

# CSE331 Introduction to Algorithm

## Lecture 14: Deterministic Algorithm for The Selection Problem

Antoine Vigneron  
[antoine@unist.ac.kr](mailto:antoine@unist.ac.kr)

Ulsan National Institute of Science and Technology

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## 1 Introduction

## 2 Selection in worst-case linear time

## 3 Concluding remarks

# Course Organization

- In the previous lecture, I gave a randomized algorithm for the selection problem that runs in expected linear time.
- Today, I present a deterministic algorithm that also runs in linear time.
- Reference: Section 9 of the textbook [Introduction to Algorithms](#) by Cormen, Leiserson, Rivest and Stein. (Available online from the UNIST library website.)

# Problem Statement

## Problem

Given an array  $A[1 \dots n]$  of  $n$  distinct numbers and an integer  $i$ , the **selection problem** is to find the  $i$ th smallest number in  $A$ .

## Example

Given  $A = [8, 4, 5, 6, 12, 9, 7, 1]$  and  $i = 3$ , then the answer is **5** because  $A$  in sorted order is  $B = [1, 4, 5, 6, 7, 8, 9, 12]$  and  $B[3] = 5$ .

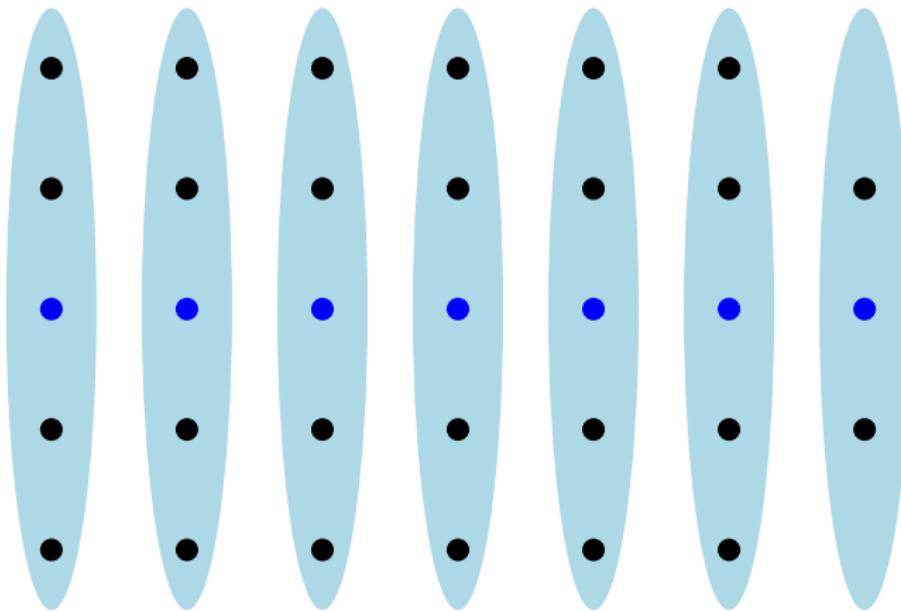
## Special cases

- $i = 1$  gives the **minimum** in  $A$ .
- $i = n$  gives the **maximum** in  $A$ .
- The middle element is the **median**. More precisely:
  - ▶  $i = \lfloor (n + 1)/2 \rfloor$  gives the **lower median**.
  - ▶  $i = \lceil (n + 1)/2 \rceil$  gives the **upper median**.

# Selection in Worst-Case Linear Time

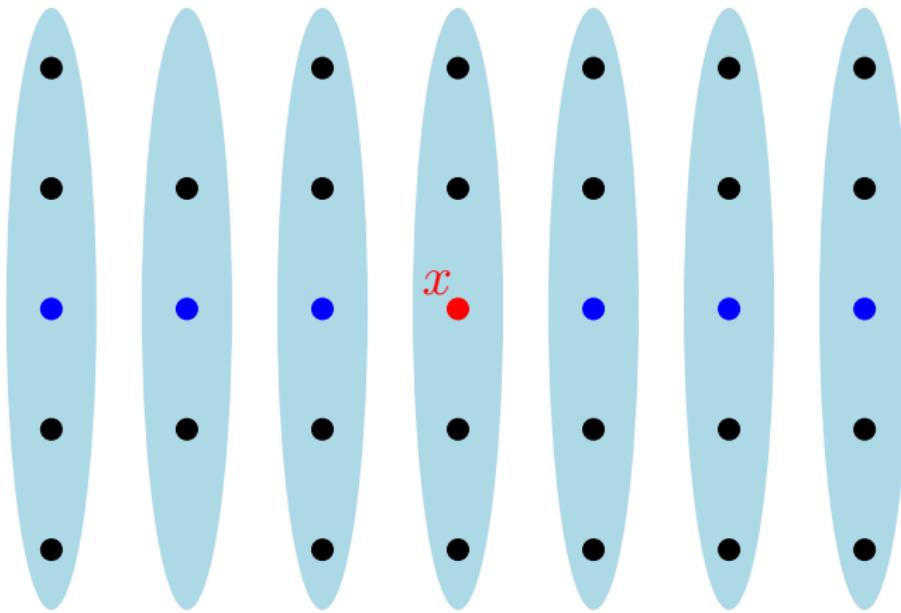
- The randomized selection algorithm picks a pivot at random.
- It gives a linear expected running time because, with probability  $1/2$ , the pivot is *central*, i.e. it splits the array into two parts, each part being of size at least  $n/4$ .
- Here we give a *deterministic* algorithm, that does not need randomization to find a central pivot.
- As we did in the lectures on QUICKSORT and randomized selection, we assume that no two input numbers are equal.

