

# Lecture - 15

## 1.7 Computation of DFT by Linear Filtering Approach :

We have studied that Radix-2 FFT algorithms are used to compute 'N'-point DFT. In this case,  $\frac{N}{2} \log_2 N$  complex multiplications and  $N \log_2 N$  complex additions are required to compute N-point DFT.

But there are certain applications in which selected number of DFT values are required; but not the entire DFT. In such cases the direct computation of DFT is more efficient than FFT algorithms. The direct computation of DFT for selected values can be realized using linear filtering approach.

This type of computation can be performed using the algorithm, called as Goertzel algorithm.

### 1.7.1 Goertzel Algorithm :

In this algorithm the periodicity property of twiddle factor ( $W_N^k$ ) is used.

Now recall the definition of DFT,

$$X(k) = \sum_{m=0}^{N-1} x(m) W_N^{km} \quad \dots(1)$$

Note that here we have used notation  $x(m)$  for input sequence.

Now the twiddle factor is given by,

$$W_N = e^{-\frac{j2\pi}{N}}$$
$$\therefore W_N^{-kN} = \left( e^{-\frac{j2\pi}{N}} \right)^{-kN} = e^{+j2\pi k} = \cos 2\pi k + j \sin 2\pi k$$
$$= 1 + j 0 = 1$$

Since  $W_N^{-kN} = 1$  we can multiply Equation (1) by  $W_N^{-kN}$

$$\therefore X(k) = \sum_{m=0}^{N-1} x(m) W_N^{km} \cdot W_N^{-kN}$$
$$\therefore X(k) = \sum_{m=0}^{N-1} x(m) W_N^{-k(N-m)} \quad \dots(2)$$

Now according to the equation of linear convolution,

$$y_k(n) = \sum_{m=-\infty}^{\infty} x(m) h_k(n-m) \quad \dots(3)$$

We know that  $h_k(n)$  is an impulse response. Now let,

$$h_k(n) = \sum_N^{-kn} u(n) \quad \dots(4)$$

Replacing  $n$  by ' $n-m$ ' we get,

$$h_k(n-m) = \sum_N^{-k(n-m)} u(n-m) \quad \dots(5)$$

Put this value in Equation (3). Here  $u(n-m)$  is unit step. So because of multiplication by  $u(n-m)$  the limits of summation will become  $m=0$  to  $m=\infty$ .

$$\therefore y_k(n) = \sum_{m=0}^{N-1} x(m) W_N^{-k(n-m)} \quad \dots(6)$$

Now we will calculate  $y_k(n)$  at  $n=N$ .

$$\therefore y_k(n) \Big|_{n=N} = \sum_{m=0}^{N-1} x(m) W_N^{-k(N-m)} \quad \dots(7)$$

Equations (2) and (7) are identical. Thus we can conclude that DFT,  $X(k)$  can be obtained as output of LTI system at  $n=N$ . ( $y_k(n)$ ). But in this case, we will get  $X(k)$  only at one value of  $K$ . So we need to use parallel systems to obtain  $X(k)$  at different values of  $K$ .

**Note :** Goertzel Algorithm is efficient when  $X(k)$  is to be computed at points less than  $\log_2^N$ .

### 1.7.2 Computation of DFT of Two Real Sequences using Only One FFT Flow Graph :

Suppose  $x_1(n)$  and  $x_2(n)$  are two real valued sequences of length  $N$ . We will define a complex valued sequence  $x(n)$  as,

$$x(n) = x_1(n) + j x_2(n), \quad 0 \leq n \leq N-1 \quad \dots(1)$$

Since DFT operation is linear, we can write DFT of Equation (1) as,

$$X(k) = X_1(k) + j X_2(k) \quad \dots(2)$$

In Equation (1),  $x_1(n)$  represents real part of  $x(n)$  and  $x_2(n)$  represents imaginary part of  $x(n)$ . Thus we can write,

$$x_1(n) = \frac{x(n) + x^*(n)}{2} \quad \dots(3)$$

$$\text{and } x_2(n) = \frac{x(n) - x^*(n)}{2j} \quad \dots(4)$$

The DFTs of Equations (3) and (4) can be written as,

$$X_1(k) = \frac{1}{2} \{ \text{DFT}[x(n)] + \text{DFT}[x^*(n)] \} \quad \dots(5)$$

$$\text{and } X_2(k) = \frac{1}{2j} \{ \text{DFT}[x(n)] - \text{DFT}[x^*(n)] \} \quad \dots(6)$$

