# CS4311 Design and Analysis of Algorithms

Lecture 8: Order Statistics

#### About this lecture

- Finding max, min in an unsorted array (upper bound and lower bound)
- Finding both max and min (upper bound)
- Selecting the k<sup>th</sup> smallest element

 $k^{th}$  smallest element  $\equiv k^{th}$  order statistics

# Finding Maximum in unsorted array

# Finding Maximum (Method I)

- Let S denote the input set of n items
- To find the maximum of S, we can:

```
Step 1: Set max = item 1
Step 2: for k = 2, 3, ..., n
if (item k is larger than max)
Update max = item k;
```

Step 3: return max;

# comparisons = n-1

# Finding Maximum (Method II)

- Define a function Find-Max as follows:
- Find-Max(R, k) /\* R is a set with k items \*/
  - 1. if  $(k \le 2)$  return maximum of R;
  - 2. Partition items of R into |k/2| pairs;
  - 3. Delete smaller item from R in each pair;
  - 4. return Find-Max(R, k [k/2]);

Calling Find-Max(S,n) gives the maximum of S

# Finding Maximum (Method II)

Let T(n) = # comparisons for Find-Max with problem size n

So, 
$$T(n) = T(n - \lfloor n/2 \rfloor) + \lfloor n/2 \rfloor$$
 for  $n \ge 3$   
 $T(2) = 1$ 

Solving the recurrence (by substitution), we get T(n) = n - 1

#### Lower Bound

Question: Can we find the maximum using fewer than n - 1 comparisons?

Answer: No! To ensure that an item x is not the maximum, there must be at least one comparison in which x is the smaller of the compared items

So, we need to ensure n-1 items not max

→ at least n - 1 comparisons are needed

# Finding Both Max and Min in unsorted array

Can we find both max and min quickly?

#### Solution 1:

First, find max with n - 1 comparisons

Then, find min with n - 1 comparisons

 $\rightarrow$  Total = 2n - 2 comparisons

Is there a better solution??

Better Solution: (Case 1: if n is even)

First, partition items into n/2 pairs;

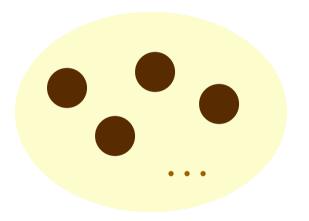


Next, compare items within each pair;

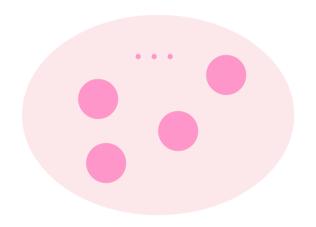




Then, max = Find-Max in larger items min = Find-Min in smaller items







Find-Min

# comparisons = 3n/2 - 2

```
Better Solution: (Case 2: if n is odd)

We find max and min of first n - 1 items;

if (last item is larger than max)

Update max = last item;

if (last item is smaller than min)

Update min = last item;
```

# comparisons = 3(n-1)/2

#### Conclusion:

To find both max and min:

if n is odd: 3(n-1)/2 comparisons

if n is even: 3n/2 - 2 comparisons

Combining: at most 3[n/2] comparisons

> better than finding max and min separately

#### Lower Bound

Textbook Ex 9.1-2 (Very challenging):

Show that we need at least

 $\lceil 3n/2 \rceil - 2$  comparisons

to find both max and min in worst-case

Hint: Consider how many numbers may be max or min (or both). Investigate how a comparison affects these counts

# Selecting kth smallest item in unsorted array

#### Selection in Linear Time

- In next slides, we describe a recursive call Select(5,k)
  - which supports finding the kth smallest element in S
- Recursion is used for two purposes:
  - (1) selecting a good pivot (as in Quicksort)
  - (2) solving a smaller sub-problem

#### Select(S, k)

```
/* First, find a good pivot */
1. Partition S into [|S|/5] groups, each
  group has five items (one group may
  have fewer items);
```

- 2. Sort each group separately;
- 3. Collect median of each group into 5';
- 4. Find median m of S':

```
m = Select(S', \lceil |S|/5 \rceil/2 \rceil);
```

```
4. Let q = \# items of S smaller than m;
5. If (k == q + 1)
       return m:
/* Partition with pivot */
6. Else partition S into X and Y
     X = {items smaller than m}
     Y = {items larger than m}
/* Next, form a sub-problem */
7. If (k < q + 1)
       return Select(X, k)
8. Else
       return Select(Y, k-(q+1));
```

#### Selection in Linear Time

#### Questions:

Why is the previous algorithm correct?
 (Prove by Induction)

2. What is its running time?

### Running Time

- In our selection algorithm, we chose m, which is the median of medians, to be a pivot and partition S into two sets X and Y
- In fact, if we choose any other item as the pivot, the algorithm is still correct
- Why don't we just pick an arbitrary pivot so that we can save some time??

