



南京大學

NANJING UNIVERSITY

Introduction to

# *Algorithm Design and Analysis*

[7] Selection



*Yu Huang*

<http://cs.nju.edu.cn/yuhuang>  
Institute of Computer Software  
Nanjing University



# In the last class...

- **MergeSort**
  - Design
  - Cost – time & space
  - MergeSort DC
- **Lower bounds for comparison-based sorting**
  - Worst-case
  - Average-case



# Selection

- **Selection – warm-ups**
  - Finding *max* and *min*
  - Finding the *second largest* key
- **Lower bound and *adversary argument***
- **Selection – select the *median***
  - Expected linear time
  - Worst-case linear time
- **A Lower Bound for Finding the Median**



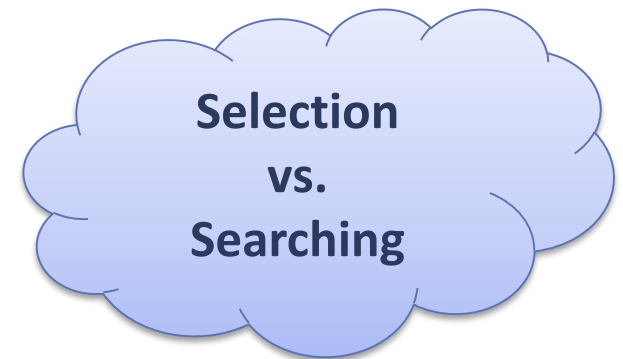
# The Selection Problem

- **Problem definition**

- Suppose  $E$  is an array containing  $n$  elements with keys from some linearly order set, and let  $k$  be an integer such that  $1 \leq k \leq n$ . The selection problem is to find an element with the  $k^{\text{th}}$  smallest key in  $E$ .

- **Special cases**

- Find the max/min –  $k=n$  or  $k=1$
- Find the *median* ( $k = \frac{n}{2}$ )



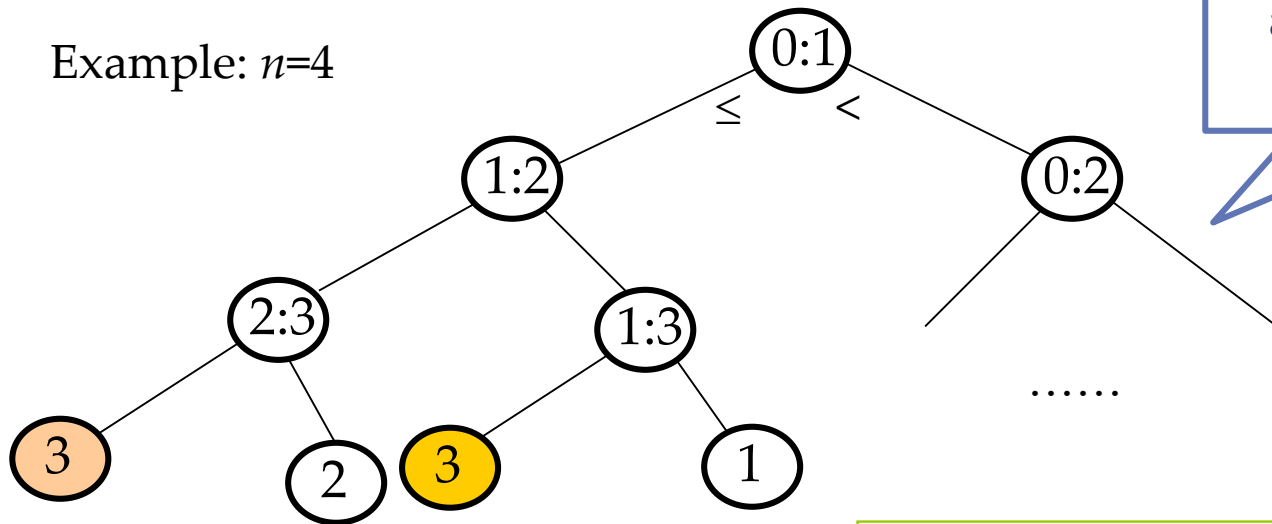
# Lower Bound of Finding the Max

- For **any** algorithm  $\mathcal{A}$  that can compare and copy numbers exclusively, in the worst case,  $\mathcal{A}$  cannot do fewer than  **$n-1$**  comparisons to find the largest entry in an array with  $n$  entries.
  - Proof: an array with  $n$  distinct entries is assumed. We can exclude a specific entry from being the largest entry only after it is determined to be “loser” to at least one entry. So,  $n-1$  entries must be “losers” in comparisons done by the algorithm. However, each comparison has only one loser, so at least  $n-1$  comparisons must be done.

