

CSE 326: Data Structures

Quicksort

Comparison Sorting Bound

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Announcements (5/14/08)

- Homework due at beginning of class on Friday.
- Section tomorrow:
 - Graded homeworks returned
 - More discussion of project, Java and hash tables
 - ...
- Reading for this lecture: Chapter 7.

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Quicksort

Quicksort uses a divide and conquer strategy, but does not require the $O(N)$ extra space that MergeSort does.

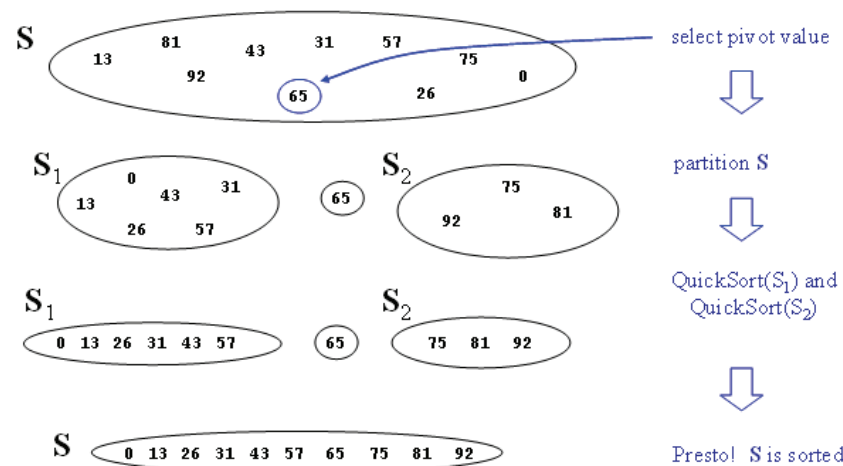
Here's the idea for sorting array **S**:

1. Pick an element v in **S**. This is the **pivot** value.
2. Partition **S** - $\{v\}$ into two disjoint subsets, **S**₁ and **S**₂ such that:
 - elements in **S**₁ are all $\leq v$
 - elements in **S**₂ are all $\geq v$
3. Return concatenation of QuickSort(**S**₁), v , QuickSort(**S**₂)

Recursion ends when QuickSort() receives an array of length 0 or 1.

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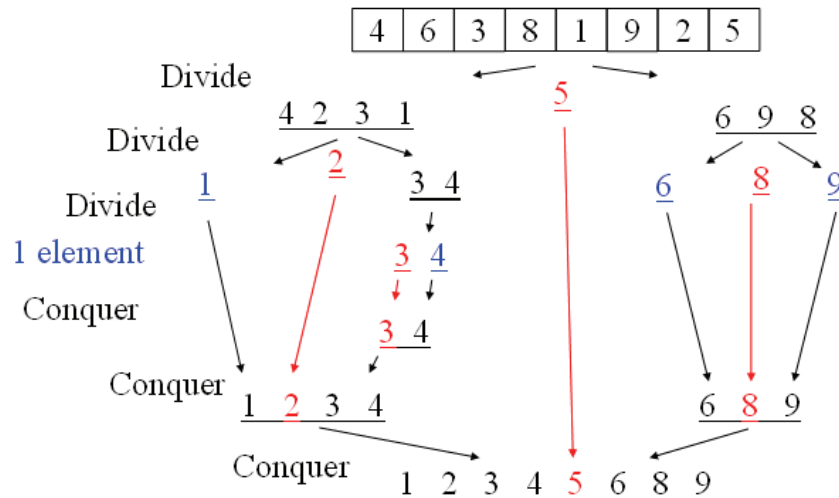
The steps of Quicksort



[Weiss]

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Quicksort Example



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Pivot Picking and Partitioning

The tricky pieces are:

- **Picking the pivot**
 - Goal: pick a pivot value that will cause $|S_1|$ and $|S_2|$ to be roughly equal in size.
- **Partitioning**
 - Preferably in-place
 - Dealing with duplicates.

