

## Sorting

CS211  
Fall 2000

## Insertion Sort

- Corresponds to how most people sort cards
- Invariant: everything to left is already sorted
- Works especially well when input is *nearly sorted*
- Runtime
  - Worst-case
    - ▲  $O(n^2)$
    - ▲ Consider reverse-sorted input
  - Best-case
    - ▲  $O(n)$
    - ▲ Consider sorted input

```
// Code for sorting a[] an array of int
for (int i = 1; i < a.length; i++) {
    int temp = a[i];
    int k = i;
    for (; k > 0 && a[k-1] > temp; k--)
        a[k] = a[k-1];
    a[k] = temp;
}
```

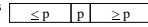
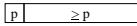
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## Merge Sort

- Uses recursion (Divide & Conquer)
- Outline (text has detailed code)
  - Split array into two halves
  - Recursively sort each half
  - Merge the two halves
- Merge = combine two sorted arrays to make a single sorted array
  - Rule: Always choose the smallest item
  - Time:  $O(n)$
- Runtime recurrence
  - Let  $T(n)$  be the time to sort an array of size  $n$
  - $T(n) = 2T(n/2) + O(n)$
  - $T(1) = O(1)$
  - Can show by induction that  $T(n) = O(n \log n)$
- Alternately, can show  $T(n) = O(n \log n)$  by looking at tree of recursive calls

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## Quick Sort

- Also uses recursion (Divide & Conquer)
- Outline
  - *Partition* the array
  - Recursively sort each piece of the partition
- *Partition* = divide the array like this 
- $p$  is the *pivot* item
- Best pivot choices
  - middle item
  - random item
  - median of leftmost, rightmost, and middle items
- Runtime analysis (worst-case)
  - Partition can work badly producing this: 
  - Runtime recurrence  $T(n) = T(n-1) + O(n)$
  - This can be solved by induction to show  $T(n) = O(n^2)$
- Runtime analysis (expected-case)
  - More complex recurrence
  - Can solve by induction to show expected  $T(n) = O(n \log n)$
- Can improve constant factor by avoiding QSort on small sets

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## Heap Sort

- Not recursive
- Outline
  - Build heap
  - Perform `removeMax` on heap until empty
  - Note that items are removed from heap in sorted order
- Heap Sort is the only  $O(n \log n)$  sort that uses *no* extra space
  - Merge Sort uses extra array during merge
  - Quick Sort uses recursive stack
- Runtime analysis (worst-case)
  - $O(n)$  time to build heap (using bottom-up approach)
  - $O(\log n)$  time (worst-case) for each removal
  - Total time:  $O(n \log n)$

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## Sorting Algorithm Summary

- The ones we have discussed
  - Insertion Sort
  - Merge Sort
  - Quick Sort
  - Heap Sort
- Why so many? Do Computer Scientists have some kind of sorting fetish or what?
  - Stable sorts: *Ins, Mer*
  - Worst-case  $O(n \log n)$ : *Mer, Hea*
  - Expected-case  $O(n \log n)$ : *Mer, Hea, Qui*
  - Best for nearly-sorted sets: *Ins*
  - No extra space needed: *Ins, Hea*
  - Fastest in practice: *Qui*
  - Least data movement: *Sel*
- Other sorting algorithms
  - Selection Sort
  - Shell Sort (in text)
  - Bubble Sort
  - Radix Sort
  - Bin Sort
  - Counting Sort

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## Lower Bounds on Sorting: Goals

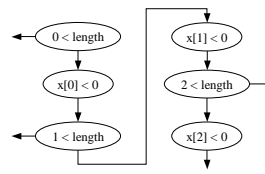
- Goal: Determine the minimum time *required* to sort  $n$  items
- Note: we want *worst-case* not *best-case* time
  - Best-case doesn't tell us much; for example, we know Insertion Sort takes  $O(n)$  time on already-sorted input
  - We want to determine the *worst-case* time for the *best-possible* algorithm
- But how can we prove anything about the *best possible* algorithm?
  - We want to find characteristics that are common to *all* sorting algorithms
  - Let's try looking at *comparisons*

