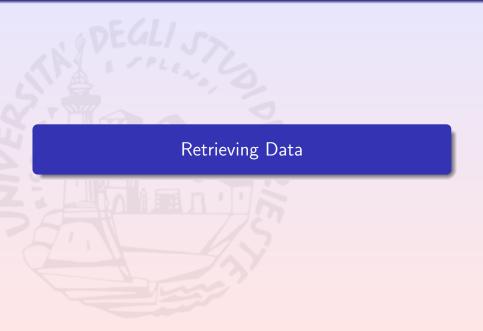
# Retrieving Data and Sorting Algorithmic Design

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a.y. 2020/2021



## Retrieving Data

 $A = \langle a_1, \dots, a_n \rangle$  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  $A[\emph{i}].\emph{id}=\mathrm{id}_1$ 

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

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

Can we do better?

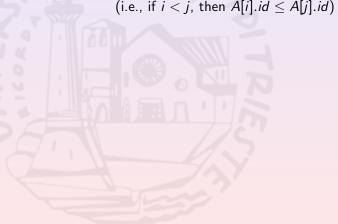
Hint: How do you search a page in a book? Why?

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

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

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

(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!

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

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$$i < j$$
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Look at element in the middle A[n/2]

if 
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Done!

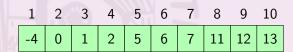
if 
$$A[n/2].id > id_1$$

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

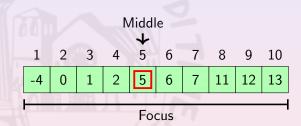
if 
$$A[n/2].id < id_1$$

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

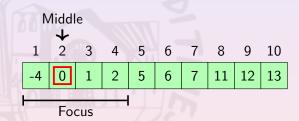
Repeat until A is not empty

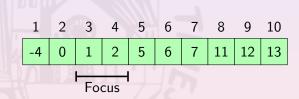


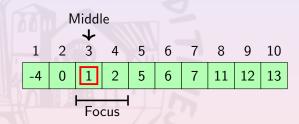














**Found:** A[4] = 2

## 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
```

0000000

At each iteration, I - r is halved.

So, if  $|A| = 2^m$ , di\_find ends after m iterations at most.

The while-block takes time  $\Theta(1)$ .

The di\_find 's complexity is

 $O(\log n)$ 

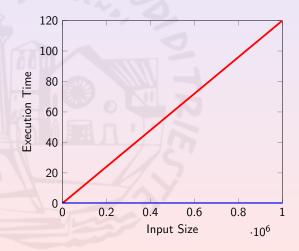
## Dichotomic Search vs Linear Search: Experiments

Execution time per  $1\times10^5$  random searches.

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

#### Dichotomic Search vs Linear Search: Experiments

Execution time per  $1 \times 10^5$  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.,

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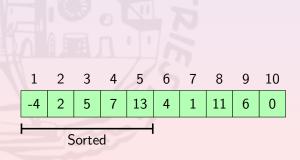
**Output:** The array A sorted i.e., if i < j, then  $A[i] \le A[j]$ 

E.g.,

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

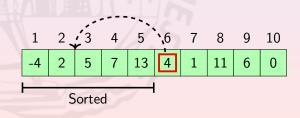


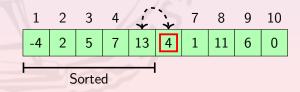
If the first fragment of the array is already sorted

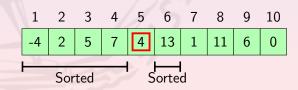


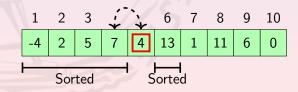
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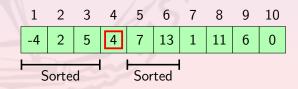
we can "enlarge" it by inserting next value in the right place

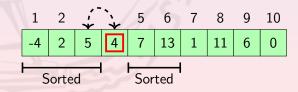




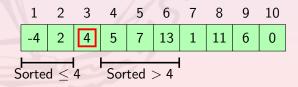






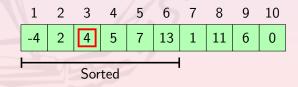


If the first fragment of the array is already sorted
we can "enlarge" it by inserting next value in the right place
by swapping it and the previous value in the array
until the previous one (if exists) is greater than it



#### Insertion Sort: Intuition

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



# Insertion Sort: Code and Complexity

