



Linear Constant Coefficient Difference Equations (LCCDE)

ECE 3640 Discrete-Time Signals and Systems

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Linear Constant Coefficient Difference Equations (LCCDE)

$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m] \quad (\text{both sides look like convolution})$$



 (output)  (input)

To compute output, solve for $y[n]$.

$$a[0]y[n] = \sum_{m=0}^M b[m]x[n-m] - \sum_{k=1}^N a[k]y[n-k] \quad (\text{recursive formula})$$

 (usually $a[0] = 1$)

Note that current output $y[n]$ depends on:

- Current input: $x[n]$
 - Past inputs: $x[n-1], \dots, x[n-M]$ 
 - Past outputs: $y[n-1], \dots, y[n-N]$ 
- (initial conditions)

Linear Constant Coefficient Difference Equations (LCCDE)

Forward recursion leads to causal output signal.

$$a[0]y[n] = \sum_{m=0}^M b[m]x[n-m] - \sum_{k=1}^N a[k]y[n-k]$$

Backward recursion leads to anti-causal output signal.

$$a[N]y[n-N] = \sum_{m=0}^M b[m]x[n-m] - \sum_{k=0}^{N-1} a[k]y[n-k]$$

- For LCCDE to represent LTI system, need zero initial conditions.
- All LTI systems representable by difference equations.

Linear Constant Coefficient Difference Equations (LCCDE)

$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m]$$

Finite Impulse Response (FIR)

Infinite Impulse Response (IIR)

$$a[0] = 1, N = 0, M < \infty$$

$$N > 0$$

$$y[n] = \sum_{k=0}^M b[k]x[n-k]$$

$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m]$$

- No feedback
- Feed forward
- Convolution
- `conv` command in Matlab

- Feedback
- Recursive
- Difference equation
- `filter` command in Matlab

Linear Constant Coefficient Difference Equations in Matlab

$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m]$$

Finite Impulse Response (FIR)

$$\mathbf{a} = 1$$

$$\mathbf{b} = [b_0, b_1, \dots, b_M]$$

$$\mathbf{x} = [x_0, x_1, \dots, x_L]$$

$$\mathbf{y} = [y_0, y_1, \dots, y_L]$$

$$\mathbf{y} = \text{filter}(\mathbf{b}, 1, \mathbf{x})$$

$$\mathbf{y} = \text{conv}(\mathbf{b}, \mathbf{x})$$

Infinite Impulse Response (IIR)

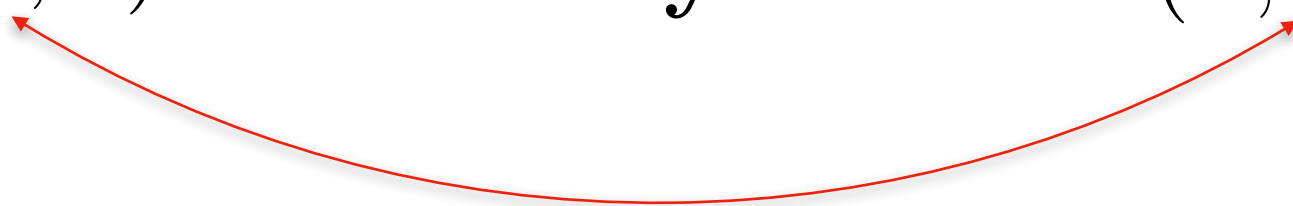
$$\mathbf{a} = [a_0, a_1, \dots, a_N]$$

$$\mathbf{b} = [b_0, b_1, \dots, b_M]$$

$$\mathbf{x} = [x_0, x_1, \dots, x_L]$$

$$\mathbf{y} = [y_0, y_1, \dots, y_L]$$

$$\mathbf{y} = \text{filter}(\mathbf{b}, \mathbf{a}, \mathbf{x})$$



Example: Accumulator

Consider an LTI system having impulse response $u[n]$.
The output is

$$\begin{aligned} y[n] &= x[n] * u[n] \\ &= \sum_{k=-\infty}^{\infty} u[k]x[n-k] = \sum_{k=0}^{\infty} x[n-k] \\ &= \sum_{k=-\infty}^{\infty} x[k]u[n-k] = \sum_{k=-\infty}^n x[k] \quad (\text{accumulation}) \end{aligned}$$

The output $y[n]$ is the accumulated sum of the current and all past inputs.

