Retrieving Data and Sorting Advanced Programming and Algorithmic Design

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a.a. 2018/2019



Retrieving Data

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 $A = \langle a_1, \dots, a_n \rangle$ contains some data, e.g., patient records

Each element is associated to an identifier, A[i].id, e.g., SSN

How to find the data associated to the identifier id_1 ?

A Naïve Solution and Outlook

Scan all the database searching for $A[i].id = \mathrm{id}_1$

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What is the asymptotic complexity in terms of big-O?

A Naïve Solution and Outlook

Scan all the database searching for $A[i].id = id_1$

What is the asymptotic complexity in terms of big-O? O(n)

Can we do better?

Hint: How do you search a page in a book? Why?

A Better Technique: Dichotomic Search

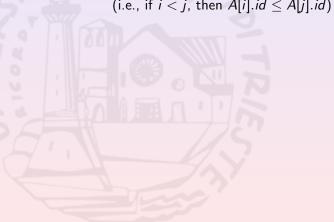
If $A = \langle a_1, \ldots, a_n \rangle$ is sorted w.r.t. the id's...



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(i.e., if i < j, then $A[i].id \le A[j].id$)



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(i.e., if i < j, then $A[i].id \le A[j].id$)

Look at element in the middle A[n/2]

If
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(i.e., if
$$i < j$$
, then $A[i].id \le A[j].id$)

Look at element in the middle A[n/2]if $A[n/2].id = id_1$ Done!

A Better Technique: Dichotomic Search

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Look at element in the middle A[n/2]

if
$$A[n/2].id = id_1$$

Done!

if
$$A[n/2].id > id_1$$

Focus on the 1st half A, i.e, $\langle a_1, \ldots, a_{n/2-1} \rangle$

A Better Technique: Dichotomic Search

If $A = \langle a_1, \ldots, a_n \rangle$ is sorted w.r.t. the id's...

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Done!

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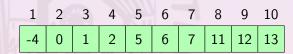
if
$$A[n/2].id > id_1$$

Focus on the 1st half A , i.e, $\langle a_1, \ldots, a_{n/2-1} \rangle$

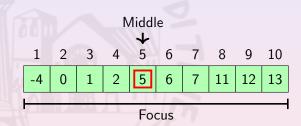
if
$$A[n/2].id < id_1$$

Focus on the 2nd half A , i.e, $a_{n/2+1}, \ldots, a_n > 0$

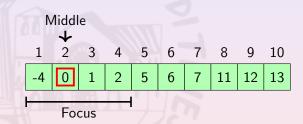
Repeat until A is not empty

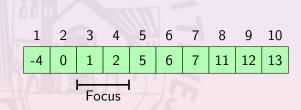


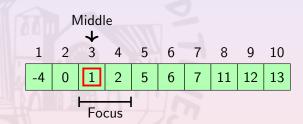


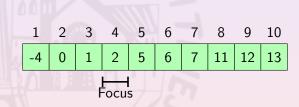












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Search for 2 in < -4, 0, 1, 2, 5, 6, 7, 11, 12, 13 >.

Found: A[4] = 2

Dichotomic Search: Pseudo-Code and Complexity

```
def di_find(A, a):
     (1, r) \leftarrow (1, |A|)
     while r > 1:
          m \leftarrow (1+r)/2
           if A[m]==a:
                return m
          endif
          if A[m]>a:
                r \leftarrow m-1
           else
                I \leftarrow m+1
           endif
     endwhile
     return 0
```

enddef

Sorting

Retrieving Data

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At each iteration, l - r is halved.

So, if $|A| \leq 2^m$, di_find ends after m iterations.

The while-block takes time O(1).

