

Lecture 3: Divide-and-Conquer Algorithms

We just derived an $O(n \log n)$ divide-and-conquer algorithm for solving the Maximum Contiguous Subarray problem.

In COMP171 you already saw [Mergesort](#), an $O(n \log n)$ time divide-and-conquer sorting algorithm.

Divide-and-Conquer is not a trick. It is a very useful general purpose tool for designing efficient algorithms.

The Basic Divide-and-Conquer Approach

Divide: Divide a given problem into two subproblems (ideally of approximately equal size).

Conquer: Solve each subproblem (directly or **recursively**), and

Combine: Combine the solutions of the two subproblems into a global solution.

Note: the hard work and cleverness is usually in the **Combine** step.

MERGESORT

Mergesort(A, i, j) : Sort $A[i \dots j]$
 If ($i \neq j$)
 { Mergesort ($A, i, \lfloor \frac{i+j}{2} \rfloor$) $M\left(\frac{j-i}{2}\right)$
 Mergesort ($A, 1 + \lfloor \frac{i+j}{2} \rfloor, j$) $M\left(\frac{j-i}{2}\right)$
 Merge the two sorted lists $O(j-i)$
 $A[i \dots \lfloor \frac{i+j}{2} \rfloor]$ and $A[1 + \lfloor \frac{i+j}{2} \rfloor, j]$
 and return complete sorted list
 }

The algorithm sorts an array of size N by splitting it into two parts of (almost) equal size, recursively sorting each of them, and then merging the two sorted subarrays back together into a fully sorted list in $O(N)$ time (how?).

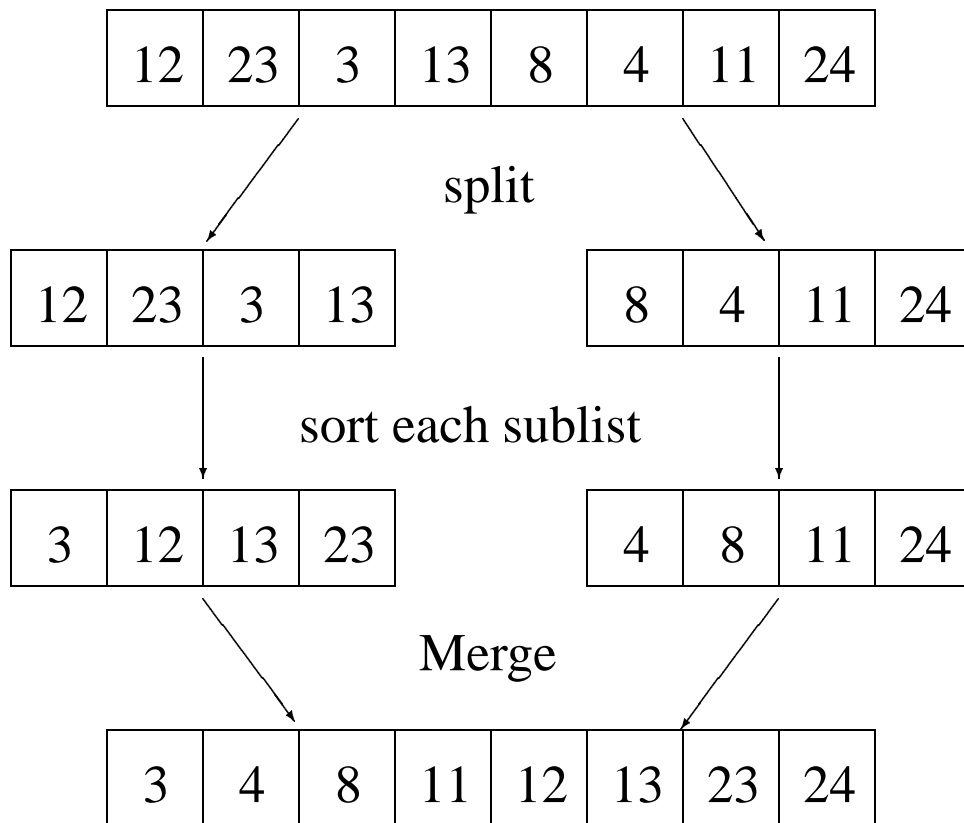
The running time of the algorithm satisfies

$$\forall N > 1, \quad M(N) \leq 2M(N/2) + O(N)$$

which we previously saw implies

$$M(N) = O(N \log N).$$

Mergesort Example



A More General Divide-and-Conquer Approach

Divide: Divide a given problem into subproblems (ideally of approximately equal size).

No longer only TWO subproblems

Conquer: Solve each subproblem (directly or **recursively**), and

Combine: Combine the solutions of the subproblems into a global solution.

