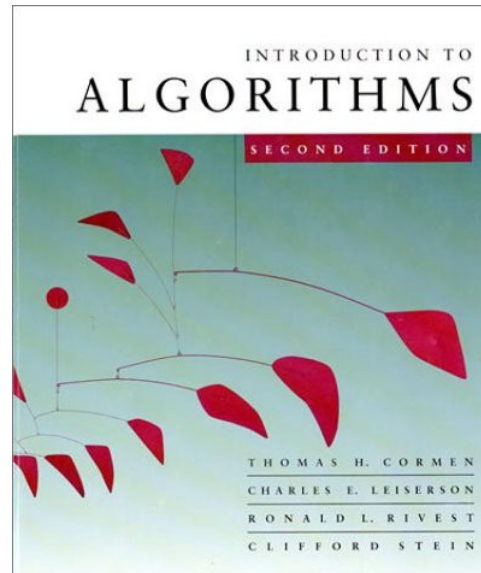


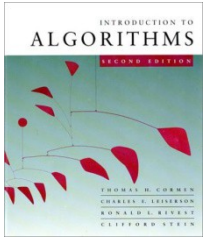
Introduction to Algorithms

6.046J/18.401J/SMA5503



Lecture 2

Based on slides by Prof. Erik Demaine



How fast can we sort?

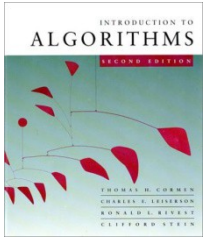
All the sorting algorithms we have seen so far are *comparison sorts*: only use comparisons to determine the relative order of elements.

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

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

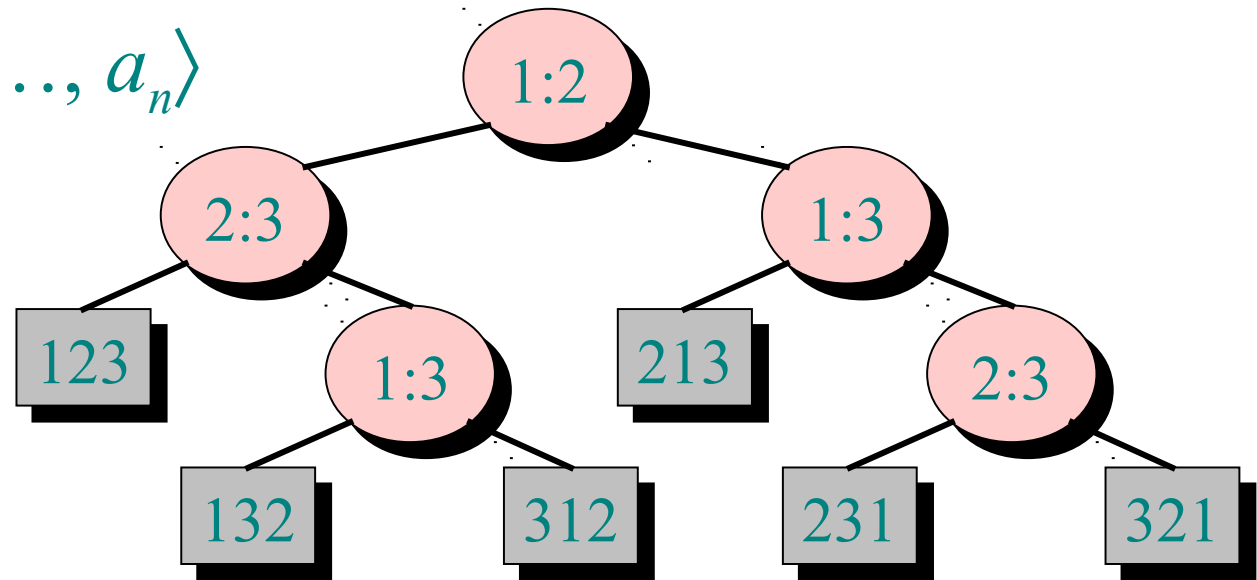
Is $O(n \lg n)$ the best we can do?

Decision trees can help us answer this question.



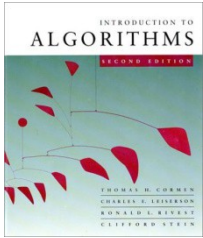
Decision-tree example

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



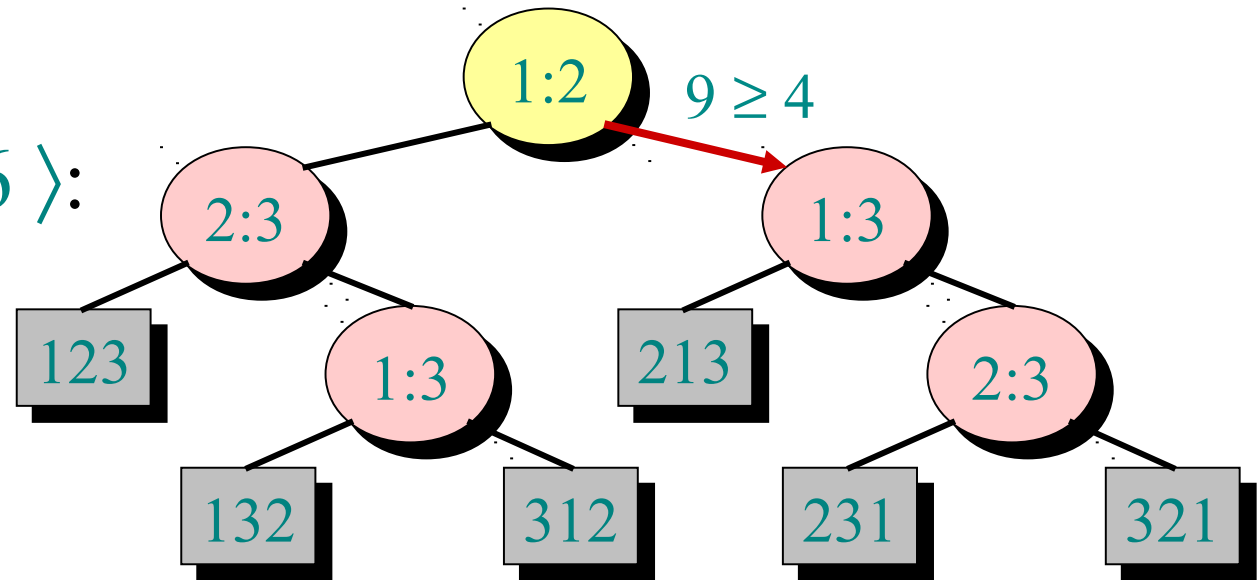
Each internal node is labeled $i:j$ for $i, j \in \{1, 2, \dots, n\}$.

