

# 6.3000: Signal Processing

## Circular Convolution

*October 24, 2024*

## 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

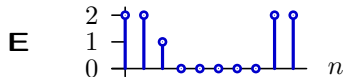
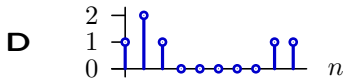
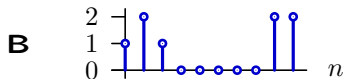
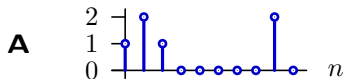
The plots on the right show the **first ten samples** of five signals.

Match the signals on the left with the corresponding plots on the right.

$$y_1 = (x * x) \quad \square$$

$$y_2 = \text{DTFT}^{-1}(X^2(\Omega)) \quad \square$$

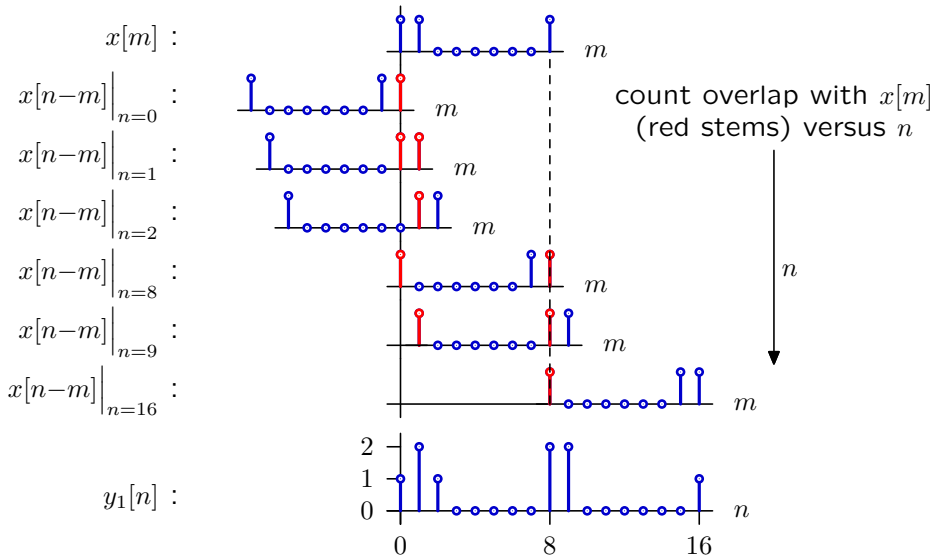
$$y_3 = N \times \text{DFT}^{-1}(X^2[k]) \quad \square$$



## Convolution: Three Ways

Calculate  $(x*x)[n]$  by direct convolution: flip and shift.

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



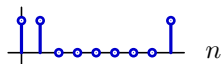
## Convolution: Three Ways

---

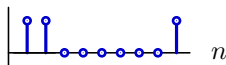
Calculate  $(x*x)[n]$  by direct convolution: superposition.

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

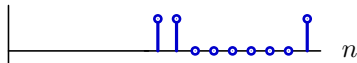
$$x[0] \times x[n-0] :$$



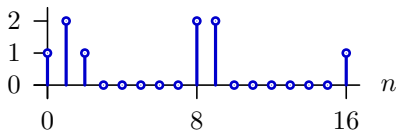
$$x[1] \times x[n-1] :$$



$$x[8] \times x[n-8] :$$



$$y_1[n] :$$



Note: Superposition and flip-and-shift are equivalent methods. They always give the same answer.

## Convolution: Three Ways

Plots on the left show the **first ten samples** of five signals.

