

# Discrete convolution

And how to calculate it with the FFT

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# Discrete convolution

$$h(t) * x(t) = \int_{-\infty}^{\infty} h(\tau) \cdot x(t-\tau) \cdot d\tau$$

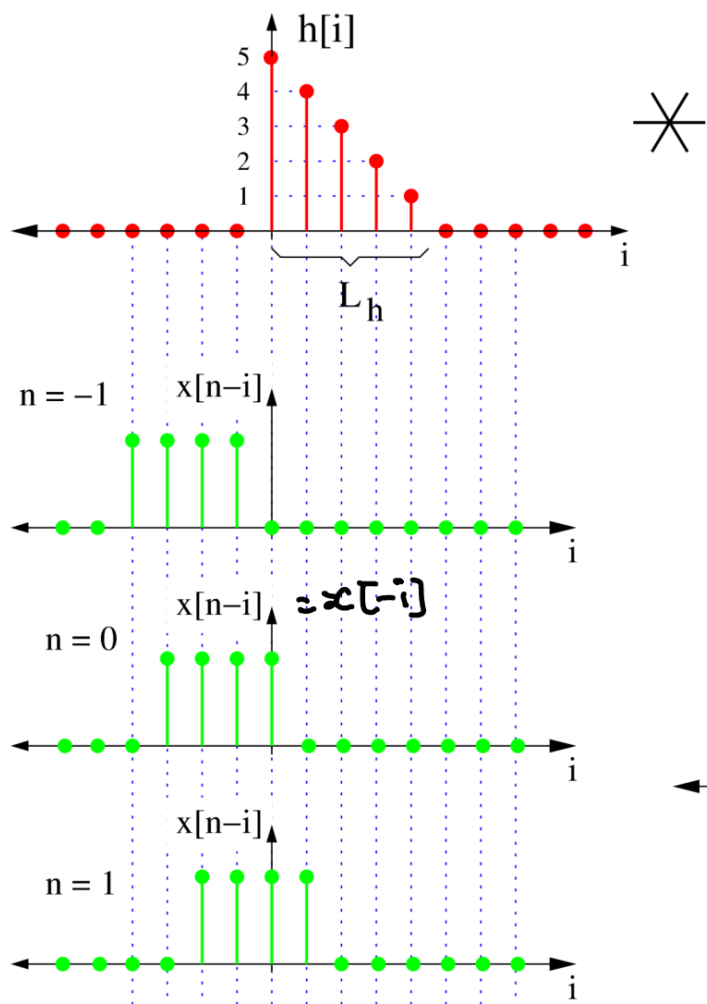
Definition:

$$h[n] * x[n] = \sum_{i=-\infty}^{\infty} h[i]x[n-i]$$

Properties:

- Commutative:  $x[n] * h[n] = h[n] * x[n]$
- Associative:  $[x[n] * h_1[n]] * h_2[n] = x[n] * [h_1[n] * h_2[n]]$
- Distributive:  $x[n] * [h_1[n] + h_2[n]] = x[n] * h_1[n] + x[n] * h_2[n]$

# Discrete convolution example

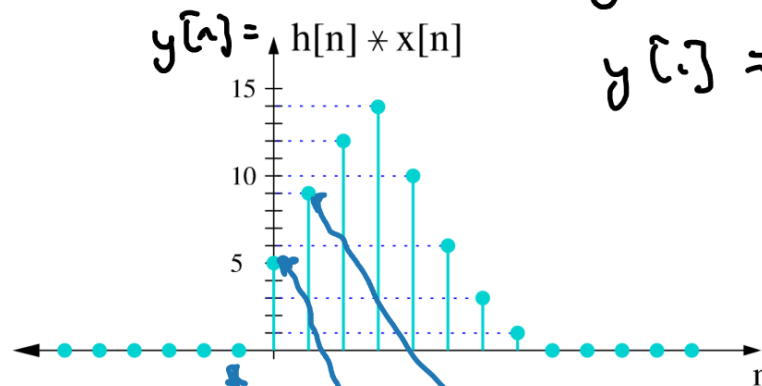


$$y[n] = h[n] * x[n] = \sum_{i=-\infty}^{\infty} h[i]x[n-i]$$

$$y[-1] = \sum_{i=-\infty}^{\infty} h[i] \cdot x[-1-i] = 0$$

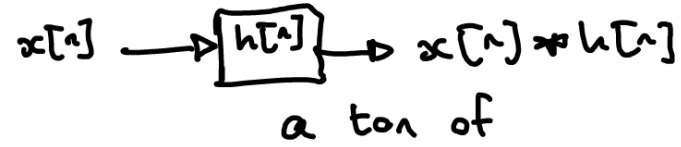
$$y[0] = \sum_{i=-\infty}^{\infty} h[i] \cdot x[-i] = 5$$

$$y[1] = 4 + 5 = 9$$



# Why and how?

- Why this obsession with convolution?



