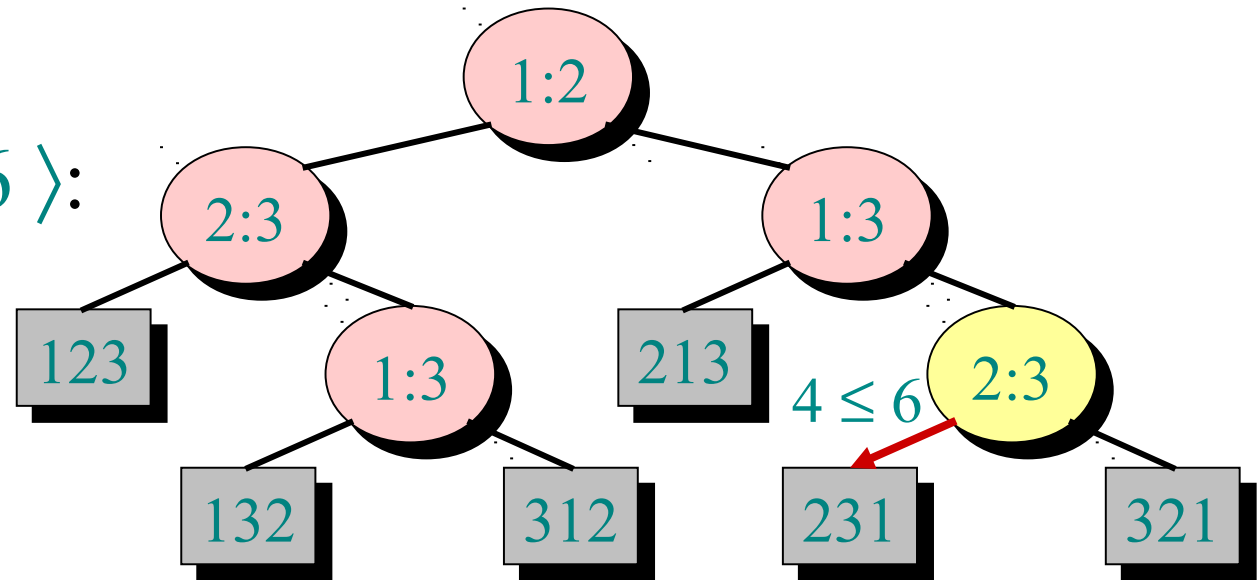
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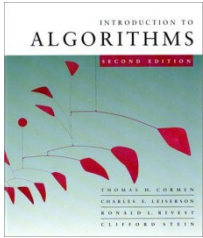
# Decision-tree example

Sort  $\langle a_1, a_2, a_3 \rangle = \langle 9, 4, 6 \rangle$ :



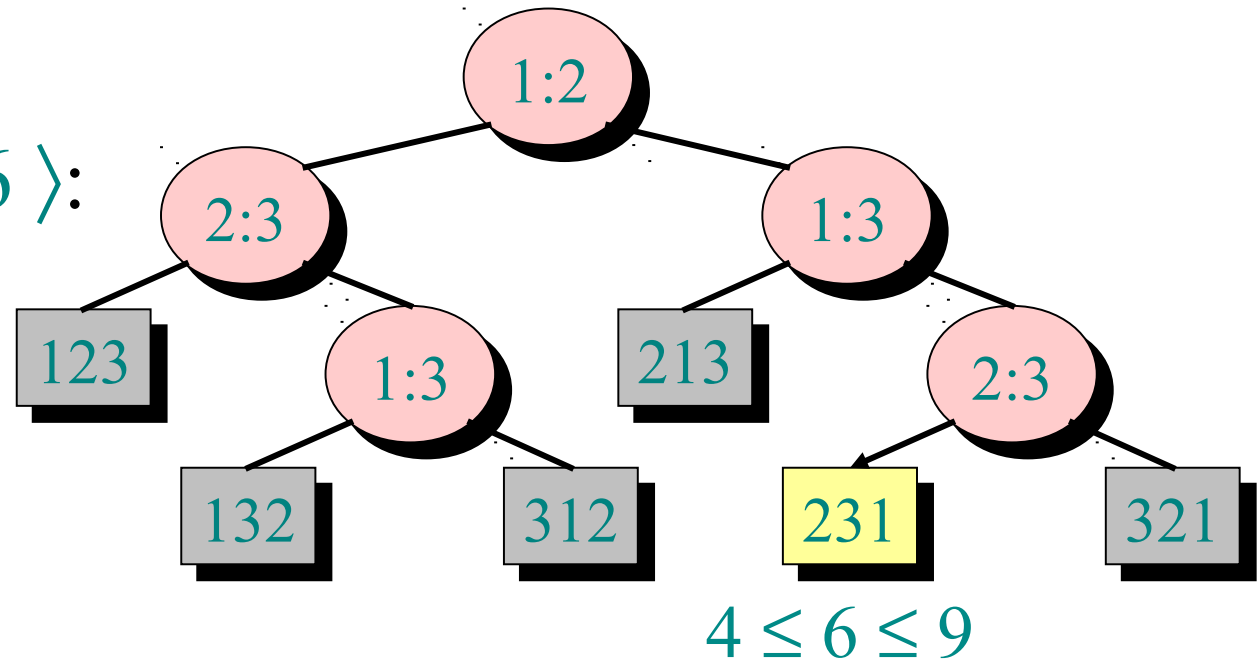
Each internal node is labeled  $i:j$  for  $i, j \in \{1, 2, \dots, n\}$ .

- The left subtree shows subsequent comparisons if  $a_i \leq a_j$ .
- The right subtree shows subsequent comparisons if  $a_i \geq a_j$ .

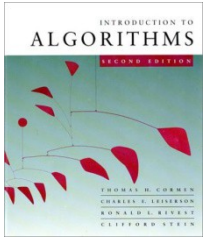


# Decision-tree example

Sort  $\langle a_1, a_2, a_3 \rangle = \langle 9, 4, 6 \rangle$ :



Each leaf contains a permutation  $\langle \pi(1), \pi(2), \dots, \pi(n) \rangle$  to indicate that the ordering  $a_{\pi(1)} \leq a_{\pi(2)} \leq \dots \leq a_{\pi(n)}$  has been established.

