

- ◆ The Fourier transform of an analogue signal  $x(t)$  is given by:

$$X(\omega) = \int_{-\infty}^{+\infty} x(t) e^{-j\omega t} dt$$

- ◆ The Discrete Fourier Transform (DFT) of a discrete-time signal  $x(nT)$  is given by:

$$X(k) = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N}nk}$$

Where:

$$k = 0, 1, \dots, N-1$$

$$x(nT) = x[n]$$

$$X(k) = \sum_{n=0}^{N-1} x[n] W_N^{nk}$$

$x[n]$  = input

$X[k]$  = frequency bins

$W$  = twiddle factors

$$X(0) = x[0]W_N^0 + x[1]W_N^{0*1} + \dots + x[N-1]W_N^{0*(N-1)}$$

$$X(1) = x[0]W_N^0 + x[1]W_N^{1*1} + \dots + x[N-1]W_N^{1*(N-1)}$$

:

$$X(k) = x[0]W_N^0 + x[1]W_N^{k*1} + \dots + x[N-1]W_N^{k*(N-1)}$$

:

$$X(N-1) = x[0]W_N^0 + x[1]W_N^{(N-1)*1} + \dots + x[N-1]W_N^{(N-1)(N-1)}$$

Note: For N samples of x we have N frequencies representing the signal.

### Performance of the DFT Algorithm

- ◆ The DFT requires  $N^2$  ( $N \times N$ ) complex multiplications:
  - ◆ Each  $X(k)$  requires N complex multiplications.
  - ◆ Therefore to evaluate all the values of the DFT (  $X(0)$  to  $X(N-1)$  )  $N^2$  multiplications are required.
- ◆ The DFT also requires  $(N-1)*N$  complex additions:
  - ◆ Each  $X(k)$  requires N-1 additions.
  - ◆ Therefore to evaluate all the values of the DFT  $(N-1)*N$  additions are required.

## The Discrete Fourier Transform (DFT)

- DFT of an  $N$ -point sequence  $x_n$ ,  $n = 0, 1, 2, \dots, N - 1$  is defined as

$$X_k = \sum_{n=0}^{N-1} x_n e^{-j \frac{2\pi k}{N} n} \quad k = 0, 1, 2, \dots, N - 1$$

- An  $N$ -point sequence yields an  $N$ -point transform
- $X_k$  can be expressed as an *inner product*:

$$X_k = \begin{bmatrix} 1 & e^{-j \frac{2\pi k}{N}} & e^{-j \frac{2\pi k}{N} 2} & \dots & e^{-j \frac{2\pi k}{N} (N-1)} \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_{N-1} \end{bmatrix}$$

## The Discrete Fourier Transform (DFT)

- Notation:  $W_N = e^{-j\frac{2\pi}{N}}$ . Hence,

$$X_k = \begin{bmatrix} 1 & W_N^k & W_N^{2k} & \dots & W_N^{(N-1)k} \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_{N-1} \end{bmatrix}$$

- By varying  $k$  from 0 to  $N - 1$  and combining the  $N$  inner products, we get the following:

$$\mathbf{X} = \mathbf{W}\mathbf{x}$$

- $\mathbf{W}$  is an  $N \times N$  matrix, called as the “DFT Matrix”

## The DFT Matrix

$$\mathbf{W} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N & W_N^2 & \dots & W_N^{N-1} \\ 1 & W_N^2 & W_N^4 & \dots & W_N^{2(N-1)} \\ & & & \ddots & \\ 1 & W_N^{N-1} & W_N^{2(N-1)} & \dots & W_N^{(N-1)(N-1)} \end{bmatrix}_{N \times N}$$

- The notation  $\mathbf{W}_N$  is used if we want to make the size of the DFT matrix explicit

## How Many Complex Multiplications Are Required?

- Each inner product requires  $N$  complex multiplications
  - There are  $N$  inner products
- Hence we require  $N^2$  multiplications
- However, the first row and first column are all 1s, and *should not be counted as multiplications*
  - There are  $2N - 1$  such instances
- Hence, the number of complex multiplications is  $N^2 - 2N + 1$ , i.e.,  $(N - 1)^2$

## How Many Complex Additions Are Required?

- Each inner product requires  $N - 1$  complex additions
  - There are  $N$  inner products
- Hence we require  $N(N - 1)$  complex additions

