

CSE 221: Algorithms

Order statistics

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References

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- 2 Erik Demaine and Charles Leiserson, *6.046J Introduction to Algorithms*. MIT OpenCourseWare, Fall 2005. Available from: ocw.mit.edu/OcwWeb/Electrical-Engineering-and-Computer-Science/6-046JFall-2005/CourseHome/index.htm

Last modified: June 20, 2009



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Contents

- 1 Order statistics
 - What is order statistics
 - Selection algorithm
 - Conclusions

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- Using $O(n \lg n)$ algorithm (such as heapsort or mergesort, but *not* quicksort), it runs in $O(n \lg n) + \Theta(1)$ or $\Theta(n \lg n)$ time in the worst case.

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- Can we do better?

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Idea behind the selection algorithm

Basic idea

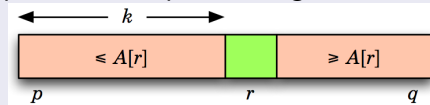
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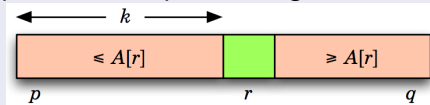


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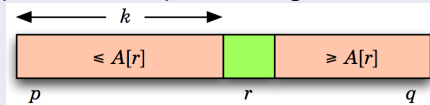


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- 2 If $k = i$, then return the value at that index.
- 3 Otherwise, if $k > i$, then the i^{th} value must be *left* of k , so recursively partition the subarray left of k until $k = i$.
- 4 Otherwise, if $k < i$, then the i^{th} value must be *right* of k , so recursively partition the subarray right of k until $k = i$.

Partition based selection example

Given the following sequence (and using the first element as the *pivot*):

6	10	13	5	8	3	2	11
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The selection algorithm

RANDOMIZED-SELECT(A, p, q, i) $\triangleright i^{th}$ smallest of $A[p..q]$

if $p = q$

then return $A[q]$

$r \leftarrow$ RANDOMIZED-PARTITION(A, p, q)

$k \leftarrow r - p + 1$

$\triangleright k = \text{rank}(A[r])$

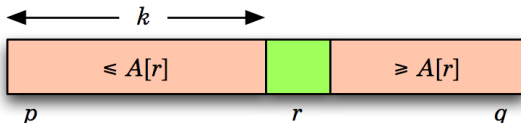
if $i = k$

then return $A[r]$

if $i < k$

then return RANDOMIZED-SELECT($A, p, r - 1, i$)

else return RANDOMIZED-SELECT($A, r + 1, q, i - k$)



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- Given this large constant before the n in the linear-time algorithm, **the randomized algorithm works better in practice.**