## Finding the Pivot



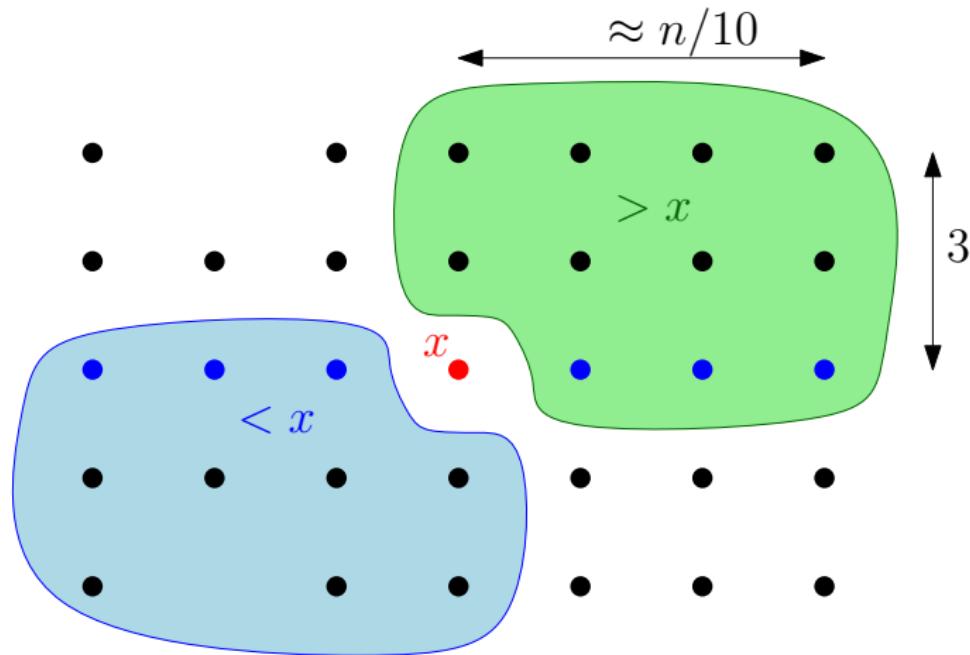
**Figure:** Group the elements of the array by 5, and compute the median of each group of 5 by brute force (blue).

## Finding the Pivot



**Figure:** Recursively compute the median  $x$  of the  $\lceil n/5 \rceil$  medians (red). Imagine they are sorted from left to right

## Finding the Pivot



**Figure:** The keys in the green region are larger than  $x$ . There are about  $3n/10$  of them. Similarly, there are about  $3n/10$  keys in the blue regions, and they are smaller than  $x$ .