## Total Operation Count

- No. of complex multiplications:  $(N - 1)^2$
- No. of complex additions:  $N(N - 1)$
- The operation count for multiplications and additions assumes that  $W_N^k$  has been computed offline and is available in memory
  - If pre-computed values of  $W_N^k$  are not available, then the operation count will increase
- We will assume that all the required  $W_N^k$  have been pre-computed and are available

## Operation Count Makes DFT Impractical

- For large  $N$ ,  
 $(N - 1)^2 \approx N^2$   
 $N(N - 1) \approx N^2$
- Hence both multiplications and additions are  $O(N^2)$
- If  $N = 10^3$ , then  $O(N^2) = 10^6$ , i.e., a million!
- This makes the straightforward method **slow** and **impractical** even for a moderately long sequence

## The Decimation in Time (DIT) Algorithm

- From  $\{x_n\}$  form two sequences as follows:

$$\{g_n\} = \{x_{2n}\} \quad \{h_n\} = \{x_{2n+1}\}$$

- $\{g_n\}$  contains the **even**-indexed samples, while  $\{h_n\}$  contains the **odd**-indexed samples
- The DFT of  $\{x_n\}$  is

$$\begin{aligned} X_k &= \sum_{n=0}^{N-1} x_n W_N^{nk} \\ &= \sum_{r=0}^{\frac{N}{2}-1} x_{2r} W_N^{(2r)k} + \sum_{r=0}^{\frac{N}{2}-1} x_{2r+1} W_N^{(2r+1)k} \\ &= \sum_{r=0}^{\frac{N}{2}-1} g_r W_N^{(2r)k} + W_N^k \sum_{r=0}^{\frac{N}{2}-1} h_r W_N^{(2r)k} \end{aligned}$$

