

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;
}
```

2

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

3

Quick Sort

- Also uses recursion (Divide & Conquer)
- Outline
 - *Partition* the array
 - Recursively sort each piece of the partition
- *Partition* = divide the array like this $\leq p \quad p \quad \geq p$
- 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: $p \quad \geq p$
 - 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

4

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: $O(n \log n)$

5

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

6

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*

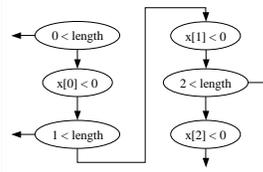
7

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];
```



8

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 $B[1]$

9

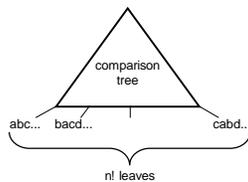
The Answer to a Sorting Problem

- 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

10

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:



11

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?
 - $\text{Height} \geq \log(n!) = \Theta(n \log n)$
- 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

12

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)

13

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

14