Introduction to

Algorithm Design and Analysis

[07] Selection

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In the last class ...

- MergeSort
 - Worst-case analysis of MergeSort
- Lower Bounds for comparison-based sorting
 - Worst-case
 - Average-case

The Selection

- Selection warm-ups
 - Finding max and min
 - Finding the second largest key
- Adversary argument and lower bound
- 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<=k<=n. The selection problem is to find an element with the kth smallest key in E.

Special cases

- Find the max/min -k=n or k=1
- Find the median (k=n/2)

Selection v.s.
Searching

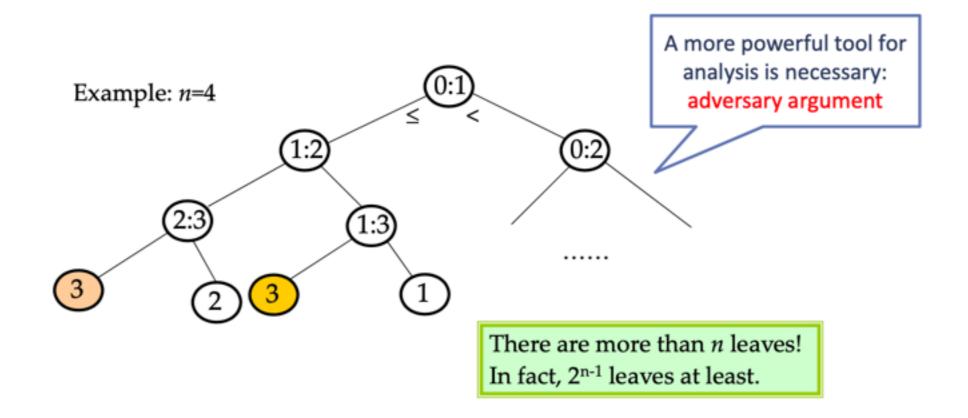
Lower Bound of Finding the Max

- For any algorithm A that can compare and copy numbers exclusively, in the worst case, 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.

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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's not a good lower bound.



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 $n:(n-1)/2 + 2((n-1)/2 + 1 1) = \lceil 3n/2 \rceil 2$

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.

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Adversary Strategy

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

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)

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

Find the 2nd Largest Key

- Brute force using FindMax twice
 - Need 2n-3 comparisons.

