

## 0.1 Discrete Fourier Transform

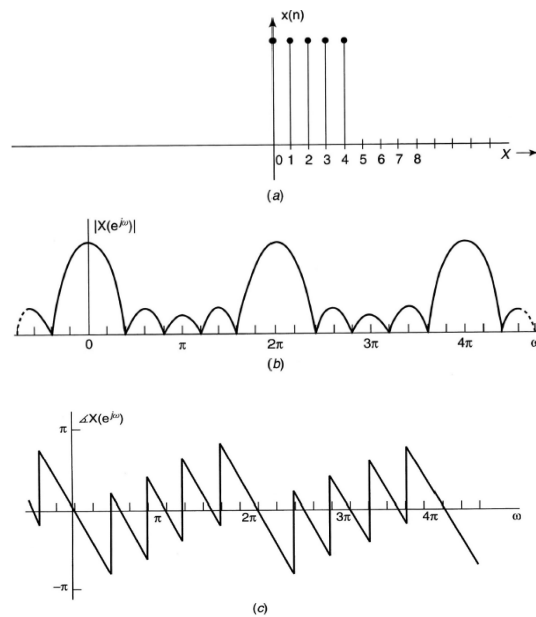
We have discussed the DTFT for a discrete-time function given by:

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} \quad (1)$$

and the IDTFT:

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})e^{j\omega n} d\omega \quad (2)$$

The pair and their properties and applications have some limitations. The input signal is usually aperiodic and may be finite in length.



**Figure 1:** In order from top to bottom, a finite-length signal, its magnitude spectrum, its phase spectrum.

Moreover, we often do not have an infinite amount of data which is required by DTFT. For example in a computer we cannot calculate uncountable infinite (continuum) of frequencies as required by DTFT. Thus, we use DTF to look at finite segment of data. We only observe the data through a window:

$$x_0[n] = x[n]w_R[n] \quad (3)$$

$$w_R[n] = \begin{cases} 1 & n = 0, 1, \dots, N-1 \\ 0 & \text{otherwise} \end{cases} \quad (4)$$

In this case, the  $x_0[n]$  is just a sampled data between  $n = 0, n = N-1$  (so,  $N$  points). The solution to our problems is given by the Discrete Fourier Transform (DFT).

### Definition 1: Discrete Fourier Transform (DFT)

The simplest relation between a length- $N$  sequence  $x[n]$ , defined for  $0 \leq n \leq N-1$ , and its DTFT  $X(e^{j\omega})$  is obtained by uniformly sampling on the  $\omega$ -axis

between  $0 \leq \omega \leq 2\pi$  at  $\omega_k = \frac{2\pi k}{N}$ , for  $0 \leq k \leq N-1$ . From the definition of the DTFT we thus have:

$$X[k] = [X(e^{j\omega})]_{\omega=\frac{2\pi k}{N}} = \sum_{n=0}^{N-1} x[n]e^{-j2\pi k \frac{n}{N}} \quad (5)$$

Note that  $X[k]$  is also a length- $N$  sequence in the frequency domain and it is called the Discrete Fourier Transform (DFT) of the sequence  $x[n]$ . Using the notation  $W_N = e^{-j\frac{2\pi}{N}}$ , the DFT is usually expressed as:

$$X[k] = \sum_{n=0}^{N-1} x[n]W_N^{kn}, \quad 0 \leq k \leq N-1 \quad (6)$$

### Definition 2: Inverse Discrete Fourier Transform (IDFT)

The Inverse Discrete Fourier Transform (IDFT) is given by:

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k]W_N^{-kn}, \quad 0 \leq n \leq N-1 \quad (7)$$

To verify the above expression we multiply both sides of the above equation by  $W_N^{\ell n}$  and sum the result from  $n=0$  to  $n=N-1$ , resulting in:

$$\begin{aligned} \sum_{n=0}^{N-1} x[n]W_N^{\ell n} &= \sum_{n=0}^{N-1} \left( \frac{1}{N} \sum_{k=0}^{N-1} X[k]W_N^{-kn} \right) W_N^{\ell n} \\ &= \frac{1}{N} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} X[k]W_N^{-(k-\ell)n} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \sum_{n=0}^{N-1} X[k]W_N^{-(k-\ell)n} \end{aligned} \quad (8)$$