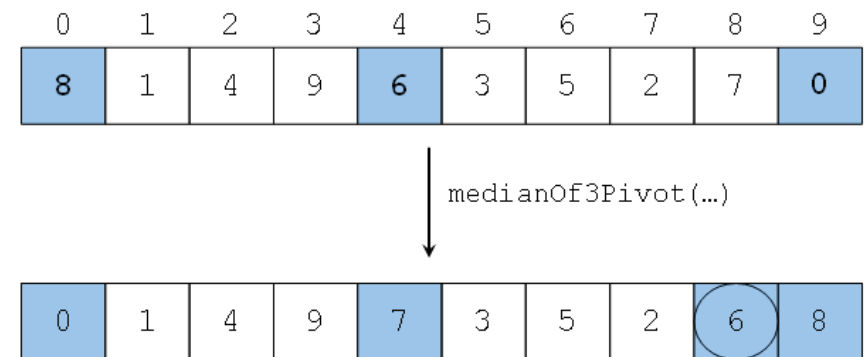
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Picking the Pivot

- Choose the median
 - expensive
- Choose $A[\text{left}]$
 - fast, enables worst case (on sorted input)
- Choose $A[\text{random}]$
 - ... $\text{Median}(A[\text{random}_1], \dots, A[\text{random}_m])$
 - rand calls often not cheap
- Choose $A[\text{middle}]$
- $\text{Median}(A[\text{left}], A[\text{mid}], A[\text{right}])$
 - does a little "free" sorting for you

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Median of Three Pivot



Choose the pivot as the median of three.

Place the pivot and the largest at the right and the smallest at the left.

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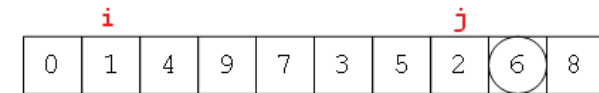
Quicksort Partitioning

- Need to partition the array into left and right sub-arrays such that:
 - elements in left sub-array are $\leq$ pivot
 - elements in right sub-array are $\geq$ pivot
- Can be done in-place with another “two pointer method”
 - Sounds like mergesort, but here we are *partitioning*, not sorting...
 - ... and we can do it in-place.

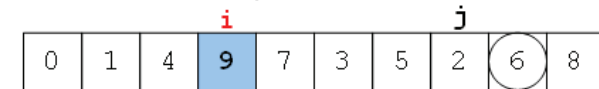
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Partitioning In-place

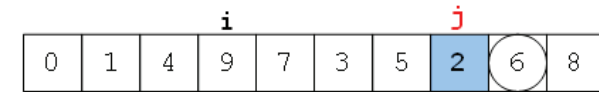
Setup: i = start and j = end of un-partioned elements:



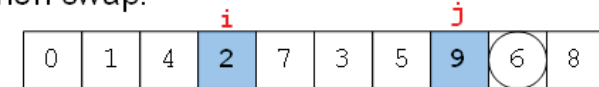
Advance i until element $\geq$ pivot:



Advance j until element $\leq$ pivot:

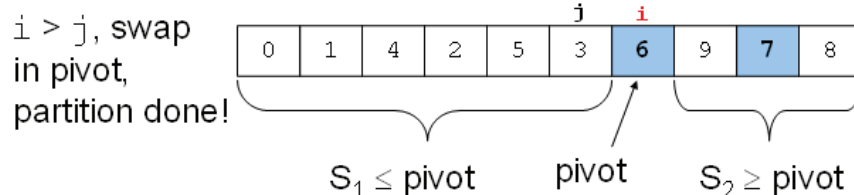
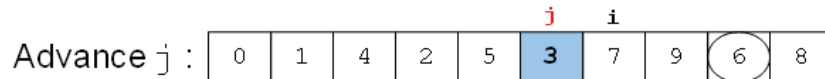
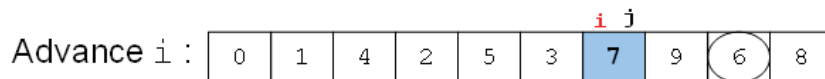
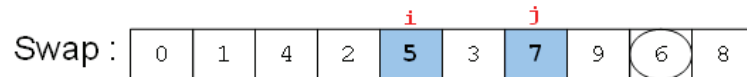
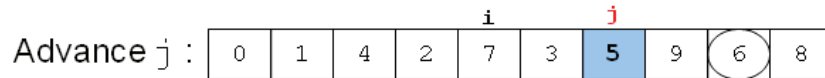
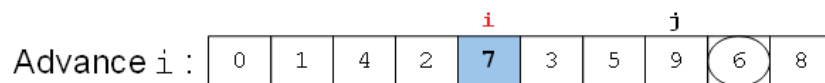