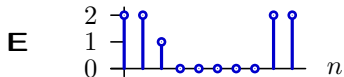
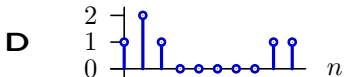
Match signals on the left with corresponding samples on the right.

$$y_1 = (x * x)$$

B

$$y_2 = \text{DTFT}^{-1}(X^2(\Omega))$$

$$y_3 = N \times \text{DFT}^{-1}(X^2[k])$$



## Convolution: Three Ways

---

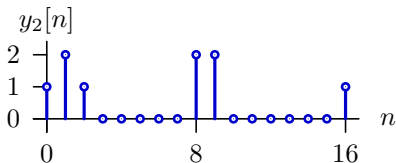
Calculate  $(x*x)[n]$  using DTFTs.



$$X(\Omega) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega n} = 1 + e^{-j\Omega} + e^{-j8\Omega}$$

$$X^2(\Omega) = \left(1 + e^{-j\Omega} + e^{-j8\Omega}\right)^2 = 1 + 2e^{-j\Omega} + e^{-j2\Omega} + 2e^{-j8\Omega} + 2e^{-j9\Omega} + e^{-j16\Omega}$$

$$\begin{aligned} y_2[n] &= \frac{1}{2\pi} \int_{2\pi} X^2(\Omega) e^{j\Omega n} d\Omega \\ &= \frac{1}{2\pi} \int_{2\pi} \left(1 + 2e^{-j\Omega} + e^{-j2\Omega} + 2e^{-j8\Omega} + 2e^{-j9\Omega} + e^{-j16\Omega}\right) e^{j\Omega n} d\Omega \\ &= \delta[n] + 2\delta[n-1] + \delta[n-2] + 2\delta[n-8] + 2\delta[n-9] + \delta[n-16] \end{aligned}$$



Multiplying DTFTs is always equivalent to direct convolution.

## Convolution: Three Ways

Plots on the left show the **first ten samples** of five signals.

Match signals on the left with corresponding samples on the right.

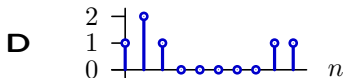
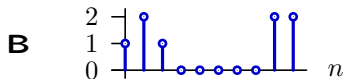
$$y_1 = (x * x)$$

B

$$y_2 = \text{DTFT}^{-1}(X^2(\Omega))$$

B

$$y_3 = N \times \text{DFT}^{-1}(X^2[k])$$





## Convolution: Three Ways

Calculate  $(x*x)[n]$  using DFTs ( $N = 16$ ).



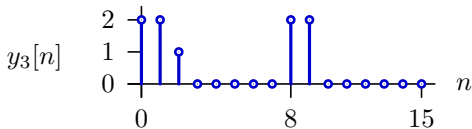
$$X[k] = \frac{1}{16} \sum_{n=0}^{15} x[n] e^{-j \frac{2\pi k}{16} n} = \frac{1}{16} \left( 1 + e^{-j \frac{2\pi k}{16}} + e^{-j \frac{2\pi k}{16} 15} \right)$$

$$X^2[k] = \frac{1}{256} \left( 1 + 2e^{-j \frac{2\pi k}{16}} + e^{-j 2 \frac{2\pi k}{16}} + 2e^{-j 8 \frac{2\pi k}{16}} + 2e^{-j 9 \frac{2\pi k}{16}} + \underbrace{e^{-j 16 \frac{2\pi k}{16}}}_{=1} \right)$$

$$y_3[n] = 16 \sum_{k=0}^{15} X^2[k] e^{j \frac{2\pi k}{16} n}$$

$$= \frac{16}{256} \sum_{k=0}^{15} \left( 2 + 2e^{-j \frac{2\pi k}{16}} + e^{-j 2 \frac{2\pi k}{16}} + 2e^{-j 8 \frac{2\pi k}{16}} + 2e^{-j 9 \frac{2\pi k}{16}} \right) e^{j \frac{2\pi k}{16} n}$$

$$= 2\delta[n] + 2\delta[n-1] + \delta[n-2] + 2\delta[n-8] + 2\delta[n-9]$$



Since  $N=16$ , the sample at  $n=16$  in direct convolution **aliases** to  $n=0$ .

## Convolution: Three Ways

Plots on the left show the **first ten samples** of five signals.