Making use of the identity:

$$\sum_{n=0}^{N-1} W_N^{-(k-\ell)n} = \begin{cases} N & k - \ell = rN, \ n \in \mathbb{Z} \\ 0 & \text{otherwise} \end{cases} \quad (9)$$

we observe that the right-hand-side of the last equation is equal to  $X[\ell]$ . Hence:

$$\sum_{n=0}^{N-1} x[n]W_N^{\ell n} = X[\ell] \quad (10)$$

### Example 1: Discrete Fourier Transform

Consider the length- $N$  sequence:

$$x[n] = \begin{cases} 1 & n = 0 \\ 0 & 1 \leq n \leq N-1 \end{cases} \quad (11)$$

Its  $N$ -point DFT is given by:

$$X[k] = \sum_{n=0}^{N-1} x[n]W_N^{kn} = x[0]W_N^0 = 1 \quad (12)$$

with  $0 \leq k \leq N - 1$ .

### Example 2: Discrete Fourier Transform

Consider the length- $N$  sequence:

$$y[n] = \begin{cases} 1 & n = m \\ 0 & 0 \leq n \leq m - 1, m + 1 \leq n \leq N - 1 \end{cases} \quad (13)$$

Its  $N$ -point DFT is given by:

$$Y[k] = \sum_{n=0}^{N-1} y[n] W_N^{kn} = y[m] W_N^{km} = W_N^{km} \quad (14)$$

with  $0 \leq k \leq N - 1$ .

### Example 3: Discrete Fourier Transform

Consider the length- $N$  sequence defined for  $0 \leq n \leq N - 1$ :

$$g[n] = \cos\left(\frac{2\pi r n}{N}\right), \quad 0 \leq r \leq N - 1 \quad (15)$$

Using trigonometric identities, we can rewrite:

$$g[n] = \frac{1}{2} \left( e^{j2\pi r \frac{n}{N}} + e^{-j2\pi r \frac{n}{N}} \right) = \frac{1}{2} (W_N^{-rn} + W_N^{rn}) \quad (16)$$

The  $N$ -point DFT of  $g[n]$  is thus given by:

$$G[k] = \sum_{n=0}^{N-1} g[n] W_N^{kn} = \frac{1}{2} \left( \sum_{n=0}^{N-1} W_N^{-(r-k)n} + \sum_{n=0}^{N-1} W_N^{(r+k)n} \right) \quad (17)$$

with  $0 \leq k \leq N - 1$ . Making use of the identity:

$$\sum_{n=0}^{N-1} W_N^{-(k-\ell)n} = \begin{cases} N & k - \ell = rN, r \in \mathbb{Z} \\ 0 & \text{otherwise} \end{cases} \quad (18)$$

we get:

$$\begin{cases} \frac{N}{2} & k = r \\ \frac{N}{2} & k = N - r \\ 0 & \text{otherwise} \end{cases} \quad (19)$$

with  $0 \leq k \leq N - 1$ .

#### 0.1.1 Matrix relations

The DFT samples defined by:

$$X[k] = \sum_{n=0}^{N-1} x[n] W_N^{kn}, \quad 0 \leq k \leq N - 1 \quad (20)$$

can be expressed in matrix form as:

$$\mathbf{X} = \mathbf{D}_N \mathbf{x} \quad (21)$$

where:

$$\mathbf{X} = [X[0] \ X[1] \ \dots \ X[N-1]]^T \quad (22)$$

$$\mathbf{x} = [x[0] \ x[1] \ \dots \ x[N-1]]^T \quad (23)$$

and  $\mathbf{D}_N$  is the  $N \times N$  DFT matrix given by:

$$\mathbf{D}_N = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N^1 & W_N^2 & \dots & W_N^{(N-1)} \\ 1 & W_N^2 & W_N^4 & \dots & W_N^{2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & W_N^{(N-1)} & W_N^{2(N-1)} & \dots & W_N^{(N-1)^2} \end{bmatrix} \quad (24)$$

Likewise, the IDFT relation given by:

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-kn}, \quad 0 \leq n \leq N-1 \quad (25)$$

can be expressed in matrix form as:

$$\mathbf{x} = \mathbf{D}_N^{-1} \mathbf{X} \quad (26)$$

where  $\mathbf{D}_N^{-1}$  is the  $N \times N$  IDFT matrix, given by:

$$\mathbf{D}_N^{-1} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N^{-1} & W_N^{-2} & \dots & W_N^{-(N-1)} \\ 1 & W_N^{-2} & W_N^{-4} & \dots & W_N^{-2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & W_N^{-(N-1)} & W_N^{-2(N-1)} & \dots & W_N^{-(N-1)^2} \end{bmatrix} = \frac{1}{N} \mathbf{D}_N^* \quad (27)$$