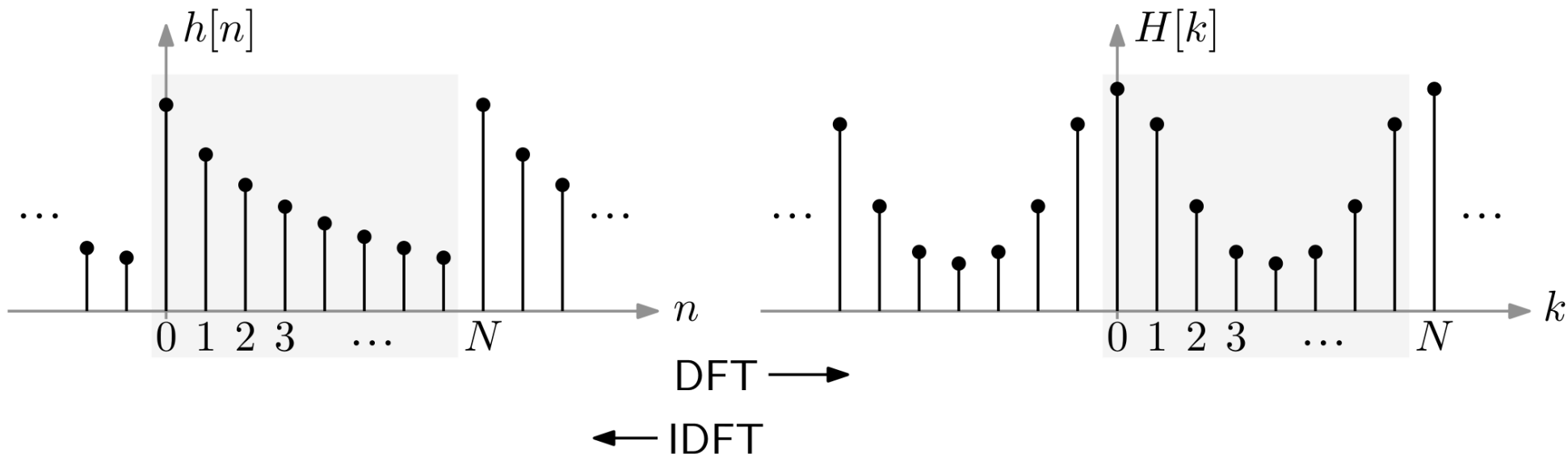
- Direct computation of discrete convolution takes roughly  ~~$N^2$~~  multiplications

- Can we do this using the FFT?

