

Lecture 12: Lower Bound for Sorting, Countingsort, Radixsort

COMS10007 - Algorithms

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Can we sort faster than $O(n \log n)$ time?

Recall: Fastest runtime of any sorting algorithm seen is $O(n \log n)$

Can we sort faster?

- For example in $O(n \log \log n)$ time?
- Or even $O(n)$ time?

Yes! we can sometimes sort faster

But in general, **no**, we cannot

Example: Sort an array of length n of bits, i.e., every array element is either 0 or 1, in time $O(n)$?

- Count number of 0s n_0
- Write n_0 0s followed by $n - n_0$ 1s
- Both operations take time $O(n)$

Comparison-based Sorting

- Order is determined solely by comparing input elements
- All information we obtain is by asking “Is $A[i] \leq A[j]$?”, for some i, j , in particular, we may not inspect the elements
- Quicksort, mergesort, insertionsort, heapsort are comparison-based sorting algorithms
- Algorithm on last slide can be turned into a comparison-based algorithm. How? (restricted domain)

Lower Bound for Comparison-based Sorting

- We will prove that every comparison-based sorting algorithm requires $\Omega(n \log n)$ comparisons
- This implies that $O(n \log n)$ is an optimal runtime for comparison-based sorting

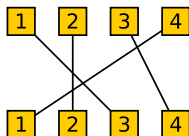
Lower Bound for Comparison-based Sorting

Problem

- A : array of length n , all elements are different
- We are only allowed to ask: Is $A[i] < A[j]$, for any $i, j \in [n]$
- How many questions are needed until we can determine the order of all elements?

Permutations

- A *bijective* function $\pi : [n] \rightarrow [n]$ is called a permutation



$$\pi(1) = 3$$

$$\pi(2) = 2$$

$$\pi(3) = 4$$

$$\pi(4) = 1$$

- A reordering of $[n]$

Lower Bound for Comparison-based Sorting (2)

How many permutations are there?

Let Π be the set of all permutations on n elements

Lemma

$$|\Pi| = n! = n \cdot (n - 1) \dots 3 \cdot 2 \cdot 1$$

Proof. The first element can be mapped to n potential elements. The second can only be mapped to $(n - 1)$ elements. etc. \square

Rephrasing our Task: Find permutation $\pi \in \Pi$ such that:

$$A[\pi(1)] < A[\pi(2)] < \dots < A[\pi(n - 1)] < A[\pi(n)]$$

Example:

Sort 3 elements by asking queries: $A[i] < A[j]$, for $i, j \in [3]$

How many Queries are needed? (worst case)

Lemma

At least 3 queries are needed to sort 3 elements.

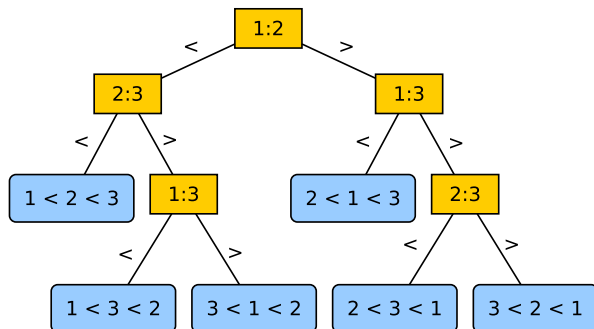
Proof. Let the three elements be a, b, c . Suppose that the first query is $a < b$ and suppose that the answer is yes. (if it is not then relabel the elements a, b, c). We are left with 3 scenarios:

$$1. a < b < c \quad 2. a < c < b \quad 3. c < a < b$$

Next we either ask $a < c$ or $b < c$. Suppose that we ask $a < c$. Then, if the answer is yes then we are left with cases 1 and 2 and we need an additional query. Suppose that we ask $b < c$. Then, if the answer is no then we are left with cases 2 and 3 and we need an additional query. □

Decision-tree Model (2)

Every Guessing Strategy is a Decision-tree

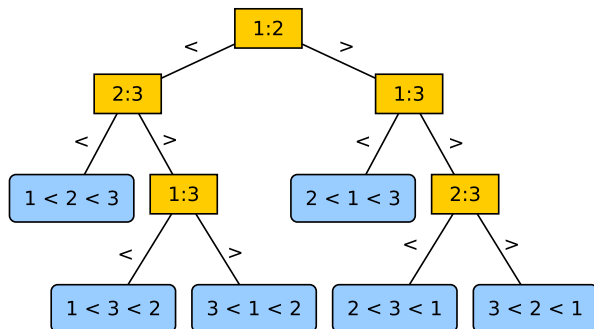


Observe:

- Every leaf is a permutation
- An execution is a root-to-leaf path

Decision-tree Model (2)

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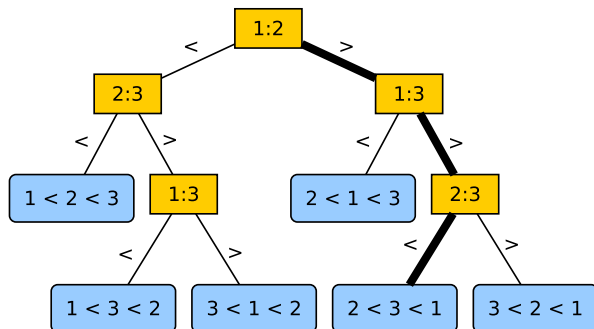


Observe:

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Decision-tree Model (2)

Every Guessing Strategy is a Decision-tree



Observe:

- Every leaf is a permutation
- An execution is a root-to-leaf path

Lemma

Any comparison-based sorting algorithm requires $\Omega(n \log n)$ comparisons.

Proof Observe that decision-tree is a binary tree. Every potential permutation is a leaf. There are $n!$ leaves. A binary tree of height h has no more than 2^h leaves. Hence:

$$\begin{aligned}2^h &\geq n! \\ h &\geq \log(n!) = \Omega(n \log n) .\end{aligned}$$



Comment: Stirling's approximation for $n!$ can be used for proving $\log(n!) = \Omega(n \log n)$

Counting Sort

Input is an array A of integers from $\{0, 1, 2, \dots, k\}$, for some integer k

Idea

- For each element x , count number of elements $< x$
- Put x directly into its position
- **Difficulty:** Multiple elements have the same value

Algorithm

```
Require: Array  $A$  of  $n$  integers from  $\{0, 1, 2, \dots, k\}$ , for some integer  $k$   
Let  $C[0 \dots k]$  be a new array with all entries equal to 0  
Store output in array  $B[0 \dots n - 1]$   
  
for  $i = 0, \dots, n - 1$  do {Count how often each element appears}  
     $C[A[i]] \leftarrow C[A[i]] + 1$   
for  $i = 1, \dots, k$  do {Count how many smaller elements appear}  
     $C[i] \leftarrow C[i] + C[i - 1]$   
for  $i = n - 1, \dots, 0$  do  
     $B[C[A[i]] - 1] \leftarrow A[i]$   
     $C[A[i]] \leftarrow C[A[i]] - 1$   
return  $m$ 
```

- Last loop processes A from right to left
- $C[A[i]]$: Number of *smaller* elements than $A[i]$
- Decrementing $C[A[i]]$: Next element of value $A[i]$ should be left of the current one

Counting Sort: Example

Example: $n = 8$, $k = 5$

	0	1	2	3	4	5	6	7
A	2	5	3	0	2	3	0	3

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C	2	2	4	7	7	8

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B								

```
for  $i = n - 1, \dots, 0$  do  
   $B[C[A[i]] - 1] \leftarrow A[i]$   
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for  $i = n - 1, \dots, 0$  do  
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```

Runtime:

$$O(n) + O(k) + O(n) = O(n + k)$$

- Counting Sort has runtime $O(n)$ if $k = O(n)$
- This beats the lower bound for comparison-based sorting

```
for  $i = 0, \dots, n - 1$  do
     $C[A[i]] \leftarrow C[A[i]] + 1$ 
for  $i = 1, \dots, k$  do
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for  $i = n - 1, \dots, 0$  do
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```

Stable? In-place? Yes, it is stable (important!) No, not in-place

Correctness Loop Invariant

Radix Sort

Input is an array A of d digits integers, each digit is from the set $\{0, 1, \dots, b - 1\}$

Examples

- $b = 2, d = 5$. E.g. 01101 (binary numbers)
- $b = 10, d = 4$. E.g. 9714

Idea

- Iterate through the d digits
- Sort according to the current digit

Radix Sort (2)

Radix Sort Algorithm

```
for  $i = 1, \dots, d$  do  
    Use a stable sort algorithm to  
    sort array  $A$  on digit  $i$ 
```

(least significant digit is digit 1)

Example

329		720		720		329
457		355		329		355
657		436		436		436
839	→	457	→	839	→	457
436		657		355		657
720		329		457		720
355		839		657		839

Analysis

Lemma

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

Proof Runtime is obvious. Correctness follows by induction on the columns being sorted. □

Observe: If $d = O(1)$ and $b = O(n)$ then the runtime is $O(n)$!