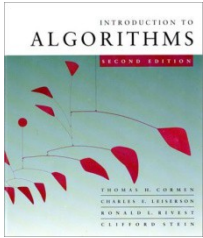


# Decision-tree model

*A decision tree can model the execution of any comparison sort:*

- One tree for each input size  $n$ .
- View the algorithm as splitting whenever it compares two elements.
- The tree contains the comparisons along all possible instruction traces.
- The running time of the algorithm = the length of the path taken.
- Worst-case running time = height of tree.



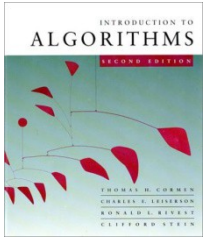


# Lower bound for decision-tree sorting

**Theorem.** Any decision tree that can sort  $n$  elements must have height  $\Omega(n \lg n)$ .

*Proof.* The tree must contain  $\geq n!$  leaves, since there are  $n!$  possible permutations. A height- $h$  binary tree has  $\leq 2^h$  leaves. Thus,  $n! \leq 2^h$ .

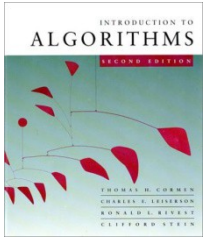
$$\begin{aligned} \therefore h &\geq \lg(n!) && (\lg \text{ is mono. increasing}) \\ &\geq \lg((n/e)^n) && (\text{Stirling's formula}) \\ &= n \lg n - n \lg e \\ &= \Omega(n \lg n). \quad \square \end{aligned}$$



# Lower bound for comparison sorting

**Corollary.** Merge sort is asymptotically optimal comparison sorting algorithm.

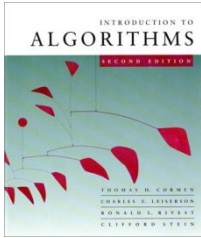




# Sorting in linear time

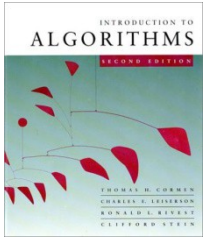
**Counting sort:** No comparisons between elements.

- **Input:**  $A[1 \dots n]$ , where  $A[j] \in \{1, 2, \dots, k\}$ .
- **Output:**  $B[1 \dots n]$ , sorted.
- **Auxiliary storage:**  $C[1 \dots k]$ .

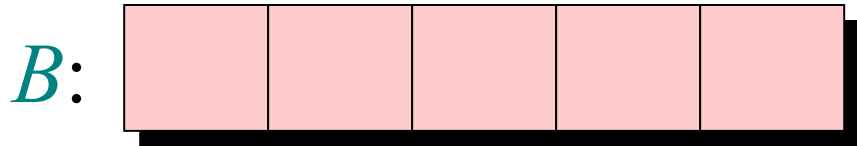
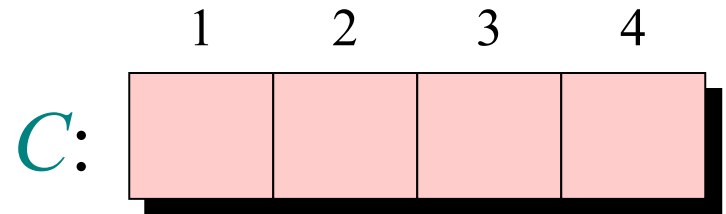
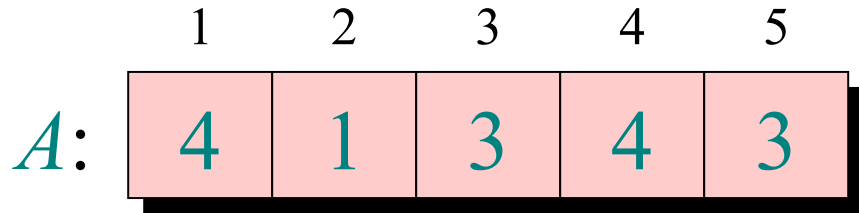


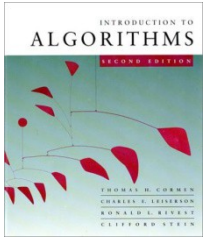
# Counting sort

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



# Counting-sort example





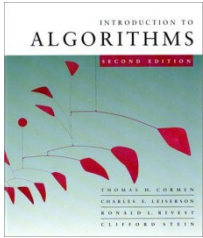
# Loop 1

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

	1	2	3	4
<i>C</i> :	0	0	0	0

<i>B</i> :					
------------	--	--	--	--	--

**for**  $i \leftarrow 1$  **to**  $k$   
    **do**  $C[i] \leftarrow 0$



# Loop 2

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

	1	2	3	4
<i>C</i> :	0	0	0	1

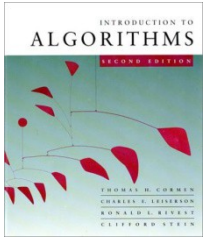
<i>B</i> :					
------------	--	--	--	--	--

**for**  $j \leftarrow 1$  **to**  $n$

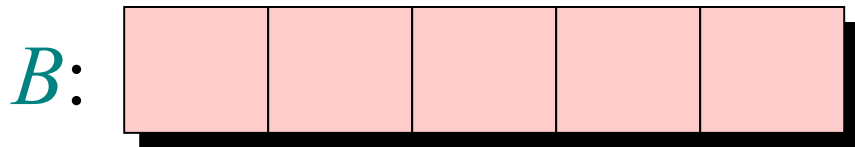
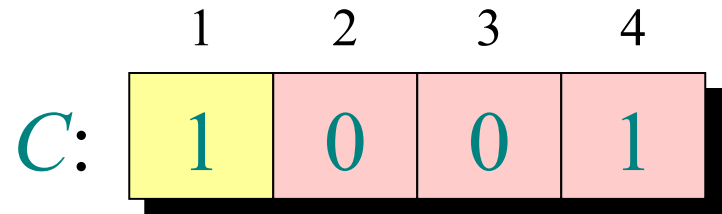
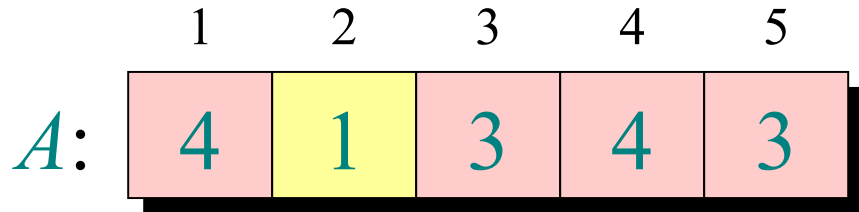
**do**  $C[A[j]] \leftarrow C[A[j]] + 1$