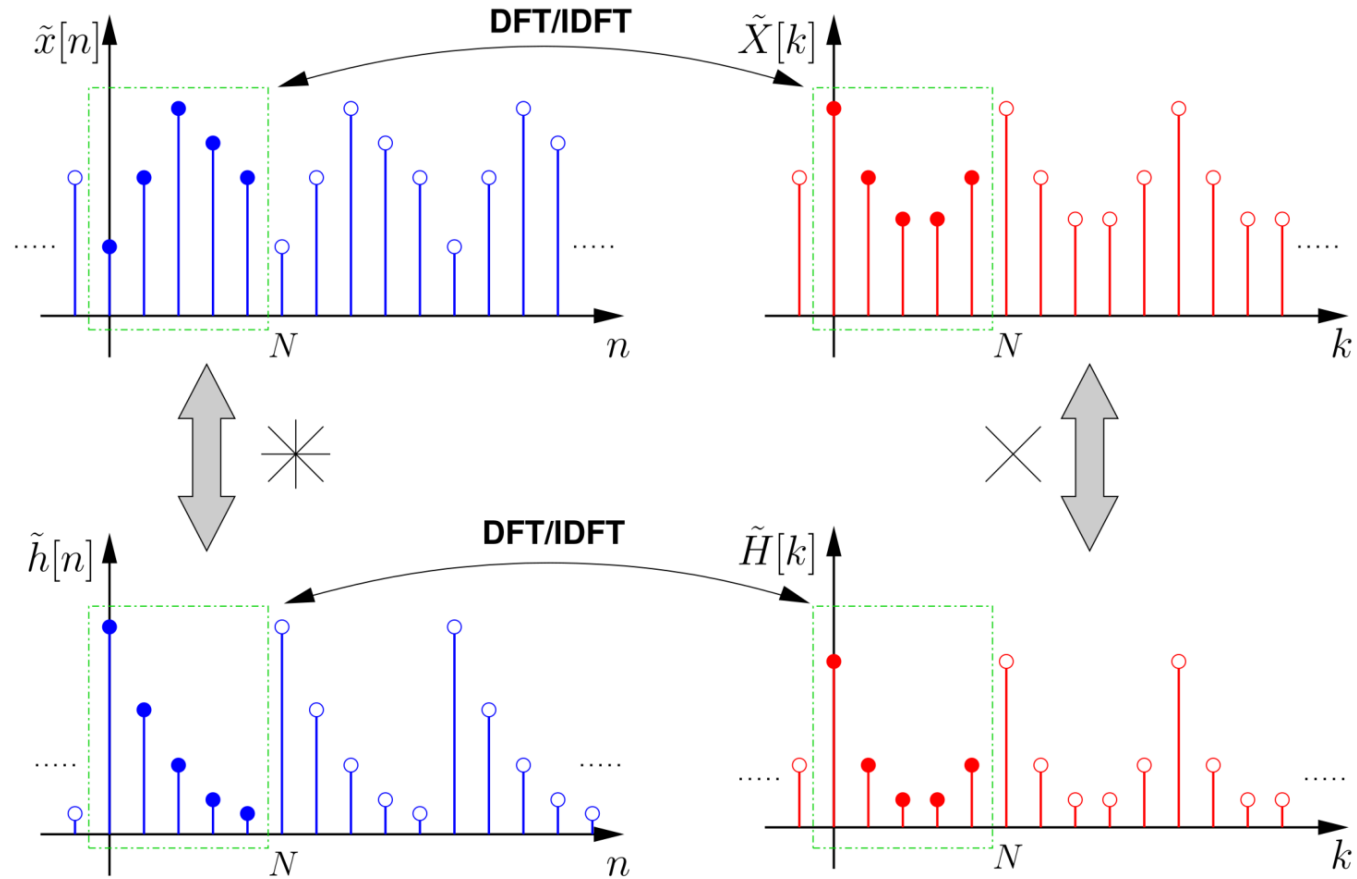
$$h(t) * x(t) \Leftrightarrow H(f) \cdot X(f)$$

- Calculate  $H[k]$  and  $X[k]$
- Multiply in frequency domain
- Take  $N$ -point IDFT
- Would be more efficient:  $H[k]$ ,  $X[k]$ , and IDFT each take  $N/2 \log_2 N$  mults
- But the following is unfortunately not actually true:

