# Digital Signal Processing Properties of the Discrete Fourier Transform

D. Richard Brown III

## Some Useful Properties of the DFT

TABLE 8.2 SUMMARY OF PROPERTIES OF THE DFT

Finite-Length Sequence (Length N)	N-point DFT (Length $N$ )
1. x[n]	X[k]
2. $x_1[n], x_2[n]$	$X_1[k], X_2[k]$
3. $ax_1[n] + bx_2[n]$	$aX_1[k] + bX_2[k]$
4. X[n]	$Nx[((-k))_N]$
$5.  x[((n-m))_N]$	$W_N^{km}X[k]$
6. $W_N^{-\ell n} x[n]$	$X[((k-\ell))_N]$
7. $\sum_{m=0}^{N-1} x_1[m] x_2[((n-m))_N]$	$X_1[k]X_2[k]$
8. $x_1[n]x_2[n]$	$\frac{1}{N} \sum_{\ell=0}^{N-1} X_1[\ell] X_2[((k-\ell))_N]$
9. $x^*[n]$	$X^*[((-k))_N]$
0. $x^*[((-n))_N]$	$X^*[k]$

Notation:  $((n))_N$  means n modulo N and  $W_N = e^{-j2\pi/N}$ .

#### Time-Shifting Property: DTFT vs. DFS vs. DFT

Recall the time-shifting property of the DTFT for the sequence  $x[n] \leftrightarrow X(e^{j\omega})$ :

$$x[n-m] \quad \leftrightarrow \quad e^{-j\omega m}X(e^{j\omega}) \quad \text{for } \omega \in \mathbb{R}$$

For the DFS of a periodic sequence  $\tilde{x}[n] \leftrightarrow \tilde{X}[k]$ , we have

$$\tilde{x}[n-m] \quad \leftrightarrow \quad e^{-j2\pi km/N}\tilde{X}[k] \quad \text{for } k \in \mathbb{Z}$$

Now denote  $\tilde{x}[n]=x[n]*\tilde{p}[n]$  with x[n] a length-N sequence and  $\tilde{p}[n]=\sum_{r=-\infty}^\infty \delta[n-rN]$  a periodic impulse train. Since

$$\tilde{x}[n] = x[((n))_N] \qquad \text{and} \qquad \tilde{X}[k] = X[k] \quad k = 0, 1, \dots, N-1$$

we see a linear shift of  $\tilde{x}[n]$  corresponds to a **circular shift** of x[n]. Hence

$$x[((n-m))_N] \quad \leftrightarrow \quad e^{-j2\pi km/N}X[k] \quad \text{for } k=0,1,\ldots,N-1$$

for the DFT.

## Convolution Property: DTFT vs. DFT

Recall the convolution property of the DTFT:

$$x_1[n] * x_2[n] \leftrightarrow X_1(e^{j\omega})X_2(e^{j\omega})$$

for all  $\omega \in \mathbb{R}$  if the DTFTs both exist. Note this relation holds for infinite length or finite length sequences (the sequences don't need to have the same length.)

What about

$$x_1[n] * x_2[n] \quad \stackrel{?}{\leftrightarrow} \quad X_1[k]X_2[k]$$

for  $k=0,\ldots,N-1$ . One problem is that  $x_1[n]$  and  $x_2[n]$  must be the same length for this to make sense at all. Another problem is that the IDFT of  $X_1[k]X_2[k]$  can't have the correct length of 2N-1.

