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### Description of the content of th
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InsertionSort //sort a[], an array of int for (int i = 1; i < a.length; i++) { int temp = a[i]; int k; for (k = i; 0 < k && temp < a[k-1]; k--) a[k] = a[k-1];a[k] = temp; Worst-case is O(n²) • Many people sort cards this way Consider reverse-sorted input • Invariant: everything to left of i is already sorted • Best-case is O(n) Works especially well when input Consider sorted input Expected case is O(n²) is nearly sorted Expected number of inversions is n(n-1)/4

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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 [1] to a [n-1]; find the smallest one and swap it with a [1]

• This is the other common way for people to sort cards

• Runtime

• Worst-case O(n²)

• Est-case O(n²)

• Expected-case O(n²)
```

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?

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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 ${\tt A}$ or ${\tt B}$ becomes empty, copy remaining elements from the other array (${\tt B}$ or ${\tt A}$, respectively) into ${\tt C}$

Merging Sorted Arrays

Language Array

C = merged array

B

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) = 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

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MergeSort Notes

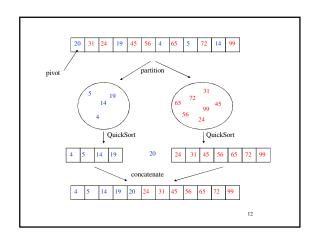
- Asymptotic complexity: O(n log n)
 - Much faster than O(n²)
- Disadvantage
 - Need extra storage for temporary arrays
 - In practice, this can be a disadvantage, even though MergeSort is asymptotically optimal for sorting
 - Can do MergeSort in place, but this is very tricky (and it slows down the algorithm significantly)
- Are there good sorting algorithms that do not use so much extra storage?
 - Yes: QuickSort

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QuickSort

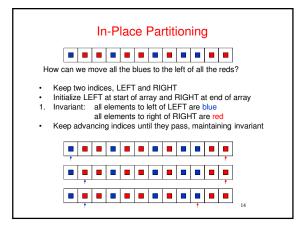
- Intuitive idea
- Given an array A to sort, choose a pivot value p
- Partition A into two subarrays, AX and AY
 - AX contains only elements ≤ p
- AY contains only elements $\geq p$
- Sort subarrays **AX** and **AY** separately
- Concatenate (not merge!) sorted AX and AY to get sorted A
 - Concatenation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

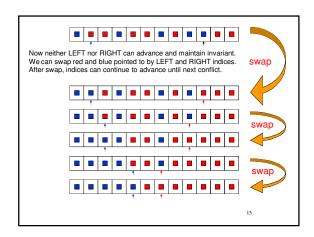
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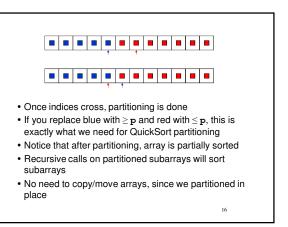


*Computions Computing the median, since this splits array in half Computing the median of an unsorted array is O(n), but algorithm is quite complicated Partitioning in place Can be done in O(n) time (next slide) Can be done in O(n) time (next slide) Partitioning in place Can be done in O(n) time (next slide) Can be done in O(n) time (next slide)

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QuickSort Analysis • Runtime analysis (worst-case) • Partition can work badly, producing this: P ≥ P • 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 sets • Switch to InsertionSort (for example) for sets of size, say, ≤ 9 • Definition of small depends on language, machine, etc.

Sorting Algorithm Summary • Why so many? Do computer • The ones we have scientists have some kind of discussed sorting fetish or what? ■ BubbleSort Stable sorts: Ins, Sel, Mer • InsertionSort ■ Worst-case O(n log n): Mer, Hea • SelectionSort Expected O(n log n): ■ MergeSort Mer, Hea, Qui • OuickSort Best for nearly-sorted sets: Ins • No extra space needed: Ins, • Other sorting algorithms ■ Fastest in practice: Qui HeapSort (will revisit this) ■ Least data movement: Sel • ShellSort (in text) RadixSort • BinSort CountingSort

Lower Bound for Comparison 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 worst-case time for the best possible
- 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 comparisonbased 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 worst-case number of comparisons for that algorithm
- Can 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, BI1 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! ~ 2n 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 n log n (since it is only binary branching, the number of nodes at most doubles at every depth)
- \bullet 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 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