$$x[n] * h[n] \not\Leftrightarrow X[k] \cdot H[k]$$



# Discrete convolution and the DFT

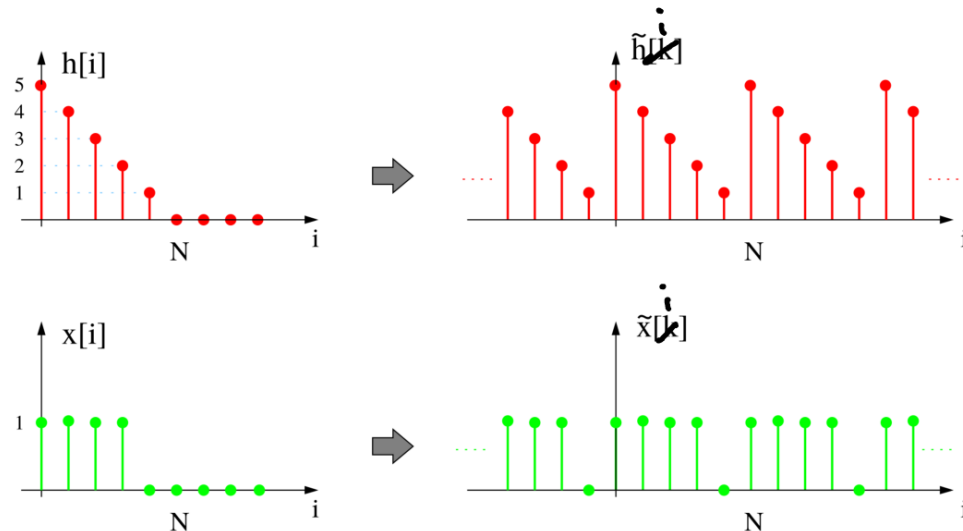


# So what happens in time domain when we multiply DFTs?

Circular convolution:

$$H[k] \cdot X[k] \iff h[n] \circledast_N x[n] = \sum_{i=0}^{N-1} h[i] \tilde{x}[n-i]$$
$$= \sum_{i=0}^{N-1} x[i] \tilde{h}[n-i]$$

where  $\tilde{x}[n]$  and  $\tilde{h}[n]$  are periodic extensions of  $x[n]$  and  $h[n]$ :



Multiplication in the time-domain leads to circular convolution in the frequency domain:

$$\text{DFT}\{x[n]y[n]\} = \frac{1}{N}\text{DFT}\{x[n]\} \circledast_N \text{DFT}\{y[n]\}$$

Multiplication in the frequency domain leads to circular convolution in the time-domain:

$$\text{DFT}\{x[n] \circledast_N y[n]\} = \text{DFT}\{x[n]\} \cdot \text{DFT}\{y[n]\}$$



