## The Discrete Fourier Transform (DFT) and Its Properties

Recall the DFT and IDFT:

$$X(k) = \sum_{n=0}^{N-1} x(n)e^{-j2\pi kn/N}, \qquad k = 0, 1, 2, ..., N-1$$
$$x(n) = \sum_{n=0}^{N-1} X(k)e^{j2\pi kn/N}, \qquad n = 0, 1, 2, ..., N-1$$

or,

$$X(k) = \sum_{n=0}^{N-1} x(n) W_N^{kn}, \qquad k = 0, 1, 2, ..., N-1$$
$$x(n) = \sum_{n=0}^{N-1} X(k) W_N^{-kn}, \qquad n = 0, 1, 2, ..., N-1$$

where  $W_N = e^{-j2\pi/N}$  and  $W_N^{kn} = e^{-j2\pi kn/N}$ .

We use the notation:  $x(n) \stackrel{DFT}{\leftrightarrow} X(k)$ N

The DFT is really the transform of a periodic extension of x(n):

if	x(n+N) = x(n)	$\forall n$
then	X(k+N) = X(k)	$\forall k$

The DFT obeys the principles of linearity and superposition:

$$if: \begin{array}{ccc} DFT & DFT \\ if: x_1(n) & \leftrightarrow & X_1(k) \\ N & N \end{array} \begin{array}{c} & DFT \\ \leftrightarrow & X_2(n) \\ N \\ N \end{array}$$

then: 
$$a_1 x_1(n) + a_2 x_2(n) \leftrightarrow a_1 X_1(k) + a_2 X_2(k)$$
  
N

What are the implications of these properties?

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#### **Symmetry Properties**

Recall our expression for the periodic extension of x(n):



We define this process as a circular shift:  $\begin{aligned} x'(n) &= (x(n-k), modulo \ N) \\ &\equiv x((n-k))_N \end{aligned}$ 

What should the impact of this operation be on the DFT?

Several interesting properties result:

$$(x((-n))_N = x(N-n)), \qquad 0 \le n \le N-1$$

and:

conjugate even: 
$$x_p(n) = x_p^*(N-n)$$
  
conjugate odd:  $x_p(n) = -x_p^*(N-n)$ 



This symmetry gives rise to general properties of the DFT:

N-Point Sequence	N-Point DFT			
<i>x</i> ( <i>n</i> )	X(k)			
$x^*(n)$	$X^*(N-k)$			
$x^*(N-n)$	$X^{*}(k)$			
$x_R(n)$	$X_{ce}(k) = \frac{1}{2}[X(k) + X^{*}(N-k)]$			
$jx_I(n)$	$X_{co}(k) = \frac{1}{2} [X(k) - X^*(N-k)]$			
$x_{ce}(n) = \frac{1}{2}[x(n) + x^{*}(N-n)]$	$X_{R}(k)$			
$x_{ce}(n) = \frac{1}{2}[x(n) + x^{*}(N-n)]$	$jX_{I}(k)$			
Real Signals				
Any real signal x(n)	$X(k) = X^*(N-k)$ $X_R(k) = X_R(N-k)$ $X_I(k) = -X_I(N-k)$ $ X(k)  =  X(N-k) $ $\angle X(k) = -\angle X(N-k)$			
$x_{ce}(n) = \frac{1}{2}[x(n) + x(N-n)]$	$X_{R}(k)$			
$x_{ce}(n) = \frac{1}{2}[x(n) - x^{*}(N-n)]$	$jX_I(k)$			



## **Circular Convolution (Finally - Something Different!)**

Suppose we have two finite duration sequences  $x_1(n)$  and  $x_2(n)$  of length N. Suppose we multiply the two DFTs together:  $X_3(k) = X_1(k)X_2(k)$ .

What happens when I take the IDFT of  $X_3(k)$ ?

$$\begin{aligned} x_{3}(m) &= \frac{1}{N} \sum_{k=0}^{N-1} X_{3}(k) e^{j2\pi km/N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} X_{1}(k) X_{2}(k) e^{j2\pi km/N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \left[ \sum_{n=0}^{N-1} X_{1}(k) e^{-j2\pi kn/N} \right] \left[ \sum_{k=0}^{N-1} X_{2}(k) e^{-j2\pi kn/N} \right] e^{j2\pi km/N} \\ &\cdots \\ &= \sum_{n=0}^{N-1} x_{1}(n) x_{2}((m-n))_{N}, \qquad m = 0, 1, ..., N-1 \end{aligned}$$

This is known as circular convolution, and has a simple graphical interpretation (see Figure 9.2):

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

$$x_{2}(n) = \{1, 2, 3, 4\}$$

$$x_{3}(0) = \begin{cases} 2, 1, 2, 1\\ 1, 4, 3, 2 \end{cases} = 14$$

$$x_{3}(1) = \begin{cases} 2, 1, 2, 1\\ 2, 1, 4, 3 \end{cases} = 16$$

$$x_{3}(2) = \begin{cases} 2, 1, 2, 1\\ 3, 2, 1, 4 \end{cases} = 14$$

$$x_{3}(3) = \begin{cases} 2, 1, 2, 1\\ 3, 2, 1, 4 \end{cases} = 16$$

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# Additional Properties of the DFT

Property	Time Domain	Frequency Domain
Notation	x(n), y(n)	X(k),Y(k)
Periodicity	x(n) = x(n+N)	X(k) = X(k+N)
Linearity	$a_1 x_1(n) + a_2 x_2(n)$	$a_1 X_1(k) + a_2 X_2(k)$
Time Reversal	x(N-n)	X(N-k)
Circular Time Shift	$x((n-l))_N$	$X(k)e^{-j2\pi kl/N}$
Circular Frequency Shift	$x(n)e^{j2\pi nl/N}$	$X((k-l))_N$
Complex Conjugate	$x^*(n)$	$X^*(N-k)$
Circular Convolution	$x_1(n) \oplus x_2(n)$	$X_1(k)X_2(k)$
Circular Correlation	$x(n) \oplus y^*(-n)$	$X(k)Y^{*}(k)$
Multiplication	$x_1(n)x_2(n)$	$\frac{1}{N}X_1(k) \oplus X_2(k)$
Parseval's Theorem	$\sum_{n=0}^{N-1} x(n)y^*(n)$	$\frac{1}{N} \sum_{n=0}^{N-1} X(k) Y^{*}(k)$





### Using The DFT For Linear Filtering - Overlap-Save

Consider the problem of performing linear filtering in the frequency domain by processing signals a frame at a time:



In this example, we desire to filter the signal in the frequency domain using an M point filter where  $L \gg M$ . Out strategy is to compute a DFT of the signal, multiply this DFT by the pre-computed DFT of the filter, and compute an inverse DFT to obtain the output signal.

In the **Overlap and Save** method, we use N = L + M - 1 points in our DFT. First, we pad the filter impulse response to *N* points and take an *N*-point DFT. Next, we process the data as follows:



*N* points are processed each frame, using M - 1 points from the previous frame. The output signal is constructed by retaining the *L* last points.

Note that *L* new data points are introduced each frame. What does this imply for the case  $L \approx M$  when the signal is time-varying?