The di_find 's complexity is $O(\log n)$

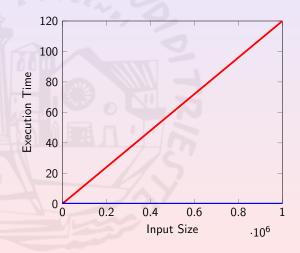
Dichotomic Search vs Linear Search: Experiments

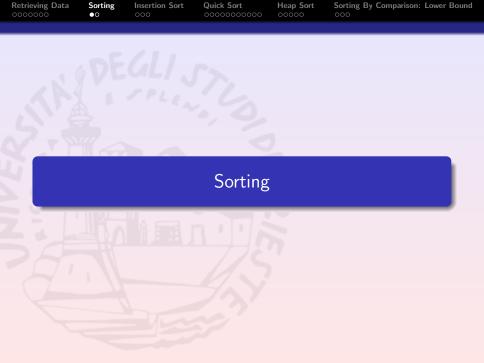
Execution time per 1×10^5 random searches.

Input size	Linear Search	Dichotomic Search
$1 imes 10^1$	$3.3 \times 10^{-3} \text{ s}$	$3.2 \times 10^{-3} \text{ s}$
1×10^2	$1.4 \times 10^{-2} \text{ s}$	$4.3 \times 10^{-3} \text{ s}$
1×10^3	$1.2 \times 10^{-1} \text{ s}$	$5.9 imes 10^{-3}$ s
1×10^4	1.2 s	$7.8 \times 10^{-3} \text{ s}$
$1 imes 10^5$	1.2×10^1 s	8.7×10^{-3} s
$1 imes 10^6$	$1.2 imes 10^2$ s	$1.2 \times 10^{-2} \text{ s}$

Dichotomic Search vs Linear Search: Experiments

Execution time per 1×10^5 random searches.





The Sorting Problem

Input: An array A of numbers

Output: The array A sorted i.e., if i < j, then $A[i] \le A[j]$

E.g.,

The Sorting Problem

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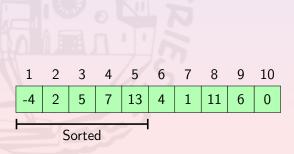
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Retrieving Data

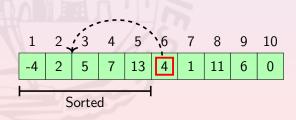
Any idea for a sorting algorithm? What is expected complexity?

If the first fragment of the array is already sorted



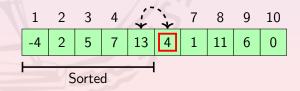
If the first fragment of the array is already sorted

we can "enlarge" it by inserting next element



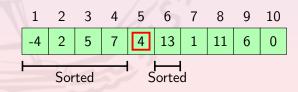
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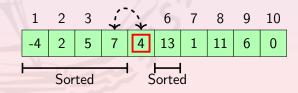
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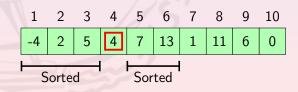
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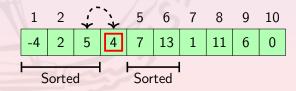
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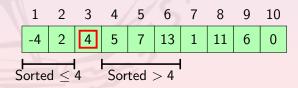


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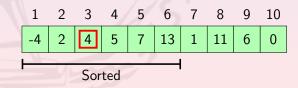


If the first fragment of the array is already sorted we can "enlarge" it by inserting next element by swapping it and the previous one in the array until the previous one (if exists) is greater than it



Insertion Sort: Intuition

If the first fragment of the array is already sorted we can "enlarge" it by inserting next element by swapping it and the previous one in the array until the previous one (if exists) is greater than it



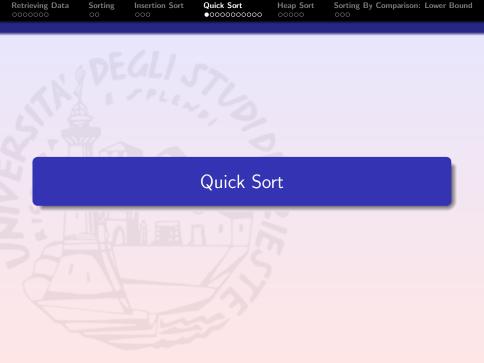
Insertion Sort: Code and Complexity

```
def insertion_sort(A):
   for i in 2.. | A | :
       while (j>1) and
               A[i] < A[i-1]:
          swap(A, j-1, j)
          i\leftarrow i-1
       endwhile
   endfor
enddef
```

