

# CS 580: Algorithm Design and Analysis

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## Order Statistics

The **selection problem** is the problem of computing, given a set  $A$  of  $n$  distinct numbers and a number  $i$ ,  $1 \leq i \leq n$ , the  $i^{th}$  **order statistics** (i.e., the  $i^{th}$  smallest number) of  $A$ .

We will consider some special cases of the order statistics problem:

- the **minimum**, i.e. the first,
- the **maximum**, i.e. the last, and
- the **median**, i.e. the “halfway point.”

## Order Statistics

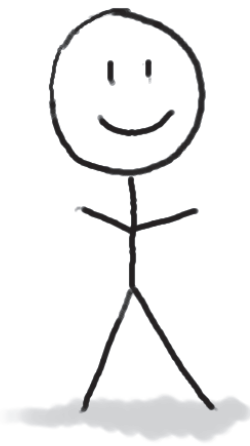
Medians occur at  $i = \lfloor (n + 1)/2 \rfloor$  and  $i = \lceil (n + 1)/2 \rceil$ . If  $n$  is odd, the median is unique, and if  $n$  is even, there are two medians.

## Recall

We work in the **comparison model**:

- Given input array  $x = (x_1, \dots, x_n)$ , we can access the array via comparison queries: “Is  $x_i < x_j$ ?”

Is  bigger than  ?



Algorithm

**YES**



There is a **genie** that knows what the right order is.

The genie can answer YES/NO questions of the form:  
**is [this] bigger than [that]?**

## Finding the Min

How many comparisons are necessary and sufficient for finding the minimum?

## Finding the Min

### Algorithm:

*Given vector  $x = (x_1, \dots, x_n)$  as input, consider the standard algorithm.*

Let  $q = x_1$ .

For each  $i = 2, \dots, n$ :

    If  $x_i < q$ :

$q = x_i$

Return  $q$

This algorithm makes  $n - 1$  comparisons.

## Finding the Min

Can we do better than  $n - 1$  comparisons?

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If not, then show a lower bound of the form:

- Every deterministic algorithm for finding the min (which is correct on every input) makes at least  $n - 1$  comparisons.



## Finding the Min

### Lower Bound:

Consider any (correct) deterministic algorithm  $A$  for finding the min.

Construct a graph  $G$  with:

- vertices  $x_1, \dots, x_n$ .
- edge  $(x_i, x_j)$  if  $A$  compares elements  $x_i$  and  $x_j$  at some point during its execution.

What happens if the graph  $G$  is not connected at the end of  $A$ 's execution?



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If  $G$  is not connected at the end of  $A$ 's execution, then the algorithm can output the wrong answer on some inputs.

For  $G$  to be connected, it must have at least  $n - 1$  edges (if it has exactly  $n - 1$ , it is a tree)  $\Rightarrow A$  makes at least  $n - 1$  comparisons.

## Selection *(Find s-th smallest element)*

Selection is a trivial problem **if the input numbers are sorted**. If we use a sorting algorithm having  $O(n \lg n)$  worst-case running time, then the selection problem can be solved in  $O(n \lg n)$  time.

But using a sorting is more like using a cannon to shoot a fly since only one number needs to be computed.

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But using a sorting is more like using a cannon to shoot a fly since only one number needs to be computed.

**Task:** Design a selection algorithm inspired by QuickSort but with fewer comparisons.

## $O(n)$ expected-time selection using the randomized partition

**Idea:** In order to find the  $s$ -th order statistics in a region of size  $n$ , use the **randomized partition** to split the region into two subarrays. Let  $k - 1$  and  $n - k$  be the size of the left subarray and the size of the right subarray. If  $k = s$ , the pivot is the key that's looked for. If  $s \leq k - 1$ , look for the **?-th element in the left subarray**.



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# Analysis

## Define

- $T(n, s)$  = expected # comparisons for selection of s-th statistic
- $T(n) = \max_s T(n, s)$  is the expected # comparisons of selection for the worst case index s.



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**Task:** Write an inequality to upper bound  $T(n)$  as a function of the amount of work done in Partition and in recursive selection calls.

## Analysis

**Recall:**  $T(n)$  is the expected # comparisons of selection for the worst case index  $s$ .

For each  $i$ ,  $0 \leq i \leq n - 1$ , the size of the left subarray is equal to  $i$  with probability  $1/n$ .

Assuming that the larger interval is taken, for some  $\alpha > 0$ ,  $T(n)$  is at most

$$\underbrace{\alpha n}_{\text{Work for partition}} + \frac{1}{n} \sum_{1 \leq k \leq n-1, k \neq s} T(\max(k, n-k)).$$

This is at most

*Expected work for recursive call*

$$\alpha n + \frac{2}{n} \left( \sum_{k=\lceil n/2 \rceil}^{n-1} T(k) \right).$$

## Analysis (cont'd)

Assume that there is  $c > 0$  such that  $T(k) \leq ck$  for all  $k < n$ .

Then the sum  $\sum_{k=\lceil n/2 \rceil}^{n-1} T(k)$  is at most  $\sum_{k=\lceil n/2 \rceil}^{n-1} ck$ . This is at most

$$\begin{aligned} & \sum_{k=1}^{n-1} ck - \sum_{k=1}^{\lceil n/2 \rceil - 1} ck \\ &= \frac{cn(n-1)}{2} - \frac{c}{2} \left( \left\lceil \frac{n}{2} \right\rceil - 1 \right) \left\lceil \frac{n}{2} \right\rceil \\ &\leq \frac{cn(n-1)}{2} - \frac{c}{2} \left( \frac{n}{2} - 1 \right) \frac{n}{2} \\ &= cn \left( \frac{3n}{8} - \frac{1}{4} \right). \end{aligned}$$

## Analysis (cont'd)

So, if  $c$  is sufficiently large,

$$T(n) \leq \alpha n + c \left( \frac{3}{4}n - \frac{1}{2} \right).$$

By making the constant  $c$  at least  $4\alpha$ , we have that  $\alpha n$  is at most  $\frac{cn}{4}$ . Then  $T(n) \leq c \cdot n$ .

## Min and Max

*How many comparisons are necessary and sufficient for computing both the minimum and the maximum?*

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Well, to compute the maximum  $n - 1$  comparisons are necessary and sufficient. The same is true for the minimum. So, the number should be  $2n - 2$  for computing both.

## Min and Max

**Hint:** *Actually you can do better by  
processing the input numbers  
in pairs*

## Min and Max Algorithm

Simultaneous computation of max and min can be done in  $\leq 3n/2$  comparisons

Assume  $n$  is even:

- Divide the numbers in pairs and find the larger and smaller one in each pair
- From the  $n/2$  larger items, find the maximum
- From the  $n/2$  smaller items, find the minimum

For  $n$  odd, compare the remaining ( $n$ -th item) with both the min and max.