$i\}$

▷  $C[i] = |\{\text{key} =$



# Loop 2



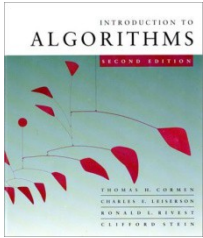
**for**  $j \leftarrow 1$  **to**  $n$

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$i\}$

▷  $C[i] = |\{\text{key} =$





# Loop 2

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

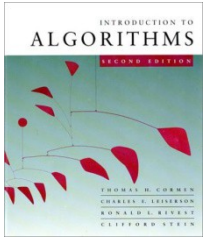
	1	2	3	4
<i>C</i> :	1	0	1	1

<i>B</i> :					
------------	--	--	--	--	--

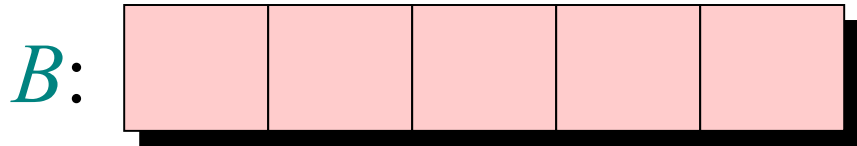
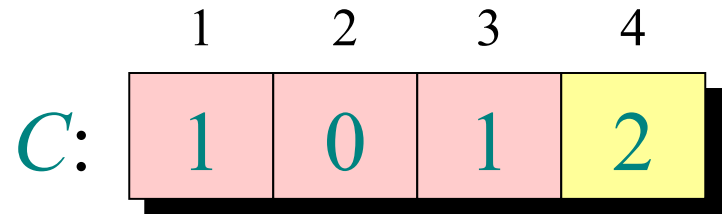
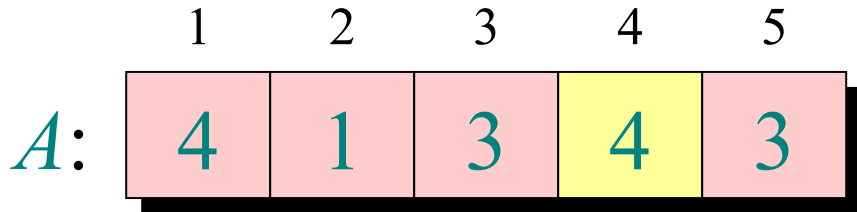
**for**  $j \leftarrow 1$  **to**  $n$

**do**  $C[A[j]] \leftarrow C[A[j]] + 1$   
 $i\}$

▷  $C[i] = |\{\text{key} =$



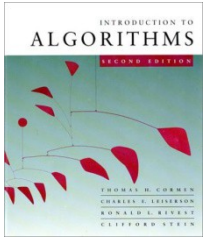
# Loop 2



**for**  $j \leftarrow 1$  **to**  $n$

**do**  $C[A[j]] \leftarrow C[A[j]] + 1$   
 $i\}$

▷  $C[i] = |\{\text{key} =$



# Loop 2

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

	1	2	3	4
<i>C</i> :	1	0	2	2

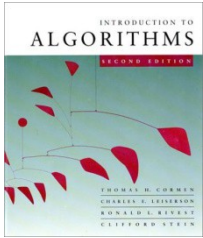
<i>B</i> :					
------------	--	--	--	--	--

**for**  $j \leftarrow 1$  **to**  $n$

**do**  $C[A[j]] \leftarrow C[A[j]] + 1$

$i\}$

▷  $C[i] = |\{\text{key} =$



# Loop 3

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

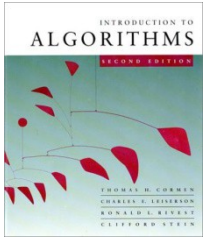
<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	1	0	2	2

<i>C'</i> :	1	1	2	2
-------------	---	---	---	---

**for**  $i \leftarrow 2$  **to**  $k$

**do**  $C[i] \leftarrow C[i] + C[i-1]$   $\triangleright C[i] = |\{\text{key} \leq i\}|$



# Loop 3

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

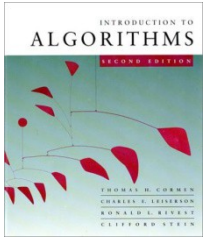
<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	1	0	2	2

<i>C'</i> :	1	1	3	2
-------------	---	---	---	---

**for**  $i \leftarrow 2$  **to**  $k$

**do**  $C[i] \leftarrow C[i] + C[i-1]$   $\triangleright C[i] = |\{\text{key} \leq i\}|$



# Loop 3

	1	2	3	4	5
<i>A</i> :	4	1	3	4	3

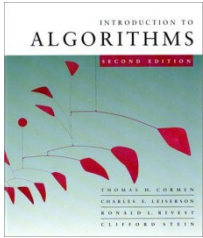
<i>B</i> :					
------------	--	--	--	--	--