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



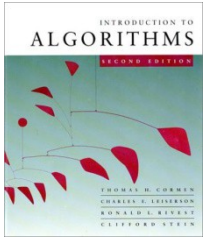
Decision-tree example

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



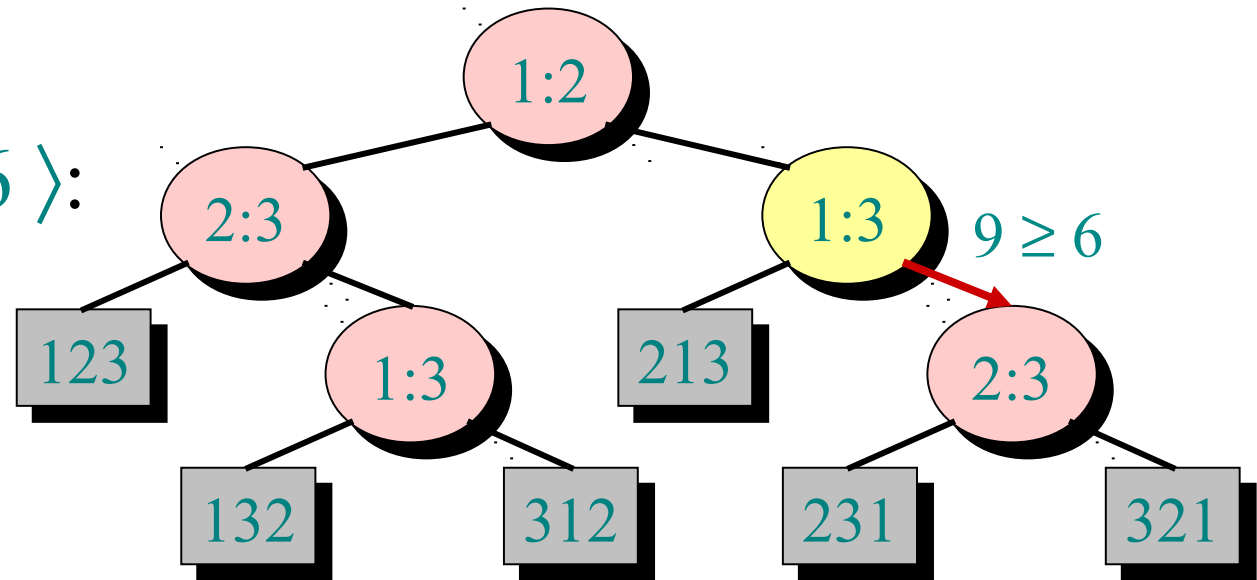
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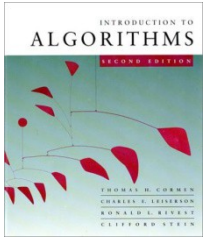
Decision-tree example

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



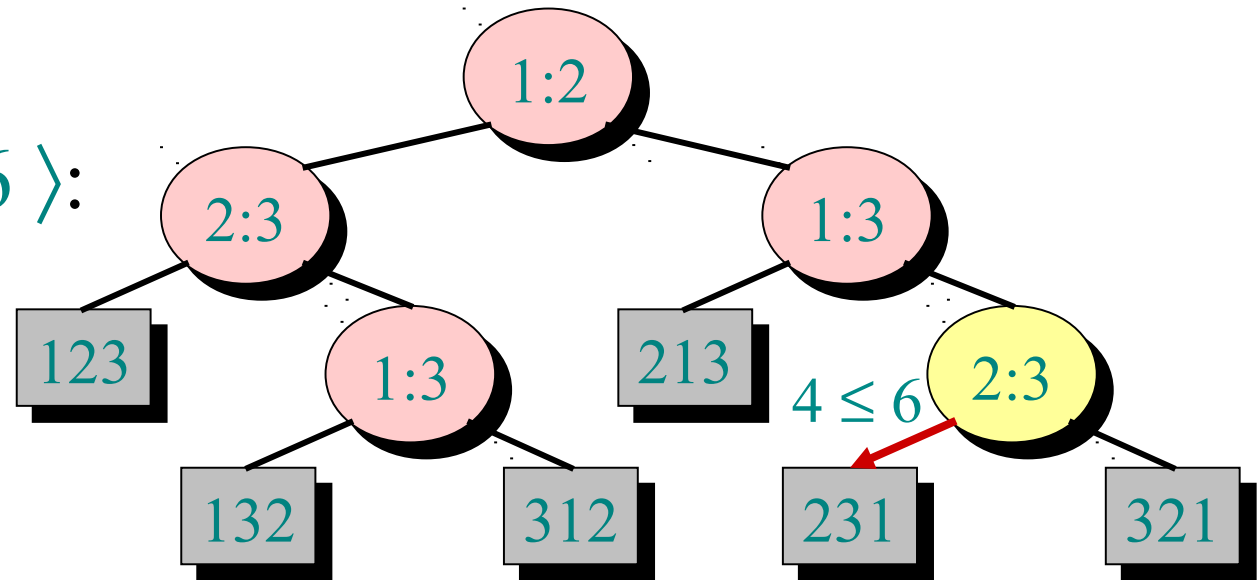
Each internal node is labeled $i:j$ for $i, j \in \{1, 2, \dots, n\}$.