The Polynomial Multiplication Problem

another divide-and-conquer algorithm

Problem:

Given two polynomials of degree $n - 1$

$$A(x) = a_0 + a_1x + \cdots + a_nx^{n-1}$$

$$B(x) = b_0 + b_1x + \cdots + b_nx^{n-1},$$

compute the product $A(x)B(x)$.

Example:

$$A(x) = 1 + 2x + 3x^2$$

$$B(x) = 3 + 2x + 2x^2$$

$$A(x)B(x) = 3 + 5x + 15x^2 + 10x^3 + 6x^4$$

Question: How can we **efficiently** calculate the coefficients of $A(x)B(x)$?

Assume that the coefficients a_i and b_i are stored in arrays $A[0 \dots n - 1]$ and $B[0 \dots n - 1]$.

Cost of any algorithm is number of scalar multiplications and additions performed.

Convolutions

Let $A(x) = \sum_{i=1}^n a_i x_i$ and $B(x) = \sum_{i=0}^m b_i x_i$.

Set $C(x) = \sum_{k=0}^{n+m} c_k x^k = A(x)B(x)$.

Then

$$c_k = \sum_{i=0}^k a_i b_{k-i}$$

for all $0 \leq k \leq m + n$.

Definition: The vector $(c_0, c_1, \dots, c_{m+n})$ is the **convolution** of the vectors (a_0, a_1, \dots, a_n) and (b_0, b_1, \dots, b_m) .

Calculating convolutions (and thus polynomial multiplication) is a major problem in digital signal processing.

The Direct (Brute Force) Approach

Let $A(x) = \sum_{i=1}^{n-1} a_i x_i$ and $B(x) = \sum_{i=0}^{n-1} b_i x_i$.

Set $C(x) = \sum_{k=0}^{2n-2} c_k x^k = A(x)B(x)$ with

$$c_k = \sum_{i=0}^k a_i b_{k-i}$$

for all $0 \leq k \leq 2n - 2$.

The direct approach is to compute all c_k using the formula above. The total number of multiplications and additions needed are n^2 and $(n - 1)^2$ respectively. Hence the complexity is $\Theta(n^2)$.

Questions: Can we do better?

Can we apply the divide-and-conquer approach to develop an algorithm?

The Divide-and-Conquer Approach

The Divide Step: Define

$$A_0(x) = a_0 + a_1x + \cdots + a_{\lfloor \frac{n}{2} \rfloor - 1}x^{\lfloor \frac{n}{2} \rfloor - 1},$$
$$A_1(x) = a_{\lfloor \frac{n}{2} \rfloor} + a_{\lfloor \frac{n}{2} \rfloor + 1}x + \cdots + a_nx^{n - \lfloor \frac{n}{2} \rfloor}.$$

Then $A(x) = A_0(x) + A_1(x)x^{\lfloor \frac{n}{2} \rfloor}$.

Similarly we define $B_0(x)$ and $B_1(x)$ such that

$$B(x) = B_0(x) + B_1(x)x^{\lfloor \frac{n}{2} \rfloor}.$$

Then

$$A(x)B(x) = A_0(x)B_0(x) + A_0(x)B_1(x)x^{\lfloor \frac{n}{2} \rfloor} + A_1(x)B_0(x)x^{\lfloor \frac{n}{2} \rfloor} + A_1(x)B_1(x)x^{2\lfloor \frac{n}{2} \rfloor}.$$

Remark: The original problem of size n is divided into 4 problems of input size $\frac{n}{2}$.

Example:

$$\begin{aligned}A(x) &= 2 + 5x + 3x^2 + x^3 \\B(x) &= 1 + 2x + 2x^2 + 3x^3 \\A(x)B(x) &= 2 + 9x + 17x^2 + 23x^3 + 23x^4 + 11x^5 + 3x^6\end{aligned}$$

$$\begin{aligned}A_0(x) &= 2 + 5x, & A_1(x) &= 3 + x, & A(x) &= A_0(x) + A_1(x)x^2 \\B_0(x) &= 1 + 2x, & B_1(x) &= 2 + 3x, & B(x) &= B_0(x) + B_1(x)x^2\end{aligned}$$

$$\begin{aligned}A_0(x)B_0(x) &= 2 + 9x + 10x^2 \\A_1(x)B_1(x) &= 6 + 11x + 3x^2 \\A_0(x)B_1(x) &= 4 + 16x + 15x^2 \\A_1(x)B_0(x) &= 3 + 7x + 2x^2 \\A_0(x)B_1(x) + A_1(x)B_0(x) &= 7 + 23x + 17x^2\end{aligned}$$

$$\begin{aligned}& A_0(x)B_0(x) \\+ & (A_0(x)B_1(x) + A_1(x)B_0(x))x^2 \\+ & A_1(x)B_1(x)x^4 = 2 + 9x + 17x^2 \\& + 23x^3 + 23x^4 + \\& + 11x^5 + 3x^6\end{aligned}$$