```
\begin{array}{ccccc} \textbf{def} & insertion\_sort (A): \\ & \textbf{for} & i & \textbf{in} & 2...|A|: \\ & j \leftarrow i \\ & \textbf{while} & (j>1 & \textbf{and} \\ & & A[j] < A[j-1]): \\ & & swap(A,j-1,j) \\ & & j \leftarrow j-1 \\ & & \textbf{endwhile} \\ & & \textbf{endfor} \end{array}
```

enddef

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

It iterates O(i) and  $\Omega(1)$  times for all  $i \in [2, n]$ 

$$\sum_{i=2}^{n} O(i) * O(1) = O(\sum_{i=2}^{n} i)$$

$$= O(n^{2})$$

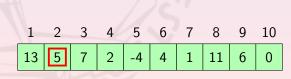
$$\sum_{i=2}^{n} \Omega(1) * \Omega(1) = \Omega(\sum_{i=2}^{n} 1)$$

$$= \Omega(n)$$



### Quick Sort: Intuition

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

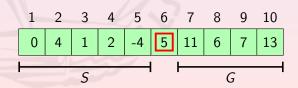


## Quick Sort: Intuition

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

#### partition A in:

- sub-array S of the values smaller or equal to the pivot
- the pivot itself
- sub-array G of the values greater than the pivot



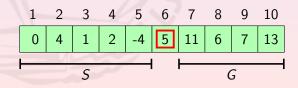
### **Quick Sort: Intuition**

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Repeat on the sub-arrays having more than 1 elements



# Quick Sort: Intuition (Cont'd)

At the end of every iteration of above steps:

- the values in S stay in S even after sorting A
- the values in G stay in G even after sorting A
- the pivot is in its "sorted" position
- S and G are shorter than A

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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

```
\label{eq:def_QUICKSORT} \begin{array}{l} \text{def} \  \, \text{QUICKSORT}(A, \ | = 1, \ r = |A|): \\ \text{if} \  \, | < r: \\ p \leftarrow \text{PARTITION}(A, |, r, |) \\ \\ \text{QUICKSORT}(A, |, p - 1) \\ \text{QUICKSORT}(A, p + 1, r) \\ \text{endfi} \\ \text{enddef} \end{array}
```

## Quick Sort: Pseudo-Code

The last recursion call is a tail recursion

# Quick Sort: Complexity

The time complexity  $T_Q$  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** 

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Is the pivot selection relevant?

# Quick Sort: Complexity

The time complexity  $T_O$  of quick sort will be

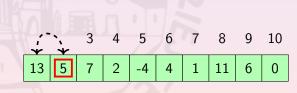
$$T_Q(|A|) = \left\{ egin{array}{ll} \Theta(1) & ext{if } |A| = 1 \ T_Q(|S|) + T_Q(|G|) + T_P(|A|) \end{array} 
ight. ext{ otherwise}$$

 $T_P$  is the complexity of partition

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

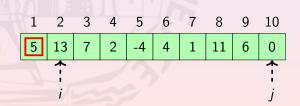
Which algorithm is the best for partition?

Switch the pivot p and the first element in A



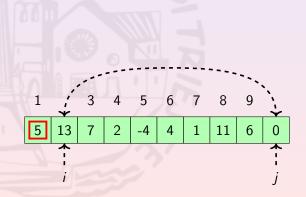
Switch the pivot p and the first element in A

If A[i] > p,



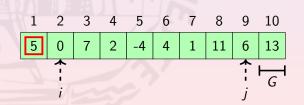
Switch the pivot p and the first element in A

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



Switch the pivot p and the first element in A

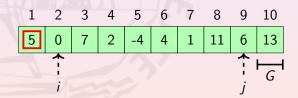
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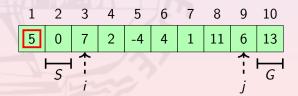
else  $(A[i] \le p)$ ,



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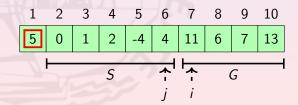


Switch the pivot p and the first element in A

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

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

Repeat until  $i \leq j$ 

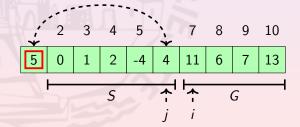


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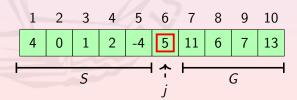
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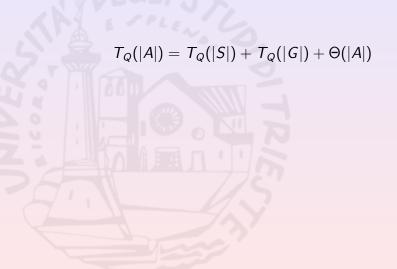
Repeat until  $i \leq j$  and swap p and A[j]

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



#### Partition: Pseudo-Code