If $j > i$, then swap:



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Partitioning In-place



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Partition Pseudocode

```
Partition(A[], left, right) {
    v = A[right]; // Assumes pivot value currently at right
    i = left;     // Initialize left side, right side pointers
    j = right-1;

    // Do i++, j-- until they cross, swapping values as needed
    while (1) {
        while (A[i] < v) i++;
        while (A[j] > v) j--;
        if (i < j) {
            Swap(A[i], A[j]); i++; j--;
        }
        else
            break;
    }

    Swap(A[i], A[right]); // Swap pivot value into position
    return i;             // Return the final pivot position
}
```

Complexity for input size n ? $O(n)$

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Quicksort Pseudocode

Putting the pieces together:

```
Quicksort(A[], left, right) {  
    if (left  $\geq$  right) return;  
  
    medianOf3Pivot(A, left, right);  
    pivotIndex = Partition(A, left+1, right-1);  
  
    Quicksort(A, left, pivotIndex - 1);  
    Quicksort(A, pivotIndex + 1, right);  
}
```

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QuickSort: Best case complexity

```
Quicksort(A[], left, right) {  
    if (left  $\geq$  right) return;  
  
    medianOf3Pivot(A, left, right);  
    pivotIndex = Partition(A, left+1, right-1);  
  
    Quicksort(A, left, pivotIndex - 1);  
    Quicksort(A, pivotIndex + 1, right);  
}
```

$$T(1) = a$$

$$T(n) = 2T(n/2) + bn + c$$

$$O(n \log n)$$

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QuickSort: Worst case complexity

```
Quicksort(A[], left, right) {  
    if (left  $\geq$  right) return;  
  
    medianOf3Pivot(A, left, right);  
    pivotIndex = Partition(A, left+1, right-1);  
  
    Quicksort(A, left, pivotIndex - 1);  
    Quicksort(A, pivotIndex + 1, right);  
}
```

$$T(1) = a$$

$$T(n) = T(n-1) + bn + c$$

$$= [T(n-2) + b(n-1) + c] + bn + c \Rightarrow b \sum_{i=1}^n i$$

$$= [T(n-3) + b(n-2) + c] + b(n-1) + c + bn + c$$

$$O(n^2)$$

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QuickSort: Average case complexity

Turns out to be $O(n \log n)$.

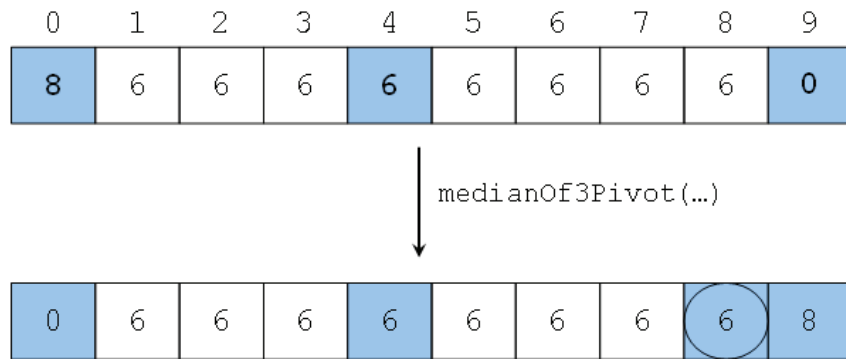
See Section 7.7.5 for an idea of the proof.

