

Analysis of Algorithms

Medians and Order Statistics

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Outline

- 1 Introduction
 - Finding the k^{th} statistics
 - Selection problem
 - Minimum-Maximum
- 2 Selection in Expected Linear Time
 - Using Randomization
 - RANDOMIZED-SELECT
- 3 Selection in worst-case linear time
 - Introduction
 - Explanation
- 4 SELECT the i th element in n elements
 - The Final Algorithm
 - Complexity Analysis
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Fact:

The i th order statistic of a set of n elements is the i th smallest element.



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Selection Problem

Input:

A set A of n (distinct) numbers and an integer i , with $1 \leq i \leq n$.

Output:

The element $x \in A$ that is larger than exactly $i - 1$ other elements of A .



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Minimum-Maximum

Minimum using $n - 1$ comparisons

Minimum(A)

- 1 $min = A[1]$
- 2 for $i = 2$ to $A.length$
- 3 if $min > A[i]$
- 4 $min = A[i]$
- 5 return min



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Using an $O(n \log n)$ Algorithm

Assumption

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- Minimum: The first element.
- Maximum: The last element.
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- For the median:
 - ▶ If n is odd, then the median is equal to the $\frac{n+1}{2}$ th element.
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 - ★ The lower median is equal to the $\lfloor \frac{n+1}{2} \rfloor$ th element.
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Question!!

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Minimum and Maximum at the same time... better choice

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- Compare the min in the tuple with the global min and do the same with the tuple max.



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This will give you

$3\lfloor \frac{n}{2} \rfloor$ comparisons.

Why?

Let's see!



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Selection in expected linear time $O(n)$

Selecting in expected linear time implies:

- Selecting the i th element.
- Use the divide and conquer algorithm RANDOMIZED-SELECT.
 - ▶ Similar to Quicksort, partition the input array recursively.
 - ▶ Unlike Quicksort, which works on both sides of the partition, just work on one side of the partition. This is called PRUNE-AND-SEARCH, prune one side, just search the other.



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Homework

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Strategy

First partition the set of n elements

- How? Remember Randomized Quicksort!!!

Then

We get a q from the random partition!!!



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Thus

We get a q from the random partition!!!

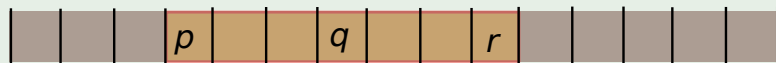


Example

Positions

Global Index Position

1 2 3 4 5 6 7 8 9 10



1 2 3 4 5 6 7

Local Index Position

$$p-q=3 \text{ then } k=p-q+1 =4$$



Thus, the Strategy

If $i < k$

The possible i th smallest element is between p and $q - 1$.



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If $i > k$

- The possible i th smallest element is between $q + 1$ and r .

• But the new $i' = i - k$, this will work because we start at $p = 1$ and $r = n$.



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If $i = k$ — We need to convert this to local index, too

- return $A[q]$



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If $i == k \leftarrow$ We need to convert this to local index too

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RANDOMIZED-SELECT

RANDOMIZED-SELECT Algorithm

Randomized-Select(A, p, r, i)

- 1 if $p == r$
- 2 return $A[p]$
- 3 $q = \text{Randomized-Partition}(A, p, r)$
- 4 $k = q - p + 1$ // Local Index
- 5 if $i == k$ // The Answer
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Analysis of RANDOMIZED-SELECT

Worst-case running time $\Theta(n^2)$. Why?

An empty side and a side with remaining elements. So every partitioning of m elements will take $\Theta(m)$ time where $m = n, n - 1, \dots, 2$. Thus in total

$$\Theta(n) + \Theta(n-1) + \dots + \Theta(2) = \Theta\left(\frac{n(n-1)}{2} + 1\right) = \Theta(n^2).$$



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Select the i th smallest element of $S = \{a_1, a_2, \dots, a_n\}$.

Solution:

- Use the so called PRUNE-AND-SEARCH technique:
 - ▶ Let $x \in S$, and partition S into three subsets.
 - ▶ $S_1 = \{a_j | a_j < x\}$, $S_2 = \{a_j | a_j = x\}$, $S_3 = \{a_j | a_j > x\}$.
 - ▶ If $|S_1| > i$, search i th smallest elements in S_1 recursively, (prune S_2 and S_3 away).
 - ▶ Else if $|S_1| + |S_2| > i$, then return x (the i th smallest element).
 - ▶ Else search the $(i - (|S_1| + |S_2|))$ th element in S_3 recursively (prune S_1 and S_2 away).

A question arises

How to select x such that S_1 and S_3 are nearly equal in cardinality? Force an even search!!!

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How to select x such that S_1 and S_3 are nearly equal in cardinality? Force an even search!!!