### 0.1.2 DTFT from DFT by interpolation

The  $N$ -point DFT  $X[k]$  of a length- $N$  sequence  $x[n]$  is simply the frequency samples of its DTFT  $X(e^{j\omega})$  evaluated at  $N$  uniformly spaced frequency points:

$$\omega = \omega_k = \frac{2\pi k}{N}, \quad 0 \leq k \leq N-1 \quad (28)$$

Given the  $N$ -point DFT  $X[k]$  of a length- $N$  sequence  $x[n]$ , its DTFT  $X(e^{j\omega})$  can be uniquely determined from  $X[k]$ . Thus:

$$\begin{aligned} X(e^{j\omega}) &= \sum_{n=0}^{N-1} x[n] e^{-j\omega n} \\ &= \sum_{n=0}^{N-1} \left[ \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-kn} \right] e^{-j\omega n} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} X[k] \underbrace{\sum_{n=0}^{N-1} e^{-j(\omega - \frac{2\pi k}{N})n}}_S \end{aligned} \quad (29)$$

To develop a compact expression for the sum  $S$ , let  $r = e^{-j(\omega - \frac{2\pi k}{N})}$ . Then:

$$S = \sum_{n=0}^{N-1} r^n \quad (30)$$

From the above:

$$\begin{aligned}
 rS &= \sum_{n=1}^N r^n = 1 + \sum_{n=1}^{N-1} r^n r^N - 1 \\
 &= \sum_{n=0}^{N-1} r^n + r^N - 1 = S + r^N - 1
 \end{aligned} \tag{31}$$

or, equivalently:

$$S - rS = (1 - r)S = 1 - r^N \tag{32}$$

Hence:

$$\begin{aligned}
 S &= \frac{1 - r^N}{1 - r} \\
 &= \frac{1 - e^{-j(\omega N - 2\pi k)}}{1 - e^{-j(\omega - \frac{2\pi k}{N})}} \\
 &= \frac{\sin\left(\frac{\omega N 2\pi k}{2}\right)}{\sin\left(\frac{\omega N - 2\pi k}{2N}\right)} e^{-j\left(\frac{\omega - 2\pi k}{N}\right)\left(\frac{N-1}{2}\right)}
 \end{aligned} \tag{33}$$

Therefore:

$$X(e^{j\omega}) = \frac{1}{N} \sum_{k=0}^{N-1} X[k] \frac{\sin\left(\frac{\omega N 2\pi k}{2}\right)}{\sin\left(\frac{\omega N - 2\pi k}{2N}\right)} e^{-j\left(\frac{\omega - 2\pi k}{N}\right)\left(\frac{N-1}{2}\right)} \tag{34}$$

### 0.1.3 Sampling the DTFT

Consider a sequence  $x[n]$  with a DTFT  $X(e^{j\omega})$ . We sample  $X(e^{j\omega})$  at  $N$  equally spaced points  $\omega_k = \frac{2\pi k}{N}$ ,  $0 \leq k \leq N-1$ , developing the  $N$  frequency samples  $\{X(e^{j\omega_k})\}$ . These  $N$  frequency samples can be considered as an  $N$ -point DFT  $Y[k]$  whose  $N$ -point IDFT is a length- $N$  sequence  $y[n]$ . Now:

$$X(e^{j\omega}) = \sum_{\ell=-\infty}^{\infty} x[\ell] e^{-j\omega\ell} \tag{35}$$

Thus:

$$Y[k] = X(e^{j\omega_k}) = X(e^{j\frac{2\pi k}{N}}) = \sum_{\ell=-\infty}^{\infty} x[\ell] e^{-j2\pi k \frac{\ell}{N}} = \sum_{\ell=-\infty}^{\infty} x[\ell] W_N^{k\ell} \tag{36}$$