The Divide-and-Conquer Approach

The Conquer Step: Solve the four subproblems, i.e., computing

$$\begin{aligned} &A_0(x)B_0(x), \quad A_0(x)B_1(x), \\ &A_1(x)B_0(x), \quad A_1(x)B_1(x) \end{aligned}$$

by recursively calling the algorithm 4 times.

The Divide-and-Conquer Approach

The Combining Step: Adding the following four polynomials

$$\begin{aligned} &A_0(x)B_0(x) \\ &+ A_0(x)B_1(x)x^{\lfloor \frac{n}{2} \rfloor} \\ &+ A_1(x)B_0(x)x^{\lfloor \frac{n}{2} \rfloor} \\ &+ A_1(x)B_1(x)x^{2\lfloor \frac{n}{2} \rfloor}. \end{aligned}$$

takes $\Theta(n)$ operations. Why?

The First Divide-and-Conquer Algorithm

PolyMulti1($A(x), B(x)$)

{

$$A_0(x) = a_0 + a_1x + \cdots + a_{\lfloor \frac{n}{2} \rfloor - 1}x^{\lfloor \frac{n}{2} \rfloor - 1};$$

$$A_1(x) = a_{\lfloor \frac{n}{2} \rfloor} + a_{\lfloor \frac{n}{2} \rfloor + 1}x + \cdots + a_nx^{n - \lfloor \frac{n}{2} \rfloor};$$

$$B_0(x) = b_0 + b_1x + \cdots + b_{\lfloor \frac{n}{2} \rfloor - 1}x^{\lfloor \frac{n}{2} \rfloor - 1};$$

$$B_1(x) = b_{\lfloor \frac{n}{2} \rfloor} + b_{\lfloor \frac{n}{2} \rfloor + 1}x + \cdots + b_nx^{n - \lfloor \frac{n}{2} \rfloor};$$

$$U(x) = \text{PolyMulti1}(A_0(x), B_0(x));$$

$$V(x) = \text{PolyMulti1}(A_0(x), B_1(x));$$

$$W(x) = \text{PolyMulti1}(A_1(x), B_0(x));$$

$$Z(x) = \text{PolyMulti1}(A_1(x), B_1(x));$$

$$\text{return } \left(U(x) + [V(x) + W(x)]x^{\lfloor \frac{n}{2} \rfloor} + Z(x)x^{2\lfloor \frac{n}{2} \rfloor} \right);$$

}

Running Time of the Algorithm

Assume n is a power of 2, $n = 2^h$. By substitution (expansion),

$$\begin{aligned}T(n) &= 4T\left(\frac{n}{2}\right) + cn \\&= 4\left[4T\left(\frac{n}{2^2}\right) + c\frac{n}{2}\right] + cn \\&= 4^2T\left(\frac{n}{2^2}\right) + (1+2)cn \\&= 4^2\left[4T\left(\frac{n}{2^3}\right) + c\frac{n}{2^2}\right] + (1+2)cn \\&= 4^3T\left(\frac{n}{2^3}\right) + (1+2+2^2)cn \\&\vdots \\&= 4^iT\left(\frac{n}{2^i}\right) + \sum_{j=0}^{i-1} 2^j cn \quad (\text{induction}) \\&\vdots \\&= 4^hT\left(\frac{n}{2^h}\right) + \sum_{j=0}^{h-1} 2^j cn \\&= n^2 T(1) + cn(n-1) \\&\quad (\text{since } n = 2^h \text{ and } \sum_{j=0}^{h-1} 2^j = 2^h - 1 = n - 1) \\&= \Theta(n^2).\end{aligned}$$

The same order as the brute force approach!

Comments on the Divide-and-Conquer Algorithm

Comments: The divide-and-conquer approach makes no essential improvement over the brute force approach!

Question: Why does this happen.

Question: Can you improve this divide-and-conquer algorithm?