Retrieving Data

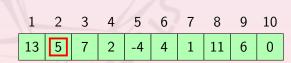
It iterates O(i) and $\Omega(1)$ times for all $i \in [2, n]$ $\sum_{i=2}^{n} O(i) * O(1) = O(\sum_{i=2}^{n} i)$ $= O(n^2)$ $\sum_{i=1}^{n}\Omega(1)*\Omega(1)=\Omega(\sum_{i=1}^{n}1)$ $=\Omega(n)$

The while-loop block costs $\Theta(1)$



Quick Sort: Intuition

Select one element of the *A*: the **pivot**

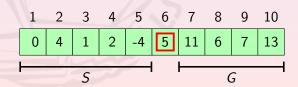


Quick Sort: Intuition

Select one element of the *A*: the **pivot**

partition A in:

- subarray S of the elements smaller or equal to the pivot
- the pivot
- subarray G of the elements greater then the pivot



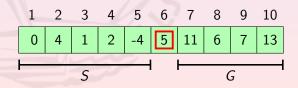
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Repeat on the subarrays having more than 1 elements



Quick Sort: Intuition (Cont'd)

At the end of every iteration of above steps:

- the elements in S stay in S if A is sorted
- the elements in G stay in G if A is sorted
- the pivot is in its "sorted" position
- S and G are shorter then A

Quick Sort: Intuition (Cont'd)

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An iteration places at least one element in the correct position

It prepares A for two recursive calls on S and G.

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Quick Sort: Pseudo-Code

Retrieving Data

```
def QUICKSORT(A, I=1, r=|A|):
     if | < r :
        p \leftarrow partition(A, I, r, I)
        QUICKSORT (A, I, p-1)
        QUICKSORT(A, p+1, r)
     endfi
enddef
```

Quick Sort: Pseudo-Code

The last recursion call is a tail recursion

```
\begin{array}{c} \textbf{def QUICKSORT}(A, \ \ | = 1, \ \ r = |A|): \\ \textbf{while} \ \ | < r: \\ p \leftarrow partition\left(A, I, r, I\right) \\ \\ \textbf{QUICKSORT}(A, I, p-1) \\ | \leftarrow p+1 \\ \textbf{endwhile} \\ \textbf{enddef} \end{array}
```

Quick Sort: Complexity

The time complexity T_Q of quick sort will be

$$T_Q(|A|) = \begin{cases} \Theta(1) & \text{if } |A| = 1 \\ T_Q(|S|) + T_Q(|G|) + T_P(|A|) & \text{otherwise} \end{cases}$$

 T_P is the complexity of **partition**

Quick Sort: Complexity

Retrieving Data

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Is the pivot selection relevant?

Quick Sort: Complexity

Retrieving Data

The time complexity T_O of quick sort will be

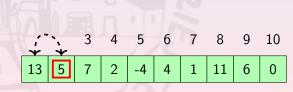
$$T_Q(|A|) = \left\{ egin{array}{ll} \Theta(1) & ext{if } |A| = 1 \ T_Q(|S|) + T_Q(|G|) + T_P(|A|) \end{array}
ight. ext{ otherwise}$$

 T_P is the complexity of partition

Is the pivot selection relevant? No, choose whatever you want

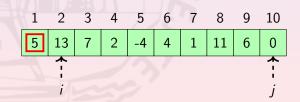
Which algorithm is the best for partition?