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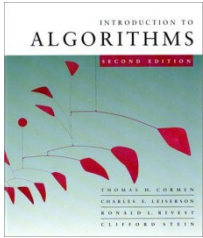
Decision-tree example

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



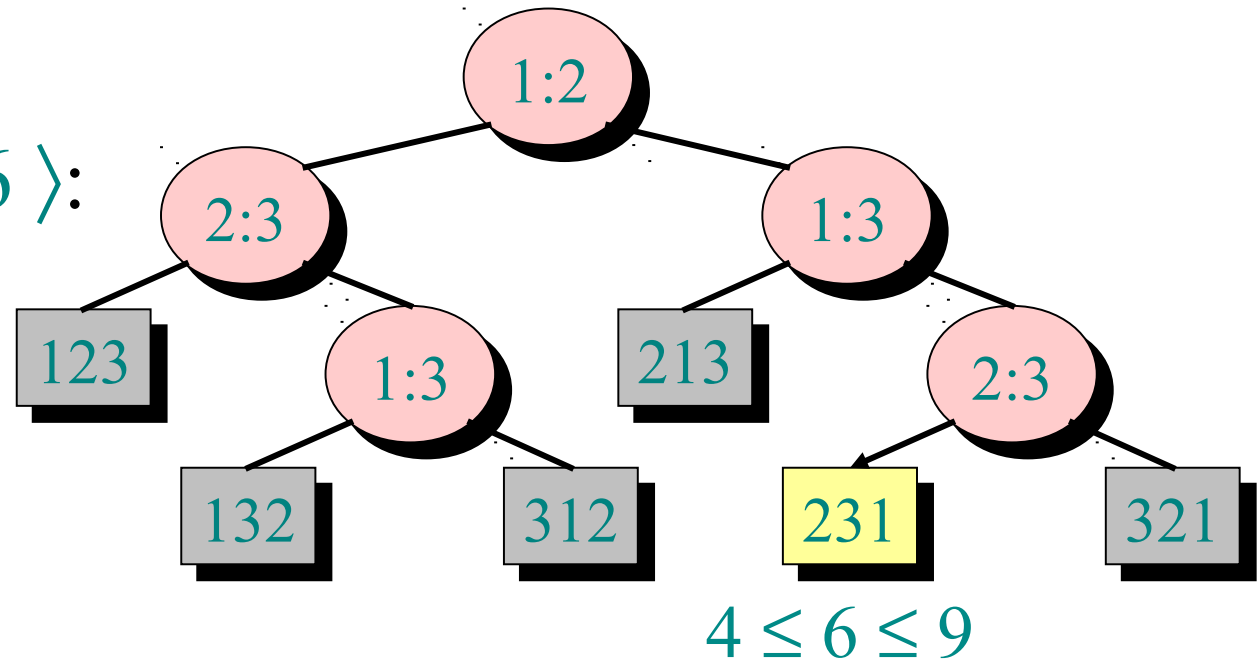
Each internal node is labeled $i:j$ for $i, j \in \{1, 2, \dots, n\}$.

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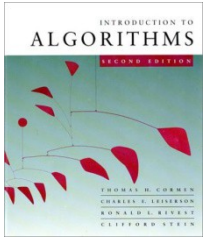


Decision-tree example

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



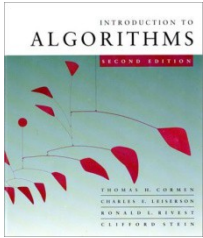
Each leaf contains a permutation $\langle \pi(1), \pi(2), \dots, \pi(n) \rangle$ to indicate that the ordering $a_{\pi(1)} \leq a_{\pi(2)} \leq \dots \leq a_{\pi(n)}$ has been established.



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.

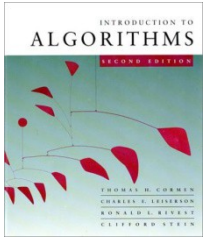


Lower bound for decision-tree sorting

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, $n! \leq 2^h$.

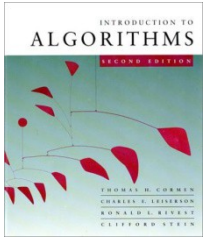
$$\begin{aligned} \therefore h &\geq \lg(n!) && (\lg \text{ is mono. increasing}) \\ &\geq \lg((n/e)^n) && (\text{Stirling's formula}) \\ &= n \lg n - n \lg e \\ &= \Omega(n \lg n). \quad \square \end{aligned}$$



Lower bound for comparison sorting

Corollary. Merge sort is asymptotically optimal comparison sorting algorithm.

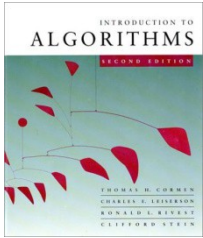




Sorting in linear time

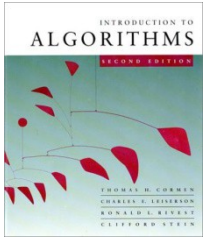
Counting sort: No comparisons between elements.

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

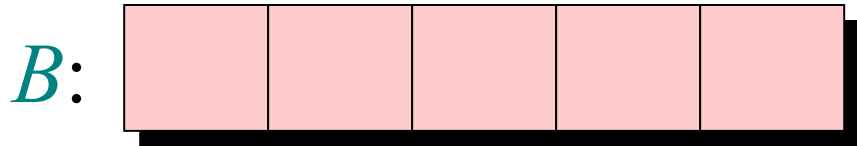
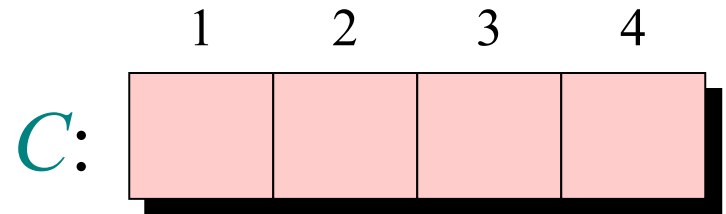
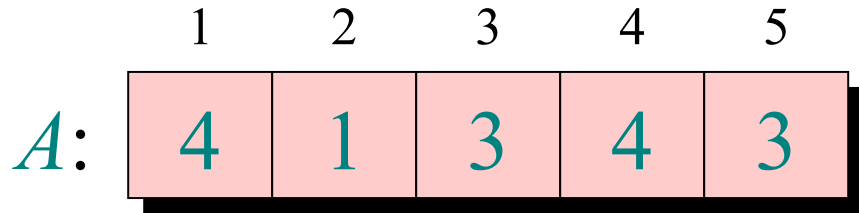