According to complex conjugate property we have,

$$\begin{array}{c} \text{DFT} \\ x^*(n) \longleftrightarrow X^*((-k))_N = X^*(N-k) \\ \text{N} \end{array}$$

$$\therefore X_1(k) = \frac{1}{2} [X(k) + X^*(N-k)] \quad \dots(7)$$

$$\text{and } X_2(k) = \frac{1}{2j} [X(k) - X^*(N-k)] \quad \dots(8)$$

Thus by performing single DFT on complex valued sequence  $x(n)$  we can obtain DFT of two real sequences  $x_1(n)$  and  $x_2(n)$ .

**Prob. 1 :** Find DFT of two real sequences using only one FFT flow graph.

$$x_1(n) = \{1, 1, 1, 1\}$$

$$x_2(n) = \{2, 1, 2, 1\}$$

Perform FFT only once.

**Soln. :**

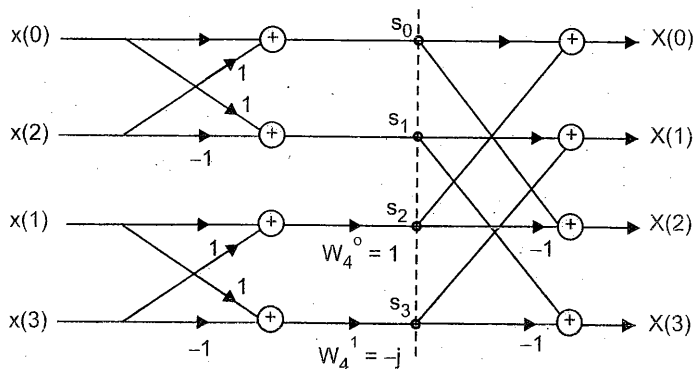
Consider a sequence  $x(n)$  such that,

$$x(n) = x_1(n) + jx_2(n)$$

$$\therefore x(n) = \{1 + j2, 1 + j1, 1 + j2, 1 + j1\}$$

Here  $x(0) = 1 + j2, x(1) = 1 + j1, x(2) = 1 + j2, x(3) = 1 + j1$ .

We will obtain DFT using DITFFT, the flow graph is as shown in Fig. G-23(a).