Example:

$$\begin{split} x_1[n] &= \{\underline{1},1\} \Leftrightarrow X_1[k] = \{\underline{2},0\} \\ x_2[n] &= \{\underline{1},-1,1,-1\} \Leftrightarrow X_2[k] = \{\underline{0},0,4,0\} \\ x_1[n] * x_2[n] &= \{\underline{1},0,0,0,-1\} \Leftrightarrow \mathsf{DFT}(x_1[n] * x_2[n]) = \{\underline{0},0.69-0.95i, \\ 1.81 - 0.59i,1.81 + 0.59i,0.69 + 0.95i\} \end{split}$$

#### Periodic Convolution (DFS) and Circular Convolution (DFT)

For the DFS, we have the **periodic convolution** property

$$\tilde{x}_3[n] = \sum_{k=0}^{N-1} \tilde{x}_1[k]\tilde{x}_2[n-k] \quad \forall n \in \mathbb{Z} \quad \leftrightarrow \quad \tilde{X}_3[k] = \tilde{X}_1[k]\tilde{X}_2[k] \quad \forall k \in \mathbb{Z}$$

where  $\tilde{x}_1[n]=x_1[n]*\tilde{p}[n]$  and  $\tilde{x}_2[n]=x_2[n]*\tilde{p}[n]$  with  $x_1[n]$  and  $x_2[n]$  both length-N sequences. Since  $\tilde{x}_i[n]=x_i[((n))_N]$  and  $\tilde{X}_i[n]=X_i[((k))_N]$  for  $n=0,\ldots,N-1$  and  $k=0,\ldots,N-1$ , we can use the periodic convolution property of the DFS to write

$$x_3[n] = \sum_{k=0}^{N-1} x_1[k] x_2[((n-k))_N] \quad n = 0, \dots, N-1$$

$$\leftrightarrow \quad X_3[k] = X_1[k] X_2[k] \quad \forall k = 0, \dots, N-1$$

$$\leftrightarrow \quad \Lambda_3[\kappa] \equiv \Lambda_1[\kappa] \Lambda_2[\kappa] \quad \forall \kappa \equiv 0, \dots, N-1$$

We can define the circular convolution operator for two length-N sequences as

$$x_3[n] = x_1[n] \widehat{N} x_2[n] = \sum_{k=0}^{N-1} x_1[k] x_2[((n-k))_N] = \sum_{k=0}^{N-1} x_1[((n-k))_N] x_2[k].$$

Note the circular convolution result  $x_3[n]$  will also be length-N.

See Matlab function cconv.

#### Circular Convolution Matrix

To illustrate the idea, suppose we want to circularly convolve  $\{a[0],a[1],a[2]\}$  and  $\{b[0],b[1],b[2]\}$ . Note N=3 here. Applying the definition, we can write

This is the same as

$$\begin{bmatrix} c[0] \\ c[1] \\ c[2] \end{bmatrix} = \underbrace{ \begin{bmatrix} a[0] & a[2] & a[1] \\ a[1] & a[0] & a[2] \\ a[2] & a[1] & a[0] \end{bmatrix} }_{\text{circular conv matrix}} \begin{bmatrix} b[0] \\ b[1] \\ b[2] \end{bmatrix} = \underbrace{ \begin{bmatrix} b[0] & b[2] & b[1] \\ b[1] & b[0] & b[2] \\ b[2] & b[1] & b[0] \end{bmatrix} }_{\text{circular conv matrix}} \begin{bmatrix} a[0] \\ a[1] \\ a[2] \end{bmatrix}$$

Like linear convolution, the circular convolution matrix has a Toeplitz (actually "circulant") structure. Unlike linear convolution, the circular convolution matrix is square  $(N \times N)$ .

## Circular Convolution Example

Suppose  ${\cal N}=5$  and

$$x_1[n] = \delta[n-1]$$
$$x_2[n] = N - n$$

for  $n=0,\ldots,4$ . Both finite-length sequences are equal to zero for all other values of n. This figure illustrates how to compute

$$x_3[n] = \sum_{m=0}^{4} x_1[m]x_2[((n-m))_N]$$

for  $n=0,\ldots,4$ . The final result is  $x_3[n]=x_2[((n-1))_4]$ , i.e., a circular shift of  $x_2[n]$  by one sample. You can confirm this result easily in MATLAB as well by computing both DFTs, computing the product  $X_1[k]X_2[k]$ , and computing the IDFT.