## The Decimation in Time (DIT) Algorithm

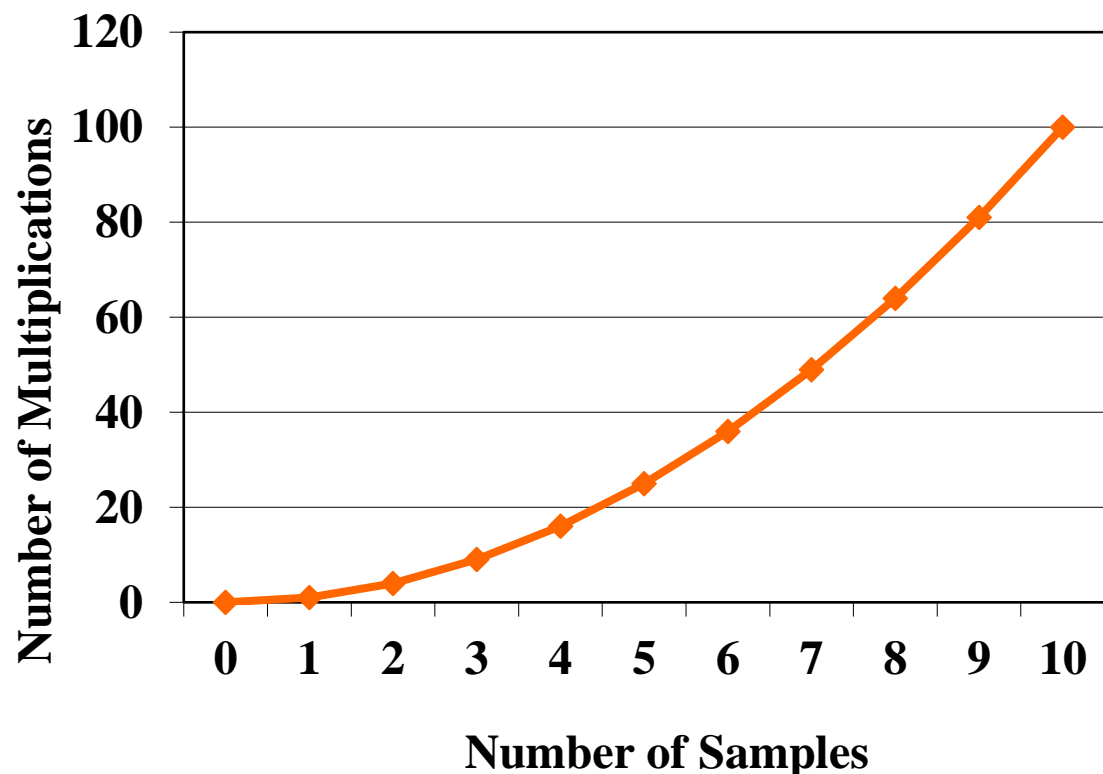
- But,

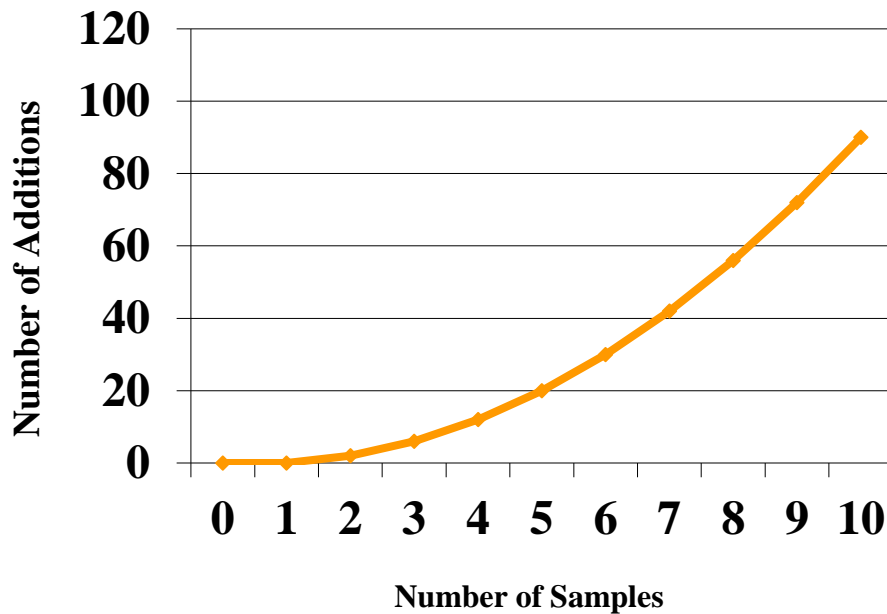
$$W_N^{2rk} = e^{-j\frac{2\pi}{N}(2rk)} = e^{-j\frac{2\pi}{N/2}(rk)} = W_{N/2}^{rk}$$

and hence

$$\begin{aligned} X_k &= \sum_{r=0}^{\frac{N}{2}-1} g_r W_{N/2}^{rk} + W_N^k \sum_{r=0}^{\frac{N}{2}-1} h_r W_{N/2}^{rk} \\ &= G_k + W_N^k H_k \quad k = 0, 1, \dots, N-1 \end{aligned}$$

- $\{G_k\}$  and  $\{H_k\}$  are  $\frac{N}{2}$  point DFTs
- The **overhead** for combining the two  $\frac{N}{2}$  point DFTs is the multiplicative factor  $W_N^k$  for  $k = 0, 1, \dots, N-1$ 
  - $W_N^k$  is called “**twiddle factor**”





- ◆ Can the number of computations required be reduced?

### DFT → FFT

- ◆ A large amount of work has been devoted to reducing the computation time of a DFT.
- ◆ This has led to efficient algorithms which are known as the Fast Fourier Transform (FFT) algorithms.

$$X(k) = \sum_{n=0}^{N-1} x[n] W_N^{nk}; \quad 0 \leq k \leq N-1 \quad (1)$$

$$x[n] = x[0], x[1], \dots, x[N-1]$$

- ◆ Lets divide the sequence  $x[n]$  into even and odd sequences:

- ◆  $x[2n] = x[0], x[2], \dots, x[N-2]$

- ◆  $x[2n+1] = x[1], x[3], \dots, x[N-1]$



◆ Equation 1 can be rewritten as:

$$X(k) = \sum_{n=0}^{\frac{N}{2}-1} x[2n]W_N^{2nk} + \sum_{n=0}^{\frac{N}{2}-1} x[2n+1]W_N^{(2n+1)k} \quad (2)$$

◆ Since:

$$\begin{aligned} W_N^{2nk} &= e^{-j\frac{2\pi}{N}2nk} = e^{-j\frac{2\pi}{N/2}nk} \\ &= W_{\frac{N}{2}}^{nk} \end{aligned}$$

$$W_N^{(2n+1)k} = W_N^k \cdot W_{\frac{N}{2}}^{nk}$$

◆ Then:

$$\begin{aligned} X(k) &= \sum_{n=0}^{\frac{N}{2}-1} x[2n]W_{\frac{N}{2}}^{nk} + W_N^k \sum_{n=0}^{\frac{N}{2}-1} x[2n+1]W_{\frac{N}{2}}^{nk} \\ &= Y(k) + W_N^k Z(k) \end{aligned}$$