**Fig. G-23(a)**

$$s_0 = x(0) + x(2) = 1 + j2 + 1 + j2 = 2 + j4$$

$$s_1 = x(0) - x(2) = 1 + j2 - 1 - j2 = 0$$

$$s_2 = [x(1) + x(3)] W_4^0 = [1 + j1 + 1 + j1] \cdot 1 = 2 + j2$$

$$s_3 = [x(1) - x(3)] W_4^1 = [1 + j1 - 1 - j1] (-j) = 0$$

The final output is,

$$X(0) = s_0 + s_2 = 2 + j4 + 2 + j2 = 4 + j6$$

$$X(1) = s_1 + s_3 = 0$$

$$X(2) = s_0 - s_2 = 2 + j4 - 2 - j2 = j2$$

$$X(3) = s_1 - s_3 = 0$$

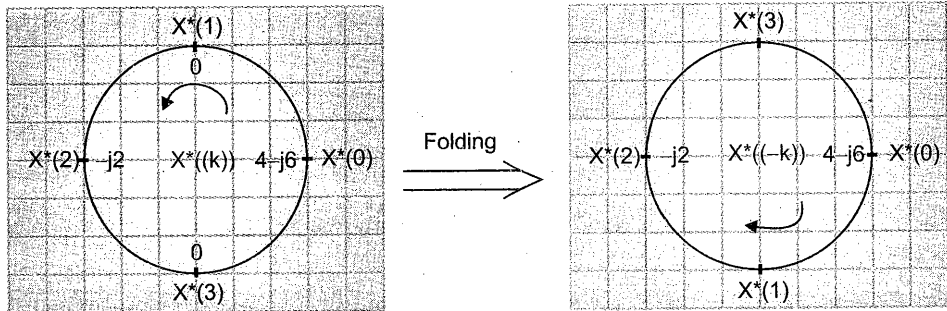
$$\therefore X(k) = \{4 + j6, 0, j2, 0\} \quad \dots(1)$$

**Calculation of  $X_1(k)$  :**

Now we have, 
$$X_1(k) = \frac{1}{2} [X(k) + X^*(N-k)] \quad \dots(2)$$

Here  $N = 4$  and  $X^*(k) = \{4 - j6, 0, -j2, 0\} \quad \dots(3)$

In Equation (2),  $X^*(N-k) = X^*((-k))$  which represents circular folding of  $X^*(k)$ . It is shown in Fig. G-23(b).



**Fig. G-23(b)**

$$\therefore X^*(-k) = \{4 - j6, 0, -j2, 0\} \quad \dots(4)$$

Using Equation (2) we can obtain DFT  $X_1(k)$  as follows :

We have, 
$$X_1(k) = \frac{1}{2} [X(k) + X^*((-k))] \quad \dots(4)$$

$$\text{For } k = 0 \Rightarrow X_1(0) = \frac{1}{2}[X(0) + X^*(0)] = \frac{1}{2}[4 + j6 + 4 - j6] = 4$$

$$\text{For } k = 1 \Rightarrow X_1(1) = \frac{1}{2}[X(1) + X^*(1)] = \frac{1}{2}[0 + 0] = 0$$

$$\text{For } k = 2 \Rightarrow X_1(2) = \frac{1}{2}[X(2) + X^*((-2))] = \frac{1}{2}[+j2 - j2] = 0$$

$$\text{For } k = 3 \Rightarrow X_1(3) = \frac{1}{2}[X(3) + X^*((-3))] = \frac{1}{2}[0 + 0] = 0$$

$$\therefore X_1(k) = \{4, 0, 0, 0\}$$

**Calculation of  $X_2(k)$  :**

We have,

$$X_2(k) = \frac{1}{2j}[X(k) - X^*(N-k)] = \frac{1}{2j}[X(k) - X^*((-k))]$$

$$\text{For } k = 0 \Rightarrow X_2(0) = \frac{1}{2j}[X(0) - X^*(0)] = \frac{1}{2j}[4 + j6 - 4 + j6] = 6$$

$$\text{For } k = 1 \Rightarrow X_2(1) = \frac{1}{2j}[X(1) - X^*((-1))] = \frac{1}{2j}[0 + 0] = 0$$

$$\text{For } k = 2 \Rightarrow X_2(2) = \frac{1}{2j}[X(2) - X^*((-2))] = \frac{1}{2j}[j2 + j2] = 2$$

$$\text{For } k = 3 \Rightarrow X_2(3) = \frac{1}{2j}[X(3) - X^*((-3))] = \frac{1}{2j}[0 - 0] = 0$$

$$\therefore X_2(k) = \{6, 0, 2, 0\}$$

## 1.9 Chirp-Z Algorithm :

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The chirp-Z algorithm is used to compute  $X(Z)$  on the other contours; in the  $Z$ -plane. It may include the unit circle.