Switch the pivot \mathbf{p} and the first element in A



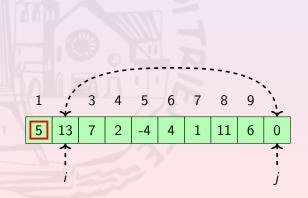
Switch the pivot \mathbf{p} and the first element in A

If A[i] > p,



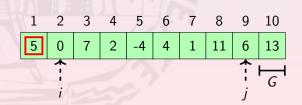
Switch the pivot \mathbf{p} and the first element in A

If A[i] > p, swap A[i] and A[j] and decrease j



Switch the pivot \mathbf{p} and the first element in A

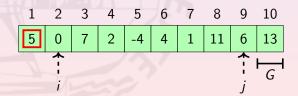
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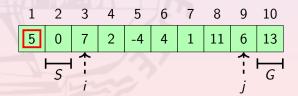
else $(A[i] \le p)$,



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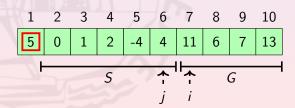


Switch the pivot \mathbf{p} and the first element in A

If A[i] > p, swap A[i] and A[j] and decrease j

else ($A[i] \leq p$), increase *i*

Repeat until $i \leq j$

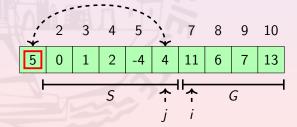


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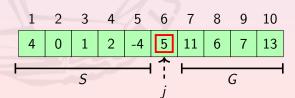
Switch the pivot \mathbf{p} and the first element in A

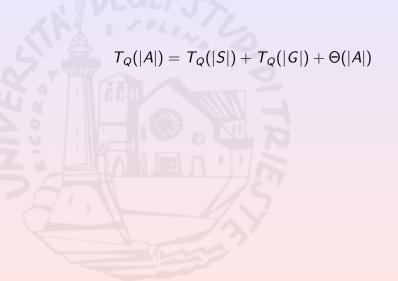
If A[i] > p, swap A[i] and A[j] and decrease j

else ($A[i] \leq p$), increase *i*

Repeat until $i \leq j$ and swap **p** and A[j]

The complexity is $\Theta(|A|)$





Retrieving Data

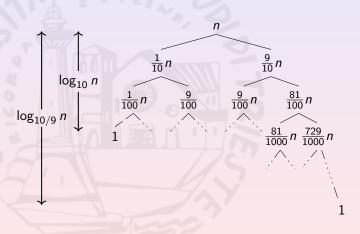
$$T_Q(|A|) = T_Q(|S|) + T_Q(|G|) + \Theta(|A|)$$

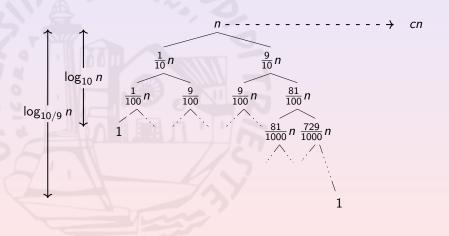
Worst Case: |G| = 0 or |S| = 0 for all recursive call.

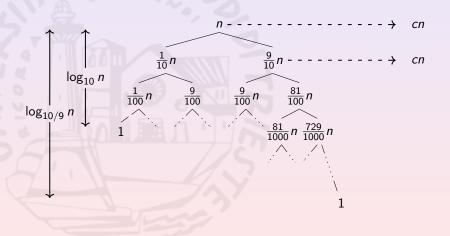
$$T_{Q}(n) = T_{Q}(n-1) + \Theta(n)$$

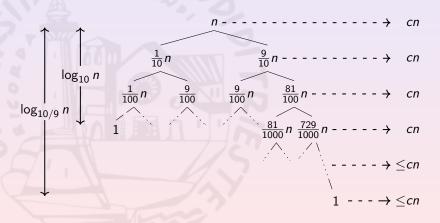
$$= \sum_{i=0}^{n} \Theta(i) = \Theta\left(\sum_{i=0}^{n} i\right)$$

