

## SelectionSort

- To sort an array of size n : -
- Examine a[0] to a[n-1]; find the smallest one and swap it with a[0]
- Examine a[1] to a[n-1]; find the smallest one and swap it with a[1]
- In general, in step $i$, examine a[i] to a[n-1]; find the smallest one and swap it with a[i]


## Divide \& Conquer?

- It often pays to
- Break the problem into smaller subproblems,
- Solve the subproblems separately, and then
- Assemble a final solution
- This technique is called divide-and-conquer
- Caveat: It won't help unless the partitioning and assembly processes are inexpensive
- Can we apply this approach to sorting?


## MergeSort

- Quintessential divide-and-conquer algorithm
- Divide array into equal parts, sort each part, then merge
- Questions:
- Q1: How do we divide array into two equal parts?
- A1: Find middle index: a.length/2
- Q2: How do we sort the parts?
- A2: call MergeSort recursively!
- Q3: How do we merge the sorted subarrays?
- A3: We have to write some (easy) code


## Merging Sorted Arrays A and B

- Create an array $C$ of size $=$ size of $A+\operatorname{size}$ of $B$
- Keep three indices:
- i into A
$-j$ into $B$
- k into C
- Initialize all three indices to 0 (start of each array)
- Compare element $A[i]$ with $B[j]$, and move the smaller element into $\mathrm{C}[\mathrm{k}]$
- Increment i or j , whichever one we took, and k
- When either A or B becomes empty, copy remaining elements from the other array (B or A, respectively) into C



## MergeSort Notes

- Asymptotic complexity: O( $\mathrm{n} \log \mathrm{n}$ )
- Much faster than $\mathrm{O}\left(\mathrm{n}^{2}\right)$
- Disadvantage
- Need extra storage for temporary arrays
- In practice, this can be a disadvantage, even though MergeSort is asymptotically optimal for sorting
- Can do MergeSort in place, but this is very tricky (and it slows down the algorithm significantly)
- MergeSort is great for huge datasets distributed over multiple computers (e.g. map-reduce)
- Are there good sorting algorithms that do not use so much extra storage?
- Yes: QuickSort

MergeSort Analysis

- Outline (detailed code on the website)
- 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: $O(n)$ where $n$ is the combined size of the two arrays
- Runtime recurrence
- Let $T(n)$ be the time to sort an array of size $n$
$-T(n)=2 T(n / 2)+O(n)$
$-T(1)=1$
- Can show by induction that $T(n)$ is $O(n \log n)$
- Alternately, can see that $T(n)$ is $O(n \log n)$ by looking at tree of recursive calls


## In-Place Partitioning



How can we move all the blues to the left of all the reds?

1. Keep two indices, LEFT and RIGHT
2. Initialize LEFT at start of array and RIGHT at end of array
3. Invariant: all elements to left of LEFT are blue all elements to right of RIGHT are red
4. Keep advancing indices until they pass, maintaining invariant


- Once indices cross, partitioning is done
- If you replace blue with $\leq p$ and red with $\geq p$, this is exactly what we need for QuickSort partitioning
- Notice that after partitioning, array is partially sorted
- Recursive calls on partitioned subarrays will sort subarrays
- No need to copy/move arrays, since we partitioned in place


## QuickSort Analysis

- Runtime analysis (worst-case)
- Partition can work badly, producing this: P
- Runtime recurrence
- $\mathrm{T}(\mathrm{n})=\mathrm{T}(\mathrm{n}-1)+\mathrm{n}$
- This can be solved to show worst-case $T(n)$ is $O\left(n^{2}\right)$
- Runtime analysis (expected-case)
- More complex recurrence
- Can solve to show expected $T(n)$ is $O(n \log n)$
- Improve constant factor by avoiding QuickSort on small sets
- Switch to InsertionSort (for example) for sets of size, say, $\leq 9$
- Definition of small depends on language, machine, etc.


## Sorting Algorithm Summary

-The ones we have
discussed

- InsertionSort
-SelectionSort
- MergeSort
-QuickSort
- Other sorting algorithms
- HeapSort (will revisit this)
- ShellSort (in text)
- BubbleSort (nice name)
-RadixSort
-BinSort
- CountingSort
-Why so many? Do computer scientists have some kind of sorting fetish or what?
- Stable sorts: Ins, Sel, Mer
- Worst-case O( $n \log n$ ): Mer, Hea
- Expected O(n logn): Mer, Hea, Qui
- Best for nearly-sorted sets: Ins
- No extra space needed: Ins, Sel, Hea
- Fastest in practice: Qui
- Least data movement: Sel
- Goal: Determine the minimum time required to sort n items
- Note: we want worstcase, not best-case time
- Best-case doesn't tell us much; for example, we know Insertion Sort takes O(n) time on alreadysorted input
- Want to know the worstcase time for the best possible algorithm


## Lower Bound for Sorting

- 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 limit attention to comparison-based algorithms and try to count number of comparisons


## Comparison Trees

- Comparison-based algorithms make decisions based on comparison of data elements
- This gives 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 worstcase number of comparisons for that algorithm
- Will show that any correct comparison-based algorithm must make at least $n \log n$ comparisons in the worst case


## Lower Bound for Comparison Sorting

- Say we have a correct comparison-based algorithm
- Suppose we want to sort the elements in an array $B[]$
- Assume the elements of $B[]$ are distinct
- Any permutation of the elements is initially possible
- When done, $B[]$ is sorted
- But the algorithm could not have taken the same path in the comparison tree on different input permutations


## Lower Bound for Comparison

Sorting

- How many input permutations are possible?

$$
n!\sim 2^{n \log n}
$$

-For a comparison-based sorting algorithm to be correct, it must have at least that many leaves in its comparison tree

- to have at least $n!\sim 2^{n \log n}$ leaves, it must have height at least $\mathrm{n} \log \mathrm{n}$ (since it is only binary branching, the number of nodes at most doubles at every depth)
-therefore its longest path must be of length at least $n \log n$, and that it its worst-case running time
java.lang.Comparable<T> Interface
- public int compareTo(T x);
- Returns a negative, zero, or positive value
- negative if this is before $\mathbf{x}$
- 0 if this. equals ( $\mathbf{x}$ )
- positive if this is after $\mathbf{x}$
- Many classes implement Comparable
- String, Double, Integer, Character, Date,..
- If a class implements Comparable, then its compareTo method is considered to define that class's natural ordering
- Comparison-based sorting methods should work with

Comparable for maximum generality