```
def PARTITION(A, i, j, p):
    swap(A,i,p)
    (p,i) \leftarrow (i,i+1)
    while i≤j:
      if A[i]>A[p]: # if A[i] is greater than the pivot
       swap(A,i,j) # place it in G
      j \leftarrow j-1 # increase G's size
                    # otherwise
     else
       i \leftarrow i+1
                       # A[i] is already in S
      endif
    endwhile
    swap(A,p,j) # place the pivot between S and G
    return i
enddef
```



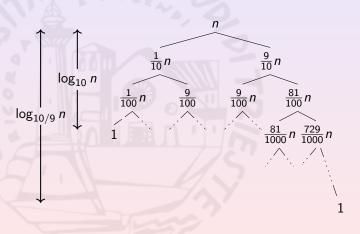
$$T_Q(|A|) = T_Q(|S|) + T_Q(|G|) + \Theta(|A|)$$

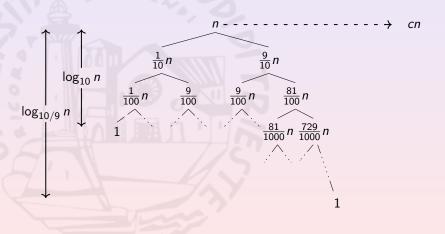
Worst Case: |G| = 0 or |S| = 0 for all recursive call.

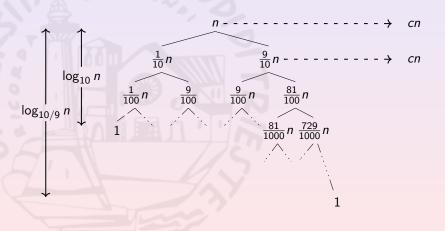
$$T_{Q}(n) = T_{Q}(n-1) + \Theta(n)$$

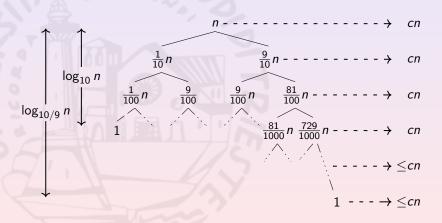
$$= \sum_{i=0}^{n} \Theta(i) = \Theta\left(\sum_{i=0}^{n} i\right)$$

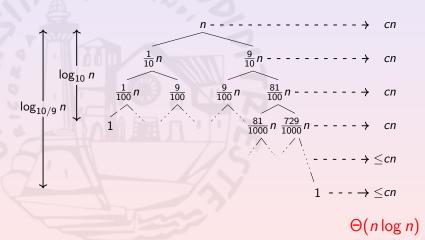
$$= \Theta(n^{2})$$











# Quick Sort Complexity: Average Case

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

If all the permutations of A are equally likely,

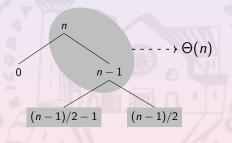
the partition has a ratio more balanced than 1/d with probability

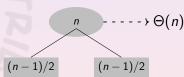
$$\frac{d-1}{d+1}$$

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

# Quick Sort Complexity: Average Case (Cont'd)

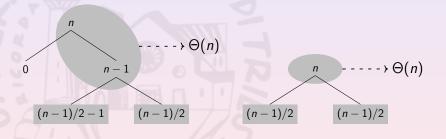
Even if "good" and "bad" cases alternate





# Quick Sort Complexity: Average Case (Cont'd)

Even if "good" and "bad" cases alternate

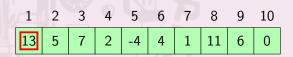


On the average  $\Theta(n \log n)$ 



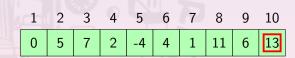
## Sorting by Searching the Maximum

Find the maximum



Find the maximum

Move the maximum at the end of the array



Find the maximum

Move the maximum at the end of the array



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Find the maximum

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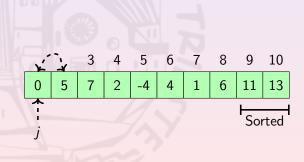
Find the maximum

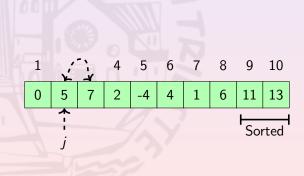
Move the maximum at the end of the array

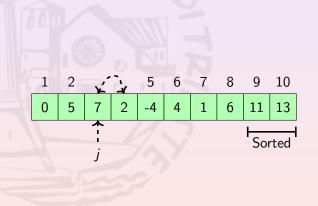
If |A| > 1, repeat on the initial fragment of A

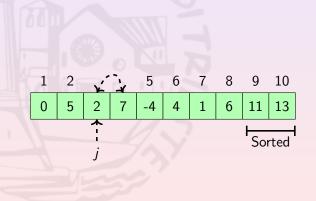
The complexity is  $\sum_{i=1}^{|A|} \left( \mathcal{T}_{\max}(i) + \Theta(1) \right)$ 

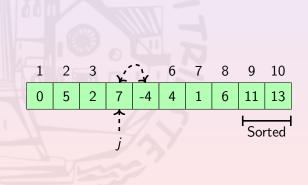












By pair-wise swapping the maximum to the right: Bubble Sort

After some swaps...



By pair-wise swapping the maximum to the right: Bubble Sort

After some swaps...

#### Bubble Sort: Code and Complexity

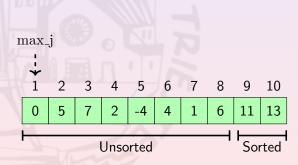
```
def BUBBLE_SORT(A):
  for i in |A|..2:
    for j in 1 \dots i-1:
       if A[i]>A[i+1]:
         swap(A, j, j+1)
      endif
    endfor
  endfor
enddef
```

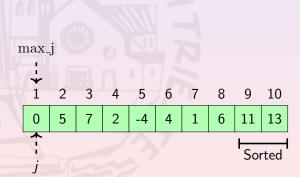
One swap-block costs  $\Theta(1)$ 

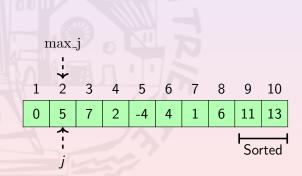
The nested for-loop costs  $\Theta(i)$ 

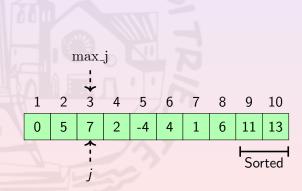
$$T_B(n) = \sum_{i=2}^n \Theta(i) * \Theta(1)$$

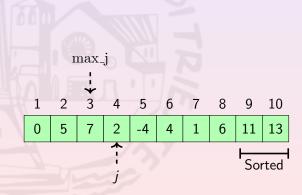
$$= \Theta(\sum_{i=2}^n i) = \Theta(n^2)$$



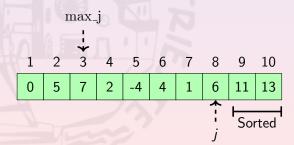




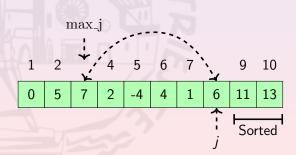




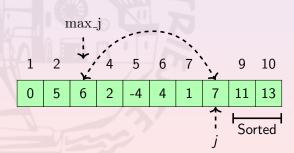
By linear scaning the unsorted part: **Selection Sort** 



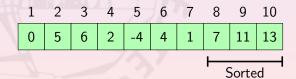
By linear scaning the unsorted part: **Selection Sort** 



By linear scaning the unsorted part: Selection Sort



By linear scaning the unsorted part: **Selection Sort** 



# Selection Sort: Code and Complexity

