The DFT and Convolution

Digital Signal Processing

September 18, 2025



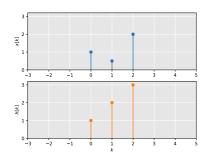
Circular Convolution

Given two periodic signals, x[n], h[n], both with period L, their **circular convolution** is

$$x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$$

The only thing we've changed is to now "wrap" the index on h.

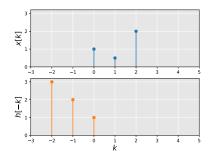
Computing
$$y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$$



$$x[n] = (1.0, 0.5, 2.0)$$

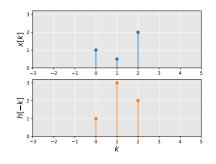
$$h[n] = (1.0, 2.0, 3.0)$$

Computing
$$y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$$



Flip h about 0

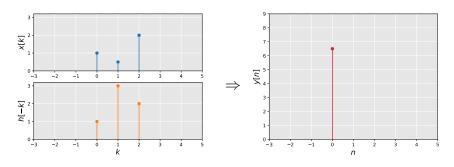
Computing
$$y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$$



Wrap h

Computing $y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$

For n = 0, flip h about 0 to get $h[-k \mod L]$.

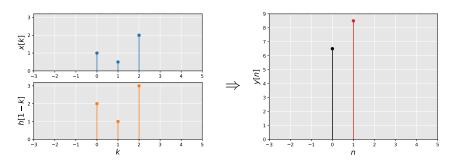


$$y[0] = x[0] \times h[0] + x[1] \times h[2] + x[2] \times h[1]$$

= 1.0 \times 1.0 + 0.5 \times 3.0 + 2.0 \times 2.0 = 6.5

Computing $y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$

For n = 1, shift h left by one to get $h[(1 - k) \mod L]$.

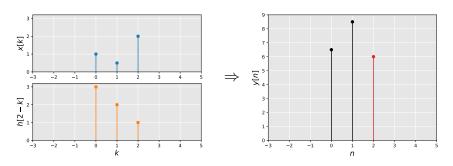


$$y[1] = x[0]h[1] + x[1]h[0] + x[2]h[2]$$

= 1.0 \times 2.0 + 0.5 \times 1.0 + 2.0 \times 3.0 = 8.5

Computing $y[n] = x[n] * h[n] = \sum_{k=0}^{L-1} x[k]h[(n-k) \mod L]$

For n = 2, shift h left again to get $h[(2 - k) \mod L]$.



$$y[2] = x[0]h[2] + x[1]h[1] + x[2]h[0]$$

= 1.0 \times 3.0 + 0.5 \times 2.0 + 2.0 \times 1.0 = 6.0

Convolution and DFT

Theorem (Convolution Theorem)

Given two periodic, complex-valued signals, x[n], y[n],

$$\mathcal{DFT}\{x[n] * y[n]\} = \sqrt{L} \left(\mathcal{DFT}\{x[n]\} \times \mathcal{DFT}\{y[n]\} \right).$$

In other words, **convolution** in the time domain becomes **multiplication** in the frequency domain.

Using notation $x[n] \stackrel{\mathcal{DFT}}{\longleftrightarrow} X[k]$ and $y[n] \stackrel{\mathcal{DFT}}{\longleftrightarrow} Y[k]$, this is:

$$x[n] * y[n] \stackrel{\mathcal{DFT}}{\longleftrightarrow} \sqrt{L}X[k]Y[k]$$

Proof of Convolution Theorem

$$\mathcal{DFT}\{x[n]*y[n]\}$$

$$= \frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} (x[n] * y[n]) e^{-i\omega_0 nk}$$

$$= \frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} \left(\sum_{m=0}^{L-1} x[m]y[n-m] \right) e^{-i\omega_0 nk}$$

$$= \frac{1}{\sqrt{L}} \sum_{m=0}^{L-1} \sum_{n=0}^{L-1} x[m]y[n-m] \left(e^{-i\omega_0 mk} e^{-i\omega_0 (n-m)k} \right)$$

$$= \sum_{m=0}^{L-1} x[m]e^{-i\omega_0 mk} \left(\frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} y[n-m]e^{-i\omega_0(n-m)k} \right)$$

$$= \left(\sum_{m=0}^{L-1} x[m]e^{-i\omega_0 km}\right) Y[k]$$

$$= \sqrt{L}X[k]Y[k]$$

Definition of \mathcal{DFT}

Definition of *

Exponential math

Rearrange

Definition of \mathcal{DFT}

Definition of \mathcal{DFT}

Convolution and DFT

Theorem (Convolution Theorem II)

Given two periodic, complex-valued signals, x[n], y[n],

$$\mathcal{DFT}\{x[n] \times y[n]\} = \frac{1}{\sqrt{L}} \left(\mathcal{DFT}\{x[n]\} * \mathcal{DFT}\{y[n]\} \right).$$

In other words, the **multiplication** in the time domain becomes **convolution** in the frequency domain.

Again, using notation $x[n] \stackrel{\mathcal{DFT}}{\longleftrightarrow} X[k]$ and $y[n] \stackrel{\mathcal{DFT}}{\longleftrightarrow} Y[k]$, this is:

$$\sqrt{L}x[n]y[n] \xleftarrow{\mathcal{DFT}} X[k] * Y[k]$$

Proof of Convolution Theorem II

$$\mathcal{DFT}^{-1}\{X[k] * Y[k]\}$$

$$= \frac{1}{\sqrt{L}} \sum_{k=0}^{L-1} (X[k] * Y[k]) e^{i\omega_0 nk}$$

$$= \frac{1}{\sqrt{L}} \sum_{k=0}^{L-1} \left(\sum_{m=0}^{L-1} X[m] Y[k-m] \right) e^{i\omega_0 nk}$$

$$= \frac{1}{\sqrt{L}} \sum_{m=0}^{L-1} \sum_{n=0}^{L-1} X[m] Y[k-m] \left(e^{i\omega_0 nm} e^{i\omega_0 n(k-m)} \right)$$

$$= \sum_{m=0}^{L-1} X[m] e^{i\omega_0 nm} \left(\frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} Y[k-m] e^{i\omega_0 n(k-m)} \right)$$

$$= \left(\sum_{m=0}^{L-1} X[m]e^{i\omega_0 km}\right) y[n]$$
$$= \sqrt{L}x[n]y[n]$$

Definition of \mathcal{DFT}^{-1}

Definition of *

Exponential math

Rearrange

Definition of \mathcal{DFT}^{-1}

Definition of \mathcal{DFT}^{-1}