

# Sorting: Lower Bounds and Linear Time

Last Revision: September 11, 2014



# Lower Bound for Sorting

- All sorting algorithms we seen so far are based on comparing elements
  - E.g., Insertion sort, Selection sort, Mergesort, Heapsort and Quicksort
- Insertion sort, Selection sort and Quicksort have worst-case running times  $\Theta(n^2)$ , while the others have worst-case running time  $\Theta(n \log n)$

## Question

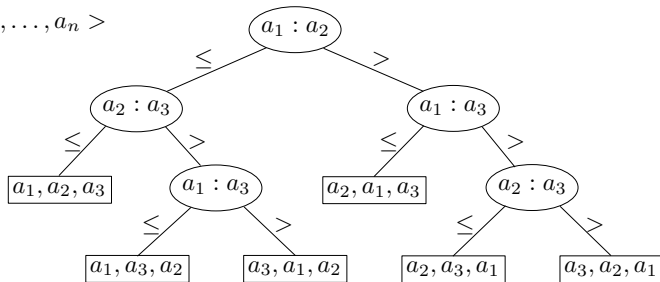
Can we do better?

## Goal

We will prove that any **comparison-based sorting algorithm** has a worst-case running time  $\Omega(n \log n)$ .

# Decision-tree Example

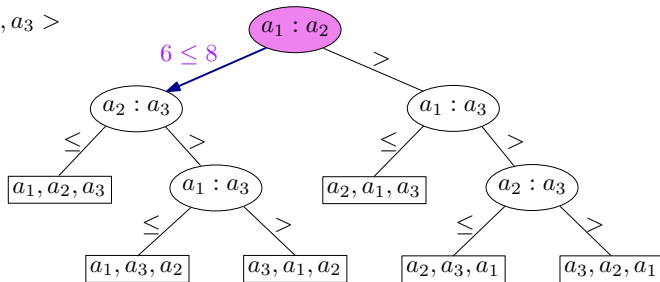
Sort  $\langle a_1, a_2, \dots, a_n \rangle$



- Each internal node is labeled  $a_i : a_j$  for  $\{1, 2, \dots, n\}$ 
  - The left subtree shows subsequent comparisons if  $a_i \leq a_j$
  - The right subtree shows subsequent comparisons if  $a_i > a_j$
- Each leaf corresponds to an input ordering

# Decision-tree Example

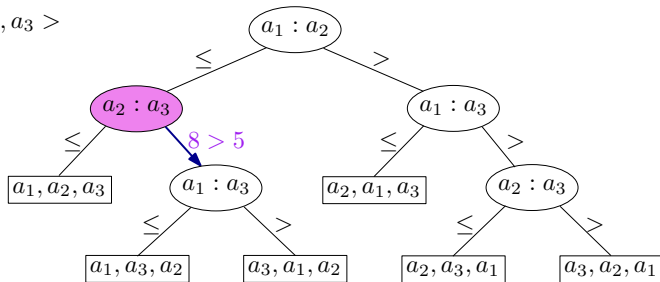
Sort  $\langle a_1, a_2, a_3 \rangle$   
 $= \langle 6, 8, 5 \rangle$ :



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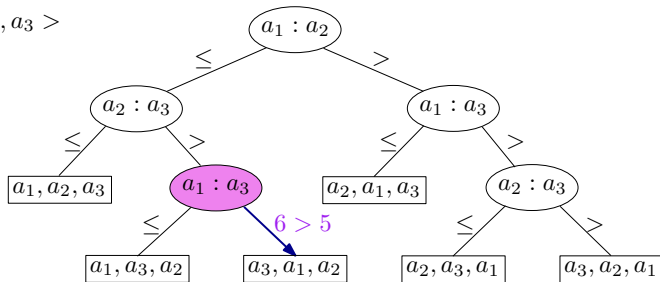
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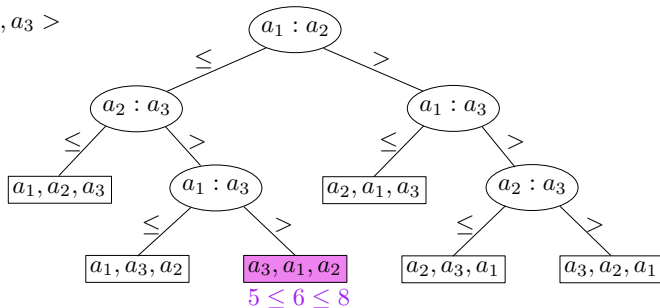
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A decision tree can model the execution of **any** comparison-based sorting algorithm