```
def SELECTION_SORT(A):
  for i in |A|..2:
     \max_{i} \leftarrow 1
     for j in 2...i:
       if A[j]>A[max_j]:
          \max_{j} \leftarrow j
       endif
     endfor
    swap(A, i, max_j)
  endfor
enddef
```

One if-block costs  $\Theta(1)$ 

The nested for-loop costs  $\Theta(i)$ 

$$T_S(n) = \Theta(1) + \sum_{i=2}^n \Theta(i) * \Theta(1)$$
$$= \Theta\left(1 + \sum_{i=2}^n i\right) = \Theta(n^2)$$



Any other idea?

What about using a max-heap  $H_{max}$ ? Heap Sort

- **①** store the elements of A in  $H_{\text{max}}$
- extract the min (i.e., the max) and place it in A
- **3** repeat from 2 until  $H_{\rm max}$  is not empty

Any other idea?

What about using a max-heap  $H_{max}$ ? Heap Sort

- **①** store the elements of A in  $H_{\text{max}}$
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Array-based representation of heaps  $\Rightarrow$  in-place algorithm.

## Heap Sort: Pseudo-Code

```
\label{eq:def-HEAPSORT} \begin{split} \text{def HEAPSORT(A):} & \quad \text{H} \leftarrow \text{BUILD\_MAX\_HEAP(A)} \ \# \ \textit{the root is the max} \\ & \quad \text{for } i \leftarrow |A| \dots 2: \\ & \quad A[i] \leftarrow \text{EXTRACT\_MIN(H)} \\ & \quad \text{endfor} \\ & \quad \text{enddef} \end{split}
```

## Heap Sort: Complexity

BUILD\_MAX\_HEAP costs  $\Theta(n)$ 

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

$$T_H(n) = \Theta(n) + \sum_{i=2}^n O(\log i)$$

$$\leq O(n) + O(\sum_{i=2}^n \log n) = O(n \log n)$$

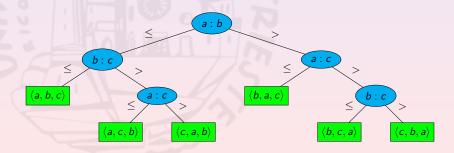
The overall complexity of heap sort is  $O(n \log n)$ 



#### 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 and b corresponds to a node which branches the computation according whether  $a \le b$  or b > a



# 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,

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

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The lower bound for comparison-based sorting is  $\Omega(n \log n)$ 

#### Sorting in Linear Time?

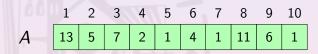
There is no general algorithm to sort in linear time by using comparisons