# Decision Tree and Lower Bound

Since the decision tree for the selection problem must have at least  $n$  leaves, the height of the tree is at least  $\lceil \log n \rceil$ . **It is not a good lower bound.**

Example:  $n=4$



A more powerful tool for analysis is necessary:  
**adversary argument**

There are more than  $n$  leaves!  
In fact,  $2^{n-1}$  leaves at least.

# Finding *max* and *min*

- The strategy
  - Pair up the keys, and do  $n/2$  comparisons (if  $n$  odd, having  $E[n]$  uncompared);
  - Doing findMax for larger key set and findMin for small key set respectively (if  $n$  odd,  $E[n]$  included in both sets)
- Number of comparisons
  - For even  $n$ :  $n/2 + 2(n/2 - 1) = 3n/2 - 2$
  - For odd  $n$ :  $(n-1)/2 + 2((n-1)/2 + 1 - 1) = \lceil 3n/2 \rceil - 2$

**How to prove this lower bound?**

**Adversary  
Argument !**



# Unit of Information

- **Max and Min**
  - That  $x$  is *max* can only be known when it is sure that every key other than  $x$  has **lost some comparison**.
  - That  $y$  is *min* can only be known when it is sure that every key other than  $y$  has **win some comparison**.
- **Each win or loss is counted as one unit of information**
  - *Any* algorithm must have at least  $2n-2$  units of information to be sure of specifying the *max* and *min*.





# Adversary Strategy

Status of keys $x$ and $y$ Compared by an algorithm	Adversary response	New status	Units of new information
N,N	$x > y$	W,L	2
W,N or WL,N	$x > y$	W,L or WL,L	1
L,N	$x < y$	L,W	1
W,W	$x > y$	W,WL	1
L,L	$x > y$	WL,L	1
W,L or WL,L or W,WL	$x > y$	No change	0
WL,WL	Consistent with Assigned values	No change	0

The principle: let the key win if it never lose, or,  
let the key lose if it never win, and

**change one value if necessary**

# Lower Bound by the Adversary Argument

- Construct an input to force *the* algorithm to do more comparisons as possible
  - To give away as few as possible units of new information with each comparison.
    - It can be achieved that 2 units of new information are given away only when the status is N,N.
    - It is *always* possible to give adversary response for other status so that at most one new unit of information is given away, *without any inconsistencies*.
- So, the **Lower Bound** is  $n/2 + n - 2$  (for even  $n$ )

$$\frac{n}{2} \times 2 + (n - 2) \times 1 = 2n - 2$$

# An Example

Comparison	$x_1$		$x_2$		$x_3$		$x_4$		$x_5$		$x_6$	
	S	V	S	V	S	V	S	V	S	V	S	V
$x_1, x_2$					N	*	N	*	N	*	N	*
$x_1, x_5$									L	5		
$x_3, x_4$												
$x_3, x_6$											L	12
$x_3, x_1$												
$x_2, x_4$												
$x_5, x_6$									WL	5	L	3
$x_6, x_4$							L	2			WL	3

Now,  $x_3$  is the only one which never loses, so, Max is  $x_3$

Now,  $x_3$  is the only

8 comparisons!  
The lower bound is 7.