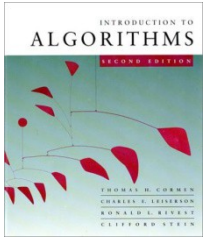
Counting sort

```
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





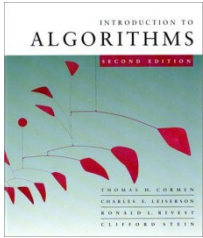
Loop 1

	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

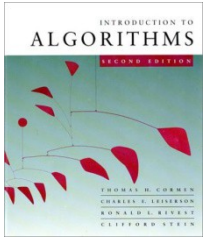
	1	2	3	4
<i>C</i> :	0	0	0	1

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

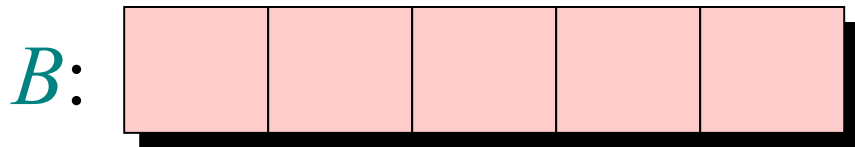
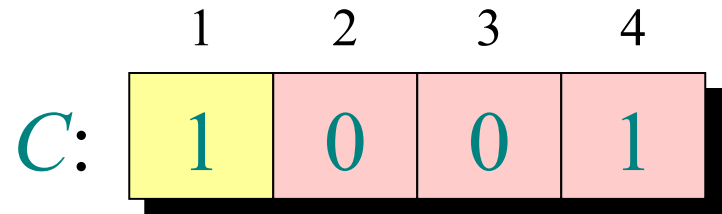
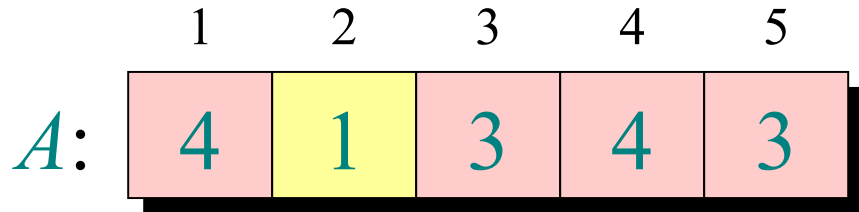
for $j \leftarrow 1$ **to** n

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

▷ $C[i] = |\{\text{key} =$



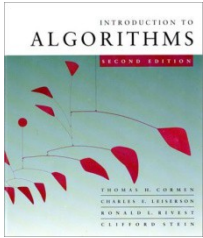
Loop 2



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

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Loop 2

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

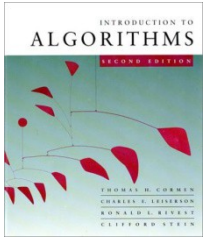
	1	2	3	4
<i>C</i> :	1	0	1	1

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

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 $i\}$

▷ $C[i] = |\{\text{key} =$



Loop 2

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

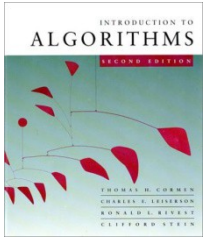
	1	2	3	4
<i>C</i> :	1	0	1	2

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

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

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

▷ $C[i] = |\{\text{key} =$



Loop 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

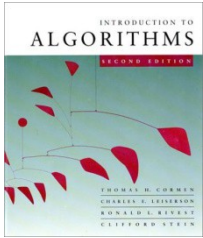
<i>B</i> :					
------------	--	--	--	--	--

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

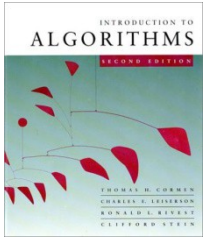
<i>B</i> :					
------------	--	--	--	--	--

	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

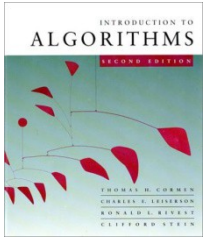
<i>B</i> :					
------------	--	--	--	--	--

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

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

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

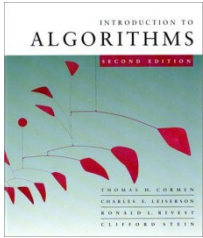
<i>B</i> :					
------------	--	--	--	--	--

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

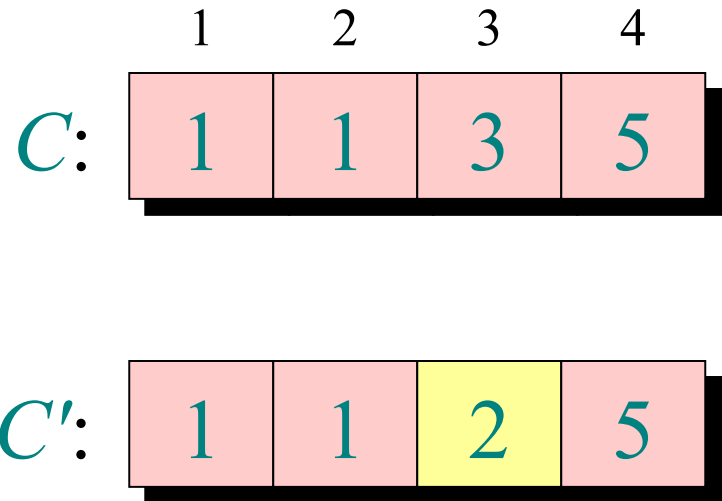
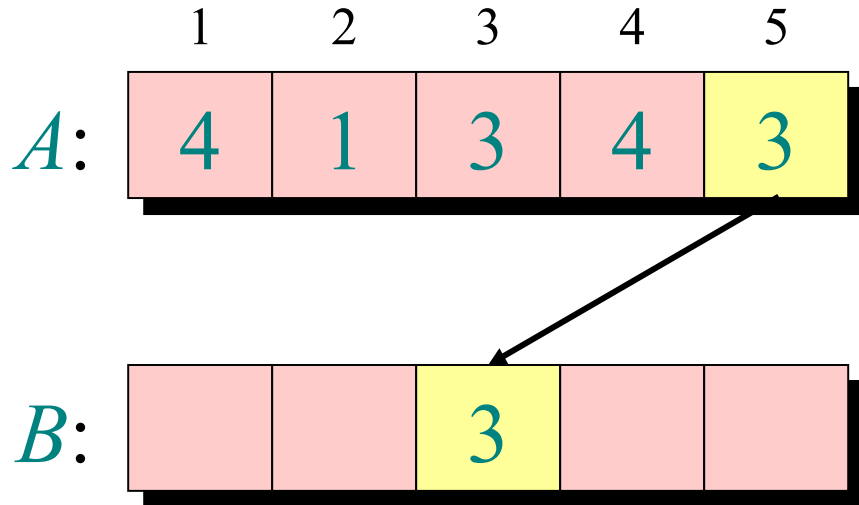
<i>C'</i> :	1	1	3	5
-------------	---	---	---	---

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

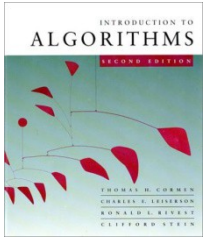
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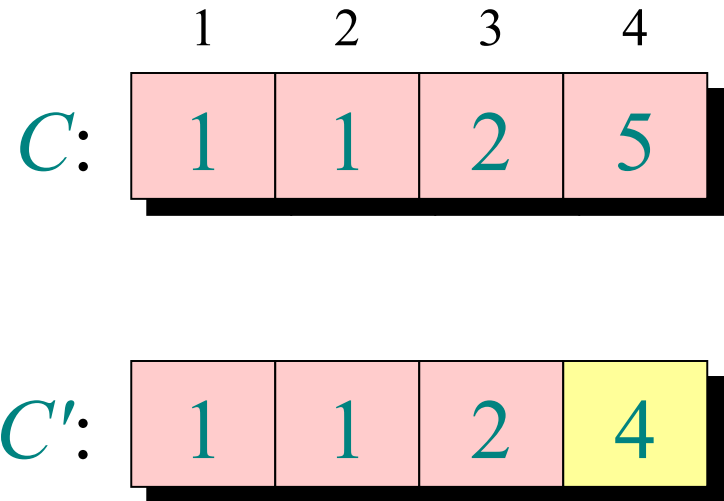
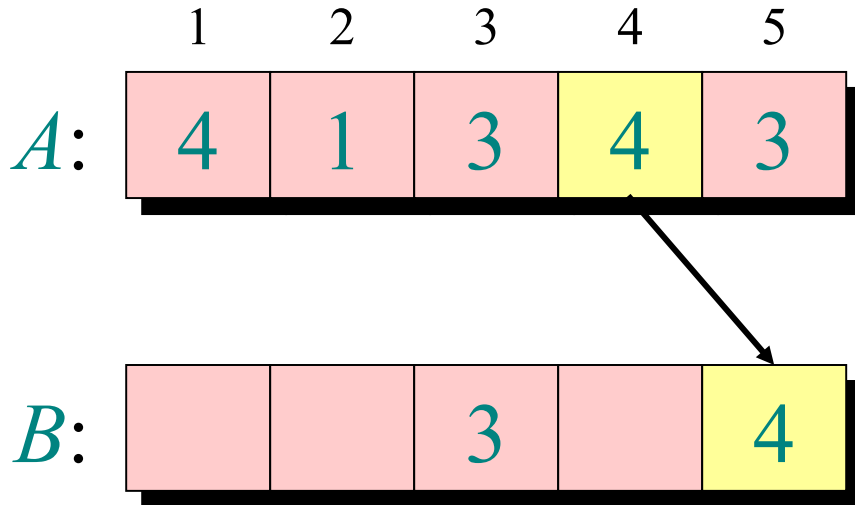
Loop 4



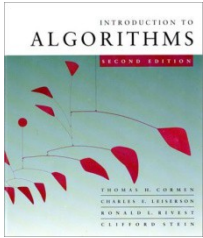
```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