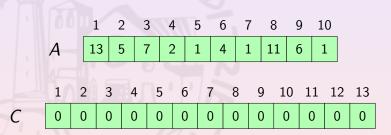
#### Sorting in Linear Time?

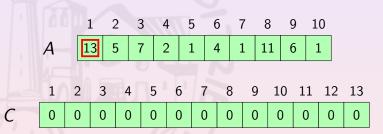
There is no general algorithm to sort in linear time by using comparisons

This bound does not hold if we introduce minor *ad-hoc* assumptions such as:

- bounded domain for the array values
- uniform distribution of the array values

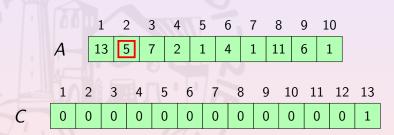


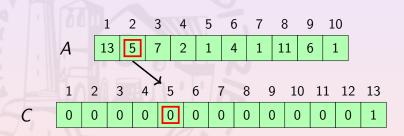


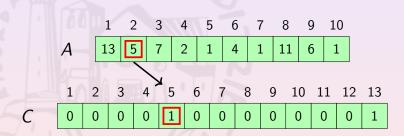


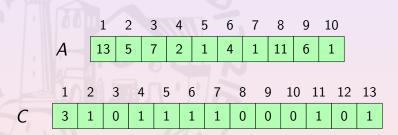




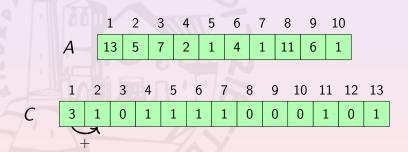




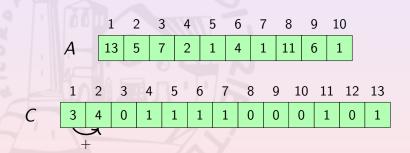




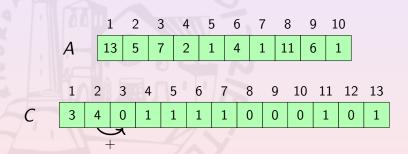
- count the occurrences of A's values and place them in C
- sums the values in C and get the # elements  $\leq$  to C's indexes



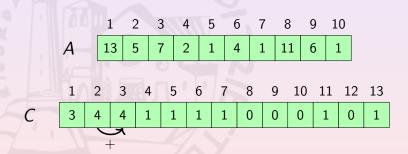
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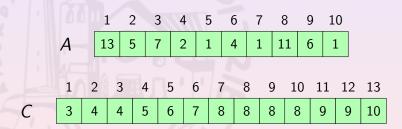
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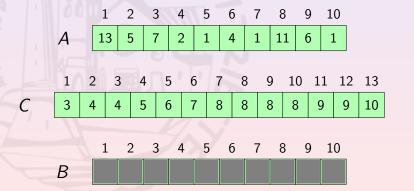
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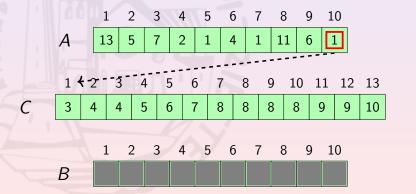
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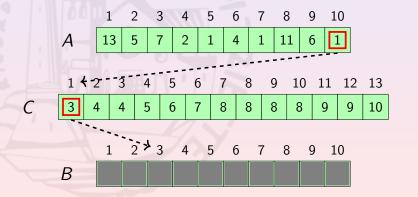
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- use C to place the elements of A in the correct positions in B



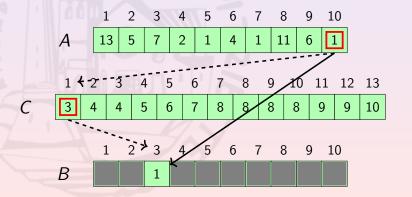
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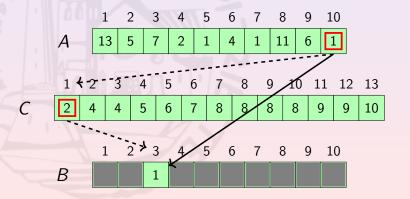
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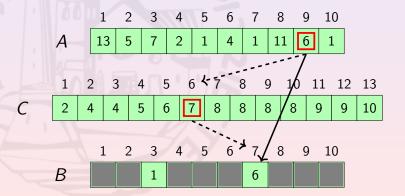
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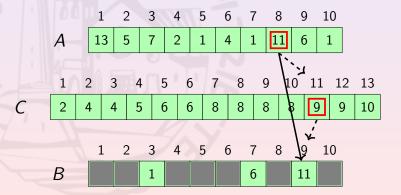
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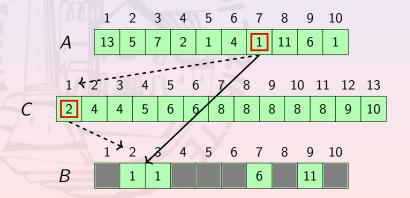
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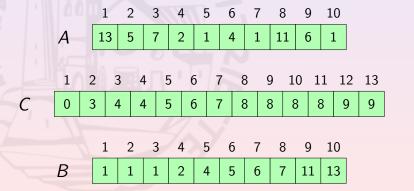
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Generalizing it to deal with any  $[k_1, k_2]$  domain is easy

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Generalizing it to deal with any  $[k_1, k_2]$  domain is easy

It is not in-place and it requires the array C

#### Counting Sort: Pseudo-Code