Raising/lowering the value according to strategy

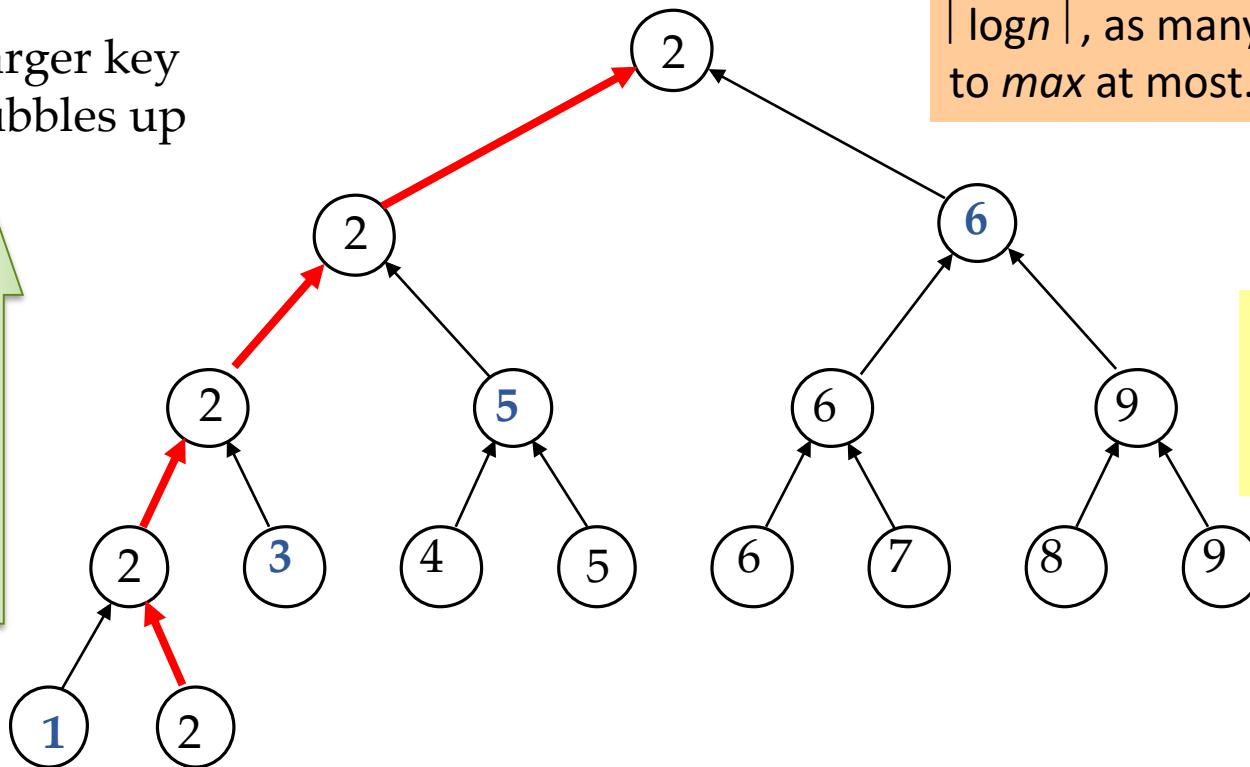
# Find the 2<sup>nd</sup> Largest Key

- **Brute force - using FindMax twice**
  - Need  $2n-3$  comparisons.
- **For a better algorithm**
  - Collect some useful information from the first FindMax
- **Observations**
  - The key which **loses to a key other than max** cannot be the 2<sup>nd</sup> largest key.
  - To check “whether you lose to max?”



# Tournament for the 2<sup>nd</sup> Largest Key

Larger key bubbles up



The length of the longest path is  $\lceil \log n \rceil$ , as many as those compared to *max* at most.

$x_2$  is *max*

Only  $x_1, x_3, x_5, x_6$  may be the second largest key.

# Analysis of Finding the 2<sup>nd</sup>

- Any algorithm that finds *secondLargest* must also find *max* before. ( $n-1$ )
- The *secondLargest* can only be in those which lose directly to *max*.
- On its path along which bubbling up to the root of tournament tree, *max* beat  $\lceil \log n \rceil$  keys at most.
- Pick up *secondLargest* ( $\lceil \log n \rceil - 1$ )
- Total cost:  $n + \lceil \log n \rceil - 2$



# Lower Bound by Adversary

- **Theorem**

- Any algorithm (that works by comparing keys) to find the second largest in a set of  $n$  keys must do at least  $n + \lceil \log n \rceil - 2$  comparisons in the worst case.

- **Proof**

- There is an adversary strategy that can force any algorithm that finds *secondLargest* to compare *max* to  $\lceil \log n \rceil$  distinct keys.



# Weighted Key

- Assigning a weight  $w(x)$  to each key
  - The initial values are all 1.
- Adversary strategy

Note: for one comparison, the weight increasing is no more than doubled.

