

Slide 2

Another Sorting Algorithm: Merge Sort Several broad algorithm design techniques exist. Insertion Sort uses "incremental" approach. (I'm not 100% sure what the textbook means by that!) Another is "divide and conquer". This approach solves problems using three basic steps: Split the problem into subproblems that are smaller instances of the whole problem, unless they're so small and simple that they can be solved directly. (Call these base cases.) Solve the subproblems by solving them recursively, to produce subsolutions. Merge the subsolutions into a solution to the whole problem. Mergesort is a classic example.

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Algorithm for Merge Sort
 Now we can write an algorithm for the full merge sort (sorting array A(p:r)). Pseudocode, using MERGE as a subprogram:
Merge-Sort(A,p,r)
if $p \ge r$ //zero or one element?
$\begin{array}{l} \text{return} \\ q = \frac{\lfloor (p+r) \rfloor}{2} \text{ //midpoint of } A[p:r] \\ \text{MERGE-SORT}(A,p,q) \text{ // recursively sort } A[p,q] \\ \text{MERGE-SORT}(A,q+1,r) \text{ // recursively sort } A[q+1,r] \\ \text{// Merge } A[p,q] \text{ and } A[q+1,r] \text{ into } A[p,r] \end{array}$



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Analysis of MERGE-SORT

- Methods of counting operations based on loops don't really work here, since the repetition is based on recursion rather than on loops. When recursion is involved, often possible to describe its running time with a *recurrence equation* or *recurrence* and then proceed using mathematical tools for solving recurrences.
- More about this in Chapter 4, but for now ...

Analysis of MERGE-SORT • Suppose: The algorithm divides a problem of size n into a subproblems, each of size n/b. The "split" part requires time D(n). The "combine" part requires time C(n). • Then worst-case execution time T(n) can be expressed with the following recurrence relation:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n < n_0 \\ D(n) + aT(n/b) + C(n) & \text{otherwise} \end{cases}$$

Analysis of MERGE-SORT, Continued For MERGE-SORT: a is 2, and b is also 2. (True that if the array size is odd, strictly speaking the two subproblems are not the same size, but their size differs by at most one, so we can reasonably say "close enough".) Solving the base case is Θ(1) (i.e., constant time). D(n) (the "split") is also Θ(1). C(n) (the "merge") is Θ(n), as discussed previously. Combining these, and simplifying a little more, we get:

$$T(n) = \begin{cases} c_1 & \text{if } n = 1 \\ 2T(n/2) + c_2n & \text{if } n > 1 \end{cases}$$

Analysis of MERGE-SORT, Continued

• We can apply the "master theorem" presented in Chapter 4 to solve this recurrence, giving

$$T(n) = \Theta(n \log n)$$

(Strictly speaking, the log function here is base-2 log, but all log functions have the same order of growth, so we can be a little sloppy.)

(If you've forgotten: $\log_2 n$ is the number m such that $2^m = n - \text{e.g.}$, $\log_2 16 = 4$.)

• Since $\log n$ grows more slowly than n, this is very much a win over insertion sort. (Compare plots for some $n \log n$ functions and some n^2 functions.)



Implementations
I wrote code to implement this algorithm; it's MergeSort.cpp under "sample programs" on the course Web site.
Worth noting that this is likely not the most efficient implementation, since it does involve allocating two new arrays and copying the whole array at every step. Better to set up two arrays to begin with, and repeatedly sort with one as input and the other as output.
Note also that applying the algorithm to small arrays involves a lot of function overhead for the recursive calls, so an efficient sort might switch to insertion sort or another O(n²) algorithm for arrays below some threshold size.

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