- One tree for each input size  $n$
- Worst-case running time = height of tree



# Lower Bound for Sorting

## Theorem

*Any comparison-based sorting algorithm requires  $\Omega(n \log n)$  comparisons in the worst case.*

## Proof.

- A decision tree to sort  $n$  elements must have at least  $n!$  leaves, since each of the  $n!$  orderings is a possible answer.
- A binary tree of height  $h$  has at most  $2^h$  leaves
- Thus,  $n! \leq 2^h$   
 $\Rightarrow h \geq \log n! = \Omega(n \log n)$  (Stirling's approximation)



## Corollary

*Heapsort and merge sort are asymptotically optimal comparison-based sorting algorithms.*

# Lower Bound for Average Running Time of Sorting

- We just proved that worst case number of comparisons used is  $\Omega(n \log n)$
- Suppose that each of the  $n!$  input permutations is equally likely. What can be said about the average case running time?

*Note: Average is taken by adding up individual running time of algorithm on each possible input and dividing total by  $n!$ .*

## Theorem

*When all input permutations are equally likely, any comparison-based sorting algorithm requires  $\Omega(n \log n)$  comparisons on average.*

# Lower Bound for Average Running Time of Sorting

## Theorem

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## Proof.

- The *External Path Length (EPL)* of a tree is the sum over all leaves of the tree, of the length of the paths from the root to the leaves.
- Average number of comparisons used by a sorting algorithm is EPL of its associated comparison tree divided by  $n!$ .
- The EPL of a binary tree with  $m$  leaves is at least  $m \log_2 m + O(m)$ .
- The comparison tree has  $m = n!$  leaves
  - $\Rightarrow$  its external path length is  $n! \log_2 n! + O(n!)$
  - $\Rightarrow$  average number of comparisons used is  $\log_2 n! + O(1)$ .
- We already saw  $\log_2 n! = \Omega(n \log n)$ .



# Can we do better?

Are there sorting algorithms which are not comparison-based?  
Can they beat the  $\Omega(n \log n)$  lower bound?

- Counting sort
  - Assumes items are in set  $\{1, 2, \dots, k\}$ .
  - Is a *stable* sort (defined soon).
- Radix sort
  - Assumes items are stored in fixed size words using finite alphabet

# Counting Sort

## Counting-sort( $A, B, k$ )

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

**Output:**  $B[1 \dots n]$ , sorted

let  $C[1 \dots k]$  be a new array;

**for**  $i \leftarrow 1$  **to**  $k$  **do**

$C[i] \leftarrow 0$ ;

**end**

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

$C[A[j]] \leftarrow C[A[j]] + 1$ ; //  $C[i] = |\{\text{key} = i\}|$

**end**

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

$C[i] \leftarrow C[i] + C[i - 1]$ ; //  $C[i] = |\{\text{key} \leq i\}|$

**end**

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

$B[C[A[j]]] \leftarrow A[j]$ ;

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**end**

## Example: Counting Sort

	1	2	3	4	5
$A$	4	2	1	4	2

	1	2	3	4
$C$				

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<i>B</i>	1		2		4
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<i>C'</i>	0	2	3	4
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# Analysis

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

**Output:**  $B[1 \dots n]$ , sorted

let  $C[1 \dots k]$  be a new array;

**for**  $i \leftarrow 1$  **to**  $k$  **do**

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**end**

Total:  $O(n + k)$

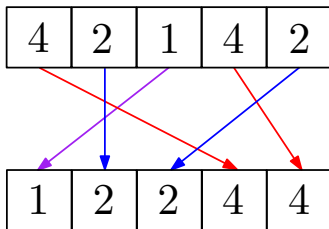
If  $k = O(n)$ , then counting sort takes  $O(n)$  time.