Problem: Given 4 numbers

$$A_0, A_1, B_0, B_1$$

how many multiplications are needed to calculate the three values

$$A_0B_0, A_0B_1 + A_1B_0, A_1B_1?$$

This can obviously be done using 4 multiplications but there is a way of doing this using only the following 3:

$$U = (A_0 + A_1)(B_0 + B_1)$$

$$V = A_0B_0$$

$$W = A_1B_1$$

Y and Z are what we originally wanted and

$$A_0B_1 + A_1B_0 = U - V - W.$$

Improving the Divide-and-Conquer Algorithm

Define

$$U(x) = (A_0(x) + A_1(x)) \times (B_0(x) + B_1(x))$$

$$V(x) = A_0(x)B_0(x)$$

$$W(x) = A_1(x)B_1(x)$$

Then

$$U(x) - V(x) - W(x) = A_0(x)B_1(x) + A_1(x)B_0(x).$$

Hence $A(x)B(x)$ is equal to

$$V(x) + [U(x) - V(x) - W(x)]x^{\lfloor \frac{n}{2} \rfloor} + W(x) \times x^{2\lfloor \frac{n}{2} \rfloor}$$

Conclusion: You need to call the multiplication procedure **3**, rather than **4** times.

The Second Divide-and-Conquer Algorithm

PolyMulti2($A(x)$, $B(x)$)

{

$$A_0(x) = a_0 + a_1x + \cdots + a_{\lfloor \frac{n}{2} \rfloor - 1}x^{\lfloor \frac{n}{2} \rfloor - 1};$$

$$A_1(x) = a_{\lfloor \frac{n}{2} \rfloor} + a_{\lfloor \frac{n}{2} \rfloor + 1}x + \cdots + a_nx^{n - \lfloor \frac{n}{2} \rfloor};$$

$$B_0(x) = b_0 + b_1x + \cdots + b_{\lfloor \frac{n}{2} \rfloor - 1}x^{\lfloor \frac{n}{2} \rfloor - 1};$$

$$B_1(x) = b_{\lfloor \frac{n}{2} \rfloor} + b_{\lfloor \frac{n}{2} \rfloor + 1}x + \cdots + b_nx^{n - \lfloor \frac{n}{2} \rfloor};$$

$$U(x) = \text{PolyMulti2}(A_0(x) + A_1(x), B_0(x) + B_1(x))$$

$$V(x) = \text{PolyMulti2}(A_0(x), B_0(x));$$

$$W(x) = \text{PolyMulti2}(A_1(x), B_1(x));$$

$$\text{return } \left(V(x) + [U(x) - V(x) - W(x)]x^{\lfloor \frac{n}{2} \rfloor} + W(x)x^{2\lfloor \frac{n}{2} \rfloor} \right);$$

}

Running Time of the Modified Algorithm

Assume $n = 2^h$. Let $\lg x$ denote $\log_2 x$.

By the substitution method,

$$\begin{aligned}
 T(n) &= 3T\left(\frac{n}{2}\right) + cn \\
 &= 3\left[3T\left(\frac{n}{2^2}\right) + c\frac{n}{2}\right] + cn \\
 &= 3^2T\left(\frac{n}{2^2}\right) + \left(1 + \frac{3}{2}\right)cn \\
 &= 3^2\left[3T\left(\frac{n}{2^3}\right) + c\frac{n}{2^2}\right] + \left(1 + \frac{3}{2} + \frac{3^2}{2^2}\right)cn \\
 &= 3^3T\left(\frac{n}{2^3}\right) + \left(1 + \frac{3}{2} + \left[\frac{3}{2}\right]^2\right)cn \\
 &\quad \vdots \\
 &= 3^hT\left(\frac{n}{2^h}\right) + \sum_{j=0}^{h-1} \left[\frac{3}{2}\right]^j cn.
 \end{aligned}$$

We have

$$3^h = (2^{\lg 3})^h = 2^{h \lg 3} = (2^h)^{\lg 3} = n^{\lg 3} \approx n^{1.586},$$

and

$$\sum_{j=0}^{h-1} \left[\frac{3}{2}\right]^j = \frac{(3/2)^h - 1}{3/2 - 1} = 2 \cdot \frac{3^h}{2^h} - 2 = 2n^{\lg 3 - 1} - 2.$$

Hence

$$T(n) = \Theta(n^{\lg 3}T(1) + 2cn^{\lg 3}) = \Theta(n^{\lg 3}).$$

Comments

- The divide-and-conquer approach doesn't always give you the best solution.
Our original D-A-C algorithm was just as bad as brute force.
- There is actually an $O(n \log n)$ solution to the polynomial multiplication problem.
It involves using the *Fast Fourier Transform* algorithm as a subroutine.
The FFT is another classic D-A-C algorithm (Chapt 30 in CLRS gives details).
- The idea of using 3 multiplications instead of 4 is used in large-integer multiplications.
A similar idea is the basis of the classic [Strassen matrix multiplication algorithm](#) (CLRS, Chapter 28).