# CSI - 3105 Design & Analysis of Algorithms Course 4

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Fall 2020

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What is the bottleneck?

Can we solve this problem without sorting?



# First Attempt for a Faster Algorithm

## **Algorithm** Select(S, k)

**Input:** Sequence S of n numbers and an integer k with  $1 \le k \le n$ .

**Output:** The k-th smallest element of S.

- 1: **if** |S| = 1 **then**
- **return** the only element in S
- 3: else
- Choose an element p in S (called the pivot) 4.
- 5: Split S into  $S_{<}$ ,  $S_{=}$  and  $S_{>}$
- if  $k \leq |S_{<}|$  then 6:
- Run Select( $S_{<}, k$ ) 7:
- else if  $k > |S_{<}| + |S_{=}|$  then 8:
- Run Select( $S_>$ ,  $k |S_<| |S_=|$ ) 9:
- else 10:
- 11: return p
- end if 12:
- 13: **end if**

The running time of Select(S, k) depends on the pivot p. In the worst case,

- *S* is sorted
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Using the Master Theorem with a=1, b=2 and d=1, we find  $\mathcal{T}(n)=\mathcal{O}(n)$ .



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How close to a good case do we need to be to get a linear-time algorithm?

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Then the running time satisfies

$$T(n) = T(\alpha n) + n$$

$$= T(\alpha^{2} n) + \alpha n + n$$

$$= T(\alpha^{3} n) + \alpha^{2} n + \alpha n + n$$

$$\vdots$$

$$= O(n)$$

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Blum, Floyd, Pratt, Rivest and Tarjan (1973) discovered the following technique.

# The Algorithm

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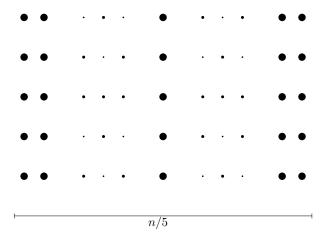
Is p a good pivot? Why?

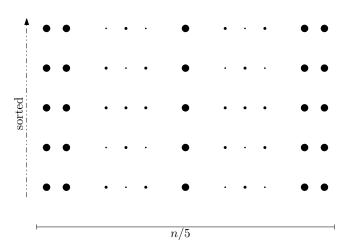
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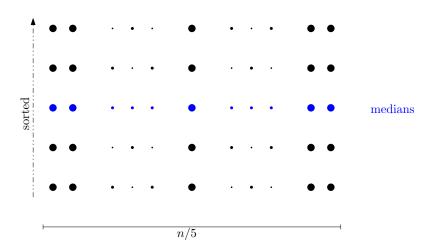
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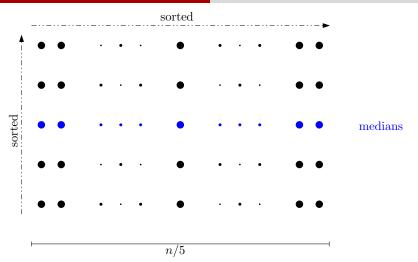
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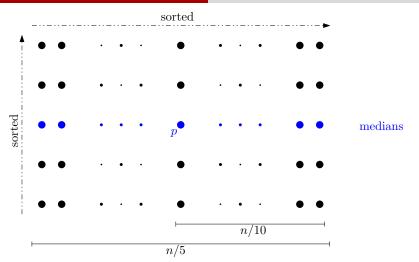
We have to figure out how many numbers in S are larger than p.

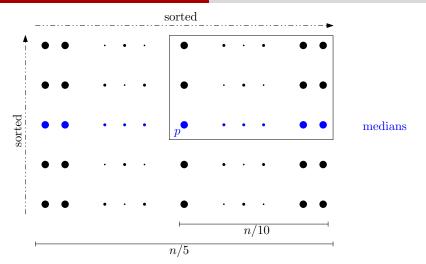


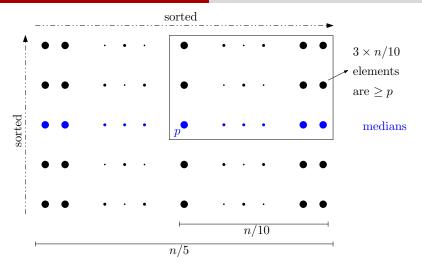


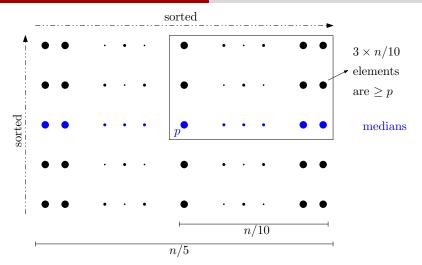




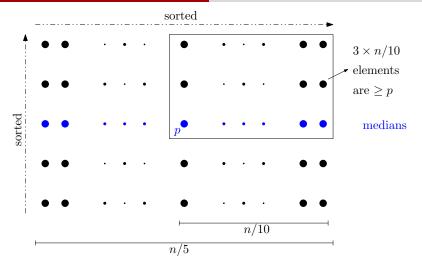






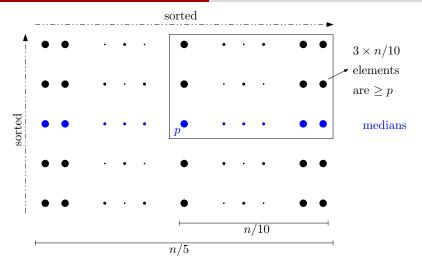


Hence, at least  $\frac{3}{10}n$  elements are  $\geq p$ .



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In other words,  $|S_{<}| \leq \frac{7}{10}n$ .

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Hence, with this choice of pivot, we have  $\alpha = \frac{7}{10}$ .



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Hence: 
$$T(n) = T(\frac{1}{5}n) + T(\frac{7}{10}n) + O(n)$$



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The Master theorem does not apply.

We use induction to show that T(n) = O(n).





T(n) = O(n) Step 1 + O(n) Step 3 + O(n) +  $T(\frac{7}{10}n)$  Step 4

How do we do Step 3: Recursively compute the n-th smallest element of the sequence m, ma, mas. This takes T(1/5) time.

We obtain the recurrence

 $T(n) = n + T(\frac{n}{5}) + T(\frac{7}{10}n)$ 

-> Unfolding get messy

-> Master theorem does not apply.

Let us use induction to show that T(n) = O(n).

Claim: T(n) = c.n for some constant e

Proof: By choosing a sufficiently large, the claim is true for "small" n (this is the base case of the induction).



Let n be "large" and assume T(m) & c.m for all 15m xn. Then

 $T(n) = n + T(\frac{n}{5}) + T(\frac{7}{10}n)$ 

\( \left\) \text{ \left\) \quad \qquad \quad \

= on + 9 c.n off words

Is n+9 c.n < cn?

Yes, provided that c>10.

Conclusion: The K-th smallest element in a sequence of n numbers can be computed in O(n) time.

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