	1	2	3	4
<i>C</i> :	1	0	2	2

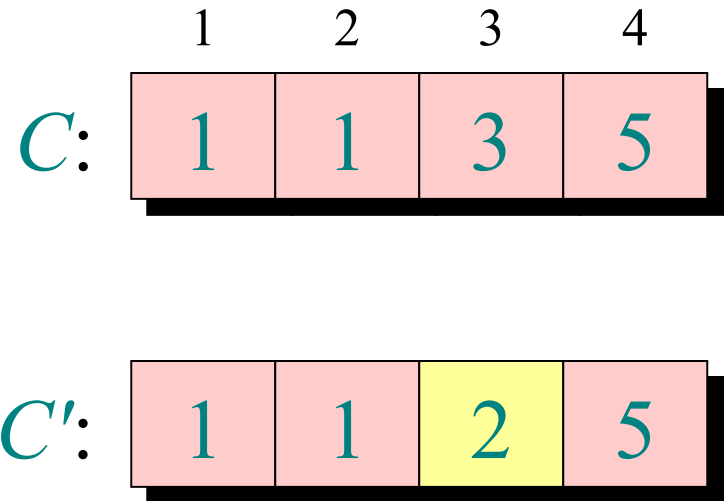
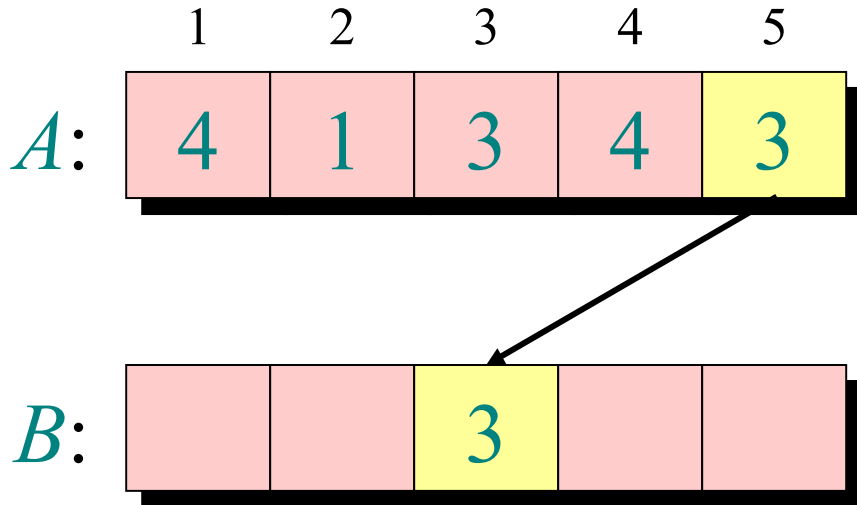
<i>C'</i> :	1	1	3	5
-------------	---	---	---	---

**for**  $i \leftarrow 2$  **to**  $k$

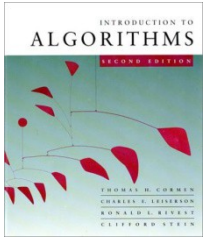
**do**  $C[i] \leftarrow C[i] + C[i-1]$   $\triangleright C[i] = |\{\text{key} \leq i\}|$



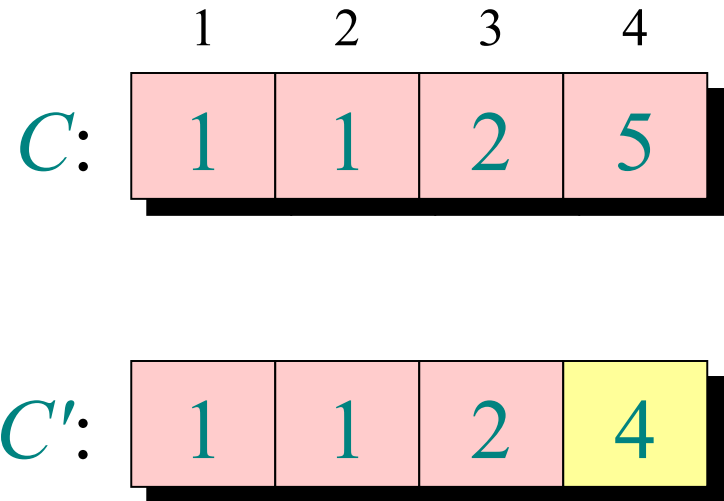
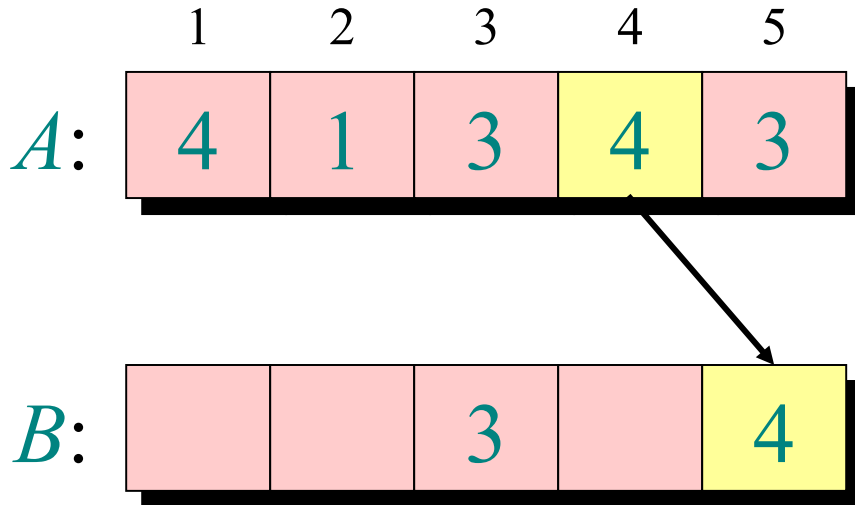
# Loop 4



```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```

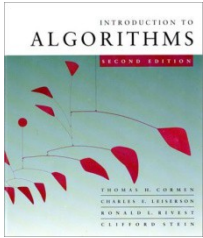


# Loop 4

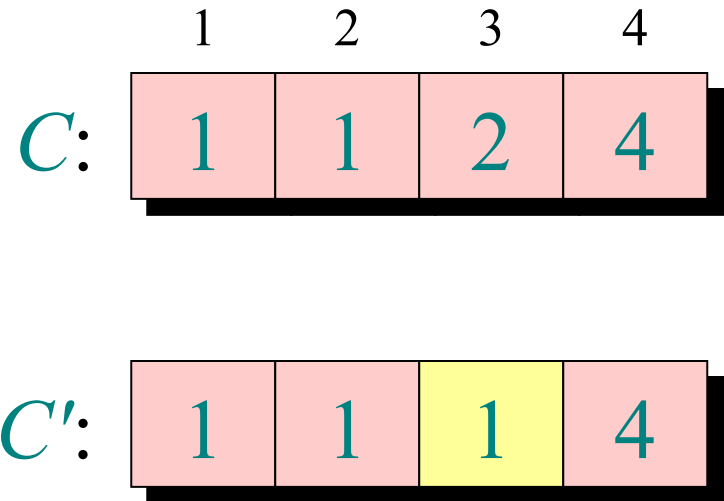
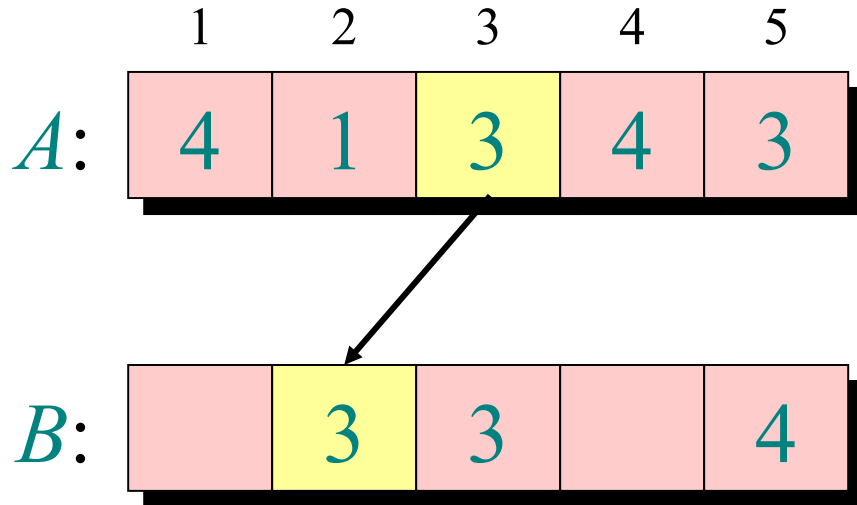


