

# Digital Signal Processing

## Linear Convolution with the Discrete Fourier Transform

D. Richard Brown III

# Linear and Circular Convolution Properties

Recall the (linear) convolution property

$$x_3[n] = x_1[n] * x_2[n] \quad \leftrightarrow \quad X_3(e^{j\omega}) = X_1(e^{j\omega})X_2(e^{j\omega}) \quad \forall \omega \in \mathbb{R}$$

if the necessary DTFTs exist. If  $x_1[n]$  is length  $N_1$  and  $x_2[n]$  is length  $N_2$ , then  $x_3[n]$  will be length  $N_3 = N_1 + N_2 - 1$ . See MATLAB function `conv`.

For the DFT, we have the **circular** convolution property

$$x_3[n] = x_1[n] \textcircled{N} x_2[n] \quad \leftrightarrow \quad X_3[k] = X_1[k]X_2[k] \quad \forall k = 0, \dots, N-1$$

where

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

Note that  $x_1[n]$  and  $x_2[n]$  must have the same length  $N = N_1 = N_2$  and the result  $x_3[n]$  will also be length  $N$ . See Matlab function `cconv`.

# Linear Convolution with the DFT?

Suppose we want to compute

$$x_3[n] = x_1[n] * x_2[n].$$

We could compute the DTFTs of  $x_1[n]$  and  $x_2[n]$ , take their product, and then compute the inverse DTFT to get  $x_3[n]$ , i.e.,

$$x_3[n] = \text{IDTFT}(\text{DTFT}(x_1[n]) \cdot \text{DTFT}(x_2[n]))$$

What if we want to use the DFT to compute the linear convolution instead? We know

$$x_3[n] = \text{IDFT}(\text{DFT}(x_1[n]) \cdot \text{DFT}(x_2[n]))$$

will not work because this performs circular convolution.

## Avoiding Time-Domain Aliasing

Recall our notation  $W_M = e^{-j2\pi/M}$ . We have seen previously that the  $M$ -point DFT of a finite-length sequence  $x_i[n]$  with length  $N_i$

$$X_i[k] = \sum_{n=0}^{N_i-1} x_i[n] W_M^{kn} \quad k = 0, 1, \dots, M-1$$

must satisfy  $M \geq N_i$  to avoid time-domain aliasing when computing  $\text{IDFT}_M(\text{DFT}_M(x_i[n]))$ .

If  $x_3[n] = x_1[n] * x_2[n]$  with  $x_1[n]$  and  $x_2[n]$  both finite length sequences, then the longest sequence is  $x_3[n]$  with length  $N_3 = N_1 + N_2 - 1$ .

This implies that our DFTs  $X_1[k]$ ,  $X_2[k]$ , and  $X_3[k]$  should all be of length  $M \geq N_1 + N_2 - 1$  to avoid time-domain aliasing. In other words,

$$x_3[n] = \text{IDFT}_M(\text{DFT}_M(x_1[n]) \cdot \text{DFT}_M(x_2[n]))$$

will result in  $x_3[n] = x_1[n] * x_2[n]$  if  $M \geq N_1 + N_2 - 1$ .

## Example

Suppose  $x_1 = [\underline{1}, 2, 3]$  and  $x_2 = [\underline{1}, 1, 1]$ . We can compute the linear convolution as

$$x_3[n] = x_1[n] * x_2[n] = [\underline{1}, 3, 6, 5, 3].$$

If we instead compute

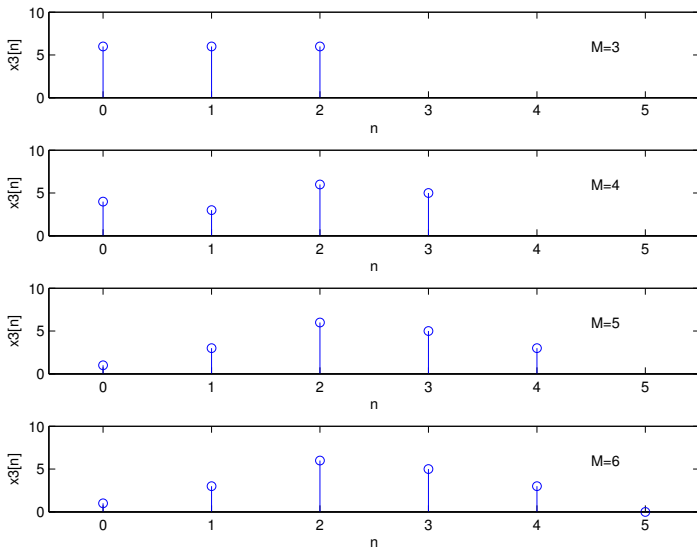
$$x_3[n] = \text{IDFT}_M(\text{DFT}_M(x_1[n]) \cdot \text{DFT}_M(x_2[n]))$$

we get

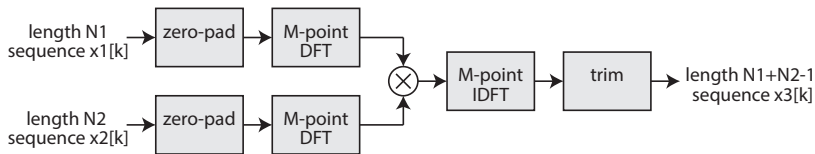
$$x_3[n] = \begin{cases} [\underline{6}, 6, 6] & M = 3 \\ [\underline{4}, 3, 6, 5] & M = 4 \\ [\underline{1}, 3, 6, 5, 3] & M = 5 \\ [\underline{1}, 3, 6, 5, 3, 0] & M = 6 \end{cases}$$

Observe that time-domain aliasing of  $x_3[n]$  is avoided for  $M \geq 5$ .

## Example (continued)



# Linear Convolution with the DFT



## Remarks:

- ▶ Zero-padding avoids time-domain aliasing and make the circular convolution behave like linear convolution.
- ▶  $M$  should be selected such that  $M \geq N_1 + N_2 - 1$ .
- ▶ In practice, the DFTs are computed with the FFT.
- ▶ The amount of computation with this method can be less than directly performing linear convolution (especially for long sequences).
- ▶ Since the FFT is most efficient for sequences of length  $2^m$  with integer  $m$ ,  $M$  is usually chosen so that  $M = 2^m \geq N_1 + N_2 - 1$ .