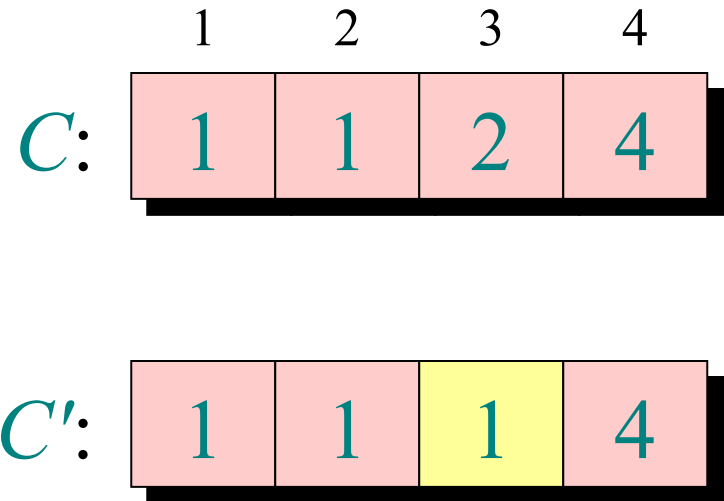
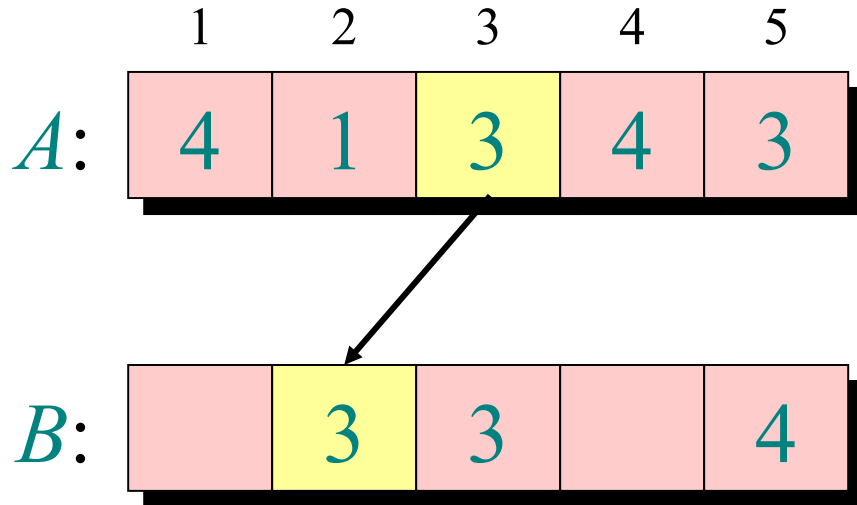
Loop 4



