

EECS 3101 A: Design and Analysis of Algorithms

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Course page: <http://www.eecs.yorku.ca/course/3101A>
Also on Moodle

Lower Bounds

- Applicable to a problem, not an algorithm
- We will prove lower bounds on the worst-case running time of an algorithm
- Warning: we must reason about all algorithms, so we have to be careful not to assume anything about how the algorithm proceeds
- We can have lower bounds on running time, memory, number of times a specific operation is used.....

Lower Bound for a Simple Problem: FindMax

- Consider only comparison-based algorithms
- Want to show any such algorithm must use $\Omega(n)$ comparisons in the worst case
- We will show a more exact result in this case
- Note that the number of comparisons is a lower bound on the running time of an algorithm

Proof of Lower Bound

- Claim: Any comparison-based algorithm for finding the maximum of n *distinct* elements must use at least $n - 1$ comparisons.
- Proof:
If x, y are compared and $x > y$, call x the winner, y the loser.
Any key that is not the maximum must have lost at least one comparison. WHY?
Each comparison produces exactly one loser and at most one NEW loser.
Therefore, at least $n - 1$ comparisons have to be made.

Observations

We proved a claim about ANY algorithm that only uses comparisons to find the maximum. Specifically, we made no assumptions about

- Nature of algorithm
- Order or number of comparisons
- Optimality of algorithm
- Whether the algorithm is “reasonable”, e.g. it could be a very wasteful algorithm, repeating the same comparisons

Lower Bounds for Sorting - Big Picture

- Can we beat the $\Omega(n \log n)$ lower bound for sorting?
- A: In general no, but in some special cases YES!
- Ch 7: Sorting in linear time
- We will prove the $\Omega(n \log n)$ lower bound.

Lower Bounds for Sorting- Details

- What (if any) are the assumptions?
- Is the model general enough?

Here we are interested in lower bounds for the WORST CASE.
So we will prove (directly or indirectly):

For any algorithm for a given problem, for each $n > 0$, there exists an input that make the algorithm take $\Omega(f(n))$ time.
Then $f(n)$ is a lower bound on the worst case running time.

Comparison-based Algorithms

- The algorithm only uses the results of comparisons, not values of elements (*)
- Very general – does not assume much about what type of data is being sorted
- However, other kinds of algorithms are possible!
- In this model, it is reasonable to count $\#$ comparisons. Note that the $\#$ comparisons is a lower bound on the running time of an algorithm.

(*) If values are used, lower bounds proved in this model are not lower bounds on the running time.

Lower Bound: Observations

- Lower bounds are rarely simple: there are virtually no known general techniques.
- So we must try ad hoc methods for each problem.
- We proved a lower bound on finding the maximum
- Sorting lower bounds:
 - Trivial: $\Omega(n)$ – every element must be in a comparison
 - Best possible result – $\Omega(n \log n)$ comparisons, since we already know several $O(n \log n)$ sorting algorithms
 - Difficulty: how do we reason about all possible comparison-based sorting algorithms?

The Decision Tree Model

Assumptions:

- All numbers are distinct
- All comparisons have form $a_i \leq a_j$ (since $a_i < a_j, a_i \leq a_j, a_i \geq a_j, a_i > a_j$ are equivalent)

Decision tree structure

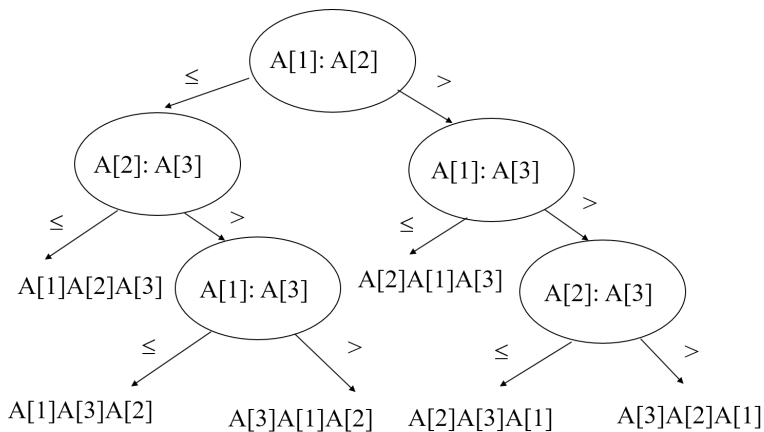
- Full binary tree
- Ignore control, movement, and all other operations, just use comparisons.
- suppose three elements $\langle a_1, a_2, a_3 \rangle$ with instance $\langle 6, 8, 5 \rangle$.

The Decision Tree Model - Example

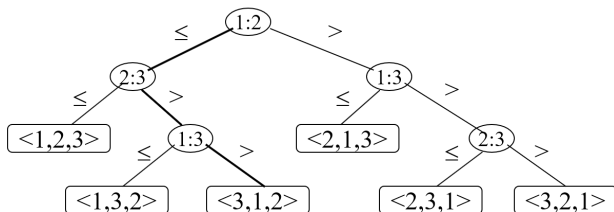
INSERTION-SORT(A)

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1..j-1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i+1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i+1] = key$ 
```

Insertion Sort: Decision Tree



Insertion Sort: Another view



Internal node $i : j$ indicates comparison between a_i and a_j .

Leaf node $\langle p_1, p_2, p_3 \rangle$ indicates ordering $a_{p_1} \leq a_{p_2} \leq a_{p_3}$

Path of bold lines indicates sorting path for $\langle 6, 8, 5 \rangle$. There are total $3! = 6$ possible permutations (paths).

The Decision Tree Model - Summary

- Only consider comparisons
- Each internal node = 1 comparison
- Start at root, make the first comparison
 - if the outcome is \leq , take the LEFT branch
 - if the outcome is $>$, take the RIGHT branch
- Repeat at each internal node
- Each LEAF represents ONE correct ordering

Lower Bound on Sorting

- Claim: The decision tree must have at least $n!$ leaves.
WHY?
- worst case number of comparisons = the height of the decision tree
- Claim: Any comparison sort in the worst case needs $\Omega(n \log n)$ comparisons
- Suppose height of a decision tree is h , number of paths (i.e., permutations) is $n!$
- Since a binary tree of height h has at most 2^h leaves,
 $n \leq 2^h$
- So $h \geq \lg n! \in \Omega(n \lg n)$

Lower Bounds: Check your understanding

- Can you prove that any algorithm that searches for an element in a sorted array of size n must have running time $\Omega(\lg n)$?