```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
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```

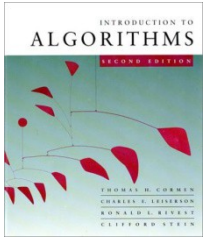




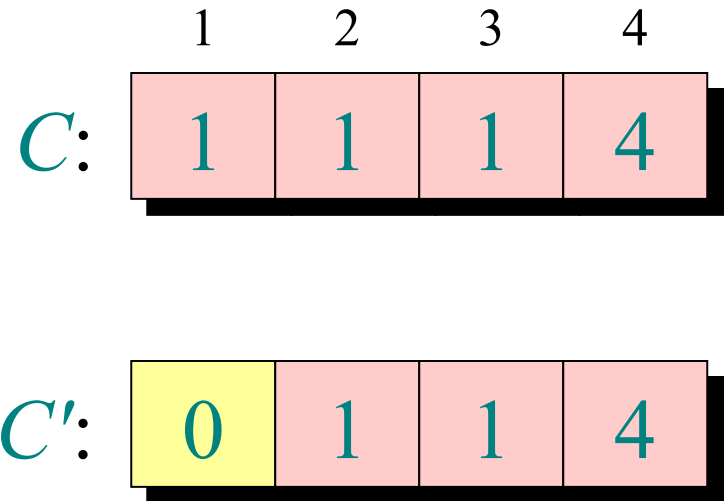
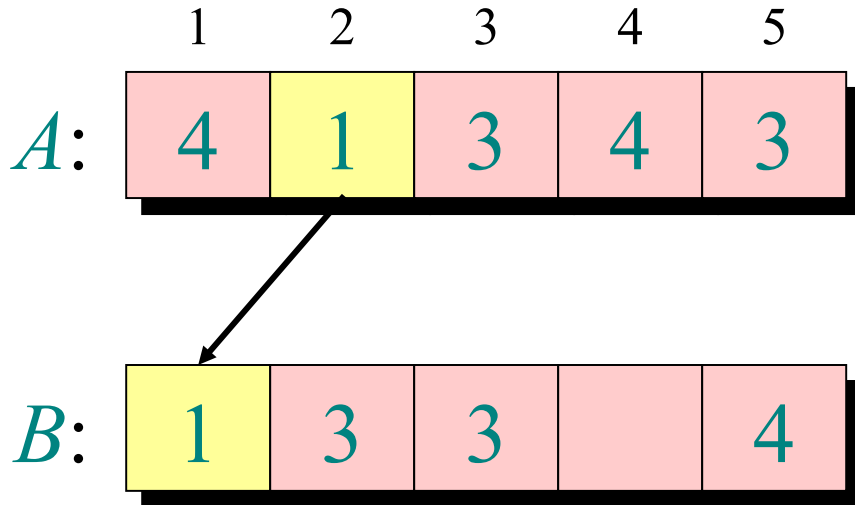
# Loop 4