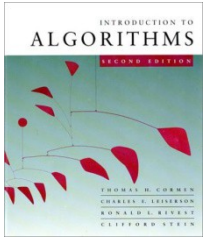
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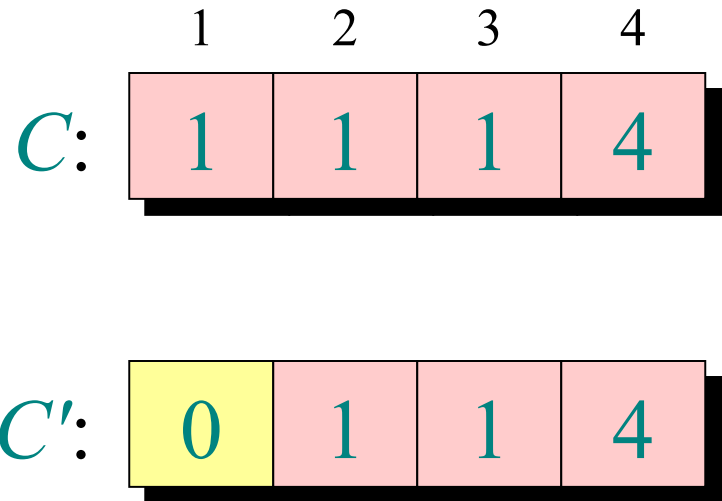
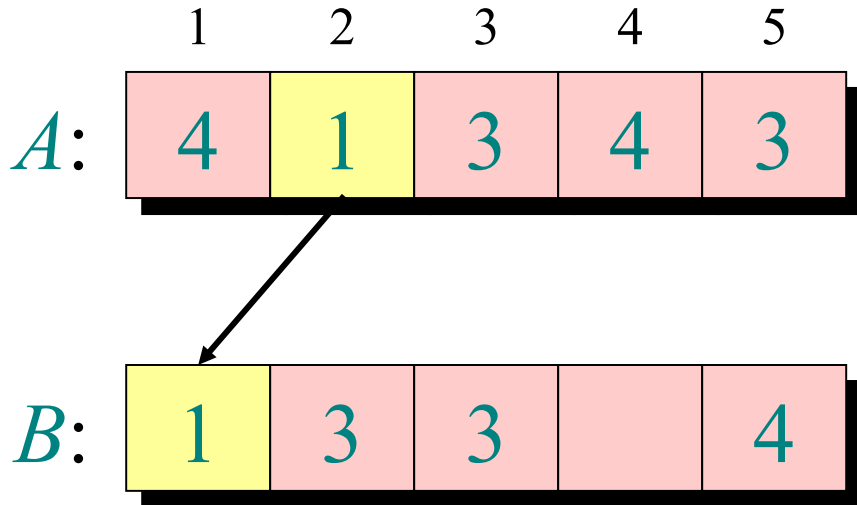
Loop 4



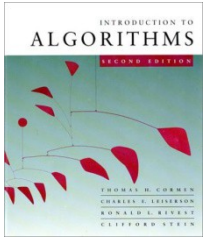
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  do  $B[C[A[j]]] \leftarrow A[j]$   
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```



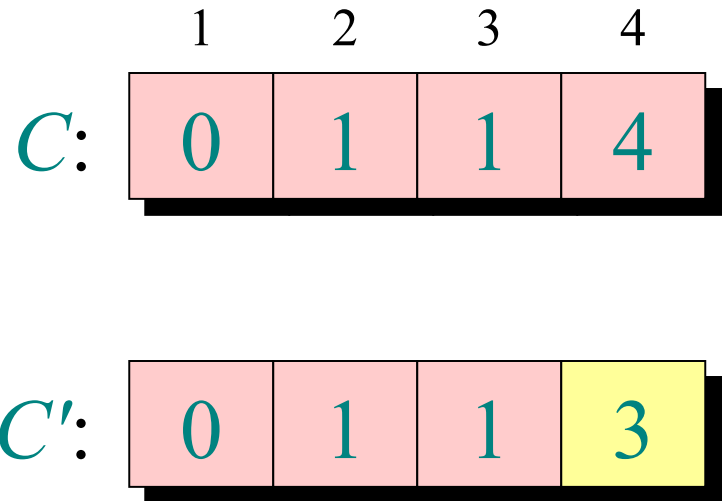
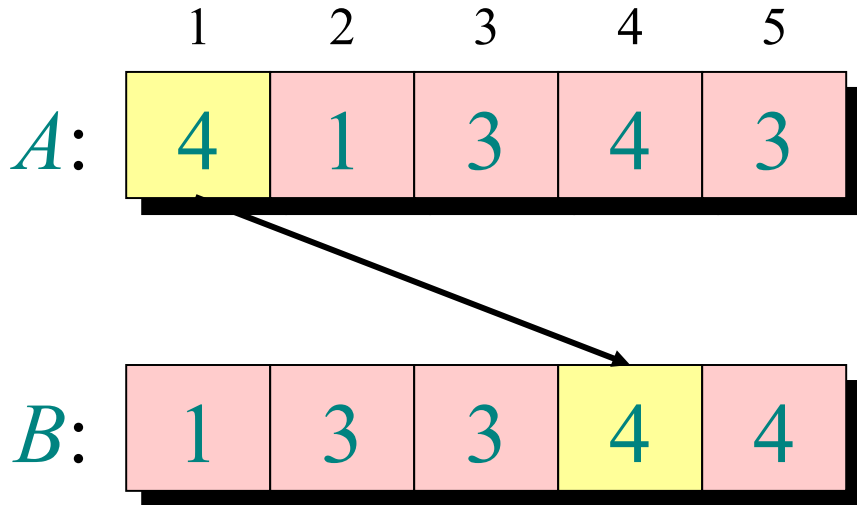
Loop 4



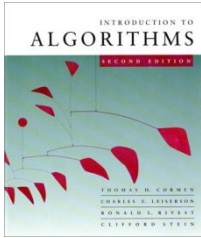
```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



Loop 4

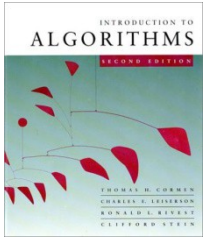