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Divide elements into $\lceil \frac{n}{5} \rceil$ groups of 5 elements each and find the median of each one

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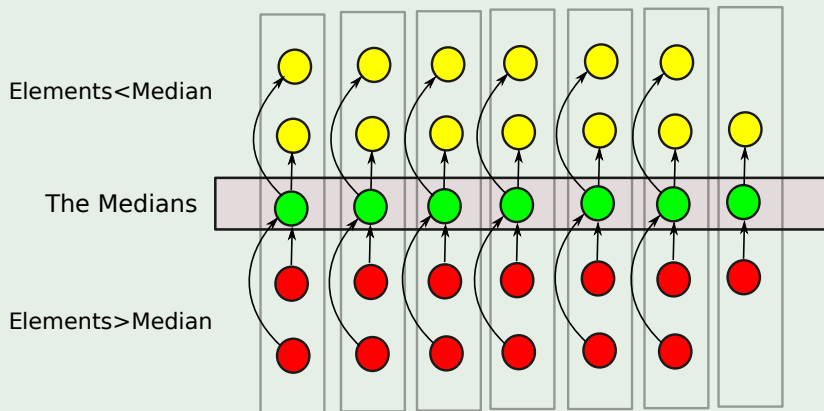
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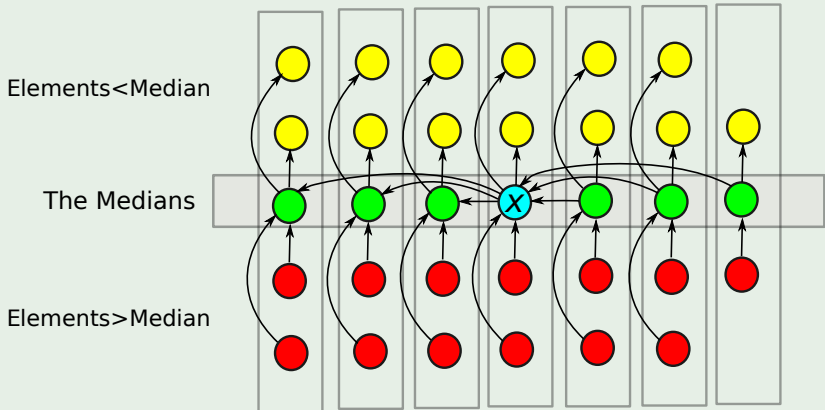
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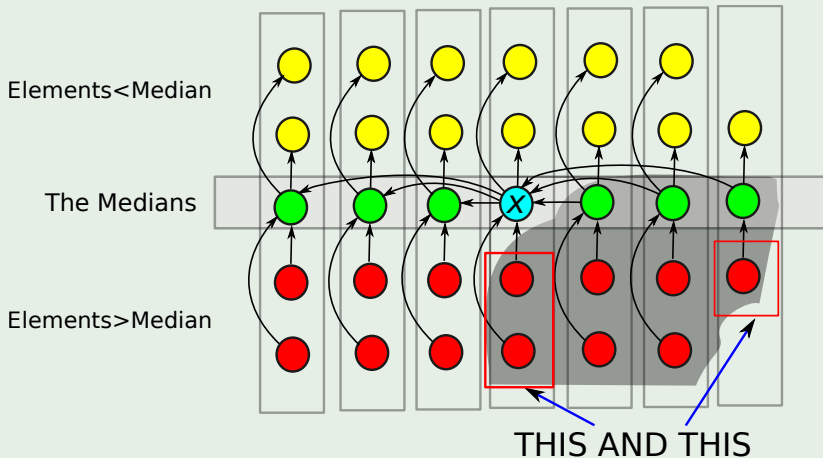
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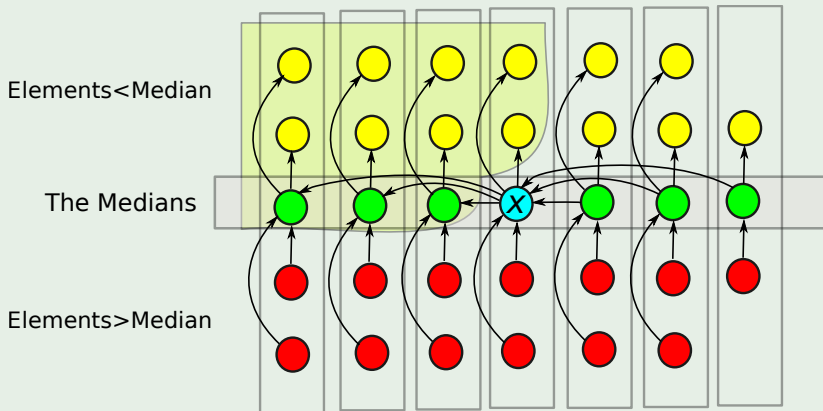
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Final Recursion

We have then

$$T(n) = \begin{cases} O(1) & \text{if } n < \text{some value (i.e. 140)} \\ T(\lceil \frac{n}{5} \rceil) + T(\frac{7n}{10} + 6) + O(n) & \text{if } n \geq \text{some value (i.e. 140)} \end{cases}$$



Solve recurrence by substitution

Suppose $T(n) \leq cn$ for some c

$$\begin{aligned}T(n) &\leq c\lceil \frac{n}{5} \rceil + c\left(\frac{7n}{10} + 6\right) + an \\ &\leq \frac{1}{5}cn + c + \frac{7}{10}cn + 6c + an \\ &\leq \frac{9}{10}cn + 7c + an \\ &\leq cn + \left(-\frac{1}{10}cn + an + 7c\right)\end{aligned}$$



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- Given a neighborhood of n members of x , you need to find the median to substitute the value in x

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Selection in worst case linear time

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- It is still unknown exactly how many comparisons are needed to determine the median.

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- Median, lower median, upper median.

Selection in expected average linear time

- Worst case running time

• PRUNE-AND-SEARCH

Selection in worst case linear time

- The fast randomized version is due to Hoare.
- It is still unknown exactly how many comparisons are needed to determine the median.

Summary.

The i th order statistic of n elements

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Selection in worst case $O(n \log n)$ time

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- **It is still unknown exactly how many comparisons are needed to determine the median.**

Exercises

From Cormen's Book Chapter 9

- 9.1-1
- 9.2-3
- 9.3-4
- 9.3-8
- 9.2