```
def COUNTING_SORT(A, B, k):
  C ← ALLOCATE_ARRAY(k, default_value=0)
  for i \leftarrow 1 upto |A|:
    C[A[i]] \leftarrow C[A[i]]+1
  endfor # C[i] is now the # of i in A
  for j \leftarrow 2 upto |C|:
    C[i] \leftarrow C[i-1] + C[i]
  endfor # C[j] is now the # of A's values < i
  for i \leftarrow |A| downto 1:
    B[C[A[i]]] \leftarrow A[i]
    C[A[i]] \leftarrow C[A[i]] - 1
  endfor
enddef
```

Allocating C and setting all its elements to 0

 $\Theta(k)$ 

Allocating C and setting all its elements to 0

 $\Theta(k)$   $\Theta(n)$ 

Counting the instances of A's values

Allocating C and setting all its elements to 0

Counting the instances of A's values

Setting in C[j] the # of A's values  $\leq j$ 

 $\Theta(k)$   $\Theta(n)$   $\Theta(k)$ 

Allocating C and setting all its elements to 0  $\Theta(k)$ 

Counting the instances of A's values  $\Theta(n)$ 

Setting in C[j] the # of A's values  $\leq j$   $\Theta(k)$ 

Copying A's values into B by using C  $\Theta(n)$ 

Allocating $C$ and setting all its elements to $0$	$\Theta(k)$
--	-------------

Counting the instances of A's values  $\Theta(n)$ 

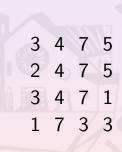
Setting in C[j] the # of A's values  $\leq j$   $\Theta(k)$ 

Copying A's values into B by using C  $\Theta(n)$ 

Total complexy  $\Theta(n+k)$ 

#### Fixed Number of Digits: Radix Sort

An array A of d-digit values can be sorted digit-by-digit



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- for each digit i from the rightmost down to the leftmost
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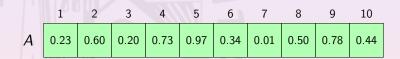
```
1 7 3 3
```

## Radix Sort: Complexity

If the digit sorting is in  $\Theta(|A|+k)$ , radix sort takes time

$$\Theta\left(d\left(|A|+k\right)\right)$$

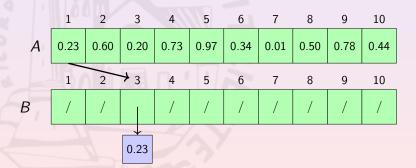
where d is the number of digits in each of A's values



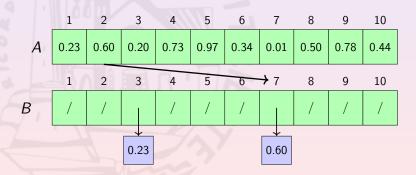
• split [0,1) in n buckets:  $\left[\frac{i-1}{n},\frac{i}{n}\right)$  for  $i\in[1,n]$ 

	1	2	3	4	5	6	7	8	9	10
Α	0.23	0.60	0.20	0.73	0.97	0.34	0.01	0.50	0.78	0.44
	1	2	3	4	5	6	7	8	9	10
В	/	/	/	/	/	/	/	/	/	/

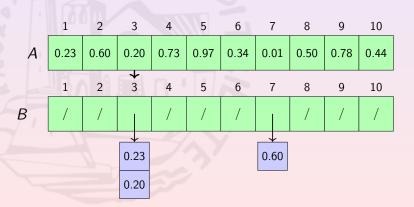
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- add each value of A to the correct bucket



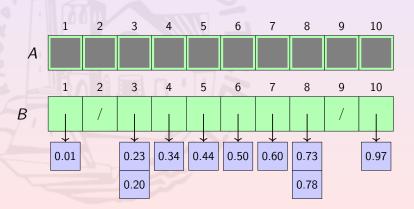
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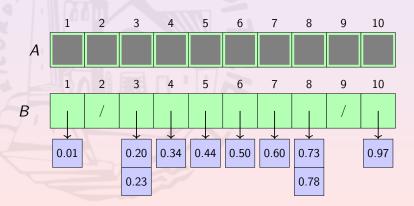
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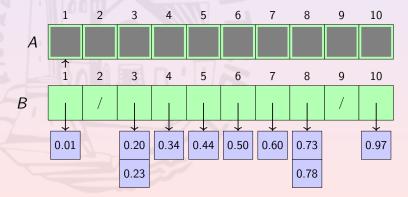
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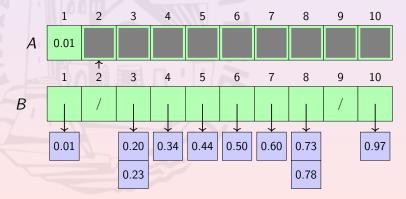
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- add each value of A to the correct bucket
- sort the buckets



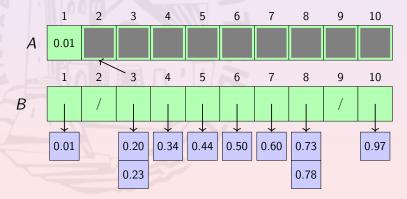
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- sort the buckets
- reverse the content of buckets in bucket order on A



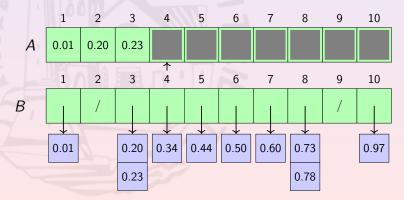
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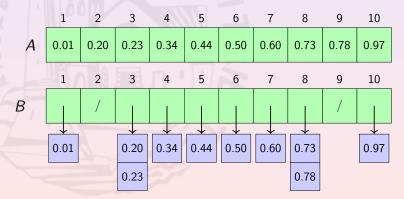
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- sort the buckets
- reverse the content of buckets in bucket order on A



#### Bucket Sort: Pseudo-Code

enddef

