10 Polynomial Multiplication and Fast Fourier Transform [KT 5.6]

Given two degree-(n-1) univariate polynomials $A(x) = \sum_{j=0}^{n-1} a_j x^j$ and $B(x) = \sum_{j=0}^{n-1} b_j x^j$, it is easy to compute their sum in linear time. However, the naïve way to compute their product

$$C(x) = A(x)B(x) = \sum_{j=1}^{2n-2} \left(\sum_{k} a_k b_{j-k}\right) x^j$$

takes $\Theta(n^2)$ time.

To compute C(x) faster, we plan to use the following strategy.

- 1. Pick (2n-1) different locations $x_1, x_2, \ldots, x_{2n-1}$, and evaluate and $A(x_1), A(x_2), \ldots, A(x_{2n-1})$ and $B(x_1), B(x_2), \ldots, B(x_{2n-1})$.
- 2. Now we know the evaluation of C(x) at locations $x_1, x_2, \ldots, x_{2n-1}$. More specifically, for each j, we have $C(x_j) = A(x_j)B(x_j)$.
- 3. Use the (2n-1) evaluations of C(x) to recover the coefficients of $C(x) = \sum_{j=0}^{2n-2} c_j x^j$.

Before talking about time complexity, let us first prove the feasibility (correctness) of such strategy. Indeed, the correctness comes from the following fact.

Fact 1 Given n points $(x_1, y_1), \ldots, (x_n, y_n)$ with distinct x_i 's, there exists a unique degree-(n-1) univariate polynomial passing through all these n points.

Proof: To prove the fact, we first prove that there exists such a polynomial and then show that the polynomial is unique.

Proof of Existence. This is done by interpolation. For each j = 1, 2, ..., n, we write down the following polynomial

$$\frac{\prod_{k=1,\dots,n,k\neq j}(x-x_k)}{\prod_{k=1,\dots,n,k\neq j}(x_j-x_k)},$$

and observe that such a degree-(n-1) polynomial takes value 1 at $x=x_j$ and value 0 at $x=x_k$ for all $k \neq j$. Therefore, it is easy to verify the following degree-(n-1) polynomial takes value y_j at $x=x_j$ for all j.

$$\sum_{i=1}^{n} \frac{\prod_{k=1,\dots,n,k\neq j} (x-x_k)}{\prod_{k=1,\dots,n,k\neq j} (x_j-x_k)} y_j.$$

To prove the uniqueness of such a polynomial, we use the Fundamental Theorem of Algebra.

Theorem 1 Every non-zero, single-variable, degree-n polynomial with complex coefficients has, counted with multiplicity, exactly n complex roots.

Proof of Uniqueness. Now let us assume there are two degree-(n-1) polynomials p(x) and q(x) passing through the given n points. Let r(x) = p(x) - q(x). We see that $r(x_j) = 0$ for all j = 1, 2, ..., n. In other words, the degree-(n-1) polynomial r(x) has n roots. According to the Fundamental Theorem of Algebra, r(x) must be constantly zero, i.e. p(x) and q(x) are the same polynomial.

Now we see that the strategy mentioned above is correct. But what about the time complexity? Step 2 takes O(n) time. However, we need a fast algorithm for Step 1 and Step 3.

While a naive implementation of Step 1 and Step 3 takes at least $\Omega(n^2)$ time, we observe that we can choose evaluation locations x_1, \ldots, x_{2n-1} at our will. We hope to choose a few locations those are easier for fast evaluation. Here is our new task.

New task. Pick n evaluation locations x_1,\ldots,x_n and design a fast algorithm to evaluate any degree-(n-1) polynomial $A(x)=\sum_{j=0}^{n-1}a_jx^j$ at these n locations. Let us assume n is a power of 2 (i.e. $n=2^m$ for some integer m) but we can always meet such a condition by padding 0's to the higher power terms of the input polynomial.

Indeed, the n locations we pick will be the n n-th roots of unity.

Fact 2 There are exactly n n-th root of unity, and they are $e^{2\pi i \cdot \frac{1}{n}}, e^{2\pi i \cdot \frac{2}{n}}, \dots, e^{2\pi i \cdot \frac{n}{n}} = 1$.

Definition 1 An n-th root of unity x is primitive if $x^k \neq 1$ for all 0 < k < n.

Fact 3 $e^{2\pi i \cdot \frac{j}{n}}$ is a primitive n-th root of unity if any only if gcd(j,n) = 1.

We will let $\omega = e^{2\pi i \cdot \frac{1}{n}}$ and evaluate the polynomials at n locations $\omega, \omega^2, \dots, \omega^n = 1$.

In order to evaluate the given polynomial $A(x) = a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$ at point $1, \omega, \dots, \omega^{n-1}$, we introduce

$$A_{even}(x) = a_0 + a_2x + a_4x^2 + \dots + a_{n-2}x^{(n-2)/2},$$

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

and we apply the following identity

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

to $x=1,\omega,\omega^2,\ldots,\omega^{n-1}$. We observe that we need to evaluate both degree- $(\frac{n}{2}-1)$ polynomials A_{even} and A_{odd} at the following n points $x=1,\omega^2,\ldots,\omega^{2n-2}$. However, since $x^n=1$, we have

$$\omega^n = 1, \omega^{n+2} = \omega^2, \dots, \omega^{2n-2} = \omega^{n-2}.$$

Therefore, there are only n/2 distinct points for evaluation. And this reduces the problem of evaluating the degree-(n-1) polynomial A(x) at n n-th roots of unity to the two subproblems of evaluating degree-(n/2-1) polynomials at (n/2) (n/2)-th roots of unity. Following this idea, we are able to design the Fast Fourier Transform algorithm in $O(n \log n)$ time. Finally, we formally describe the FFT algorithm to evaluate a polynomial at n-th roots of unity as follows.

$$FFT(n, A(x) = \sum_{j=0}^{n-1} a_j x^j)$$

IF
$$(n=1)$$
 RETURN $A(1)=a_0$
Let $\omega=e^{2\pi i\cdot \frac{1}{n}}$
 $A_{even}(x)=a_0+a_2x+\cdots+a_{n-2}x^{n/2-1}$
 $A_{odd}(x)=a_1+a_3x+\cdots+a_{n-1}x^{n/2-1}$
Recursively evaluate $A_{even}(\omega^2), A_{even}((\omega^2)^2), \ldots, A_{even}((\omega^2)^{n/2})$
Recursively evaluate $A_{odd}(\omega^2), A_{odd}((\omega^2)^2), \ldots, A_{odd}((\omega^2)^{n/2})$
For each $j=1,\ldots,n$, evaluate $A(\omega^j)=A_{even}((\omega^2)^j)+\omega^j\cdot A_{odd}((\omega^2)^j)$

Now we discuss the inverse Fast Fourier Transform in order to complete Step 3 of the whole polynomial multiplication algorithm. That is, once we have the evaluations $C(1), \ldots, C(\omega^{n-1})$, we would like to design a fast algorithm to recover the coefficients of C(x) polynomial. In order to

do this, we introduce the Fourier Transform matrix and took a matrix view of Fourier Transform. We define the Fourier Transform matrix $M_n(\omega)$ to be an $n \times n$ matrix in the following form.

$$M_n(\omega) = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots \\ 1 & \omega & \omega^2 & \omega^3 & \dots \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

That is, the *i*-th row and *j*-th column of the matrix is $\omega^{(i-1)(j-1)}$.

We can verify that for any degree-(n-1) polynomial $A(x) = a_0 + a_1x + \ldots + a_{n-1}x^{n-1}$, we have

$$\begin{pmatrix} A(1) \\ A(\omega) \\ A(\omega^2) \\ \vdots \end{pmatrix} = M_n(\omega) \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \end{pmatrix}$$

and this is the matrix view of Fourier Transform (also what was done by the FFT algorithm). In order to recover the coefficients of a polynomial form its evaluations at *n*-th roots of unity, we can simply multiply the left hand side of the equation above by the inverse of the Fourier Transform matrix, i.e.,

$$\begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ \vdots \end{pmatrix} = (M_n(\omega))^{-1} \begin{pmatrix} C(1) \\ C(\omega) \\ C(\omega^2) \\ \vdots \end{pmatrix}$$

The following claim shows that $(M_n(\omega))^{-1} = \frac{1}{n} M_n(\omega^{-1})$. Indeed, using this claim and the fact that ω^{-1} is also a primitive n-th root of unity, we see that we only need to invoke the FFT algorithm again, with ω^{-1} and $C(1), C(\omega), C(\omega^2), \ldots$ as input, and get the coefficients of C(x) in $O(n \log n)$ time.

Claim 6
$$M_n(\omega)M_n(\omega^{-1})=nI$$
.

Proof: Let us directly compute the i-th row and j-th entry of the product matrix. By the definition of matrix multiplication, this value is

$$\sum_{k=1}^{n} (M_n(\omega))_{i,k} (M_n(\omega^{-1}))_{k,j} = \sum_{k=1}^{n} \omega^{(i-1)(k-1)} w^{(-1)(k-1)(j-1)} = \sum_{k=1}^{n} (\omega^{i-j})^{(k-1)}.$$

We see this is sum of a geometric series. If i=j (and therefore $\omega^{i-j}=1$), this sum becomes n, otherwise we have $\omega^{i-j}\neq 1$, and the sum becomes

$$\frac{1 - (\omega^{i-j})^n}{1 - \omega^{i-j}} = 0,$$

where the equation is because ω^{i-j} is also an n-th root of unity.

To summarize, we have proved that if i = j, the diagonal entries are always n, otherwise, the off-diagonal entries are always 0. Therefore the product matrix is nI.