An IDFT of  $Y[k]$  yields:

$$\begin{aligned}
 y[n] &= \frac{1}{N} \sum_{k=0}^{N-1} Y[k] W_N^{-kn} \\
 &= \frac{1}{N} \sum_{k=0}^{N-1} \sum_{\ell=-\infty}^{\infty} x[\ell] W_N^{k\ell} W_N^{-kn} \\
 &= \sum_{\ell=-\infty}^{\infty} x[\ell] \left[ \sum_{k=0}^{N-1} W_N^{-k(n-\ell)} \right] \\
 &= \sum_{m=-\infty}^{\infty} x[n + mN]
 \end{aligned} \tag{37}$$

with  $0 \leq n \leq N - 1$ , where in the last passage the identity in Eq. 18 is employed. Thus,  $y[n]$  is obtained from  $x[n]$  by adding an infinite number of shifted replicas of  $x[n]$ , with each replica shifted by an integer multiple of  $N$  sampling instants, and observing the sum only for the interval  $0 \leq n \leq N - 1$ .

To apply the last result to finite-length sequences, we assume that the samples outside the specified range are zeros. Thus, if  $x[n]$  is a length- $M$  sequence with  $M \leq N$ , then  $y[n] = x[n]$  for  $0 \leq n \leq N - 1$ . If  $M > N$ , there is a time-domain aliasing of samples of  $x[n]$  in generating  $y[n]$ , and  $x[n]$  cannot be recovered from  $y[n]$ .

#### Example 4: Aliasing

Let  $x[n] = \{0, 1, 2, 3, 4, 5\}$ . By sampling its DTFT  $X(e^{j\omega})$  at  $\omega_k = \frac{2\pi k}{4}$ , with  $0 \leq k \leq 3$ , and then applying a 4-point IDFT to these samples, we arrive at the sequence  $y[n]$  given by:

$$y[n] = x[n] + x[n+4] + x[n-4], \quad 0 \leq n \leq 3 \quad (38)$$

We get  $y[n] = \{4, 6, 2, 3\}$ .  $x[n]$  cannot be recovered from  $y[n]$ .

### 0.1.4 DFT properties

Like the DTFT, the DFT also satisfies a number of properties that are useful in signal processing applications. Some of these properties are essentially identical to those of the DTFT, while some others are somewhat different. A summary of the DFT properties are given in Figures 2, 3 and 4.

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k]$
$x^*[n]$	$X^*[-k]_N$
$x^*[\langle -n \rangle_N]$	$X^*[k]$
$\text{Re}\{x[n]\}$	$X_{\text{pcs}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] + X^*[\langle -k \rangle_N]\}$
$j \text{Im}\{x[n]\}$	$X_{\text{pca}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] - X^*[\langle -k \rangle_N]\}$
$x_{\text{pcs}}[n]$	$\text{Re}\{X[k]\}$
$x_{\text{pca}}[n]$	$j \text{Im}\{X[k]\}$

Note:  $x_{\text{pcs}}[n]$  and  $x_{\text{pca}}[n]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $x[n]$ , respectively. Likewise,  $X_{\text{pcs}}[k]$  and  $X_{\text{pca}}[k]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $X[k]$ , respectively.

**Figure 2:** Symmetry relations of DFT for a complex sequence  $x[n]$ .

### 0.1.5 Circular shift of a sequence

This property is analogous to the time-shifting property of the DTFT but with a difference. Consider length- $N$  sequences defined for  $0 \leq n \leq N - 1$ . Sample values of such sequences are equal to zero for values of  $n < 0$  and  $n \geq N$ . If  $x[n]$  is such a sequence, then for any arbitrary integer  $n_0$ , the shifted sequence  $x_1[n] = x[n - n_0]$  is no longer defined for the range  $0 \leq n \leq N - 1$ . We thus need to define another type of shift that will always keep the shifted sequence in the range  $0 \leq n \leq N - 1$ . The desired shift, called the circular shift, is defined using a modulo operation:

$$x_c[n] = x[\langle n - n_0 \rangle_N] \quad (39)$$

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k] = \text{Re}\{X[k]\} + j \text{Im}\{X[k]\}$
$x_{\text{pe}}[n]$ $x_{\text{po}}[n]$	$\text{Re}\{X[k]\}$ $j \text{Im}\{X[k]\}$
Symmetry relations	$X[k] = X^*[-k]_N$
	$\text{Re } X[k] = \text{Re } X[-k]_N$
	$\text{Im } X[k] = -\text{Im } X[-k]_N$
	$ X[k]  =  X[-k]_N $
	$\arg X[k] = -\arg X[-k]_N$

Note:  $x_{\text{pe}}[n]$  and  $x_{\text{po}}[n]$  are the periodic even and periodic odd parts of  $x[n]$ , respectively.

