

## Lecture Notes 4: Fast Fourier Transformation

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## 1 Problem Definition

We have two polynomials  $p(x)$  and  $q(x)$ . We want to compute the result of  $p(x)q(x)$ .

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

$$q(x) = b_{n-1}x^{n-1} + \cdots + b_1x + b_0$$

$$r(x) = p(x)q(x)$$

We model the problem as  $A\vec{x} = \vec{b}$ .

$$\begin{pmatrix} x_2^2 & x_2 & 1 \\ x_1^2 & x_1 & 1 \\ x_0^2 & x_0 & 1 \end{pmatrix} \begin{pmatrix} C_2 \\ C_1 \\ C_0 \end{pmatrix} = \begin{pmatrix} r(X_2) \\ r(X_1) \\ r(X_0) \end{pmatrix}$$

$$\therefore \vec{x} = A^{-1}\vec{b}$$

If  $A$  and  $\vec{b}$  were known, we could find the coefficient of  $r(x)$  by solving  $\vec{x} = A^{-1}\vec{b}$ . Since the computation of  $\text{inverse}(A)$  is roughly  $O(n^3)$ , if the value of  $X$  are arbitrary values, the computational cost will also be  $O(n^3)$ . By carefully choosing the values of  $x$ , we can significantly reduce the computation cost.

## 2 Algorithm

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### Algorithm 1 Polynomial Multiplication

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1. Find  $x_0, x_1, \dots, x_{2n}$  (Find  $A$ )
  2. Evaluate  $p(x_0), p(x_1), \dots, p(x_{2n})$  ( $A\vec{x}_p = \vec{b}_p$ )
  3. Evaluate  $q(x_0), q(x_1), \dots, q(x_{2n})$  ( $A\vec{x}_q = \vec{b}_q$ )
  4. Compute  $r(x_0), r(x_1), \dots, r(x_{2n})$  (Use  $r(x_i) = p(x_i)q(x_i)$  to get  $\vec{b}_r$ )
  5. Solve  $\vec{x}_r$  in  $A\vec{x}_r = \vec{b}_r$
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If we use some random points to substitute in, it's  $O(n^3)$  time. Because there are  $n$  points and both of two polynomials. So, we use  $\omega_j, j = 1, \dots, 2n - 1$ .  $\omega_j$  are the  $2n - th$

root of 1. It can get  $\vec{b}_p$  and  $\vec{b}_q$  in  $O(n \lg n)$  time. We use  $2n = 8(\omega = e^{i\frac{2\pi}{8}})$  in the following example.

$$A = W_8 = \left[ \begin{array}{cccc|cccc} 1^0 & 1^1 & 1^2 & 1^3 & 1^4 & 1^5 & 1^6 & 1^7 \\ 1 & \omega^1 & \omega^2 & \omega^3 & \omega^4 & \omega^5 & \omega^6 & \omega^7 \\ 1 & \omega^2 & \omega^4 & \omega^6 & \omega^8 & \omega^{10} & \omega^{12} & \omega^{14} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \omega^{12} & \omega^{15} & \omega^{18} & \omega^{21} \\ \hline 1 & \omega^4 & \omega^8 & \omega^{12} & \omega^{16} & \omega^{20} & \omega^{24} & \omega^{28} \\ 1 & \omega^5 & \omega^{10} & \omega^{15} & \omega^{20} & \omega^{25} & \omega^{30} & \omega^{35} \\ 1 & \omega^6 & \omega^{12} & \omega^{18} & \omega^{24} & \omega^{30} & \omega^{36} & \omega^{42} \\ 1 & \omega^7 & \omega^{14} & \omega^{21} & \omega^{28} & \omega^{35} & \omega^{42} & \omega^{49} \end{array} \right]$$

$$W_4 = \left[ \begin{array}{cc|cc} 1^0 & 1^2 & 1^4 & 1^6 \\ 1 & \omega^2 & \omega^4 & \omega^6 \\ \hline 1 & \omega^4 & \omega^8 & \omega^{12} \\ 1 & \omega^6 & \omega^{12} & \omega^{18} \end{array} \right]$$

We want to find the relation between  $W_8$  and  $W_4$ . Then, we can use Divide-and-Conquer. First, we collect all odd columns to the “front” and put all even columns to the “back”.  $\Leftrightarrow$  multiply a permutation matrix  $P$ .

$$P_8 = \begin{pmatrix} 1 & & & & & & & \\ & & & & 1 & & & \\ & 1 & & & & & & \\ & & & & & 1 & & \\ & & 1 & & & & & \\ & & & & & & 1 & \\ & & & 1 & & & & \\ & & & & & & & 1 \end{pmatrix}$$

Then,

$$W_8 \times P_8 = \begin{pmatrix} M_1 & M_3 \\ M_2 & M_4 \end{pmatrix}$$

$$= \left[ \begin{array}{cccc|cccc} 1^0 & 1^1 & 1^2 & 1^3 & 1^4 & 1^5 & 1^6 & 1^7 \\ 1 & \omega^2 & \omega^4 & \omega^6 & \omega^1 & \omega^3 & \omega^5 & \omega^7 \\ 1 & \omega^4 & \omega^8 & \omega^{12} & \omega^2 & \omega^6 & \omega^{10} & \omega^{14} \\ 1 & \omega^6 & \omega^{12} & \omega^{18} & \omega^3 & \omega^9 & \omega^{15} & \omega^{21} \\ \hline 1 & \omega^8 & \omega^{16} & \omega^{24} & \omega^4 & \omega^{12} & \omega^{20} & \omega^{28} \\ 1 & \omega^{10} & \omega^{20} & \omega^{30} & \omega^5 & \omega^{15} & \omega^{25} & \omega^{35} \\ 1 & \omega^{12} & \omega^{24} & \omega^{36} & \omega^6 & \omega^{18} & \omega^{30} & \omega^{42} \\ 1 & \omega^{14} & \omega^{28} & \omega^{42} & \omega^7 & \omega^{21} & \omega^{35} & \omega^{49} \end{array} \right]$$

$$M_1 = M_2 = W_4$$

$$\text{Let } M_x \text{ be } \left( \begin{array}{cc|cc} 1^0 & 0 & 0 & 0 \\ 0 & \omega^1 & 0 & 0 \\ \hline 0 & 0 & \omega^2 & 0 \\ 0 & 0 & 0 & \omega^3 \end{array} \right) \text{ Let } M_y \text{ be } \left( \begin{array}{cc|cc} \omega^4 & 0 & 0 & 0 \\ 0 & \omega^5 & 0 & 0 \\ \hline 0 & 0 & \omega^6 & 0 \\ 0 & 0 & 0 & \omega^7 \end{array} \right)$$

Then,  $M_3 = M_x \times M_1$ ,  $M_4 = M_y \times M_2$

$\because M_y = (\omega^4) \times M_x$   
 $\therefore M_y = -M_x$   
 Let  $D_4$  be  $M_x$ .  
 $\because P_8$  is an orthogonal matrix.  
 $\therefore P_8 \times P_8^{-1} = I$ .  
 Then,

$$\begin{aligned}
 W_8 \vec{x} &= W_8 P_8 P_8^{-1} \vec{x} \\
 &= \begin{pmatrix} W_4 & D_4 W_4 \\ W_4 & -D_4 W_4 \end{pmatrix} P_8^{-1} \vec{x}
 \end{aligned}$$

Then we can reduce the problem( $W_8$ ) to the subproblem( $W_4$ ), whose size is half of original one.

### 3 iFFT Algorithm

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#### Algorithm 2 iFFT Algorithm

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 $\vec{c} = \text{BitReverse}(\vec{x})$ 
for  $s = 0 : \lg n - 1$  do
   $m \leftarrow 2^s$ 
   $\omega \leftarrow e^{\frac{-i\theta}{m}}$ 
  Set  $D_m$ 
  for  $k = 0 : 2m : n - 1$  do
     $C_1 \leftarrow C(k : k + m - 1)$ 
     $C_2 \leftarrow D_m C(k + m : k + 2m - 1)$ 
     $C(k : k + 2m - 1) \leftarrow [C_1 + C_2, C_1 - C_2]^T$ 
  end for
end for
  
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### 4 Conclusion

At the first, we can use FFT to get  $\vec{b}_p$  and  $\vec{b}_q$  in  $O(n \lg n)$  time. Then, we compute the array-wise multiplication( $\vec{b}_r$ ) of  $\vec{b}_p$  and  $\vec{b}_q$ . Finally, we use iFFT to get  $\vec{x}_r$  in  $O(n \lg n)$  time.

### 5 Appendix

**Definition 1** (Euler Formula).

$$e^{i\theta} = \cos \theta + i \sin \theta$$

*Proof.*  $O(1)$  Multiplication  
 Let  $x = e^{i\theta}$ .

Let  $y = e^{i\phi}$ .

Then,

$$\begin{aligned} x * y &= e^{(i\theta)+(i\phi)} \\ &= e^{i(\theta+\phi)} \\ &= \cos(\theta + \phi) + i \sin(\theta + \phi). \end{aligned}$$

□

We can compute  $x^n$  in  $O(1)$  time, if  $x = e^{i\theta}$ .

**Definition 1** (Orthogonal Matrix).

$$A = (\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n), \|\vec{a}_i\| = 1$$

$$\vec{a}_i^T \vec{a}_j \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{otherwise} \end{cases}$$

*Proof.* The inverse of an orthogonal matrix  $A$  is  $A^T$ .

$$\begin{aligned} \because A &= (\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n) \therefore A^T = \begin{pmatrix} \vec{a}_1^T \\ \vec{a}_2^T \\ \vdots \\ \vec{a}_n^T \end{pmatrix} \\ \Rightarrow A^T A &= \begin{pmatrix} \vec{a}_1^T \vec{a}_1(1) & \vec{a}_1^T \vec{a}_2(0) & \dots & \vec{a}_1^T \vec{a}_n(0) \\ \vec{a}_2^T \vec{a}_1(0) & \ddots & & \vec{a}_1^T \vec{a}_n(0) \\ \vdots & & \ddots & \vec{a}_1^T \vec{a}_n(0) \\ \vec{a}_n^T \vec{a}_1(0) & \vec{a}_n^T \vec{a}_2(0) & \dots & \vec{a}_1^T \vec{a}_n(1) \end{pmatrix} \\ \because A^T A &= I \therefore A^{-1} = A^T \end{aligned}$$

□