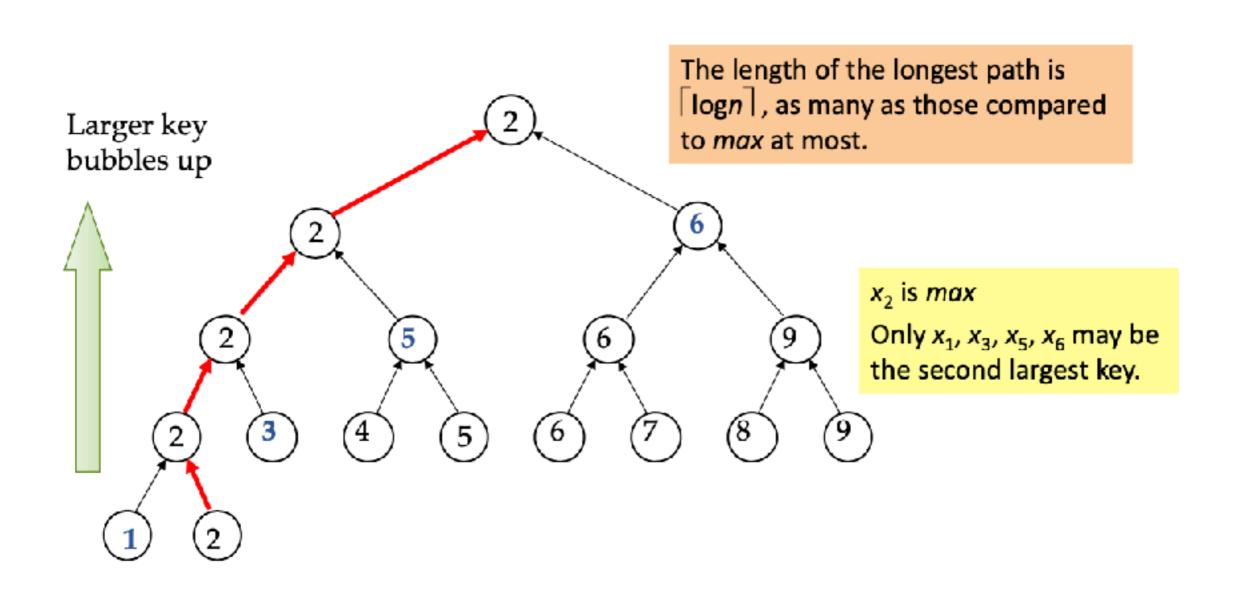
For a better algorithm

Collect some useful information from the first FindMax

Observations

- The key which loses to a key other than max cannot be the 2nd largest key.
- To check "whether you lose to max?"

Tournament for the 2nd Largest Key



Analysis of Finding the 2nd

- Any algorithm that finds secondLargest must also find max before.
- 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 $(\lceil \log n \rceil 1)$
- Totalcost: $n + \lceil \log n \rceil 2$

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	<i>x>y</i>	w(x):=w(x)+w(y); w(y):=0	
w(y)>w(x)	<i>y>x</i>	w(y):=w(x)+w(y); w(x):=0	
w(x)=w(y)=0	Consistent with previous replies	No change	

Zero loss

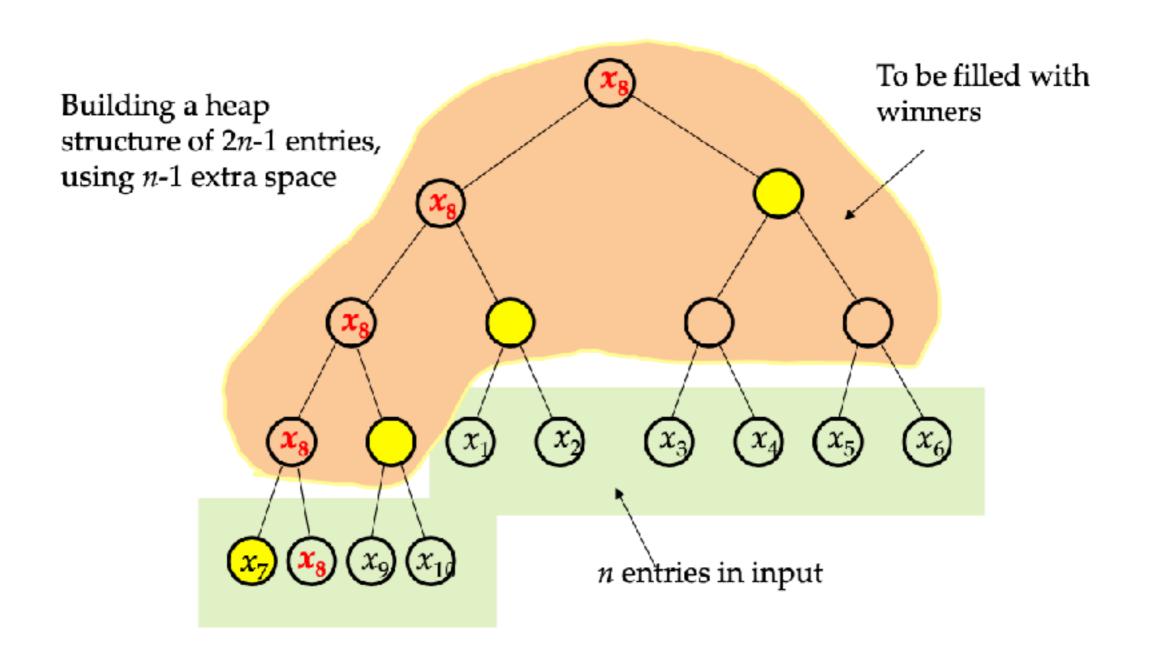
Lower Bound by Adversary: Details

- 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: $w_k(x) \le 2w_{k-1}(x)$
- Let K be the number of comparisons x wins against previously undefeated keys:

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

• So, $K \leq \lceil \log n \rceil$

Tracking the Losers to MAX



Finding the Median: the Strategy

Observation

 If we can partition the problem set of keys into 2 subsets: S1, S2, such that any key in S1 is smaller that that of S2, the median must located in the set with more elements.

Divide-and-Conquer

 Only one subset is needed to be processed recursively.