```
def BUCKET_SORT(A):
  B ← ALLOCATE_ARRAY_OF_EMPTY_LISTS ( | A | )
  for i \leftarrow 1 upto |A|:
    B[FLOOR(A[i]*n)+1]. append (A[i])
  endfor # now B contains the buckets
  i \leftarrow 0
  for j \leftarrow 1 upto |B|
    for v in B[j]: # reverse the bucket in A
      A[i] \leftarrow v
     i \leftarrow i+1
    endfor
    sort(A, i-|B[j]|, len=|B[j]|) # sort the bucket
  endfor
```



 $\Theta(n)$ 

Allocating and initializing B

Filling the buckets

 $\Theta(n)$   $\Theta(n)$ 

Allocating and initializing B

Filling the buckets

Sorting each bucket (expected)

 $\Theta(n)$   $\Theta(n)$  O(n)

Allocating and initializing $B$	$\Theta(n)$
Filling the buckets	$\Theta(n)$
Sorting each bucket (expected)	O(n)
Reversing buckets' content into A	$\Theta(n)$

Allocating and initializing $B$	$\Theta(n)$
Filling the buckets	$\Theta(n)$
Sorting each bucket (expected)	O(n)
Reversing buckets' content into A	$\Theta(n)$
Total expected complexy	O(n)



Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

1?

Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

• 1? Complexity?

Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

 $\Theta(n)$ 

- 1? Complexity?
- n?

Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

- 1? Complexity?
- n? Complexity?

 $\Theta(n)$ 

Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

- 1? Complexity?
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- $\Theta(n)$
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- n? Complexity?
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Let A be unsorted array

How to find the value that, if A was sorted, would be in position:

• 1? Complexity?

 $\Theta(n)$ 

• n? Complexity?

 $\Theta(n)$ 

•  $i \in [1, n]$ ? Complexity?

 $O(n \log n)$ 

Can we do better?

#### The Select Problem

**Input:** a potentially unsorted array A and an index  $i \in [1, |A|]$  **Output:** the value  $\bar{A}[i]$  where  $\bar{A}$  is the sorted version of A



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We reformulate it as:

**Input:** a potentially unsorted array A and an index  $i \in [1, |A|]$  **Output:** an index j s.t.  $\tilde{A}[j] = \bar{A}[i]$  where  $\tilde{A}$  and  $\bar{A}$  are A after the computation and the sorted version of A, respectively.

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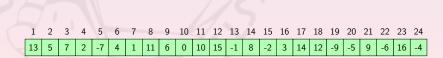
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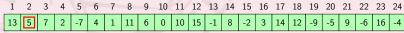
We will assume that A does not contains multiple instances of the same value (not necessary, but simplify things)





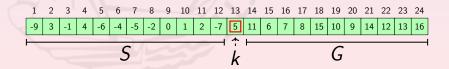
What about using PARTITION and a "dichotomic approach"?  $\bullet$  select a pivot A[j]



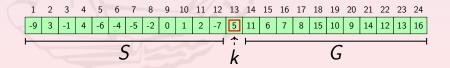




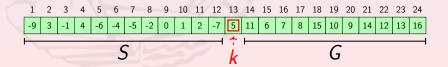
- select a pivot A[j]
- ullet compute  $k \leftarrow PARTITION(A,1,|A|,j)$  and get S and G



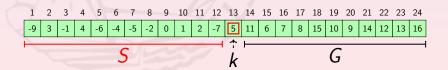
- select a pivot A[j]
- compute  $k \leftarrow PARTITION(A,1,|A|,j)$  and get S and G
- compare i and k



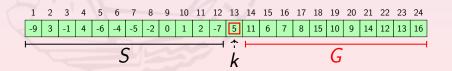
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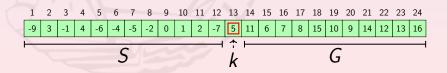
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What about using PARTITION and a "dichotomic approach"?

- select a pivot A[j]
- compute  $k \leftarrow PARTITION(A,1,|A|,j)$  and get S and G
- compare i and k
  - if i = k, then return k
  - if i < k, then  $\bar{A}[i]$  is in S
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A recursive algorithm can solve the problem!