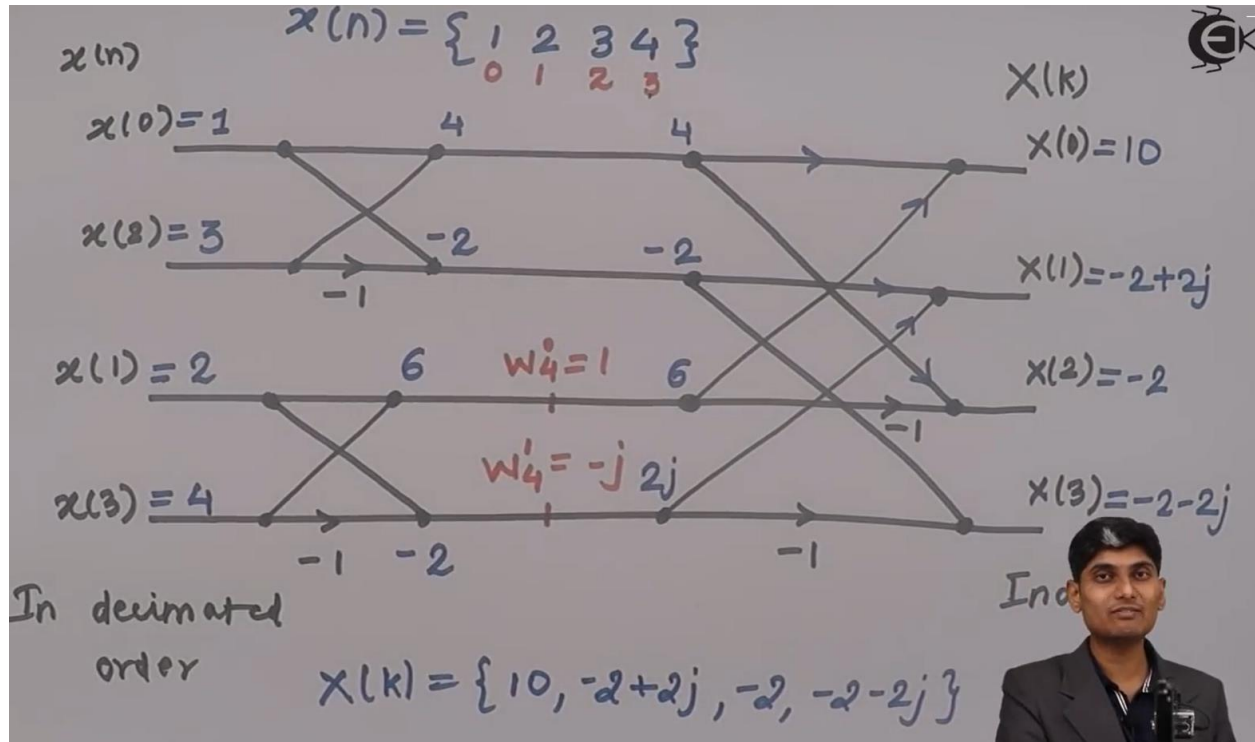
◆ The result is that an N-point DFT can be divided into two N/2 point DFT's:

$$X(k) = \sum_{n=0}^{N-1} x[n]W_N^{nk}; \quad 0 \leq k \leq N-1 \quad \text{N-point DFT}$$

◆ Where Y(k) and Z(k) are the two N/2 point DFTs operating on even and odd samples respectively:

$$\begin{aligned} X(k) &= \sum_{n=0}^{\frac{N}{2}-1} x_1[n]W_{\frac{N}{2}}^{nk} + W_N^k \sum_{n=0}^{\frac{N}{2}-1} x_2[n]W_{\frac{N}{2}}^{nk} \\ &= Y(k) + W_N^k Z(k) \end{aligned} \quad \text{Two N/2-point DFTs}$$

#### 4 point DIT-FFT



**Example 1:** Consider a sequence  $x[n] = \{1, 1, -1, -1, -1, 1, 1, -1\}$

Determine DFT  $X[k]$  of  $x[n]$  using the **decimation-in-time FFT** algorithm.