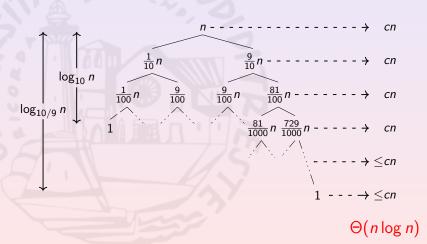
$$= \Theta(n^{2})$$











Heap Sort

Quick Sort Complexity: Average Case

"Good" and "bad" cases depend on the ordering of A

If all the permutations of A are equally likely,

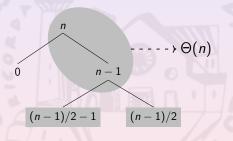
the partition has a ratio more balanced than 1/d with probability

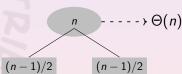
$$\frac{d-1}{d+1}$$

e.g., a partition "better" than 1/9 has probability 0.8

Quick Sort Complexity: Average Case (Cont'd)

Even if "good" and "bad" cases alternate

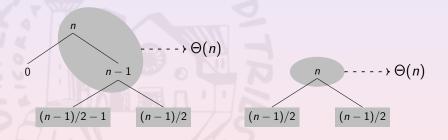




Quick Sort Complexity: Average Case (Cont'd)

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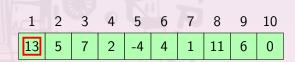
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On the average $\Theta(n \log n)$

Sorting by Searching the Maximum

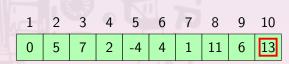
Find the maximum



Sorting by Searching the Maximum

Find the maximum

Move the maximum at the end of the array



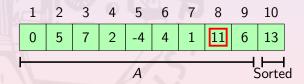
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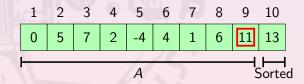
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Find the maximum

Retrieving Data

Move the maximum at the end of the array

If |A| > 1, repeat on the initial fragment of A

The complexity is $\sum_{i=1}^{|A|} \left(T_{\max}(i) + \Theta(1) \right)$

How to Find the Maximum?

By using ...

pushing the max to the right

 \Longrightarrow Bubble Sort

$$egin{aligned} \mathcal{T}(|A|) &= \sum_{i=1}^{|A|} \left(\Theta(i) + \Theta(1)
ight) \ &= \Theta(|A|^2) \end{aligned}$$

• binary heap (see here)

⇒ Heap Sort

Heap Sort: Pseudo-Code

The array-based implementation of binary heap plays a crucial role

```
def HEAPSORT(A):
    BUILD_HEAP(A)

for i ← |A| downto 2
    swap(H,1,i)

H.size ← H.size − 1
    HEAPIFY(H,1)
    endfor
enddef
```

Heap Sort: Complexity

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Building the binary heap costs $\Theta(n)$

HEAPIFY costs $O(\log i)$ per iteration and in total

$$\sum_{i=2}^n \log i \le \sum_{i=2}^n \log n \in O(n \log n)$$

The overall complexity of heap sort is $O(n \log n)$

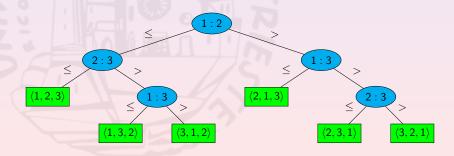


you can demonstrate that any algorithm sorting by comparison needs at least (n log n)

Sorting By Comparison: Lower Bound

The execution of a sorting-by-comparison algorithm can be modeled as a decision-tree model

Any comparison between a_i and a_j corresponds to a node which branches the computation according whether $a_i \leq a_j$ or $a_i > a_j$



Heap Sort

Sorting By Comparison: Lower Bound (Cont'd)

The decision tree's leaves are labeled by all the possible permutations of A which are n!

The height h is the maximum # of comparisons required by the algorithm

Since a binary tree has no more than 2^h leaves,

$$h \ge \log_2(n!) \in \Omega(n \log n)$$

Sorting

Heap Sort

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The lower bound for comparison-based sorting is $\Omega(n \log n)$ so there's no general algorithm using comparisono which can sort in linear time. How can we