Don't need to know proof details for this course.

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Many Duplicates?

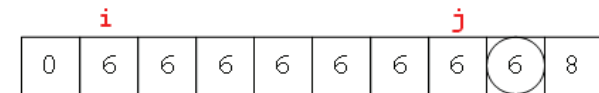
An important case to consider is when an array has many duplicates.



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Partitioning with Duplicates

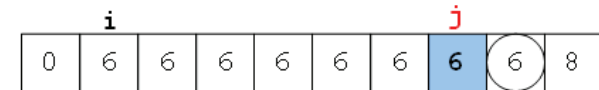
Setup: i = start and j = end of un-partioned elements:



Advance i until element $\geq$ pivot:



Advance j until element $\leq$ pivot:

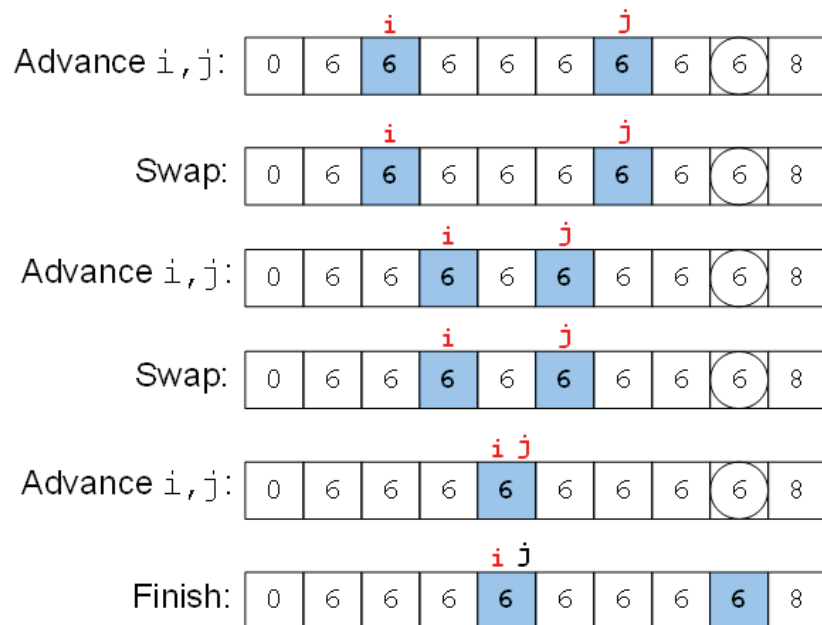


If $j > i$, then swap:



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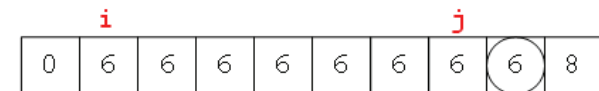
Partitioning with Duplicates



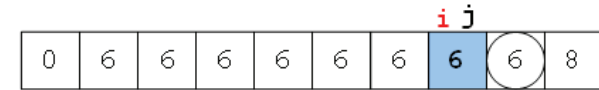
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Partitioning with Duplicates: Take Two

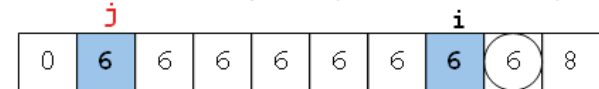
Start i = start and j = end of un-partioned elements:



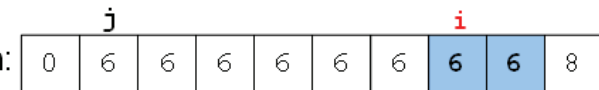
Advance i until element $>$ pivot (and in bounds):



Advance j until element $<$ pivot (and in bounds):



Finish:



Is this better?

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Partitioning with Duplicates: Upshot

It's better to stop advancing pointers when elements are equal to pivot, and then just do swaps.