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## Comparison Trees

- Any algorithm can be "unrolled" to show the comparisons that are (potentially) performed
- In general, you get a *comparison tree*
- If the algorithm fails to terminate for some input then the comparison tree is infinite
- The height of the comparison tree represents the *worst-case number of comparisons* for that algorithm

Example  
 for (int i = 0; i < x.length; i++)  
 if (x[i] < 0) x[i] = -x[i];



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## Lower Bounds on Sorting: Notation

- Suppose we want to sort the items in the array  $B[ ]$
- Let's name the items
  - $a_1$  is the item initially residing in  $B[1]$ ,  $a_2$  is the item initially residing in  $B[2]$ , etc.
  - In general,  $a_i$  is the item initially stored in  $B[i]$
- Rule: an item keeps its name forever, but it can change its location
  - Example: after  $\text{swap}(B, 1, 5)$ ,  $a_1$  is stored in  $B[5]$  and  $a_5$  is stored in  $B[1]$

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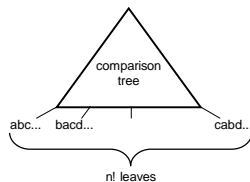
## The Answer to a Sorting Problem

- An *answer* for a sorting problem tells where each of the  $a_i$  resides when the algorithm finishes
- How many answers are possible?
- The *correct* answer depends on the actual values represented by each  $a_i$
- Since we don't know what the  $a_i$  are going to be, it has to be *possible* to produce each permutation of the  $a_i$
- For a sorting algorithm to be valid it must be possible for that algorithm to give any of  $n!$  potential answers

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## Comparison Tree for Sorting

- Every sorting algorithm has a corresponding *comparison tree*
  - Note that other stuff happens during the sorting algorithm, we just aren't showing it in the tree
- The comparison tree must have  $n!$  (or more) leaves because a valid sorting algorithm must be able to get any of  $n!$  possible answers
- Comparison tree for sorting  $n$  items:



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## Time vs. Height

- The worst-case time for a sorting method must be  $\geq$  the height of its comparison tree
  - The height corresponds to the worst-case number of comparisons
  - Each comparison takes  $\Theta(1)$  time
  - The algorithm is doing more than just comparisons
- What is the minimum possible height for a binary tree with  $n!$  leaves?
  - Height  $\geq \log(n!) = \Theta(n \log n)$
- This implies that any comparison-based sorting algorithm must have a worst-case time of  $\Omega(n \log n)$ 
  - Note: this is a lower bound; thus, the use of big-Omega instead of big-O

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## Using the Lower Bound on Sorting

Claim: I have a PQ

- Insert time:  $O(1)$
- GetMax time:  $O(1)$
- True or false?

False (for general sets)  
because if such a PQ  
existed, it could be used to  
sort in time  $O(n)$

Claim: I have a PQ

- Insert time:  $O(\log \log n)$
- GetMax time:  $O(\log \log n)$
- True or false?

False (for general sets)  
because it could be used to  
sort in time  $O(n \log \log n)$

True for items with priorities in  
range  $1..n$  [van Emde Boas]  
(Note: such a set can be  
sorted in  $O(n)$  time)

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## Sorting in Linear Time

There are several sorting  
methods that take linear  
time

- Counting Sort
  - sorts integers from a  
small range:  $[0..k]$   
where  $k = O(n)$
- Radix Sort
  - the method used by the  
old card-sorters
  - sorting time  $O(dn)$   
where  $d$  is the number  
of "digits"

■ How do these methods get  
around the  $\Omega(n \log n)$  lower  
bound?

- They don't use  
comparisons
- What sorting method works  
best?
  - QuickSort is best  
general-purpose sort
  - Counting Sort or Radix  
Sort can be best for  
*some* kinds of data

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