

EE451: SRE Presentation

Fourier Transform of PsuedoRandom Sequences

Badal Varshney
19D070015

Guided by- Prof. Virendra R. Sule
Department of Electrical Engineering
IIT Bombay

Nov. 27, 2022

Introduction

Discrete Fourier Transform (DFT):

- DFT translates the finite length time-domain discrete sequence into its frequency domain counterpart having the same length (or greater) so that no spectral information is lost and perfect reconstruction is possible
- Suppose $x(n)$ is a sequence of finite length 'N' then its DFT is

$$y(k) = \sum_{n=0}^{N-1} x(n) \cdot e^{-i2nk\pi/N} \quad k \in \{0, 1, 2, \dots, N-1\} \quad (1)$$

- The time complexity of calculating DFT results to $O(N^2)$.
- Further, Introduction of a FFT by Cooley-Tukey, Radix-2 algorithm was developed with time complexity $O(N \log N)$.
- Our aim is to further reduce the time complexity for large sequences.

- Minimal Polynomial
- Henkel Matrix of Sequences
- Primitive Nth root of unity
- Fourier Transform matrix (T)
- Formation of Ω -Matrix
- DFT of sequences (F)
- Future Work

Minimal Polynomial

A sequence $S = (s_0, s_1, s_2, \dots, s_m, \dots, s_{M-1})$ with terms in a finite field $GF(q)$ with q elements is called a linear recurring sequence over $GF(q)$ with minimal polynomial

$$X^m = a_0 + a_1X + \dots + a_{m-1}X^{m-1} \in GF(q)[X] \quad (2)$$

If,

$$a_0s_i + a_1s_{i+1} + \dots + a_ms_{i+m} = 0 \text{ for any } i \geq 1. \quad (3)$$

Take the coefficient of minimal as column vector ' α '-

$$\alpha = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{m-1} \end{pmatrix} \quad (4)$$

Henkel Matrix of Sequences

- Henkel Matrix of sequence $(s_0, s_1, s_2, \dots, s_{M-1})$ as defined as-

$$H(m) = \begin{pmatrix} s_0 & s_1 & \cdots & s_{m-1} \\ s_1 & s_2 & \cdots & s_m \\ \vdots & \vdots & \ddots & \vdots \\ s_{m-1} & s_{m-2} & \cdots & s_{2m-1} \end{pmatrix} \quad (5)$$

- Column vector $h(m+1)$ is defined as

$$h(m+1) = H(m) * \alpha \quad (6)$$

- The shifted sequence vector of our original sequence looks like-

$$h(m+1) = \begin{pmatrix} s_m \\ s_{m+1} \\ \vdots \\ s_{2m-1} \end{pmatrix} \quad (7)$$

- come up with results $m = \frac{M}{2}$
- $H(m)$ have to satisfy this relation-

$$\text{Rank}H(m) = \text{Rank}H(m+j), \quad j = \{1, 2, \dots\} \quad (8)$$

- 'm' is the degree of maximum rank minimum polynomial.

Primitive Nth root of unity

Let n be a positive integer. A primitive n^{th} root of unity is an n^{th} root of unity that is not a k^{th} root of unity for any positive $k < n$. That is, ω is a primitive n^{th} root of unity if and only if

$$\omega^n = 1, \text{ and } \omega^k \neq 1 \text{ for any positive integer } k < n. \quad (9)$$

- Here, Take the primitive n^{th} root of unity as follows-

$$\begin{aligned} \omega^n &= 1 = e^{j2\pi} = \cos(2\pi) + j\sin(2\pi) \\ \implies \omega &= e^{\frac{j2\pi}{n}} = \cos\left(\frac{2\pi}{n}\right) + j\sin\left(\frac{2\pi}{n}\right) \end{aligned} \quad (10)$$

Fourier Transform matrix (T)

- Fourier Transform matrix (T) of DFT of length m is defined as

$$T = \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \dots & \omega^{n-1} \\ 1 & \omega^2 & \omega^3 & \dots & \omega^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{n-1} & \omega^{2(n-1)} & \dots & \omega^{(n-1)^2} \end{pmatrix} \quad (11)$$