**Figure 3:** Symmetry relations of DFT for a real sequence  $x[n]$ .

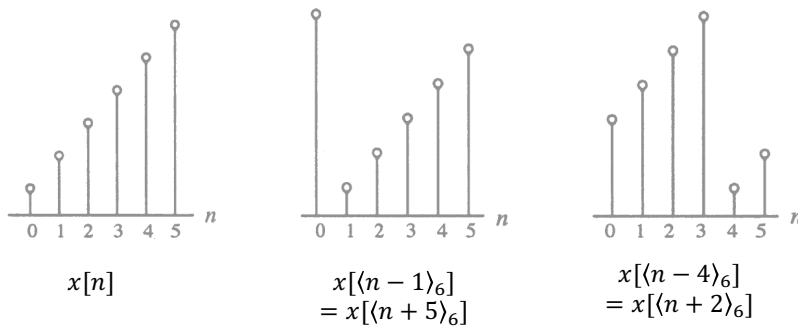
Type of Property	Length- $N$ Sequence	$N$ -point DFT
	$g[n]$ $h[n]$	$G[k]$ $H[k]$
Linearity	$\alpha g[n] + \beta h[n]$	$\alpha G[k] + \beta H[k]$
Circular time-shifting	$g[\langle n - n_0 \rangle_N]$	$W_N^{kn_0} G[k]$
Circular frequency-shifting	$W_N^{-k_0 n} g[n]$	$G[(k - k_0)_N]$
Duality	$G[n]$	$N g[-k]_N$
$N$ -point circular convolution	$\sum_{m=0}^{N-1} g[m] h[\langle n - m \rangle_N]$	$G[k] H[k]$
Modulation	$g[n] h[n]$	$\frac{1}{N} \sum_{m=0}^{N-1} G[m] H[\langle k - m \rangle_N]$
Parseval's relation	$\sum_{n=0}^{N-1}  x[n] ^2 = \frac{1}{N} \sum_{k=0}^{N-1}  X[k] ^2$	

**Figure 4:** General properties of DFT.

For  $n_0 > 0$  (right circular shift), the above equation implies:

$$x_c[n] = \begin{cases} x[n - n_0] & n_0 \leq n \leq N - 1 \\ x[N - n_0 + n] & 0 \leq n \leq n_0 \end{cases} \quad (40)$$

An illustration of the concept of circular shift is showed in Figure 5. As it is possible to observe, a right circular shift by  $n_0$  is equivalent to a left circular shift by  $N - n_0$  sample periods. A circular shift by an integer number  $n_0$  greater than  $N$  is equivalent to a circular shift by  $\langle n_0 \rangle_N$ .



**Figure 5:** Illustration of circular shift.

### 0.1.6 Circular convolution

This operation is analogous to linear convolution, but with a difference. Consider two length- $N$  sequences,  $g[n]$  and  $h[n]$ , respectively. Their linear convolution results in a length- $(2N - 1)$  sequence  $y_L[n]$  given by:

$$y_L[n] = \sum_{m=0}^{N-1} g[m]h[n-m], \quad 0 \leq n \leq 2N-2 \quad (41)$$

In computing  $y_L[n]$  we have assumed that both length- $N$  sequences have been zero-padded to extend their lengths to  $2N - 1$ . The longer form of  $y_L[n]$  results from the time-reversal of the sequence  $h[n]$  and its linear shift to the right. The first nonzero value of  $y_L[n]$  is  $y_L[0] = g[0]h[0]$  and the last nonzero value is  $y_L[2N-2] = g[N-1]h[N-1]$ .

To develop a convolution-like operation resulting in a length- $N$  sequence  $y_C[n]$ , we need to define a circular time-reversal, and then apply a circular time-shift. Resulting operation, called a circular convolution, is defined by:

$$y_C[n] = \sum_{m=0}^{N-1} g[m]h[\langle n-m \rangle_N], \quad 0 \leq n \leq N-1 \quad (42)$$

Since the operation defined involves two length- $N$  sequences, it is often referred to as an  $N$ -point circular convolution, denoted as: