Algorithms and Data Structures: Lower Bounds for Sorting
Comparison Based Sorting Algorithms

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Example 2
*Insertion-Sort, Quicksort, Merge-Sort, Heapsort* are all comparison based.
The Decision Tree Model

Abstractly, we may describe the behaviour of a comparison-based sorting algorithm $S$ on an input array $A = \langle A[1], \ldots, A[n] \rangle$ by a decision tree:

At each leaf of the tree the output of the algorithm on the corresponding execution branch will be displayed. Outputs of sorting algorithms correspond to permutations of the input array.

ADS: lect 7 – slide 3 –
A Simplifying Assumption

In the following, we assume that all keys of elements of the input array of a sorting algorithm are distinct. (It is ok to restrict to a special case, because we want to prove a lower bound.) Thus the outcome \( A[i] = A[j] \) in a comparison will never occur, and the decision tree is in fact a binary tree:

![Decision Tree Diagram]

\[
\]

\[
\]

\[
t : u
\]
In insertion sort, when we get the result of a comparison, we often swap some elements of the array. In showing decision trees, we don’t *implement* a swap. Our indices always refer to the *original* elements at that position in the array. To understand what I mean, draw the evolving array of `INSERTIONSORT` beside this decision tree.
A Lower Bound for Comparison Based Sorting

For a comparison based sorting algorithm $S$:

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Theorem 3

For all comparison based sorting algorithms $S$ we have

$$C_S(n) = \Omega(n \lg n).$$

Corollary 4

The worst-case running time of any comparison based sorting algorithm is $\Omega(n \lg n)$. 
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Proof of Theorem 3 uses Decision-Tree Model of sorting.

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- For sorting, the input can encode \( n! \) outputs.
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It is an Information-Theoretic Lower Bound:
- “Information-Theoretic” means that it is based on the amount of “information” that an instance of the problem can encode.
- For sorting, the input can encode $n!$ outputs.
- Proof does not make any assumption about how the sorting might be done (except it is comparison-based).
Observation 5

For every $n$, $C_S(n)$ is the height of the decision tree of $S$ on inputs $n$ (the longest path from the “root” to a leaf is the maximum number of comparisons that algorithm $S$ will do on an input of length $n$).

We shall prove a lower bound for the height of the decision tree for any algorithm $S$.

Remark

Maybe you are wondering ... was it really ok to assume all keys are distinct?
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It is ok - because the problem of sorting $n$ keys (with no distinctness assumption) is more general than the problem of sorting $n$ distinct keys.

The worst-case for sorting certainly is as bad as the worst-case for all-distinct keys sorting.
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- Thus

$$C_S(n) \geq \lg(n!) = \Omega(n \lg(n)).$$
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- Putting everything together, we get
  
  $n! \leq \text{number of leaves of decision tree} \leq 2^{\text{height of decision tree}} \leq 2^{C_S(n)}$.

- Thus
  
  $C_S(n) \geq \lg(n!) = \Omega(n \lg(n))$.

To obtain the last inequality, we can use the following inequality:

$n^{n/2} \leq n! \leq n^n$

This tells us that $\lg n! \geq \lg(n^{n/2}) = (n/2) \lg n = \Omega(n \lg(n))$.

Thm 3 QED
An Average Case Lower Bound

For any comparison based sorting algorithm $S$:

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Proof uses the fact that the average length of a path from the root to a leaf in a binary tree with $\ell$ leaves is $\Omega(\lg \ell)$ (Theorem 11 and 12).

Corollary 8
The average-case running time of any comparison based sorting algorithm is $\Omega(n \lg n)$.  

ADS: lect 7 – slide 10 –
Average root-leaf length in Binary tree

Definition 9
For any binary tree $T$, let $\text{AvgRL}(T)$ denote the average root-to-leaf length for $T$.

Definition 10
A near-complete binary tree $T$ is a binary tree in which every internal node has exactly two child nodes, and all leaves are either at depth $h$ or depth $h - 1$.

Theorem 11
Any “near-complete” binary tree $T$ with leaf set $L(T)$, $|L(T)| \geq 4$, has Average root-to-leaf length $\text{AvgRL}(T)$ at least $\lg(|L(T)|/2)$.

Theorem 12
For any binary tree $T$, there is a near-complete binary tree $T'$ such that $L(T) = L(T')$ (same leaf set) and such that $\text{AvgRL}(T') \leq \text{AvgRL}(T)$. Hence $\text{AvgRL}(T) \geq \lg(|L(T)|/2)$ holds for all binary trees.

Proof of Theorems 11 and 12 via BOARD notes.
Implications of These Lower Bounds

Theorem 3 and Theorem 7 are significant because they hold for all comparison-based algorithms S. They imply the following:

1. By Thm 3, any comparison-based algorithm for sorting which has a worst-case running-time of \(O(n \lg n)\) is asymptotically optimal (ie, apart from the constant factor inside the “O” term, it is as good as possible in terms of worst-case analysis). This includes algorithms like MergeSort, HeapSort.

2. By Thm 7, any comparison-based algorithm for sorting which has an average-case running-time of \(O(n \lg n)\) is the best you can hope for in terms of average-case analysis (apart from the constant factor inside the “O” term). This is accomplished by MergeSort and HeapSort. In Lecture 8, we will see that it is also true for QuickSort (for the average-case, not the worst-case).
Lecture 9 (after average-case analysis of QuickSort)

We will show how in a special case of sorting (when the inputs are numbers, coming from the range \( \{1, 2, \ldots, n^k\} \) for some constant \( k \), we can sort in linear time (NOT a comparison-based algorithm).

Reading Assignment

[CLRS] Section 8.1 (2nd and 3rd edition) or
[CLR] Section 9.1
Well-worth reading - this is a nice chapter of CLRS (not too long).
Problems

1. Draw (simplified) decision trees for **Insertion Sort** and **Quicksort** for \( n = 4 \).

2. Exercise 8.1-1 of [CLRS] (both 2nd and 3rd ed).

3. Resolve the complexity (in terms of no-of-comparisons) of sorting 4 numbers.
   3.1 Give an algorithm which sorts any 4 numbers and which uses at most 5 comparisons in the worst-case.
   3.2 Prove (using the decision-tree model) that there is no algorithm to sort 4 numbers, which uses less than 5 comparisons in the worst-case.