- 'w' is primitive Nth root of unity Primitive nth root of unity.
- Convert the binary sequence to real sequence of period n

$$c = (-1)^a, \text{ where } a \text{ is sequence element} \quad (12)$$

- $c = [(-1)^{s_0}, (-1)^{s_1}, \dots, (-1)^{s_{n-1}}]$

Formation of Ω – Matrix

- It is defined as identity matrix with first column embedded with zero vector and last row is embedded with minimal polynomial coefficient

$$\Omega = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ a_0 & a_1 & a_2 & \cdots & a_{m-1} \end{pmatrix} \quad (13)$$

- $a_0, a_1, a_2, \dots, a_{m-1}$ are the coefficients of minimal polynomial
- Ω matrix is used to get the shifted subsequences which satisfy these relation-

$$x(k+1) = \Omega * x(k), \text{ where } k = \{0, 1, 2, \dots\} \quad (14)$$

and,

$$x(k) = \Omega^k * x(0) \quad (15)$$

DFT of sequences (F)

- Here, we find the DFT of sub-sequences 'c', which is defined as

$$\begin{aligned} F &= T * c^t \\ \Rightarrow F &= \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \dots & \omega^{n-1} \\ 1 & \omega^2 & \omega^3 & \dots & \omega^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{n-1} & \omega^{2(n-1)} & \dots & \omega^{(n-1)^2} \end{pmatrix} \begin{pmatrix} (-1)^{s_0} \\ (-1)^{s_1} \\ (-1)^{s_2} \\ \vdots \\ (-1)^{s_{n-1}} \end{pmatrix} \end{aligned} \quad (16)$$

- we defined the matrix $X(f)$ as

$$X(f) = T * x(0) \quad (17)$$

DFT of sequences Contd...

- from the above relation, we get-

$$x(0) = T^{-1} * x(f) \quad (18)$$

- Combining the equation (15) and (18), we get

$$x(k) = \Omega^k * T^{-1} * x(f) \quad (19)$$

- DFT of $x(k)$ as defined as

$$\text{DFT of } x(k) = T * x(k)$$

$$\implies T * x(k) = T * (\Omega^k * T^{-1} * x(f))$$







$$\implies T * x(k) = (T * \Omega^k * T^{-1}) * x(f) \quad (20)$$

- Now, we come up with DFT of all sub-sequences.






- Find the way to merge all the DFT of subsequences to determine the DFT of large sequences to reduce the complexity of DFT
- Verify the property of this DFT techniques like linearity, time-shifting, frequency-shifting, convolution etc.

Thank You

References

-  Prof. R. Fateman, "The (finite field) Fast Fourier Transform", <https://people.eecs.berkeley.edu/fateman/282/readings/fftnotes.pdf>
-  Zhi-Han Gaoa, Fang-Wei Fu, "The minimal polynomial of a sequence obtained from the componentwise linear transformation of a linear recurring sequence", Theor Comput Sci, 2010, vol 411: 3883–3893
-  James L Massey and Shirlei Scrconek, "Linear Complexity of Periodic Sequences: A General Theory", N. Koblitz (Ed.): Advances in Cryptology - CRYPTO '96, LNCS 1109, pp. 358-371, 1996
-  Graham H. Norton, "Minimal Polynomial Algorithms for Finite Sequences", arXiv:0911.0130v3 [cs.IT]
-  Graham H. Norton. "The Berlekamp-Massey Algorithm via Minimal Polynomials", <http://arXiv.org>, 1001.1597.
-  Howard W. Levinson and Vadim A. Markel, "Binary Discrete Fourier Transform and its Inversion", arXiv:2011.10130v2 [math.NA]

References

-  lanqin Zhou, "A fast algorithm for determining the linear complexity of periodic sequences", arXiv:cs/0512040v1 [cs.CR]
-  Blackburn, S.R., A generalization of the discrete Fourier transform: determining the minimal polynomial of a periodic sequence[J]. IEEE Trans on Information Theory, 1994, 40(5): 1702-1704.
-  R. Jha, R. Prasad, R. Khemka and A. Mandpura, "A Faster DFT Algorithm for Specific Binary Pulse Sequences," 2020 International Conference on Communication and Signal Processing (ICCSP), 2020, pp. 1459-1461, doi: 10.1109/ICCSP48568.2020.9182054.
-  S. V. Fedorenko, "A Method for Computation of the Discrete Fourier Transform over a Finite Field", Probl Inf Transm 42, 139–151 (2006). <https://doi.org/10.1134/S0032946006020074>
-  J.W. Cooley and J. Tukey, "An Algorithm for the Machine Calculation of Complex Fourier Series" Math Comp. 19 29–301 (1965)