$$f[n] = x[2n] = \{x[0], x[2], x[4], x[6]\} = \{1, -1, -1, 1\}$$

$$g[n] = x[2n+1] = \{x[1], x[3], x[5], x[7]\} = \{1, -1, 1, -1\}$$

$$X = W_N X$$

$$\begin{bmatrix} F(0) \\ F(1) \\ F(2) \\ F(3) \end{bmatrix} = \begin{bmatrix} w_4^0 & w_4^0 & w_4^0 & w_4^0 \\ w_4^0 & w_4^1 & w_4^2 & w_4^3 \\ w_4^0 & w_4^2 & w_4^4 & w_4^6 \\ w_4^0 & w_4^3 & w_4^6 & w_4^9 \end{bmatrix} \begin{bmatrix} f(0) \\ f(1) \\ f(2) \\ f(3) \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 2 + j2 \\ 0 \\ 2 - j2 \end{bmatrix}$$

$$\begin{bmatrix} G(0) \\ G(1) \\ G(2) \\ G(3) \end{bmatrix} = \begin{bmatrix} W_4^0 & W_4^0 & W_4^0 & W_4^0 \\ W_4^0 & W_4^1 & W_4^2 & W_4^3 \\ W_4^0 & W_4^2 & W_4^4 & W_4^6 \\ W_4^0 & W_4^3 & W_4^6 & W_4^9 \end{bmatrix} \begin{bmatrix} g(0) \\ g(1) \\ g(2) \\ g(3) \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 0 \\ 4 \\ 0 \end{bmatrix}$$

$$X[0] = F[0] + W_8^0 G[0] = 0$$

$$X[4] = F[0] - W_8^0 G[0] = 0$$

$$X[1] = F[1] + W_8^1 G[1] = 2 + j2$$

$$X[5] = F[1] - W_8^1 G[1] = 2 + j2$$

$$X[2] = F[2] + W_8^2 G[2] = -j4$$

$$X[6] = F[2] - W_8^2 G[2] = j4$$

$$X[3] = F[3] + W_8^3 G[3] = 2 - j2$$

$$X[7] = F[3] - W_8^3 G[3] = 2 - j2$$

**Example 1:** Consider a sequence  $x[n] = \{1, 1, -1, -1, -1, 1, 1, -1\}$

Determine DFT  $X[k]$  of  $x[n]$  using the **decimation-in-frequency FFT algorithm**.

$$p[n] = x[n] + x\left[n + \frac{N}{2}\right]$$

$$= \{(1 - 1), (1 + 1), (-1 + 1), (-1 - 1)\} = \{0, 2, 0, 2\}$$

$$\begin{aligned}
q[n] &= \left( x[n] - x\left[n + \frac{N}{2}\right] \right) W_8^n \\
&= \{ (1+1)W_8^0, (1-1)W_8^1, (-1-1)W_8^2, (-1+1)W_8^3 \} \\
&= \{2, 0, j2, 0\}
\end{aligned}$$

$$X = Wx$$

$$\begin{bmatrix} P(0) \\ P(1) \\ P(2) \\ P(3) \end{bmatrix} = \begin{bmatrix} w_4^0 & w_4^0 & w_4^0 & w_4^0 \\ w_4^0 & w_4^1 & w_4^2 & w_4^3 \\ w_4^0 & w_4^2 & w_4^4 & w_4^6 \\ w_4^0 & w_4^3 & w_4^6 & w_4^9 \end{bmatrix} \begin{bmatrix} p(0) \\ p(1) \\ p(2) \\ p(3) \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} 0 \\ 2 \\ 0 \\ -2 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ -j4 \\ 0 \\ j4 \end{bmatrix}$$

$$\begin{bmatrix} Q(0) \\ Q(1) \\ Q(2) \\ Q(3) \end{bmatrix} = \begin{bmatrix} w_4^0 & w_4^0 & w_4^0 & w_4^0 \\ w_4^0 & w_4^1 & w_4^2 & w_4^3 \\ w_4^0 & w_4^2 & w_4^4 & w_4^6 \\ w_4^0 & w_4^3 & w_4^6 & w_4^9 \end{bmatrix} \begin{bmatrix} q(0) \\ q(1) \\ q(2) \\ q(3) \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} 2 \\ 0 \\ j2 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 2+j2 \\ 2-j2 \\ 2+j2 \\ 2-j2 \end{bmatrix}$$

$$X[0] = P[0] = 0$$

$$X[4] = P[2] = 0$$

$$X[1] = Q[0] = 2 + j2$$

$$X[5] = Q[2] = 2 + j2$$

$$X[2] = P[1] + W_8^2 G[2] = -j4$$

$$X[6] = P[3] = j4$$

$$X[3] = Q[1] + W_8^3 G[3] = 2 - j2$$

$$X[7] = Q[3] = 2 - j2$$