Complexity of quicksort on an array of identical values? $O(n \log n)$

Can we do better? *4-fingered partition*



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Important Tweak

Insertion sort is actually better than quicksort on small arrays. Thus, a better version of quicksort:

```
Quicksort(A[], left, right) {  
    if (left + CUTOFF ≤ right) {  
        medianOf3Pivot(A, left, right);  
        pivotIndex = Partition(A, left+1, right-1);  
  
        Quicksort(A, left, pivotIndex - 1);  
        Quicksort(A, pivotIndex + 1, right);  
    } else {  
        InsertionSort(A, left, right);  
    }  
}
```

CUTOFF = 10 is reasonable.

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Properties of Quicksort

- $O(N^2)$ worst case performance, but $O(N \log N)$ average case performance.
- Pure quicksort not good for small arrays.
- No iterative version (without using a stack).
- “In-place,” but uses auxiliary storage because of recursive calls.
- Stable? *N*
- Used by Java for sorting arrays of primitive types.

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How fast can we sort?

Heapsort, Mergesort, and Binary Tree Sort all have $O(N \log N)$ worst case running time.

These algorithms, along with Quicksort, also have $O(N \log N)$ average case running time.

Can we do any better?

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Sorting Model

- Recall our basic assumption: we can only compare two elements at a time
 - we can only reduce the possible solution space by half each time we make a comparison (see alternate proof)
- Suppose you are given N elements
 - Assume no duplicates
- How many possible orderings can you get?
 - Example: a, b, c ($N = 3$)

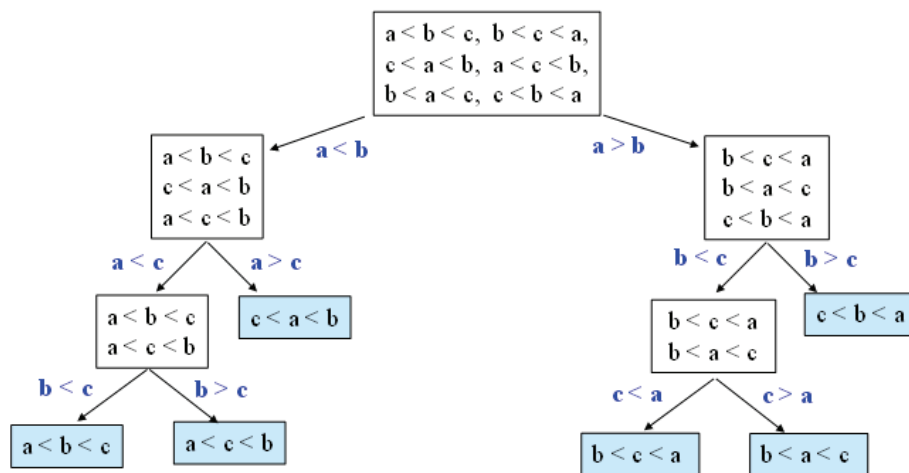
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Permutations

- How many possible orderings can you get?
 - Example: a, b, c ($N = 3$)
 - $(a b c), (a c b), (b a c), (b c a), (c a b), (c b a)$
 - 6 orderings = $3 \cdot 2 \cdot 1 = 3!$ (i.e., "3 factorial")
 - All the possible permutations of a set of 3 elements
- For N elements
 - N choices for the first position, $(N-1)$ choices for the second position, ..., (2) choices, 1 choice
 - $N(N-1)(N-2) \dots (2)(1) = \underline{N! \text{ possible orderings}}$

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Decision Tree



The leaves contain all the possible orderings of a, b, c .