```
for  $j \leftarrow n$  downto 1  
  do  $B[C[A[j]]] \leftarrow A[j]$   
      $C[A[j]] \leftarrow C[A[j]] - 1$ 
```



Analysis

$\Theta(k)$	$\left\{ \begin{array}{l} \text{for } i \leftarrow 1 \text{ to } k \\ \quad \text{do } C[i] \leftarrow 0 \end{array} \right.$
$\Theta(n)$	$\left\{ \begin{array}{l} \text{for } j \leftarrow 1 \text{ to } n \\ \quad \text{do } C[A[j]] \leftarrow C[A[j]] + 1 \end{array} \right.$
$\Theta(k)$	$\left\{ \begin{array}{l} \text{for } i \leftarrow 2 \text{ to } k \\ \quad \text{do } C[i] \leftarrow C[i] + C[i-1] \end{array} \right.$
$\Theta(n)$	$\left\{ \begin{array}{l} \text{for } j \leftarrow n \text{ downto } 1 \\ \quad \text{do } B[C[A[j]]] \leftarrow A[j] \\ \quad \quad C[A[j]] \leftarrow C[A[j]] - 1 \end{array} \right.$
<hr/>	
$\Theta(n + k)$	



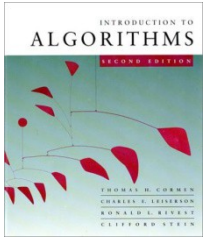
Running time

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

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

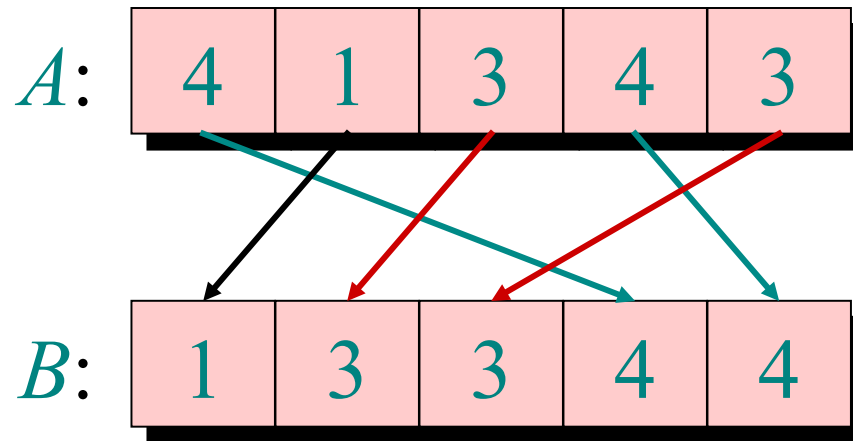
Answer:

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

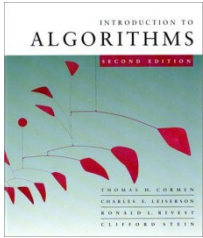


Stable sorting

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

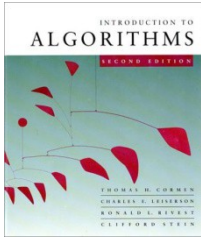


Exercise: What other sorts have this property?

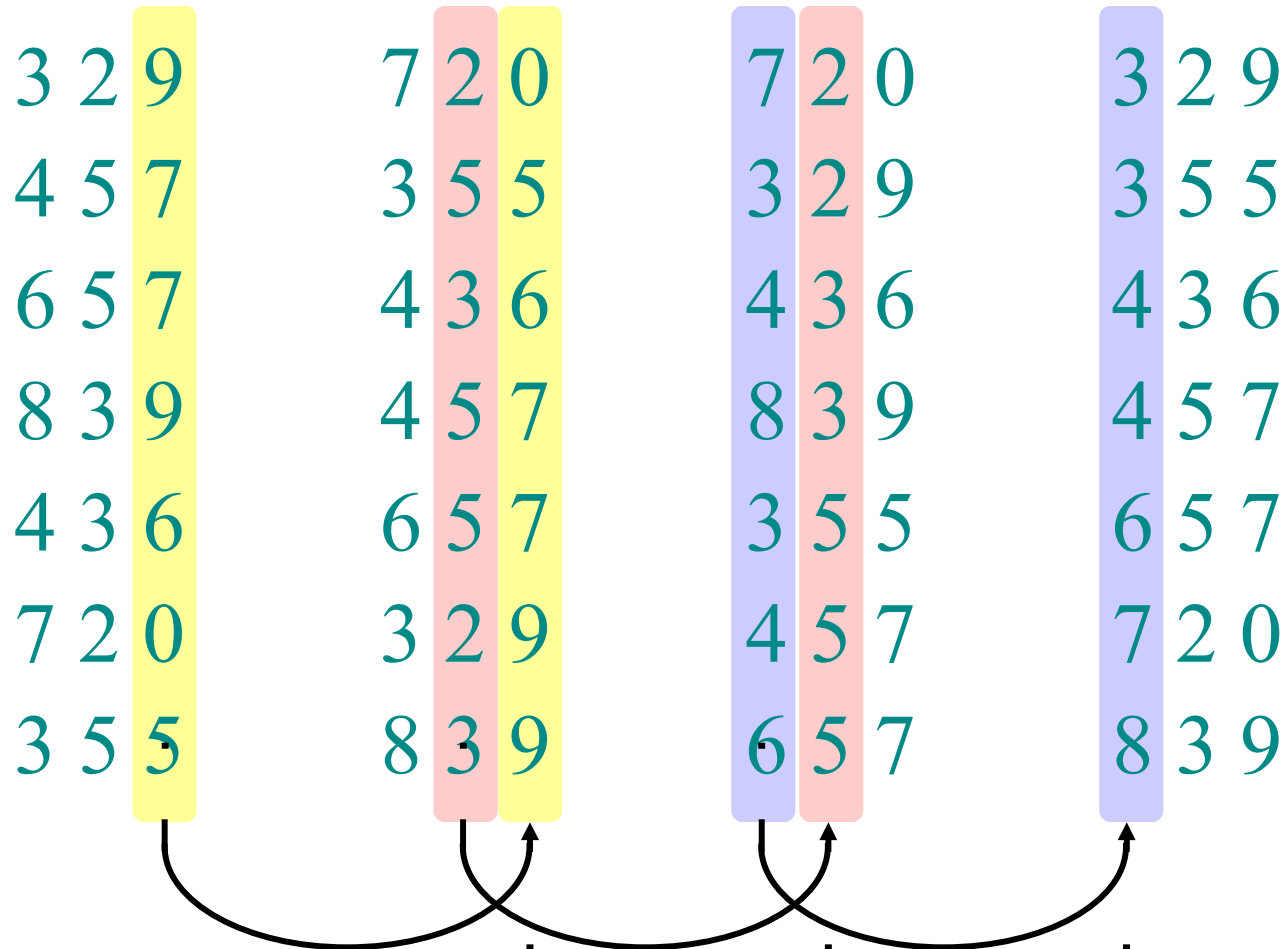