# Selection in Worst-Case Linear Time

## Pseudocode

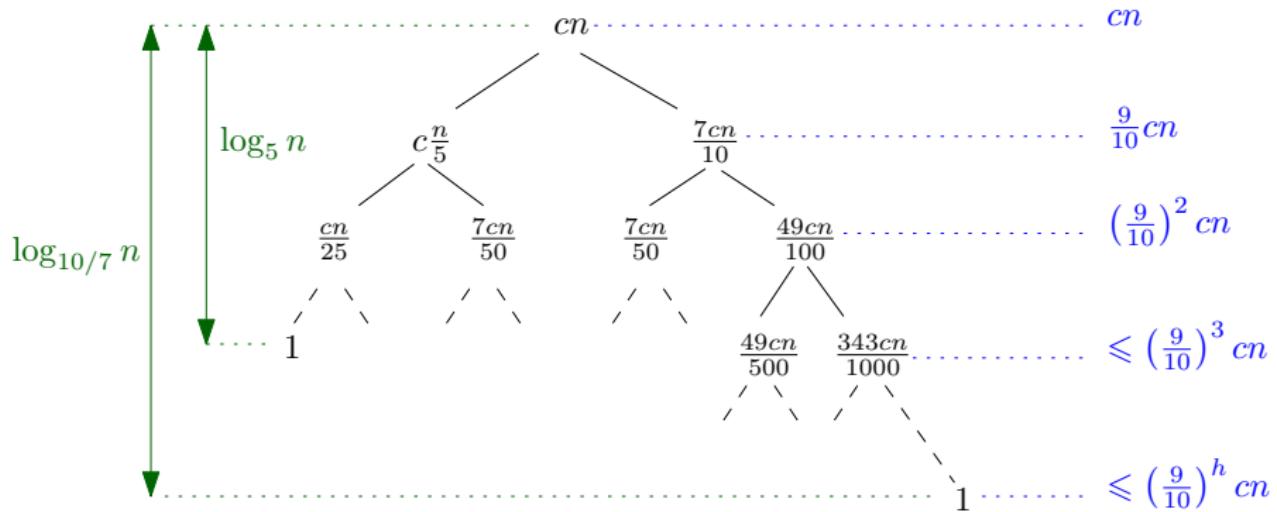
$\text{SELECT}(A, p, r, i)$

- ➊ Divide the keys into  $\lceil n/5 \rceil$  groups of size 5 each. (More precisely, one group may have size  $< 5$ .)
  - ➋ Find the median of each group using INSERTION SORT.
  - ➌ Find the median  $x$  of these  $\lceil n/5 \rceil$  medians by calling SELECT recursively.
  - ➍ Partition the array using  $x$  as the pivot. Assume that now  $x = A[q]$ , and let  $k = q - p + 1$ .
  - ➎ If  $i = k$ , then return  $x$ .
  - ➏ If  $i < k$ , then return  $\text{SELECT}(A, p, q - 1, i)$ .
  - ➐ If  $i > k$ , then return  $\text{SELECT}(A, q + 1, r, i - k)$ .
- 
- The only difference with RANDOMIZED SELECT is the choice of the pivot at lines 1–3.

# Analysis

- Rough analysis: Let  $T(n)$  be the running time of SELECT.
  - ▶ Line 1:  $\Theta(n)$ .
  - ▶ Line 2: INSERTION SORT on an array of size 5 takes  $O(1)$  time, so Line 2 takes time  $\Theta(n)$
  - ▶ Line 3:  $T(n/5)$
  - ▶ Line 4:  $\Theta(n)$
  - ▶ Line 5:  $\Theta(1)$
  - ▶ From Slide 8, after partitioning an array of size  $n$ , the resulting subarrays have size  $\leq 7n/10$
  - ▶ So Line 6 and 7 are at most  $T(7n/10)$
- We obtain  $T(n) \leq T(n/5) + T(7n/10) + \Theta(n)$ .
- The master theorem does not apply.
- Next slide: We guess a bound on  $T(n)$  using the recursion tree method.

# Analysis: Recursion Tree



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$$\begin{aligned}\text{Total: } &\leq \frac{1}{1 - \frac{9}{10}} cn \\ &= 10cn\end{aligned}$$

## Analysis: Recursion Tree

- The sum of the size of the subproblems decreases by a factor (at least)  $10/9$  at each level.
- So an upper bound on the running time should be

$$\begin{aligned}T(n) &\leq \sum_{j=0}^{\infty} \left(\frac{9}{10}\right)^j cn \\&= cn \sum_{j=0}^{\infty} \left(\frac{9}{10}\right)^j \\&= cn \frac{1}{1 - \frac{9}{10}} \\&= 10cn\end{aligned}$$

- So our conjecture is that  $T(n) = \Theta(n)$ .

# Analysis: Substitution Method

- We now give a rigorous proof that  $T(n) = \Theta(n)$  using the substitution method.
- The proof can be found in the textbook. (Page 221.)

## Lemma

*The number of elements  $\leq x$  is at least  $\frac{3n}{10} - 6$ . Similarly, the number of elements  $> x$  is at least  $\frac{3n}{10} - 6$ .*

(Proof done in class.)

Remark: These are the numbers from the textbook. The alternate proof I made during the lecture also works, and gives  $3n/10 - 2$  and  $3n/10 - 3$ , which is slightly better. But in the end the time bound remains  $\Theta(n)$ .

## Analysis: Substitution Method

- Our analysis above shows that for all  $n \geq 1$

$$T(n) \leq T(\lceil n/5 \rceil) + \max_{1 \leq i \leq 7n/10+6} \{T(i)\} + O(n).$$

- So there are positive constants  $a, b$  such that:

$$T(n) \leq \begin{cases} T(\lceil n/5 \rceil) + \max_{1 \leq i \leq 7n/10+6} \{T(i)\} + an & \text{if } n \geq 140 \\ b & \text{if } n < 140 \end{cases} \quad (1)$$

- We make the inductive hypothesis that  $T(m) \leq cm$  for all  $m < n$ .
- Basis step:** For  $n = 140$ , and thus  $m < 140$ , it suffices to choose  $c \geq b$ .

# Analysis: Substitution Method

## Inductive step:

- We assume that  $n \geq 140$  and that, for all  $m < n$ , we have  $T(m) \leq cm$ .
- We need to prove that  $T(n) \leq cn$ .
- By Equation (1), we obtain

$$T(n) \leq T(\lceil n/5 \rceil) + c \left( \frac{7n}{10} + 6 \right) + an.$$

- Rest of the proof done in class. It suffices to choose  $c \geq 20a$ .

## Concluding Remarks

- The deterministic algorithm for selection is interesting from a theoretical standpoint, but in practice it should be slower than the randomized version. In other words, the constant factor hidden in the notation  $\Theta(n)$  is larger for the deterministic version.
- This is similar to sorting algorithms: `QUICKSORT` is a simple randomized algorithm that outperforms deterministic algorithms with a high probability.