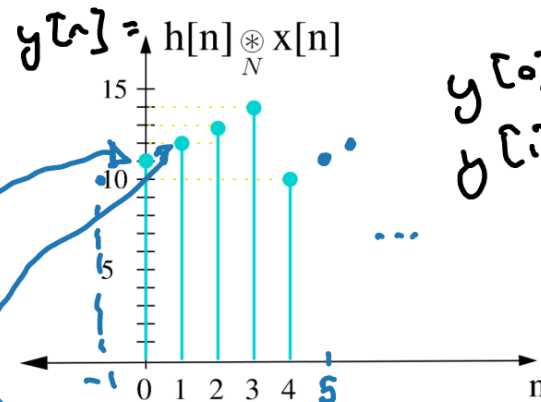
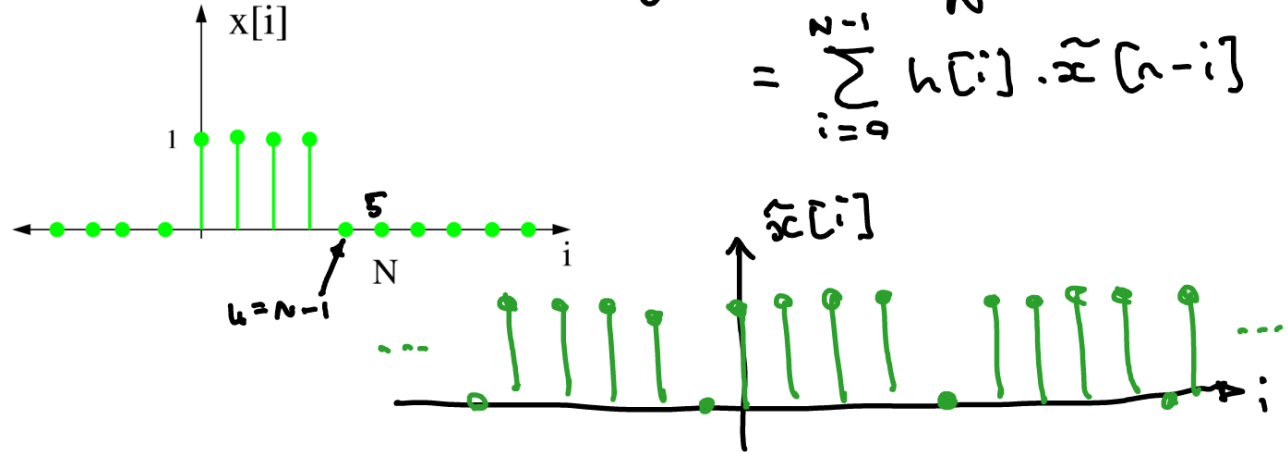
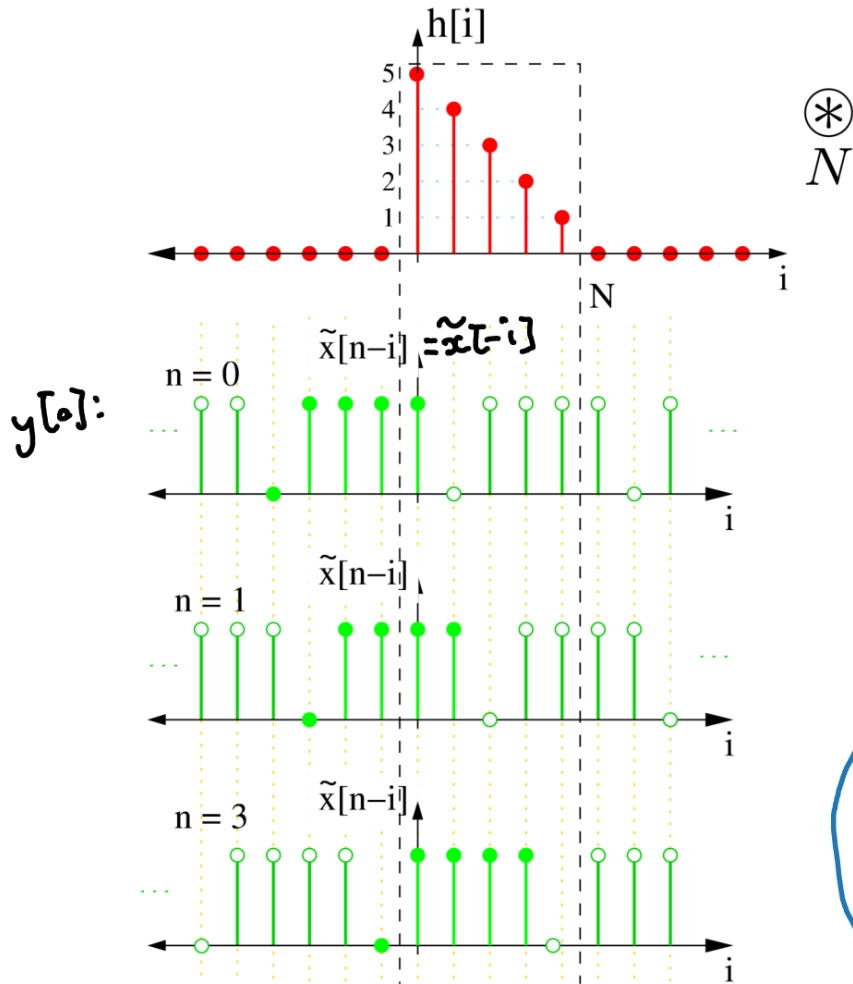
# Proof that multiplying DFTs match circular convolution in time

Proakis and Manolakis (2007, Sec. 7.2.2)

