

Title-The Polynomial Multiplication

Problem-Given two polynomials- $A(x) = a_0 + a_1x + a_2x^2 + \dots + a_{n-1}x^{n-1}$ and $B(x) = b_0 + b_1x + b_2x^2 + \dots + b_{n-1}x^{n-1}$, find the polynomial $C(x) = A(x) * B(x)$.

General Algorithm-If we think of A and B as vectors, then C vector is called '*Convolution*' of A and B (represented as $\{A \otimes B\}$). In this algorithm each coefficient in vector a must be multiplied by each coefficient in vector b this makes its time complexity $O(n^2)$.

Objective-To study an algorithm that runs faster than the naive algorithm.

Fast Fourier Transform-

FFT was invented around 1805 by Carl Friedrich Gauss but his work was not widely recognized. FFTs became popular after James Cooley of IBM and John Tukey of Princeton published a paper in 1965 reinventing the algorithm and describing how to perform it conveniently on a computer. Therefore it is also known as Cooley–Tukey FFT algorithm.

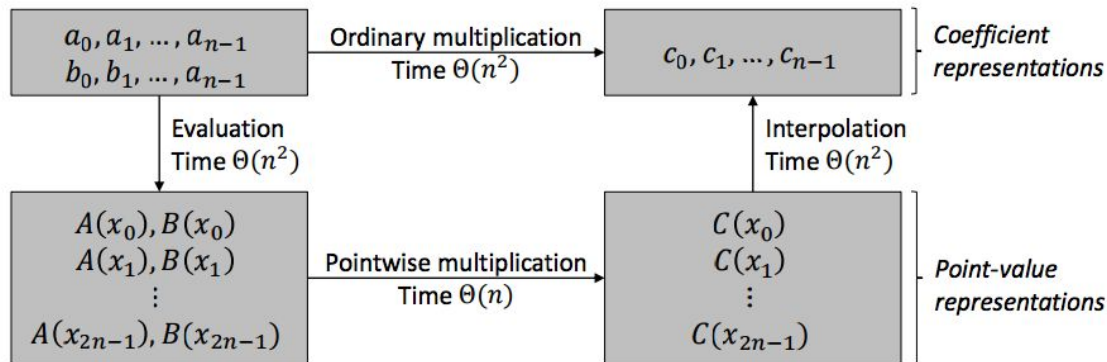
The idea is to represent polynomials in point-value form and then compute the product. Using Horner's method evaluation of the polynomial at n -points takes $O(n^2)$. Now that the polynomials are converted in point-value representation evaluation of $C(x) = A(x) * B(x)$ takes $O(n)$ time.

An important point here is that $C(x)$ has degree bound $2n$, then n points will give n points of $C(x)$, so for that case we need $2n$ different values of x to calculate $2n$ different values of y . This is because of the Fundamental Theorem of Algebra that states—*“A degree $n-1$ polynomial $A(x)$ is uniquely specified by its evaluation at n distinct values of x ”*.

Now that we have the product calculated, we can convert it back to coefficient vector form. The inverse evaluation of determining the coefficient form of a

polynomial from a point-value representation is called interpolation. Interpolation can be done by using Lagrange's formula in $O(n^2)$.

The road so far-



Now the efficiency of our algorithm depends on the efficiency of conversion between the two representations i.e coefficient representation to point-value representation and visa-versa. This conversion can be done efficiently by choosing the evaluation points carefully. If we choose “complex roots of unity” as the evaluation points, we can produce a point-value representation by taking Discrete Fourier Transform (DFT) of coeff. vector.

Discrete Fourier Transform (DFT)-

DFT is evaluating values of the polynomial at n complex roots of unity $\omega_n^0, \omega_n^1, \dots, \omega_n^{n-1}$ ($\omega_n^k = e^{2\pi i k/n}$ $k=0,1,2,\dots,n-1$). Here we assume n is power of 2, we can meet this requirement by adding high-order zero coeff.

Fast Fourier Transform (FFT) -

Fast Fourier Transform (FFT) takes advantage of the special properties of the complex roots of unity to compute DFT in time $O(n \log n)$.