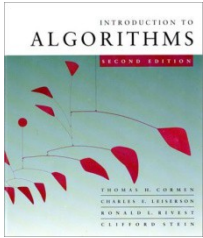
Radix sort

- *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

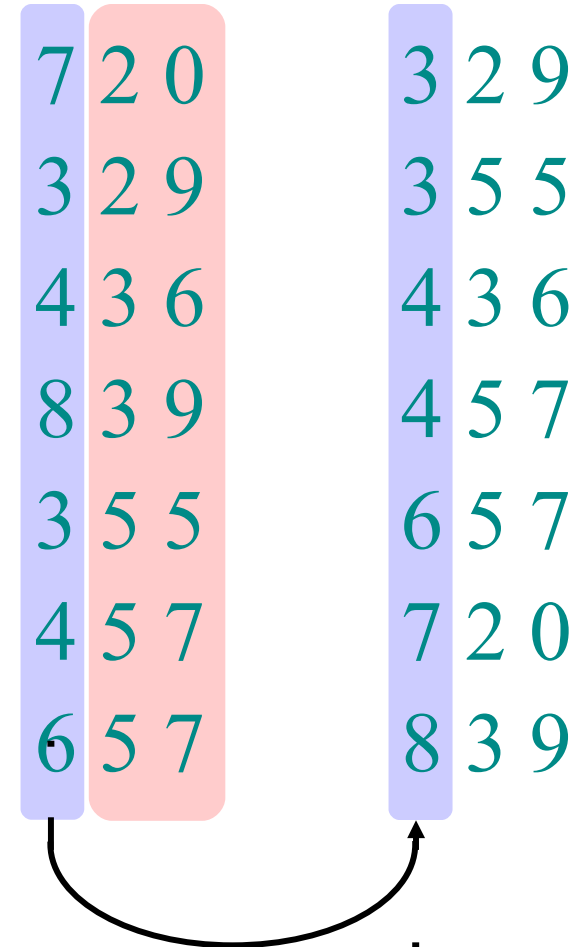


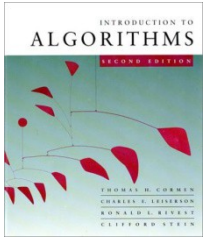


Correctness of radix sort

Induction on digit position

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

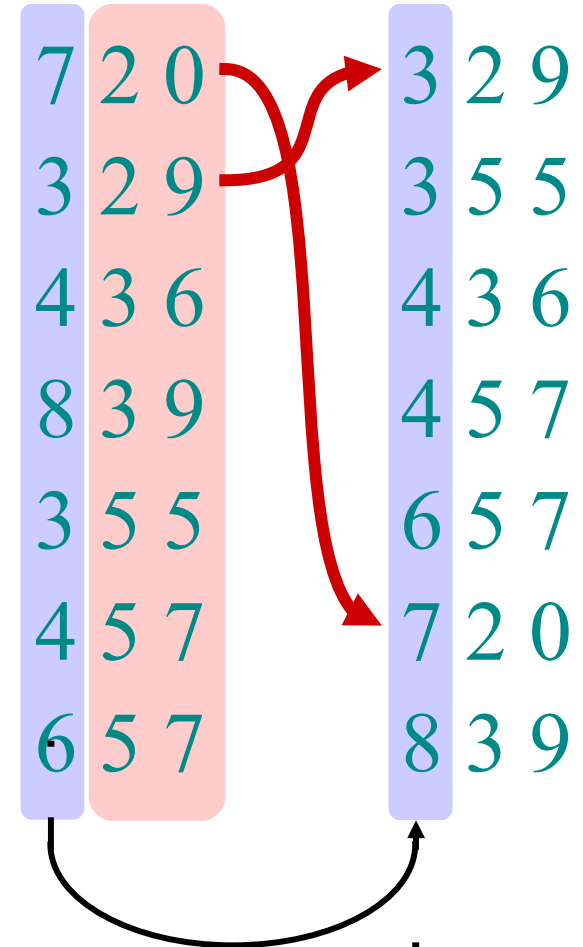


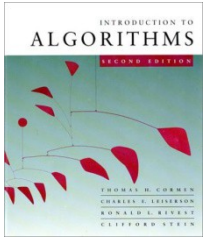


Correctness of radix sort

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.

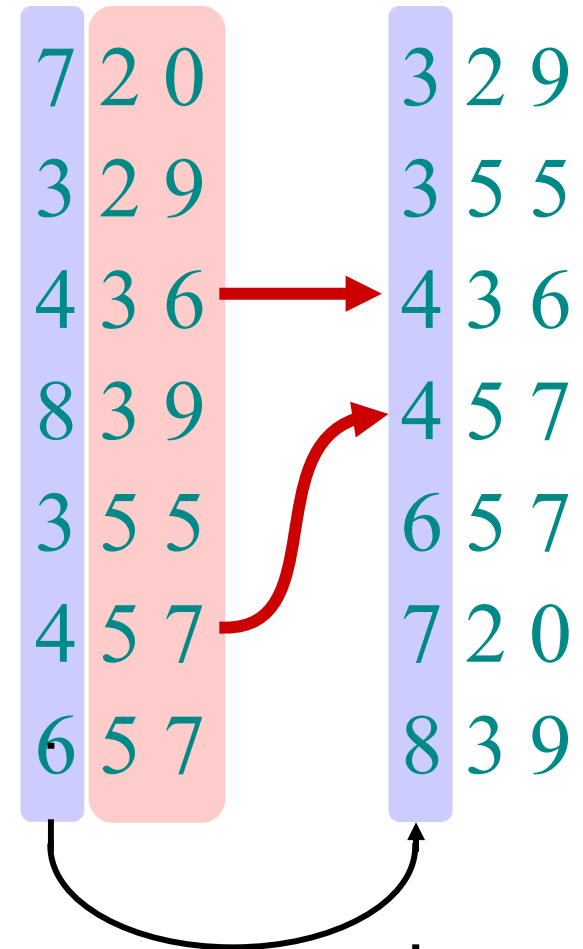


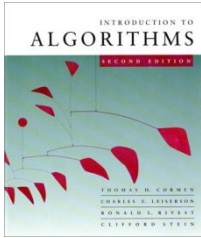


Correctness of radix sort

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

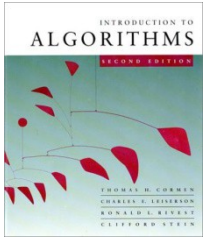
- Assume counting sort is the auxiliary stable sort.
- Sort n computer words of b bits each.
- Each word can be viewed as having b/r base- 2^r digits.

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)

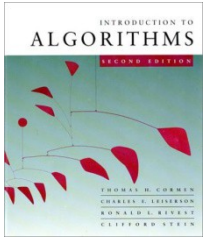
Recall: Counting sort takes $\Theta(n + k)$ time to sort n numbers in the range from 0 to $k - 1$.

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

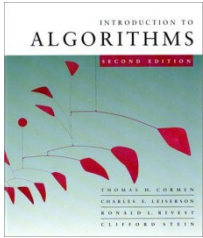
$$T(n, b) = \Theta\left(\frac{b}{r} \binom{n + 2^r}{r}\right)$$

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 ≥ 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.