If we want to calculate  $X(Z)$  at a set of points  $[Z_k]$  then the equation of  $Z$ -transform can be written as,

$$X[Z_k] = \sum_{n=0}^{N-1} x(n) Z_k^{-n} \quad \dots(1)$$

Here let  $k = 0, 1, \dots, L-1$

and  $n = 0, 1, \dots, N-1$

Now  $Z$  can be represented in the polar form with radius of contour 'r' and at  $N$  equally spaced points then

$$Z_k = r e^{j \frac{2\pi kn}{N}}, \quad k = 0, 1, \dots, N-1 \quad \dots(2)$$

Putting this value in Equation (1) we get,

$$X[Z_k] = \sum_{n=0}^{N-1} [x(n) r^{-n}] e^{-j 2\pi kn / N} \quad \dots(3)$$

Now let us say that  $Z_k$  fall on an arc which starts at some point  $Z_0 = r_0 e^{j\theta_0}$  and spirals either in; towards the origin or out, away from the origin. Then  $Z_k$  can be defined as,

$$Z_k = r_0 e^{j\theta_0} (R_0 e^{j\phi_0})^k, \quad k = 0, 1, \dots, L-1 \quad \dots(4)$$

Here if  $R_0 < 1$ ; the points fall on a contour that spirals towards the origin and if  $R_0 > 1$  then points fall on a contour that spirals away from the origin. If  $R = 1$  then the contour is a circular

arc of radius  $r_0$ . And when  $r_0 = R_0 = 1$  then the contour is an arc of unit circle.

Now if  $r_0 = R_0 = 1$ ,  $\theta = 0$ ,  $\phi = \frac{2\pi}{N}$  and  $L = N$  then the contour is the entire unit circle and frequencies are those of the DFT.

Putting Equation (4) in Equation (1) we get,

$$X[Z^k] = \sum_{n=0}^{N-1} x(n) [r_0 e^{j\theta_0}]^{-n} \cdot [R_0 e^{j\phi_0}]^{-nk}$$

$$\text{Let } V = R_0 e^{j\phi_0}$$

$$\therefore X[Z_k] = \sum_{n=0}^{N-1} x(n) [r_0 e^{j\theta_0}]^{-n} \cdot V^{-nk} \quad \dots(5)$$

Now the term  $nk$  can be expressed as,

$$nk = \frac{1}{2} [n^2 + k^2 - (k-n)^2] \quad \dots(6)$$

Putting this value in Equation (5) we get,

$$X[Z_k] = V^{-k^2/2} \sum_{n=0}^{N-1} [x(n) [r_0 e^{j\theta_0}]^{-n} \cdot V^{-n^2/2} \cdot V^{(k-n)^2/2}] \quad \dots(7)$$

Let us define,

$$g(n) = x(n) [r_0 e^{j\theta_0}]^{-n} V^{-n^2/2}$$

$$\therefore X[Z_k] = V^{-k^2/2} \sum_{n=0}^{N-1} g(n) V^{(k-n)^2/2} \quad \dots(8)$$

Let  $g(n) =$  Input sequence

and impulse response  $h(n) = V^{n^2/2} \quad \dots(9)$

$$\therefore h(k-n) = V^{(k-n)^2/2} \quad \dots(10)$$

Thus Equation (8) becomes,

$$X[Z_k] = V^{-k^2/2} \sum_{n=0}^{N-1} g(n) h(k-n) \quad \dots(11)$$

Here the summation term represents the equation of linear convolution. That means it represents the output of digital filter  $y(k)$ .



$$\therefore X[Z_k] = V^{-k^2/2} \cdot y(k) \quad \dots(12)$$

Now we have  $V^{n^2/2} = h(n)$  (from Equation (9))

$$\therefore V^{-k^2/2} = [h(k)]^{-1} = \frac{1}{h(k)}$$

Thus Equation (12) becomes,

$$X[Z_k] = \frac{y(k)}{h(k)}$$

Thus the DFT at discrete point  $Z_k$  is the ratio of  $\frac{y(k)}{h(k)}$ . Now the Z-transform evaluated as in

Equation (8) is called as, **Chirp-Z transform**.