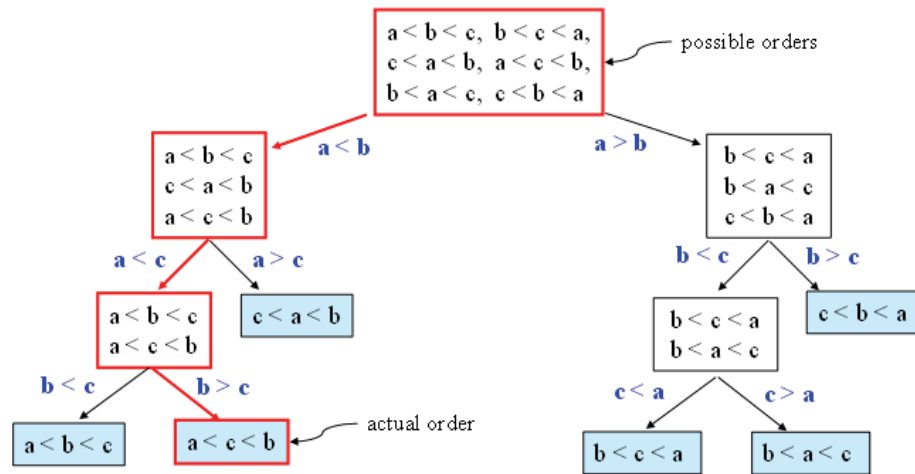
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Decision Trees

- A Decision Tree is a Binary Tree such that:
 - Each node = a set of orderings
 - i.e., the remaining solution space
 - Each edge = 1 comparison
 - Each leaf = 1 unique ordering
 - How many leaves for N distinct elements?
- $L = N!$
- Only 1 leaf has the ordering that is the desired correctly sorted arrangement

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Decision Tree Example



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Decision Trees and Sorting

- Every sorting algorithm corresponds to a decision tree
 - Finds correct leaf by choosing edges to follow
 - ie, by making comparisons
 - Each decision reduces the possible solution space by one half (see alternate proof)
- We will focus on worst case run time.

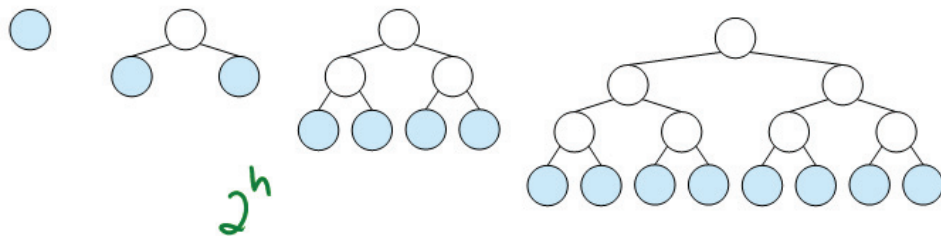
Observations:

 - Worst case run time is $\geq$ maximum number of comparisons.
 - Maximum number of comparisons is the length of the longest path in the decision tree, i.e. the height of the tree.

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How many leaves on a tree?

Suppose you have a binary tree of height h . How many leaves in a perfect tree?



We can prune a perfect tree to make any binary tree of same height. Can # of leaves increase?

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Lower bound on Height

- A binary tree of height h has at most 2^h leaves
 - Can prove formally by induction
- A decision tree has $N!$ leaves. What is its minimum height of that tree?

$$\begin{aligned}
 L &\leq 2^h \\
 L &= N! \\
 N! &\leq 2^h \\
 \log_2 N! &\leq h
 \end{aligned}$$

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Lower Bound on $\log(N!)$

$$\begin{aligned}
 \log(N!) &= \log(N \cdot (N-1) \cdot (N-2) \cdots 1) \\
 &= \log N + \log(N-1) + \log(N-2) + \cdots + 0 \\
 &= \sum_{i=1}^N \log i \\
 &= \sum_{i=\frac{N}{2}+1}^N \log i + \sum_{i=1}^{\frac{N}{2}} \log i \\
 &\geq \sum_{i=\frac{N}{2}+1}^N \log i \geq \sum_{i=\frac{N}{2}+1}^N \log \frac{N}{2} = \frac{N}{2} \log \frac{N}{2} \\
 &= \frac{N}{2} (\log N - \log 2) \\
 &= \frac{N}{2} \log N - \frac{\log 2}{2} N
 \end{aligned}$$

$\log(N!) \in \Omega(N \log N)$

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$\Omega(N \log N)$

Worst case run time of any comparison-based sorting algorithm is $\Omega(N \log N)$.

Can also show that average case run time is also $\Omega(N \log N)$.

Can we do better if we don't use comparisons? (Huh?)

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