# Circular convolution example

$$y[n] = h[n] \circledast_N x[n]$$

$$= \sum_{i=0}^{N-1} h[i] \cdot \tilde{x}[n-i]$$



$$y[0] = 5 + 0 + 3 + 2 + 1 = 11$$

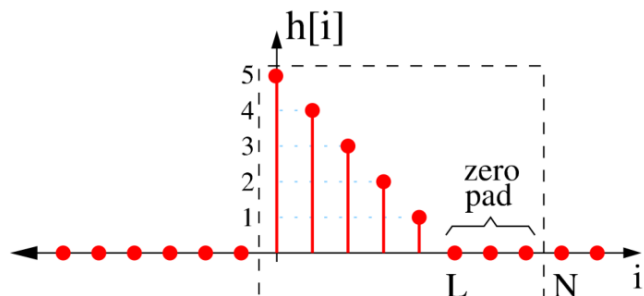
$$y[1] = 5 + 4 + 0 + 2 + 1 = 12$$

# Zero padding: Discrete convolution via circular convolution

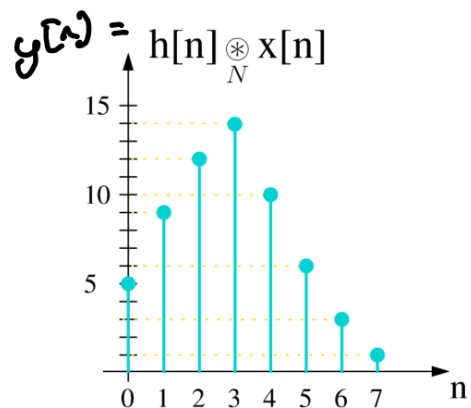
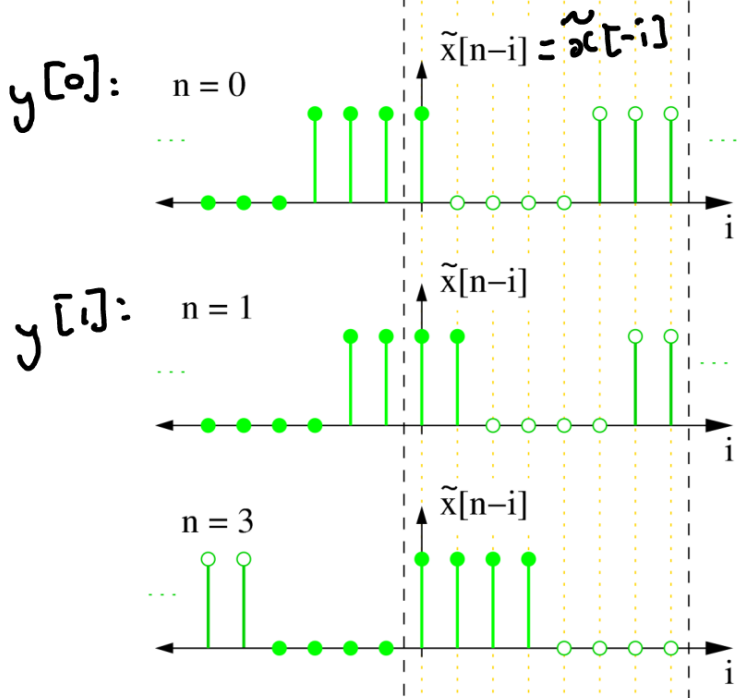
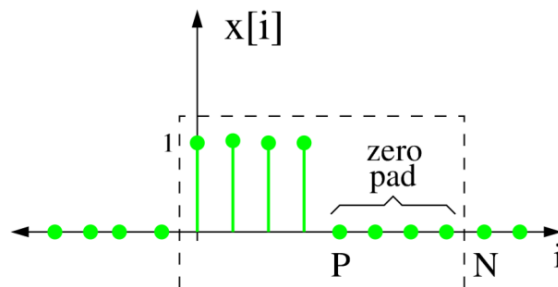
- Signal  $x[n]$  has length  $L$  and signal  $h[n]$  has length  $P$
- Zero pad both to length  $N$  so that  $N \geq L + P - 1$
- Calculate  $N$ -point DFT for both
- Multiply DFTs
- Calculate the inverse DFT: Result is the discrete convolution (not circular)

# Zero padding example

- $L = 5, P = 4$
- Zero pad to  $N \geq L + P - 1 = 8$   
Choose  $N = 8$



$\otimes$   
 $N$



$$y[0] = 5$$

$$y[1] = 5 + 4 = 9$$

# Discrete convolution example

$$x[n] = \{ \underset{\uparrow}{1} \quad 2 \quad 3 \quad 1 \} \quad y[n] = \{ \underset{\uparrow}{2} \quad 3 \quad 1 \quad 2 \}$$

What is  $x[n] * y[n]$ ?

$$x[n] * y[n] = \{ 0 \quad \underset{\uparrow}{2} \quad 7 \quad 13 \quad 15 \quad 10 \quad 7 \quad 2 \}$$

# Circular convolution example

$$x[n] = \{ \underset{\uparrow}{1} \quad 2 \quad 3 \quad 1 \} \quad y[n] = \{ \underset{\uparrow}{2} \quad 3 \quad 1 \quad 2 \}$$

What is  $x[n] \underset{N}{\circledast} y[n]$ ?

$$w[n] = x[n] \underset{4}{\circledast} y[n]$$

$$\begin{aligned} w[0] &= 12 \\ w[1] &= 14 \\ w[2] &= 15 \\ w[3] &= 15 \\ w[4] &= w[0] = 12 \\ &\vdots \end{aligned}$$