Match signals on the left with corresponding samples on the right.

$$y_1 = (x * x)$$

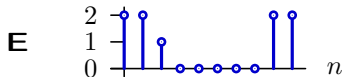
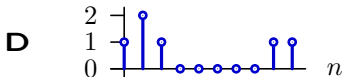
B

$$y_2 = \text{DTFT}^{-1}(X^2(\Omega))$$

B

$$y_3 = N \times \text{DFT}^{-1}(X^2[k])$$

E



## 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  $f_b[n]$ .

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

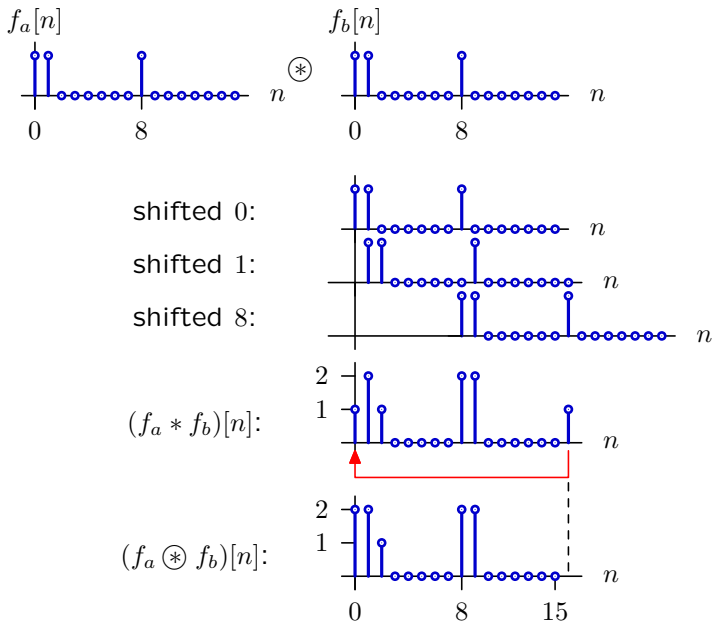
where  $f_{ap}[n] = f_a[n \bmod N]$  is a periodically extended version of  $f_a[n]$ .

We refer to this as **circular** or **periodic** convolution:

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

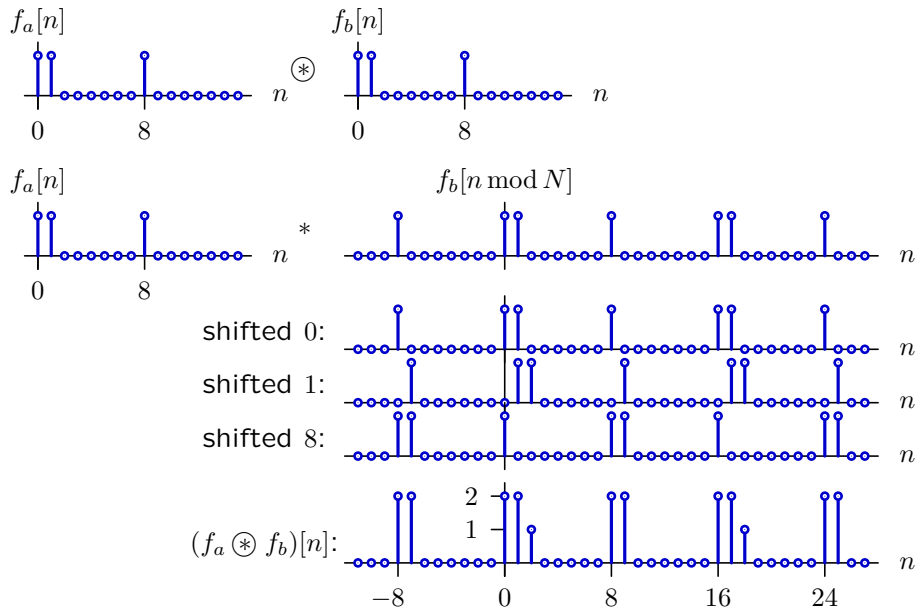
# Circular Convolution

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.

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

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

$$x[n] = x[n + mN] \quad \text{for all integers } m$$

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.”

$$\frac{1}{N}(f \circledast g)[n] \xrightarrow{\text{DFT}} F[k]G[k]$$