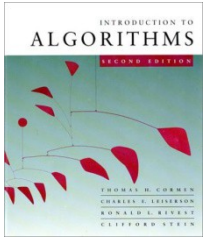
```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



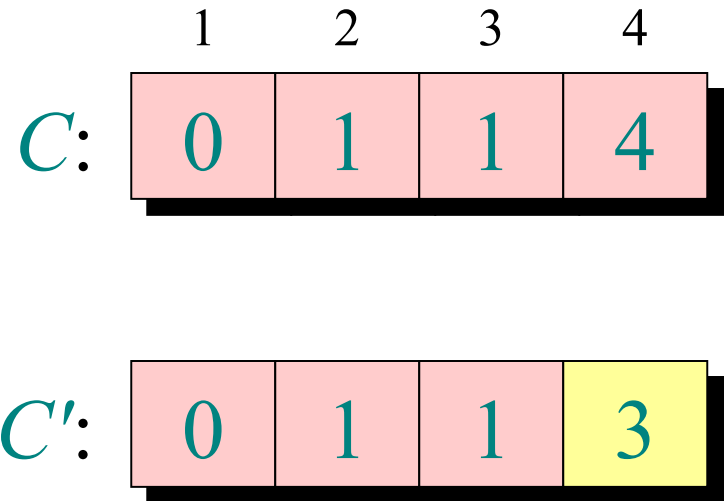
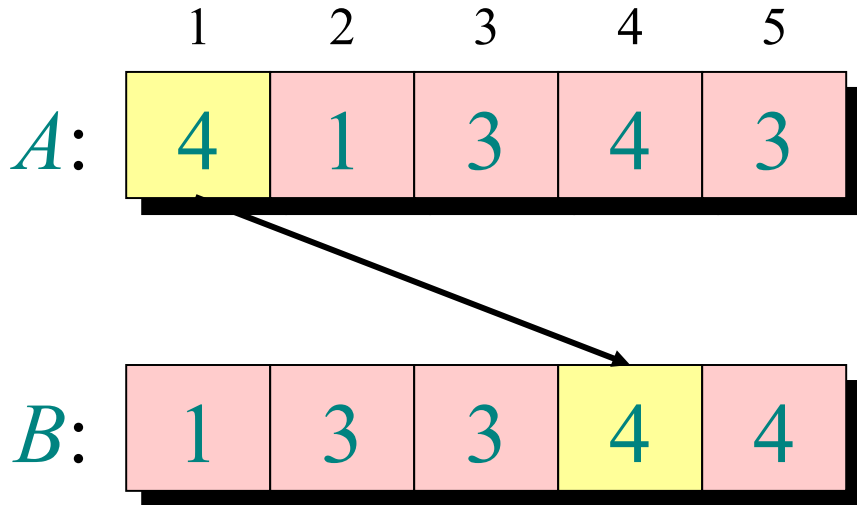
# Loop 4



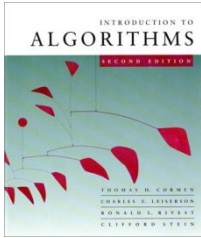
```
for  $j \leftarrow n$  downto 1
  do  $B[C[A[j]]] \leftarrow A[j]$ 
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