- But didn't we prove that sorting must take  $\Omega(n \log n)$  time?
- No, actually we proved that any comparison-based sorting algorithm takes  $\Omega(n \log n)$  time.
- Note that counting sort is *not* a comparison-based sorting algorithm.
- In fact, it makes no comparisons at all!

# Stable Sorting

Counting sort is a **stable** sort

- it preserves the input order among equal elements.

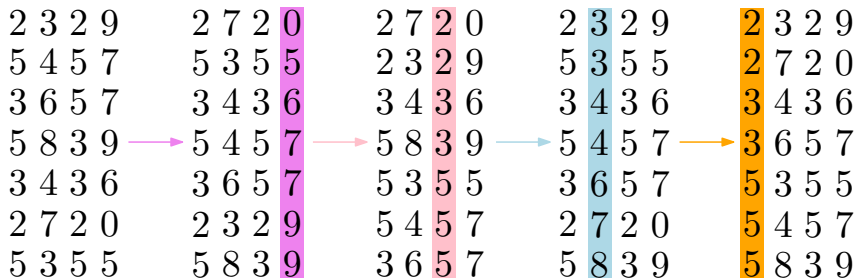


## Exercise

What other sorts have this property?

# Radix Sort

- Sort on *least significant* digit first using stable sort

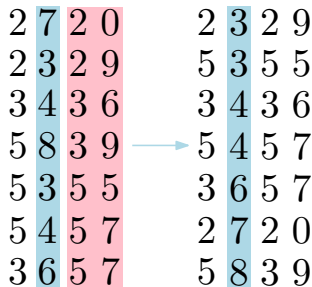




# Radix Sort: Correctness

## *Induction on digit position*

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

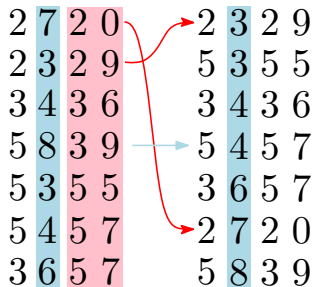


2	7	2	0		2	3	2	9
2	3	2	9		5	3	5	5
3	4	3	6		3	4	3	6
5	8	3	9	→	5	4	5	7
5	3	5	5		3	6	5	7
5	4	5	7		2	7	2	0
3	6	5	7		5	8	3	9

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## *Induction on digit position*

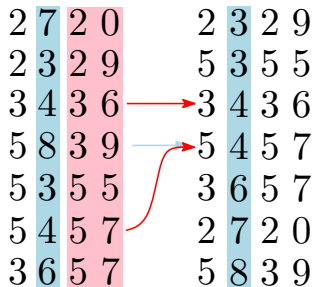
- Assume that the numbers are sorted by their low-order  $i - 1$  digits
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  - Two numbers that differ on digit  $i$  are correctly sorted by their low-order  $i$  digits



# Radix Sort: Correctness

## *Induction on digit position*

- Assume that the numbers are sorted by their low-order  $i - 1$  digits
- Sort on digit  $i$ 
  - Two numbers that differ on digit  $i$  are correctly sorted by their low-order  $i$  digits
  - Two numbers equal on digit  $i$  are put in the same order as the input  $\Rightarrow$  correctly sorted by their low-order  $i$  digits



# Radix Sort: Running Time & Application

## Lemma

*Given  $n$   $d$ -digit numbers in which each digit can take on up to  $k$  possible values, radix sort correctly sorts these numbers in  $O(d(n + k))$  time if the stable sort it uses takes  $O(n + k)$  time.*

## Application:

Sorting numbers in the range from 0 to  $n^b - 1$ , where  $b$  is a constant

- $b \log n$  bits for each number
- each number can be viewed as having  $O(b)$  digits of  $\log n$  bits each
- running time is  $O(d(n + k)) = O(b(n + 2^{\log n})) = O(bn)$
- since  $b$  is a constant, the running time is  $O(n)$ .