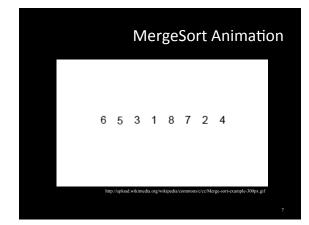


Divide & Conquer?

- 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 + 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 C[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)

Merging Sorted Arrays 4 7 7 3 9 1 3 4 4 6 7 1 3 4 6 8

MergeSort Analysis

- Outline (detailed code on the website)
- Split array into two halves
- Recursively sort each halfMerge 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) = 2T(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

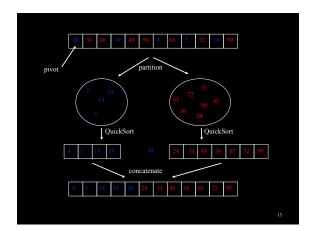
MergeSort Notes

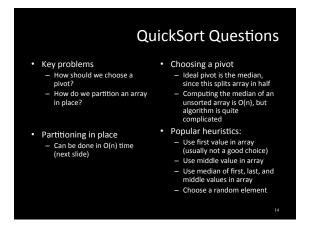
- Asymptotic complexity: O(n log n)
 - Much faster than O(n2)
- Disadvantage
 - Need extra storage for temporary arrays
 - In practice, this can be a disadvantage, even though MergeSort is asymptotically optimal for sorting
 Can do MergeSort in place, but this is very tricky (and it slows down the algorithm significantly)

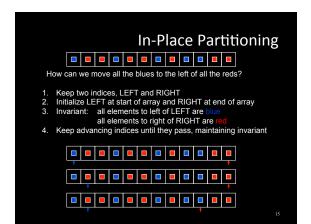
 - MergeSort is great for huge datasets distributed over multiple computers (e.g. map-reduce)
- Are there good sorting algorithms that do not use so much extra storage?
 - Yes: QuickSort

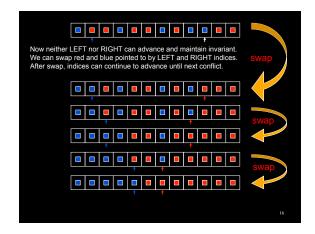
QuickSort

- · Intuitive idea
 - Given an array A to sort, choose a pivot value p
 - Partition A into two subarrays, AX and AY
 - AX contains only elements ≤ p
 - AY contains only elements ≥ p
 - Sort subarrays AX and AY separately
 - Concatenate (not merge!) sorted AX and AY to get
 - Concatenation is easier than merging O(1)

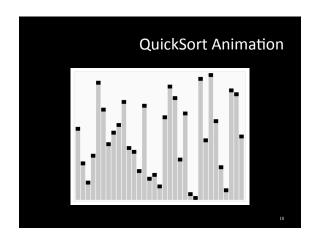








Once indices cross, partitioning is done
If you replace blue with ≤ p and red with ≥ p, this is exactly what we need for QuickSort partitioning
Notice that after partitioning, array is partially sorted
Recursive calls on partitioned subarrays will sort subarrays
No need to copy/move arrays, since we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



QuickSort Analysis

- · Runtime analysis (worst-case)
 - Partition can work badly, producing this
 - Runtime recurrence
 - T(n) = T(n-1) + n
 - This can be solved to show worst-case T(n) is O(n²)
- 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
 - Switch to InsertionSort (for example) for sets of size, say, ≤
 - 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?

Lower Bound for Sorting

- 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 already-sorted input
 - Want to know the worstcase 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 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 cor
- 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



Sorting

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

be correct, it must have at least that many leaves in its comparison tree •to have at least n! ~ 2^{n log n} leaves, it must

have height at least n log n (since it is only binary branching, the number of nodes at most doubles at every depth)

Lower Bound for Comparison

• How many input permutations are possible?

• For a comparison-based sorting algorithm to

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

java.lang.Comparable<T> Interface

- public int compareTo(T x);
 Returns a negative, zero, or positive value
 negative if this is before x
 0 if this.equals(x)
 positive if this is after 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