Case	Adversary reply	Updating of weights
$w(x) > w(y)$	$x > y$	$w(x) := w(x) + w(y); w(y) := 0$
$w(x) = w(y) > 0$	$x > y$	$w(x) := w(x) + w(y); w(y) := 0$
$w(y) > w(x)$	$y > x$	$w(y) := w(x) + w(y); w(x) := 0$
$w(x) = w(y) = 0$	Consistent with previous replies	No change



**Zero loss**



# Lower Bound by Adversary: Details

- Note: the sum of weights is always  $n$ .
- Let  $x$  is *max*, then  $x$  is the only nonzero weighted key, that is  $w(x)=n$ .
- By the adversary rules:

$$w_k(x) \leq 2w_{k-1}(x)$$

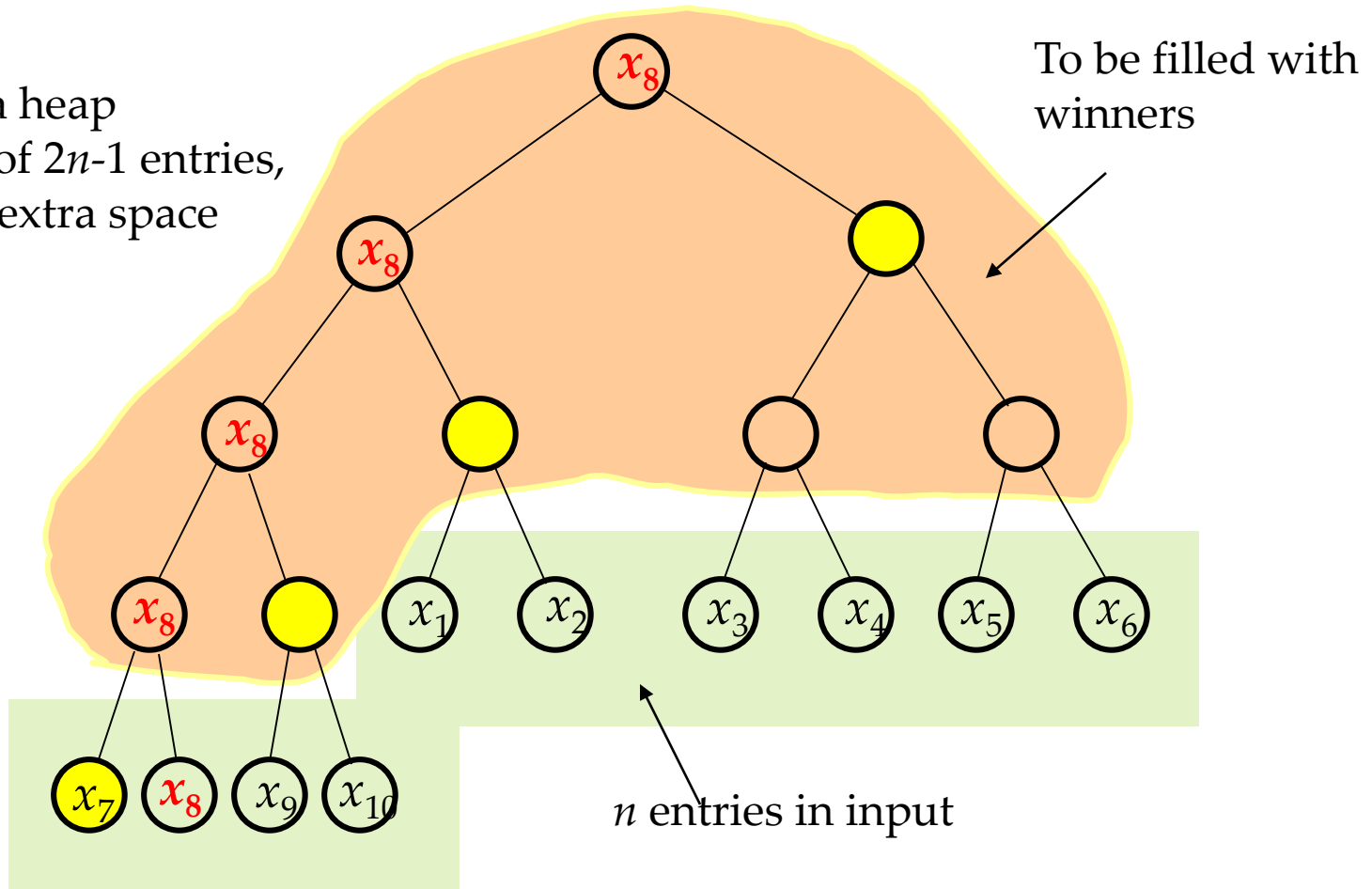
- Let  $K$  be the number of comparisons  $x$  wins against previously undefeated keys:

$$n = w_K(x) \leq 2^K w_0(x) = 2^K$$

- So,  $K \geq \lceil \log n \rceil$

# Tracking the Losers to *MAX*

Building a heap structure of  $2n-1$  entries, using  $n-1$  extra space



# Finding the Median: the Strategy

- **Observation**

- If we can partition the problem set of keys into 2 subsets:  $S_1$ ,  $S_2$ , such that any key in  $S_1$  is smaller than that of  $S_2$ , the median must be located in the set with more elements.

- **Divide-and-Conquer**

- Only one subset is needed to be processed recursively.

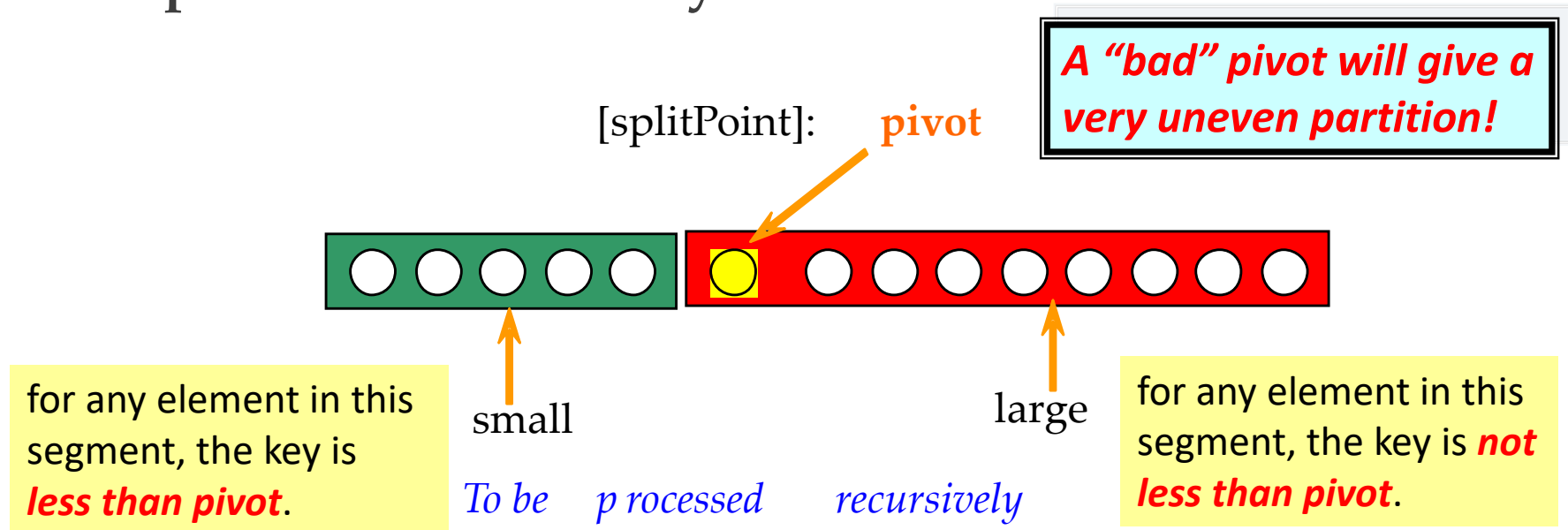


# Adjusting the Rank

- The rank of the median (of the original set) in the subset considered can be evaluated easily.
- An example
  - Let  $n=255$
  - The rank of median we want is 128
  - Assuming  $|S_1|=96$ ,  $|S_2|=159$
  - Then, the original median is in  $S_2$ , and the new rank is  $128-96=32$

# Partitioning: Larger and Smaller

- Dividing the array to be considered into two subsets: “small” and “large”, the one with more elements will be processed recursively.