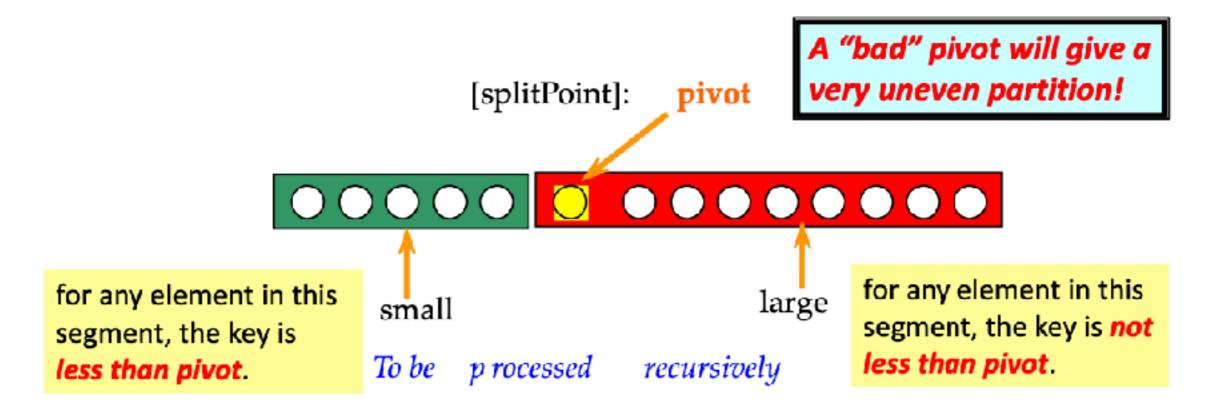
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Adjusting the Rank

- 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 |S₁|=96, |S₂|=159
 - Then, the original median is in S₂, 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.



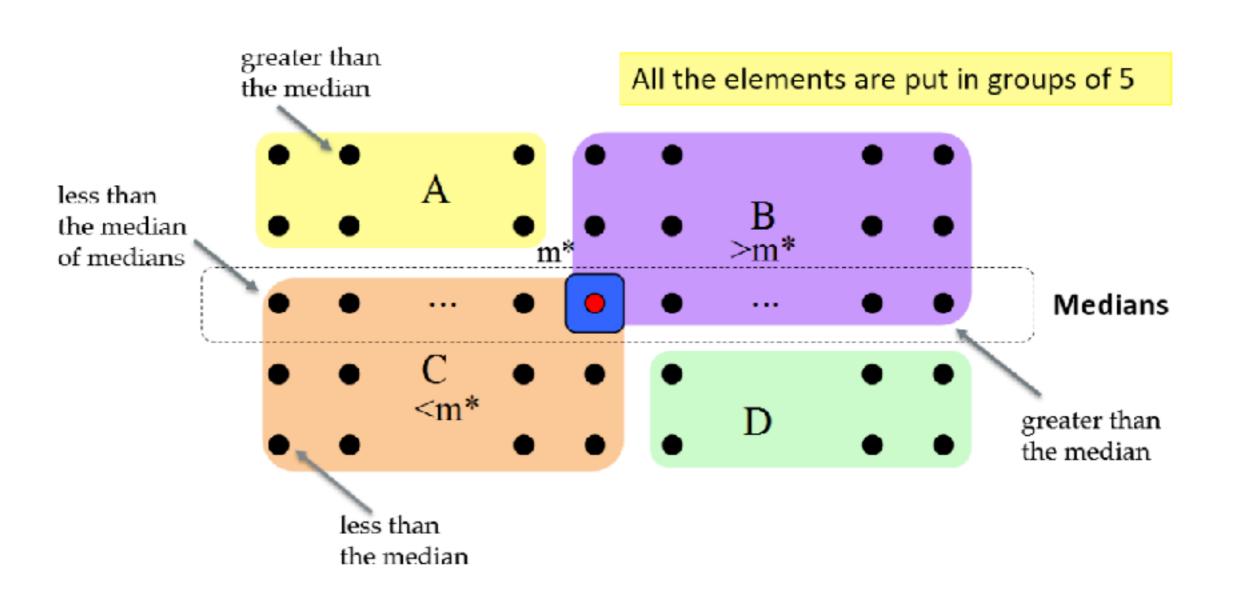
Selection: the Algorithm

- Input: S, a set of n keys; and k, an integer such that $1 \le k \le n$.
- Output: The kth 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|<=5 return direct solution; else

Key issue:

- Constructing the subsets S_1 and S_2 : How to construct the partition?
- Processing one of S₁,S₂ with more elements, recursively.