Special properties of the complex roots of unity are-

1. Cancellation lemma.
2. Halving lemma.

3. Summation lemma.

Divide-and-conquer strategy-

Define two new polynomials of degree-bound $n/2$, using even-index and odd-index coefficients of $A(x)$ separately –

$$A_{\text{even}} = a_0 + a_2x + a_4x^2 + \dots + a_{n-2}x^{n/2-1}$$

$$A_{\text{odd}} = a_1 + a_3x + a_5x^2 + \dots + a_{n-1}x^{n/2-1}$$

$$A(x) = A_{\text{even}}(x^2) + xA_{\text{odd}}(x^2)$$

The problem of evaluating $A(x)$ at $\omega^n^0, \omega^n^1, \dots, \omega^n^{n-1}$ reduces to evaluating the degree-bound $n/2$ polynomials A_{even} and A_{odd} at the points $(\omega^n^0)^2, (\omega^n^1)^2, \dots, (\omega^n^{n-1})^2$. Polynomials A_{even} and A_{odd} are recursively evaluated at the $n/2$ complex $n/2$ th roots of unity. Subproblems have exactly the same form as the original problem, but half the size.

So, the recurrence formed is $T(n) = 2T(n/2) + O(n)$ i.e $O(n \log n)$.

Interpolation-Just replace ω^n with ω^n^{-1} and divide each element of the result by n .

Pseudocode for recursive fft -

Let A be an array of length n , ω be primitive n th root of unity.

Goal: produce DFT $F(A)$: evaluation of A at $1, \omega, \omega^2, \dots, \omega^{n-1}$.

FFT(A, n, ω)

{

if ($n == 1$) return vector (a_0)

else {

$A_{\text{even}} = (a_0, a_2, \dots, a_{n-2})$

$A_{\text{odd}} = (a_1, a_3, \dots, a_{n-1})$

$F_{\text{even}} = \text{FFT}(A_{\text{even}}, n/2, \omega^2)$ // ω^2 is a $n/2$ -th root of unity

$F_{\text{odd}} = \text{FFT}(A_{\text{odd}}, n/2, \omega^2)$

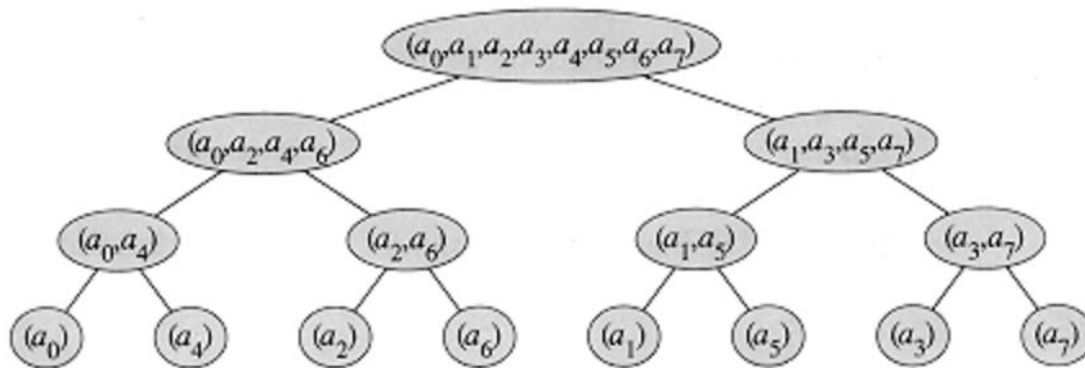
$F =$ new vector of length n

```

x = 1
for (j=0; j < n/2; ++j) {
    F[j] = F_even[j] + x*F_odd[j]
    F[j+n/2] = F_even[j] - x*F_odd[j]
    x = x * ω
}
return F
}

```

Here is the tree of input vectors having $n=8$ to the recursive calls of the FFT procedure.



A popular FFT application area is in the field of signal processing. It is also widely used in image processing to analyze data in 2 or more dimensions. There has been a recent modification of 2 dimensional fft with an analog of the Cooley–Tukey algorithm for image processing.

References-

- J. W. Cooley and J. W. Tukey, "An algorithm for the machine calculation of complex Fourier series", *Math. of Comput.*, vol. 19, pp. 297-301, April 1965.
- T. H. Cormen et al. Introduction to Algorithms. McGraw-Hill, Inc., 2nd edition, 2001.

- Cooley, James W.; Lewis, Peter A. W.; Welch, Peter D. (1967). "Historical notes on the Fast Fourier transform" (PDF). *IEEE Transactions on Audio and Electroacoustics*.
- Fast Fourier Transform retrieved from https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-046j-design-and-analysis-of-algorithms-spring-2012/lecture-notes/MIT6_046JS12_lec05.pdf
- Noskov.M & Tulatichikov V. ,(2017) Modification of a two-dimensional fast Fourier transform algorithm with an analog of the Cooley–Tukey algorithm for image processing.*Institute of Space and Information Technology* . doi 10.1134/S1054661817010096.
- Cooley–Tukey FFT algorithm wikipedia page-
https://en.wikipedia.org/wiki/Cooley%E2%80%93Tukey_FFT_algorithm