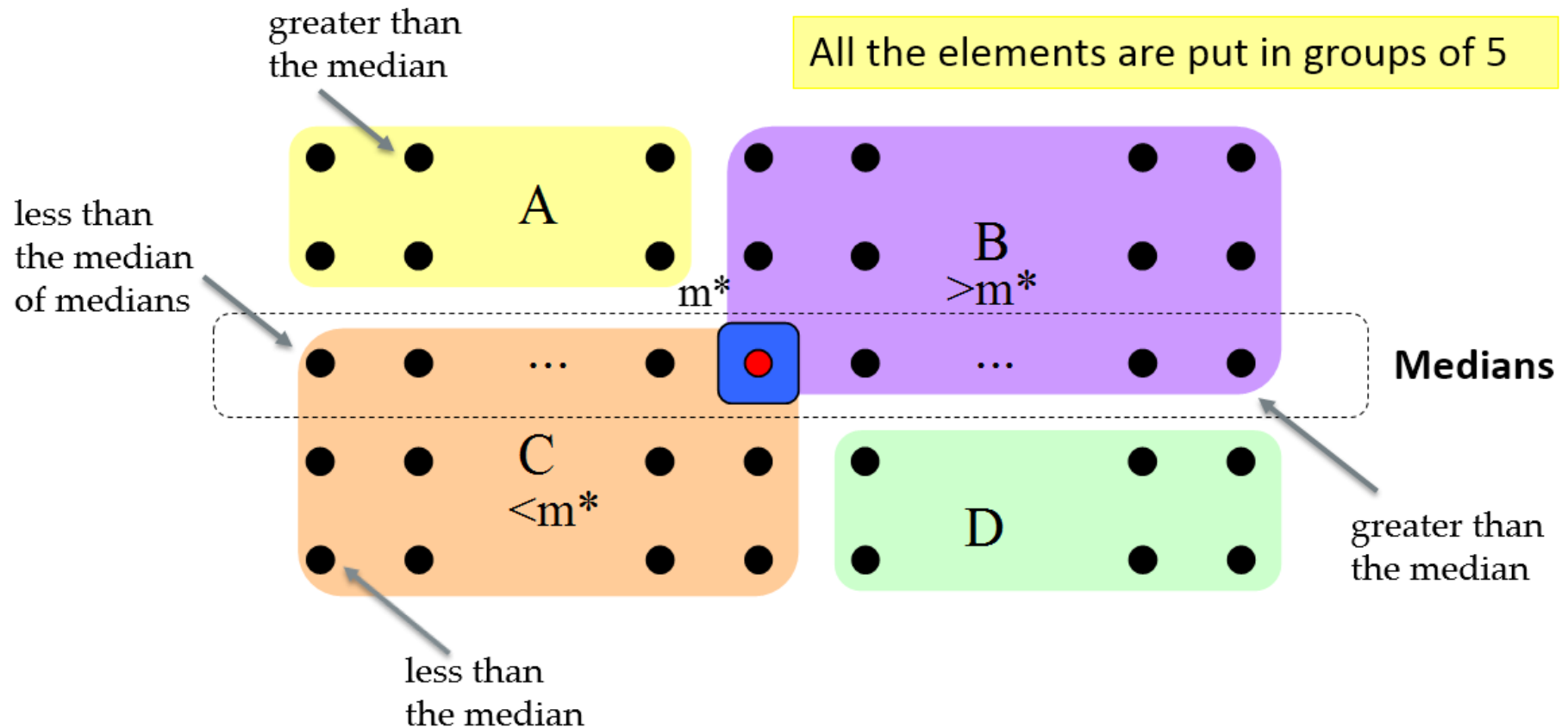
# Selection: the Algorithm

- Input:  $S$ , a set of  $n$  keys; and  $k$ , an integer such that  $1 \leq k \leq n$ .
- Output: The  $k$ th smallest key in  $S$ .
- Note: Median selection is only a special case of the algorithm, with  $k = \lceil n/2 \rceil$ .
- Procedure
- Element select(SetOfElements  $S$ , int  $k$ )
  - if ( $|S| \leq 5$ ) return direct solution; else
  - Constructing the subsets  $S_1$  and  $S_2$ ;
  - Processing one of  $S_1, S_2$  with more elements, recursively.

