# 6.3000: Signal Processing

**DFT** and Circular Convolution

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#### **Convolution: Three Ways**

The signal x[n], defined below, is zero outside the indicated range.



Consider three ways to calculate the convolution of x[n] with itself. 1. direct convolution:

$$y_1[n] = (x * x)[n] = \sum_{m=-\infty}^{\infty} x[m]x[n-m]$$

2. using DTFTs:

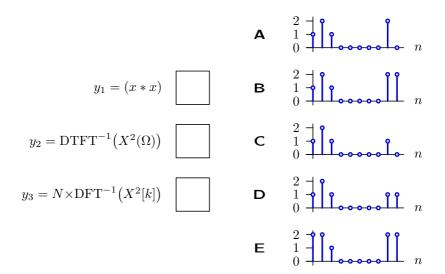
$$y_2[n] = \frac{1}{2\pi} \int_{2\pi} X^2(\Omega) e^{j\Omega n} \, d\Omega \quad \text{where} \quad X(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n}$$

3. using DFTs of length N=16:

$$y_3[n] = 16\sum_{k=0}^{15} X^2[k]e^{j\frac{2\pi k}{16}n} \quad \text{where} \quad X[k] = \frac{1}{16}\sum_{n=0}^{15} x[n]e^{-j\frac{2\pi k}{16}n}$$

## **Convolution: Three Ways**

The plots on the right show the **first ten samples** of five signals. Match the signals on the left with the corresponding plots on the right.



## **Circular Convolution**

Multiplication of DFTs corresponds to **circular** convolution in time. Assume that F[k] is the product of the DFTs of  $f_a[n]$  and  $f_b[n]$ .

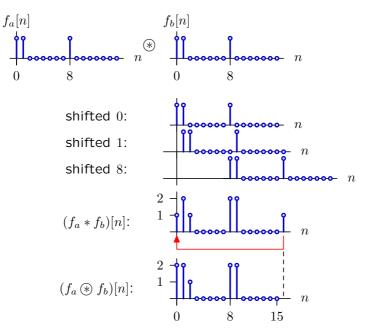
$$\begin{split} f[n] &= \sum_{k=0}^{N-1} F[k] e^{j\frac{2\pi k}{N}n} = \sum_{k=0}^{N-1} F_a[k] F_b[k] e^{j\frac{2\pi k}{N}n} \\ &= \sum_{k=0}^{N-1} F_a[k] \Big( \frac{1}{N} \sum_{m=0}^{N-1} f_b[m] e^{-j\frac{2\pi k}{N}m} \Big) e^{j\frac{2\pi k}{N}n} \\ &= \frac{1}{N} \sum_{m=0}^{N-1} f_b[m] \sum_{k=0}^{N-1} F_a[k] e^{j\frac{2\pi k}{N}(n-m)} \\ &= \frac{1}{N} \sum_{m=0}^{N-1} f_b[m] f_{ap}[n-m] \end{split}$$

where  $f_{ap}[n] = f_a[n \mod N]$  is a periodically extended version of  $f_a[n]$ . We refer to this as **circular** or **periodic** convolution:

$$\frac{1}{N} (f_a \circledast f_b)[n] \qquad \stackrel{\text{DFT}}{\Longrightarrow} \qquad F_a[k] \times F_b[k]$$

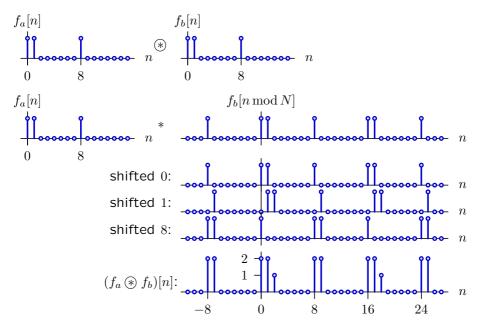
# **Circular Convolution**

Circular convolution is equivalent to conventional convolution followed by periodic summation of results back into base period.



# **Circular Convolution**

Circular convolution of two signals is equal to conventional convolution of one signal with a periodically extended version of the other.



## Summary

One of the most useful properties of the DTFT is its filter property: convolution in time corresponds to multiplication in frequency.

 $(f * g)[n] \stackrel{\text{DTFT}}{\Longrightarrow} F(\Omega)G(\Omega)$ 

The DFT is slightly more complicated since the DFT is equivalent to the DTFS of a periodically extended version of x[n]:

x[n] = x[n+mN] for all integers m

A result of this periodicity is that the convolution that results when two DFTs are multiplied is also periodic.

We refer to this type of convolution as "circular convolution."

 $\frac{1}{N}(f \circledast g)[n] \stackrel{\mathrm{DFT}}{\Longrightarrow} F[k]G[k]$