# Loop 4

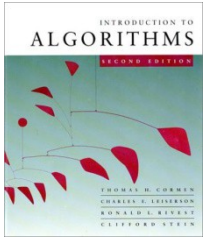


```
for  $j \leftarrow n$  downto 1
  do  $B[C[A[j]]] \leftarrow A[j]$ 
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



# Analysis

$\Theta(k)$	{	<b>for</b> $i \leftarrow 1$ <b>to</b> $k$ <b>do</b> $C[i] \leftarrow 0$
$\Theta(n)$	{	<b>for</b> $j \leftarrow 1$ <b>to</b> $n$ <b>do</b> $C[A[j]] \leftarrow C[A[j]] + 1$
$\Theta(k)$	{	<b>for</b> $i \leftarrow 2$ <b>to</b> $k$ <b>do</b> $C[i] \leftarrow C[i] + C[i-1]$
$\Theta(n)$	{	<b>for</b> $j \leftarrow n$ <b>downto</b> $1$ <b>do</b> $B[C[A[j]]] \leftarrow A[j]$ $C[A[j]] \leftarrow C[A[j]] - 1$
<hr/>		
$\Theta(n + k)$		



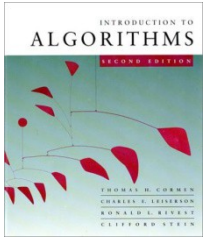
# Running time

If  $k = O(n)$ , then counting sort takes  $\Theta(n)$  time.

- But, sorting takes  $\Omega(n \lg n)$  time!
- Where's the fallacy?

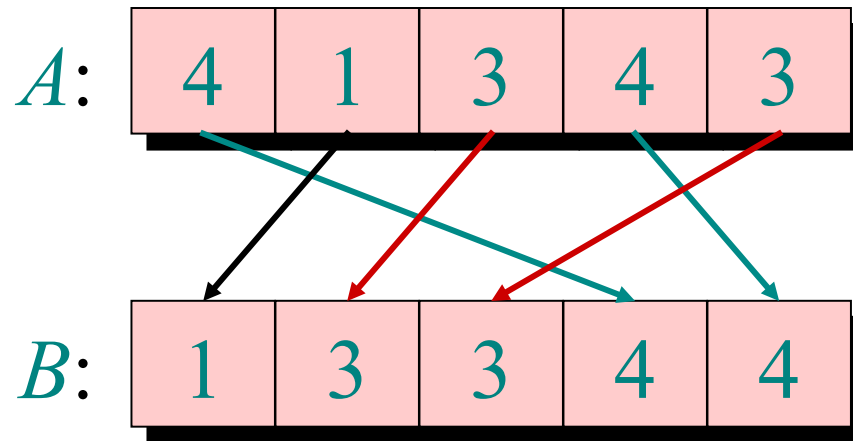
## Answer:

- *Comparison sorting* takes  $\Omega(n \lg n)$  time.
- Counting sort is not a *comparison sort*.
- In fact, not a single comparison between elements occurs!

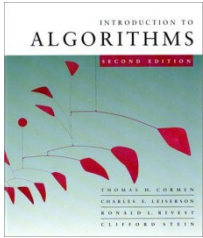


# Stable sorting

Counting sort is a *stable* sort: it preserves the input order among equal elements.

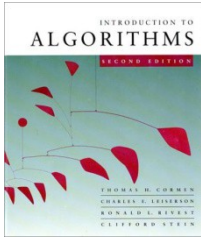


**Exercise:** What other sorts have this property?

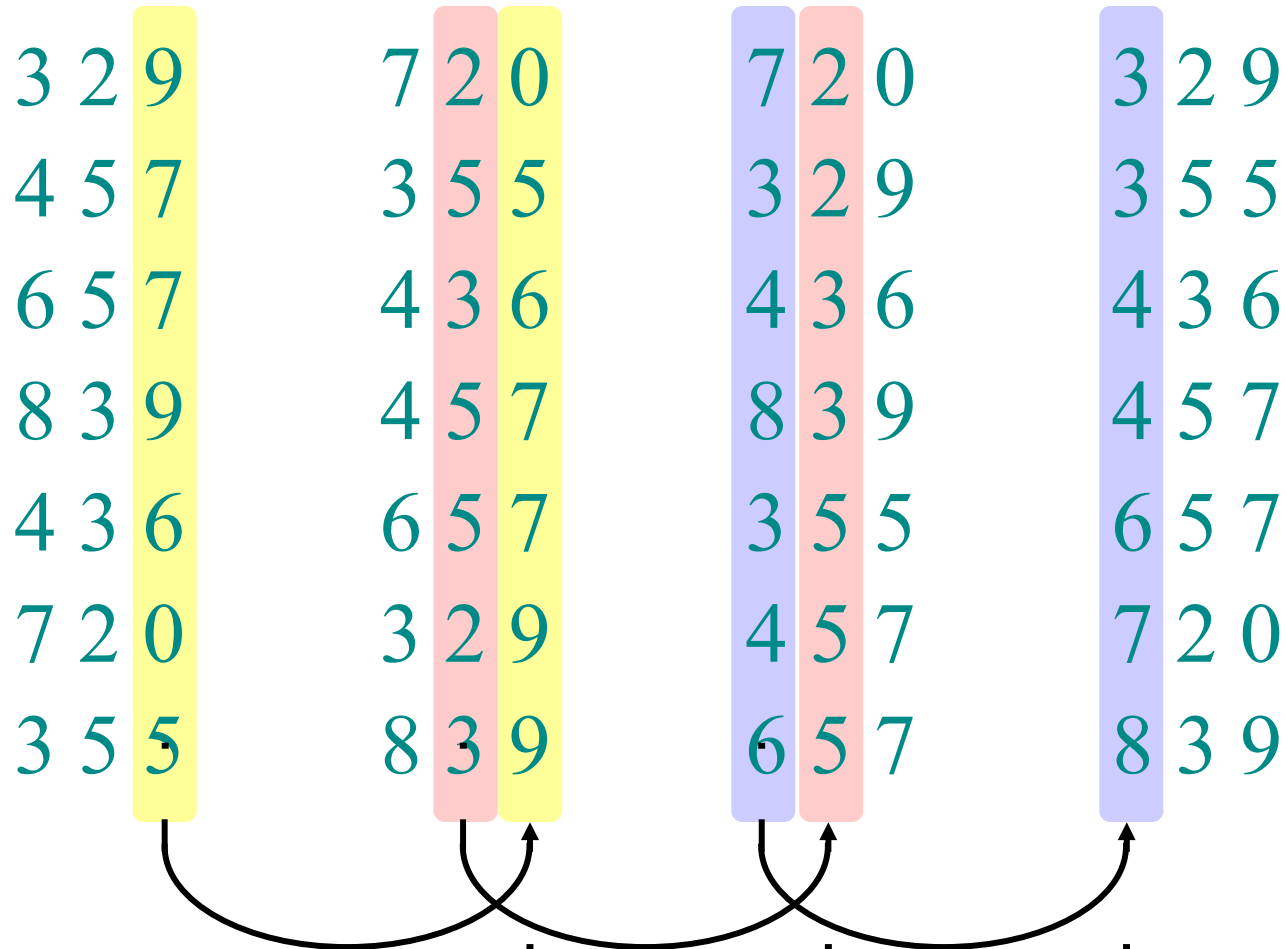


# Radix sort

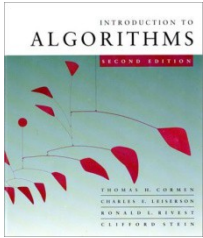
- *Origin*: Herman Hollerith's card-sorting machine for the 1890 U.S. Census. (See Book.)
- Digit-by-digit sort.
- Hollerith's original (bad) idea: sort on most-significant digit first.
- Good idea: Sort on *least-significant digit first* with auxiliary *stable* sort.



# Operation of radix sort



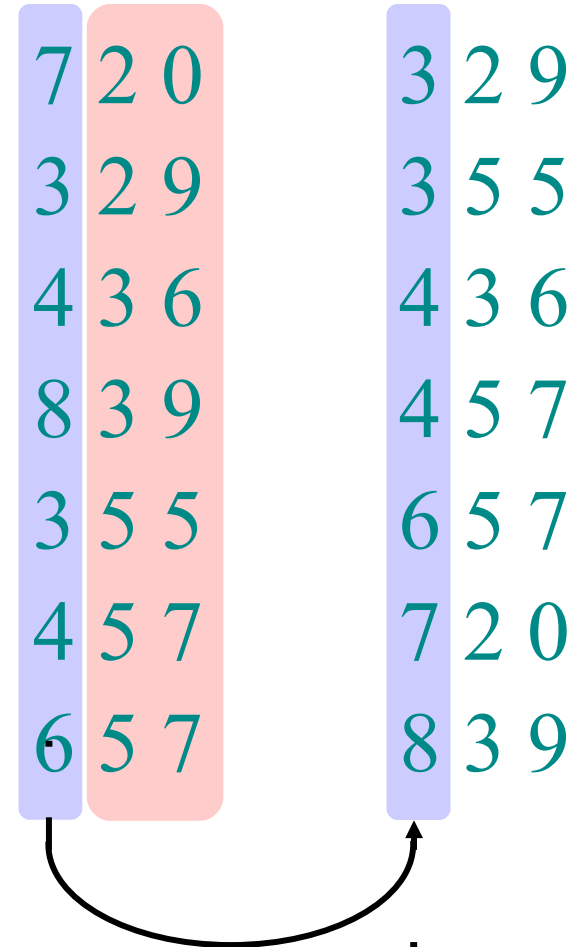


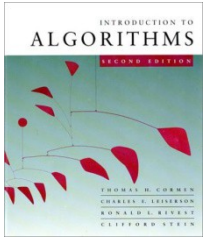


# Correctness of radix sort

*Induction on digit position*

- Assume that the numbers are sorted by their low-order  $t - 1$  digits.
- Sort on digit  $t$

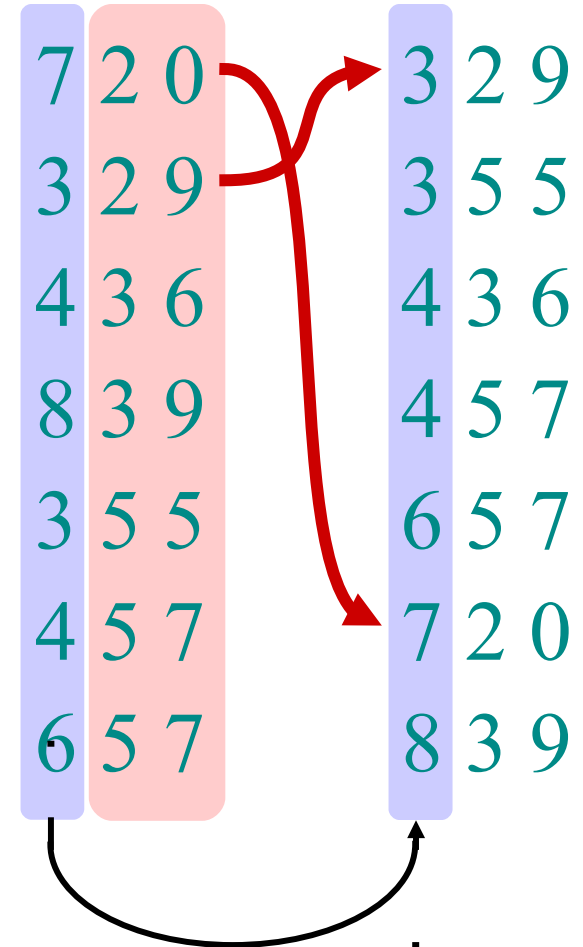


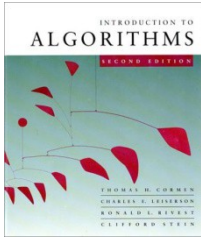


# Correctness of radix sort

*Induction on digit position*

- 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.

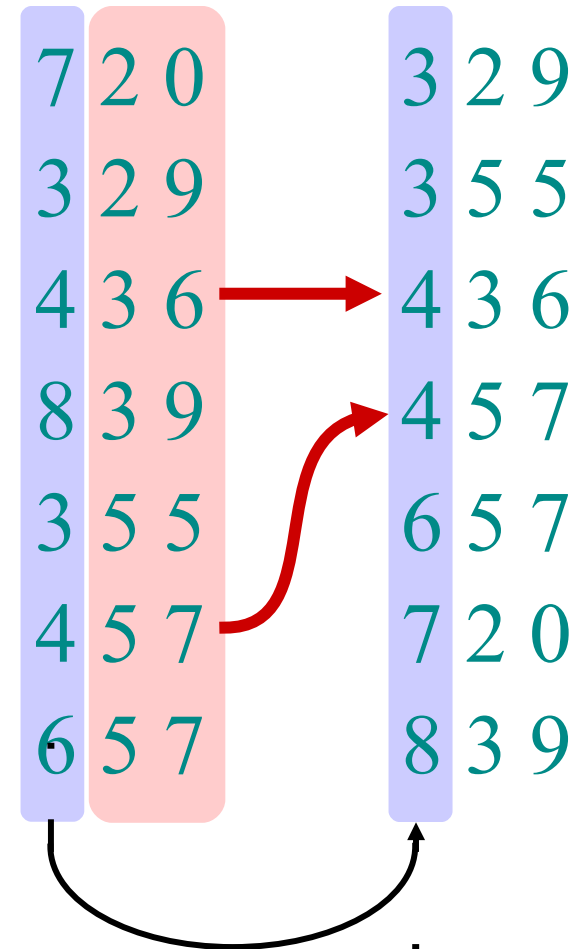


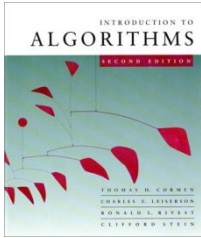


# Correctness of radix sort

*Induction on digit position*

- 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  $\Rightarrow$  correct order.





# Analysis of radix sort

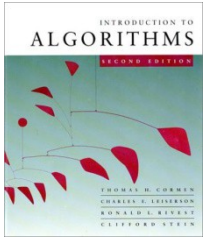
- Assume counting sort is the auxiliary stable sort.
- Sort  $n$  computer words of  $b$  bits each.
- Each word can be viewed as having  $b/r$  base- $2^r$  digits.

**Example:** 32-bit word 

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$r = 8 \Rightarrow b/r = 4$  passes of counting sort on base- $2^8$  digits; or  $r = 16 \Rightarrow b/r = 2$  passes of counting sort on base- $2^{16}$  digits.

*How many passes should we make?*



# Analysis (continued)

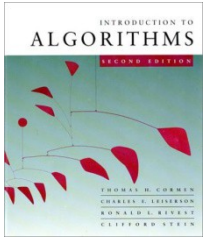
**Recall:** Counting sort takes  $\Theta(n + k)$  time to sort  $n$  numbers in the range from  $0$  to  $k - 1$ .

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  $T(n, b)$ :

- Increasing  $r$  means fewer passes, but as  $r \gg \lg n$ , the time grows exponentially.



# Choosing $r$

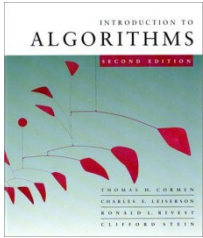
$$T(n, b) = \Theta\left(\frac{b}{r} \binom{n + 2^r}{r}\right)$$

Minimize  $T(n, b)$  by differentiating and setting to 0.

Or, just observe that we don't want  $2^r \gg n$ , and there's 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  $b = d \lg n \Rightarrow$  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):

- At most 3 passes when sorting  $\geq 2000$  numbers.
- Merge sort and quicksort do at least  $\lceil \lg 2000 \rceil = 11$  passes.

**Downside:** Unlike quicksort, radix sort displays little locality of reference, and thus a well-tuned quicksort fares better on modern processors, which feature steep memory hierarchies.