Key issue:

How to construct the **partition** ?

# Partition improved: the Strategy



# Constructing the Partition

- Find the  $m^*$ , the median of medians of all the groups of 5, as illustrated previously.
- Compare each key in sections A and D to  $m^*$ , and
  - Let  $S_1 = C \cup \{x \mid x \in A \cup D \text{ and } x < m^*\}$
  - Let  $S_2 = B \cup \{x \mid x \in A \cup D \text{ and } x > m^*\}$

( $m^*$  is to be used as the pivot for the partition)



# Divide and Conquer

```
if ( $k = |S_1| + 1$ )  
    return  $m^*$ ;  
else if ( $k \leq |S_1|$ )  
    return select( $S_1, k$ ); //recursion  
else  
    return select( $S_2, k - |S_1| - 1$ ); //recursion
```



# Analysis

- For simplicity:
  - Assuming  $n=5(2r+1)$  for all calls of *select*.

- $$W(n) \leq 6\left(\frac{n}{5}\right) + W\left(\frac{n}{5}\right) + 4r + W(7r + 2)$$

Finding the median  
in every group of 5

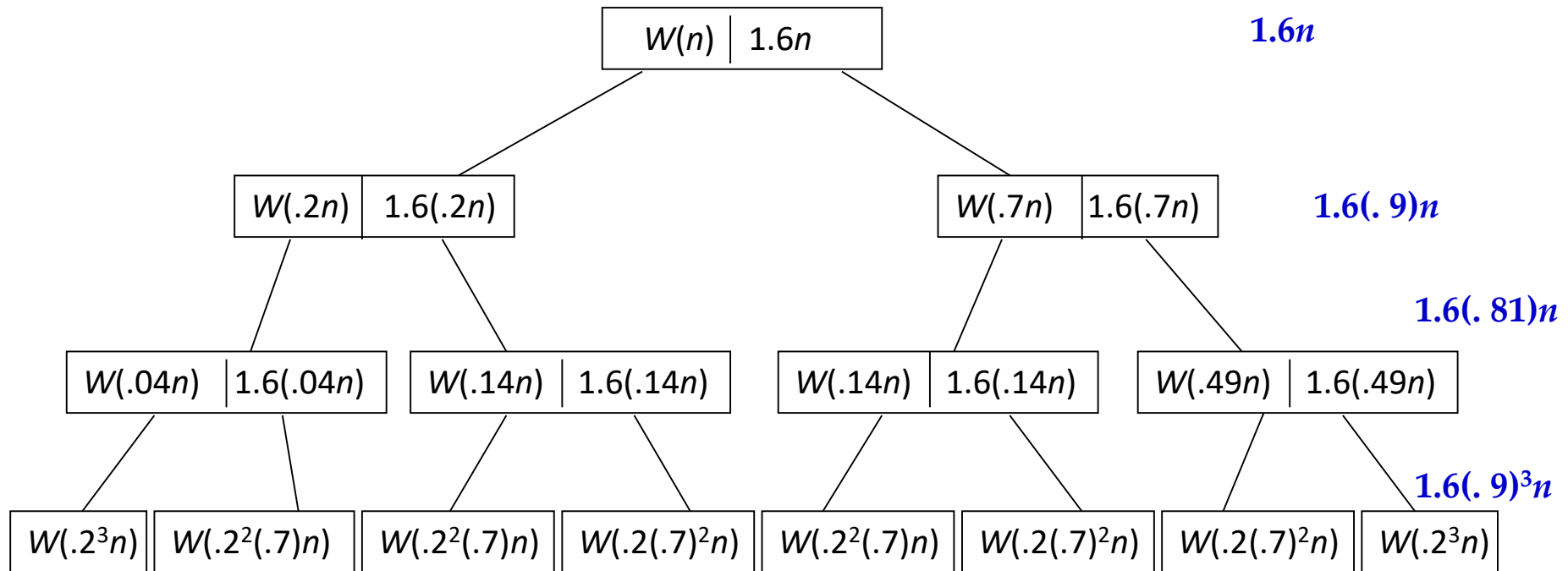
Finding the median  
of the medians

Comparing all the elements  
in  $A \cup D$  with  $m^*$

The extreme case: all  
the elements in  $A \cup D$   
in one subset.