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 $S_1 = C \cup \{x \mid x \in A \cup Dandx < m^*\}$
 - Let $S_2 = B \cup \{x \mid x \in A \cup Dandx > m^*\}$ (m* is to be used as the pivot for the partition)

Divide and Conquer

- if $(k=|S_1|+1)$ return m^* ;
- else if $(k <= |S_1|)$ return select (S_1,k) ; //recursion
- else return select(S₂,k-|S₁|-1); //recursion

Analysis

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

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

The extreme case: all the elements in A∪D in one subset.

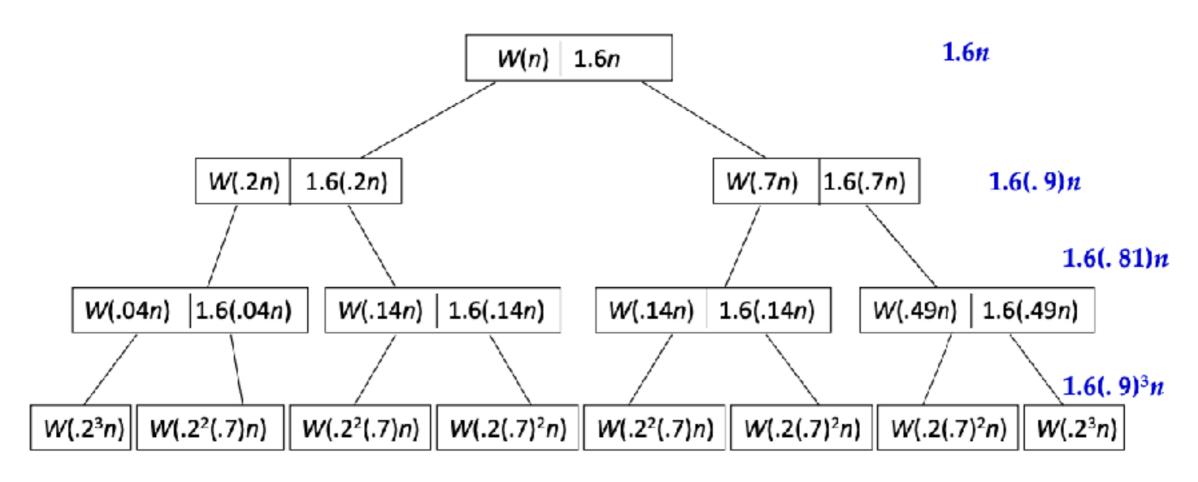
Finding the median in every group of 5

Finding the median of the medians

Comparing all the elements in A∪D with *m**

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

Worst Case Complexity of Select

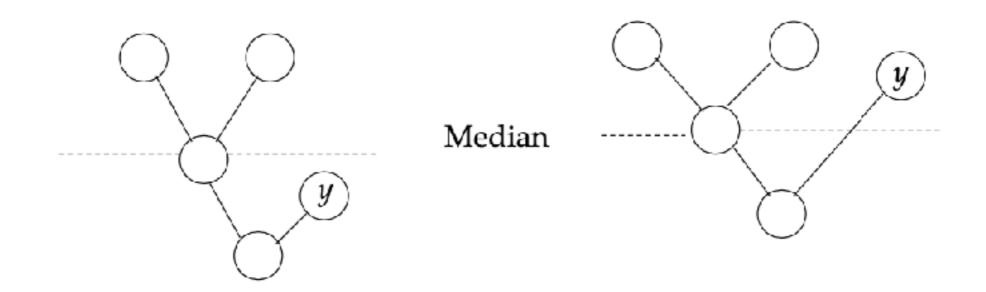


Note: Row sums is a decreasing geometric series, so $W(n) \in \Theta(n)$

Relation to Median

Observation

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



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=>median, or x<y for some y<=median
 - Non-crucial comparison: the comparison between x and y where x>median and y<median, or vise 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 one L, the another S L,N or N,L change N to S S,N or N,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 uncrucial 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 uncrucial 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