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However, constant ratio partitions are QUICKSORT's best case scenarios as well

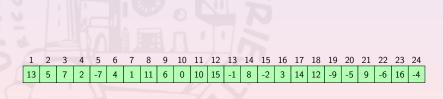
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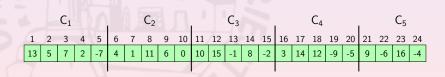
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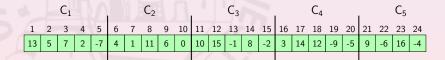
Is there a smart way to guess an almost-median value for  $\bar{A}$ ?



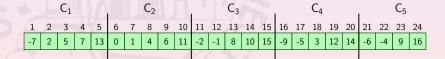
• split A in  $\lceil n/5 \rceil$  chunks  $C_1, \ldots, C_{n/5}$  each of size 5



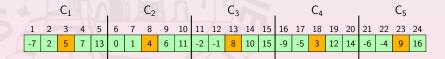
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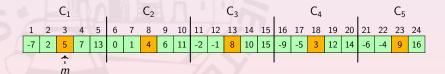
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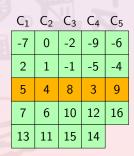
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- ullet recursively compute the median m of the  $m_i$ 's



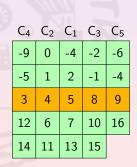
Think the chunks as they were the columns of a matrix



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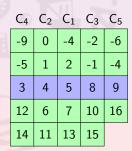


Sort the chunks according the medians



How many chunks are there?

$$\left\lceil \frac{n}{5} \right\rceil$$

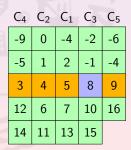


How many  $m_i$  are greater or equal to m?

$$\left\lceil \frac{1}{2} \left\lceil \frac{n}{5} \right\rceil \right\rceil$$

How many chunks at least have 3 elements greater than m?

$$\left\lceil \frac{1}{2} \left\lceil \frac{n}{5} \right\rceil \right\rceil - 2$$



How many elements at least are greater than m?

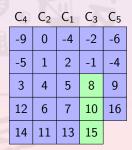
$$3\left(\left\lceil\frac{1}{2}\left\lceil\frac{n}{5}\right\rceil\right\rceil-2\right)$$

How many elements at least are greater than m?

$$3\left(\left\lceil\frac{1}{2}\left\lceil\frac{n}{5}\right\rceil\right\rceil-2\right)\geq\frac{3n}{10}-6$$

An upper bound for the # of elements smaller or equal to m is

$$n - \left(\frac{3n}{10} - 6\right) = \frac{7n}{10} + 6$$



$$T_S(n) = T_S(\lceil n/5 \rceil) + T_S(7n/10 + 6) + \Theta(n)$$

Prove by induction that  $T_S(n) \in O(n)$  (Substitution Method)

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 $\le c(n/5+1) + c(7n/10+6) + c'n$   
 $\le 9/10cn + c'n + 7c$ 

Hence,  $T_S(n) \le cn$  for  $c \ge 20c'$  and  $n \ge 140$  and

$$T_S(n) = T_S(\lceil n/5 \rceil) + T_S(7n/10 + 6) + \Theta(n)$$

Prove by induction that  $T_S(n) \in O(n)$  (Substitution Method)

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Hence,  $T_S(n) \le cn$  for  $c \ge 20c'$  and  $n \ge 140$  and  $T_S(n) \in O(n)$ 

# Select Algorithm: Pseudo-Code

```
def SELECT(A, I=1, r=|A|, i):
  if r-l≤10: # base case
    SORT(A, I, r)
    return i
  endif
  j \leftarrow SELECT_PIVOT(A, I, r)
  k \leftarrow PARTITION(A, I, r, j)
  if i=k:
                     # dichotomic approach
    return k
  endif
  if i < k:
             # search in S
    return SELECT(A, I, k-1, i)
  endif
  # search in G
  return SELECT(A, k+1, r, i)
enddef
```

# Select Pivot Algorithm: Pseudo-Code

```
def SELECT_PIVOT(A, I=1, r=|A|):
  if r−l<10:
                                  # base case
    SORT(A, I, r)
    return (1+r)/2
  endif
  chunks \leftarrow (r-1)/5
  for c in 0... chunks -1:
                                  # for each chunk
    (c_{-1}, c_{-r}) \leftarrow (1, 5) + c*5
    SORT(A, c_I, c_r)
                                 # sort it
    SWAP(A, c_1+2, 1+c)
                                  # place the middle elem
                                  # at the beginning of A
  endfor
 # recursive step
  return SELECT (A, I, I+chunks-1, chunks/2)
enddef
```