- *Note:  $r$  is about  $n/10$ , and  $0.7n+2$  is about  $0.7n$ , so*  
$$W(n) \leq 1.6n + W(0.2n) + W(0.7n)$$

# Worst Case Complexity of *Select*



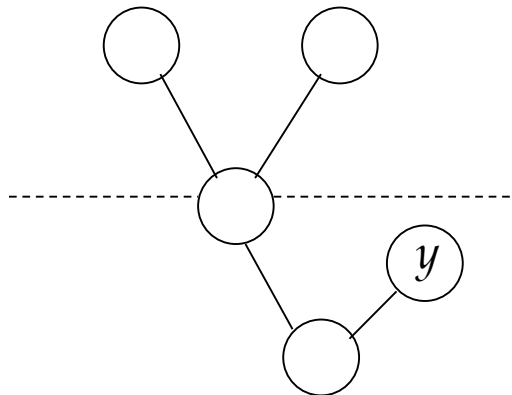
**Note:** Row sums is a decreasing geometric series, so

$$W(n) \in \Theta(n)$$

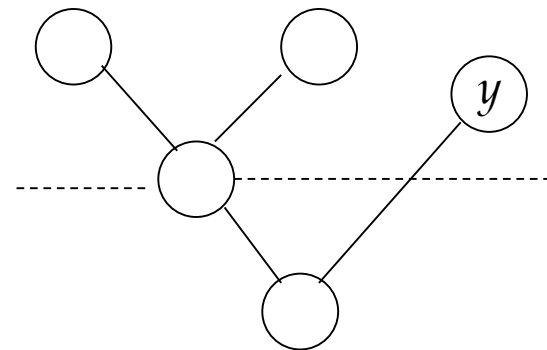
# Relation to Median

- **Observation**

- Any algorithm of selection must know the relation of every element to the *median*.



Median



The adversary makes you wrong in either case

# Crucial Comparison

- **A crucial comparison**
  - Establishing the relation of some  $x$  to the median.
- **Definition** (for a comparison involving a key  $x$ )
  - **Crucial comparison for  $x$** : the first comparison where  $x > y$ , for some  $y \geq \text{median}$ , or  $x < y$  for some  $y \leq \text{median}$
  - **Non-crucial comparison**: the comparison between  $x$  and  $y$  where  $x > \text{median}$  and  $y < \text{median}$ , or vice versa

# Adversary for Lower Bound

- Status of the key during the running of the Algorithm:
  - $L$ : Has been assigned a value *larger* than median
  - $S$ : Has been assigned a value *smaller* than median
  - $N$ : Has not yet been in a comparison

- Adversary rule:

Comparands

Adversary's action

$N, N$   
 $L, N$  or  $N, L$   
 $S, N$  or  $N, S$

one  $L$ , the another  $S$   
change  $N$  to  $S$   
change  $N$  to  $L$

(In all other cases, just keep consistency)



# Notes on the Adversary Arguments

- All actions explicitly specified above make the comparisons un-crucial.
  - At least,  $(n-1)/2$   $L$  or  $S$  can be assigned freely.
  - If there are already  $(n-1)/2$   $S$ , a value **larger** than median must be assigned to the new key, and if there are already  $(n-1)/2$   $L$ , a value **smaller** than median must be assigned to the new key. The last assigned value is the median.
- So, an adversary can force the algorithm to do  $(n-1)/2$  un-crucial comparisons at least (In the case that the algorithm start out by doing  $(n-1)/2$  comparisons involving two  $N$ ).



# Lower Bound for Selection Problem

- **Theorem:**

- Any algorithm to find the median of  $n$  keys (for odd  $n$ ) by comparison of keys must do at least  $3n/2 - 3/2$  comparisons in the worst case.

- **Argument:**

- There must be done  $n-1$  crucial comparisons at least.
- An adversary can force the algorithm to perform as many as  $(n-1)/2$  noncrucial comparisons.
  - Note: the algorithm can always start out by doing  $(n-1)/2$  comparisons involving 2  $N$ -keys, so, only  $(n-1)/2$   $L$  or  $S$  left for the adversary to assign freely as the adversary rule.





*Thank you!*

*Q & A*

*Yu Huang*

<http://cs.nju.edu.cn/yuhuang>