### Running Time

- A closer look reviews that the worst-case running time depends on |X| and |Y|
- Precisely, if T(|S|) denote the worst-case running time of the algorithm on S, then

```
T(|S|) = T(|S|/5|) + \Theta(|S|) + \max \{T(|X|), T(|Y|)\}
```

### Running Time

 Later, we show that if we choose m, the "median of medians", as the pivot,

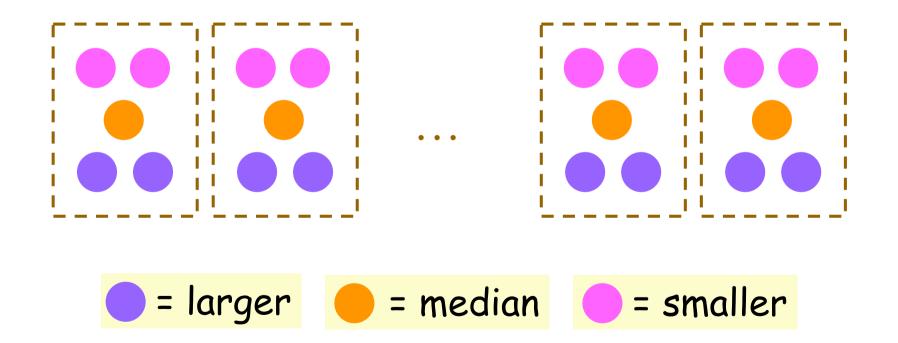
both |X| and |Y| will be at most 3|5|/4

· Consequently,

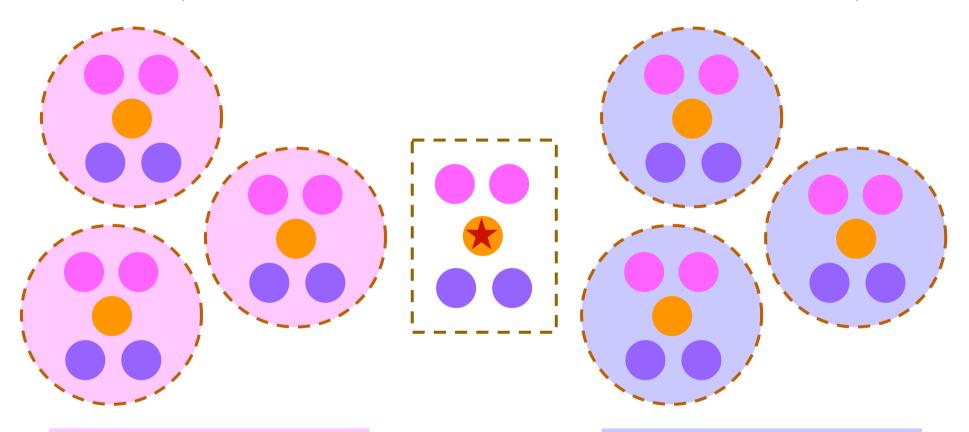
$$T(n) = T(\lceil n/5 \rceil) + \Theta(n) + T(3n/4)$$

 $\rightarrow$  T(n) =  $\Theta$ (n) (obtained by substitution)

 Let's begin with [n/5] sorted groups, each has 5 items (one group may have fewer)



· Then, we obtain the median of medians, m

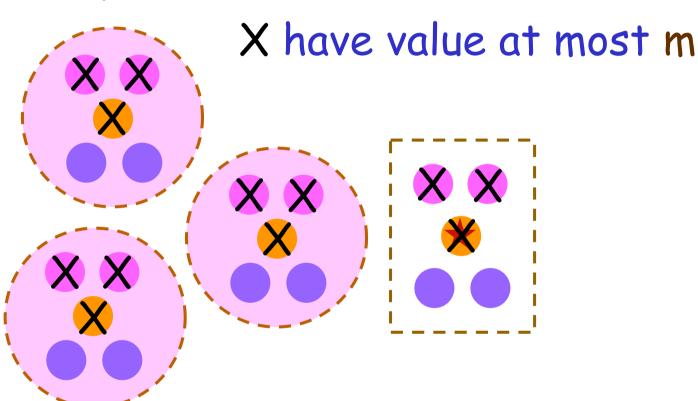


Groups with median smaller than m



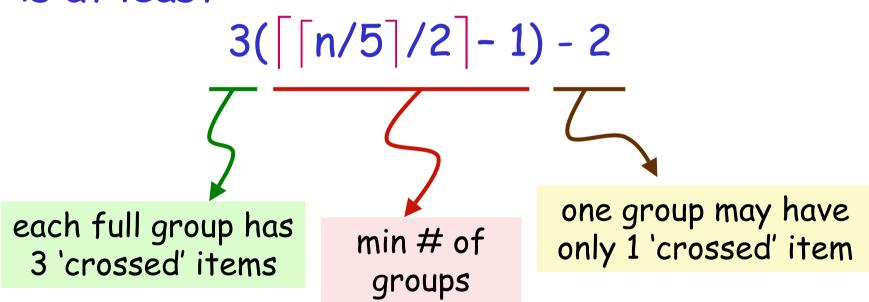
Groups with median larger than m

Then, we know that all items marked with



Groups with median smaller than m

The number of items with value at most m is at least



→ number of items: at least 3n/10 - 5

Previous page implies that at most

$$7n/10 + 5$$
 items

are greater than m

- For large enough n (say,  $n \ge 100$ )  $7n/10 + 5 \le 3n/4$
- $\rightarrow$  |Y| is at most 3n/4 for large enough n

- Similarly, we can show that at most 7n/10 + 5 items are smaller than m
- $\rightarrow$  |X| is at most 3n/4 for large enough n

#### Conclusion:

The "median of medians" helps us control the worst-case size of the sub-problem

 $\rightarrow$  without it, the algorithm runs in  $\Theta(n^2)$  time in the worst-case