Example: Accumulator

Consider an LTI system having impulse response $u[n]$.
The output is

$$y[n] = x[n] * u[n]$$

$$= \sum_{k=-\infty}^{\infty} u[k]x[n-k] = \sum_{k=0}^{\infty} x[n-k]$$

$$= \sum_{k=-\infty}^{\infty} x[k]u[n-k] = \sum_{k=-\infty}^n x[k] \quad (\text{accumulation})$$

The output $y[n]$ is the accumulated sum of the current and all past inputs.

$$y[n] = \sum_{k=0}^{\infty} x[n-k]$$

$$a[0] = 1, N = 0$$

$$b[n] = 1 \quad \text{for all } n \geq 0, M = \infty$$

$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m]$$

Example: Accumulator


$$y[n] = \sum_{k=-\infty}^n x[k]$$

$$= x[n] + \sum_{k=-\infty}^{n-1} x[k]$$

$$= x[n] + y[n-1]$$

$$y[n] - y[n-1] = x[n]$$

(diff. eq. representation
is more compact)



$$\sum_{k=0}^N a[k]y[n-k] = \sum_{m=0}^M b[m]x[n-m]$$

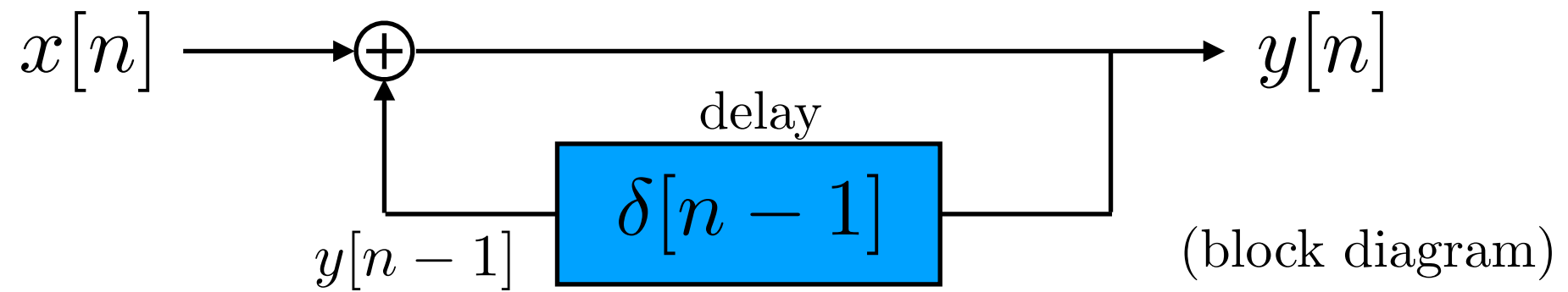
$$a[0] = 1, a[1] = -1, N = 1$$

$$b[0] = 1, M = 0$$

$$y = \text{filter}(1, [1, -1], x) \quad (\text{Matlab command})$$

Example: Accumulator

$$y[n] = x[n] + y[n - 1]$$



Example: Moving Average

(impulse response) $b[n] = \begin{cases} \frac{1}{L}, & n = 0, 1, \dots, L-1 \\ 0, & \text{otherwise} \end{cases}$

$$= \left(\frac{1}{L}\right) (u[n] - u[n-L])$$

$$= \left(\frac{1}{L}\right) (\delta[n] - \delta[n-L]) * u[n]$$

$$y[n] = \frac{1}{L} \sum_{k=0}^{L-1} x[n-k]$$

This representation uses L adds.

$$N = 0, \mathbf{a} = 1$$

$$M = L-1, \mathbf{b} = \left[\frac{1}{L}, \dots, \frac{1}{L} \right]$$

L length

This means no feedback, i.e. FIR.

`y = filter(b,a,x)` in Matlab

`y = conv(b,x)` in Matlab

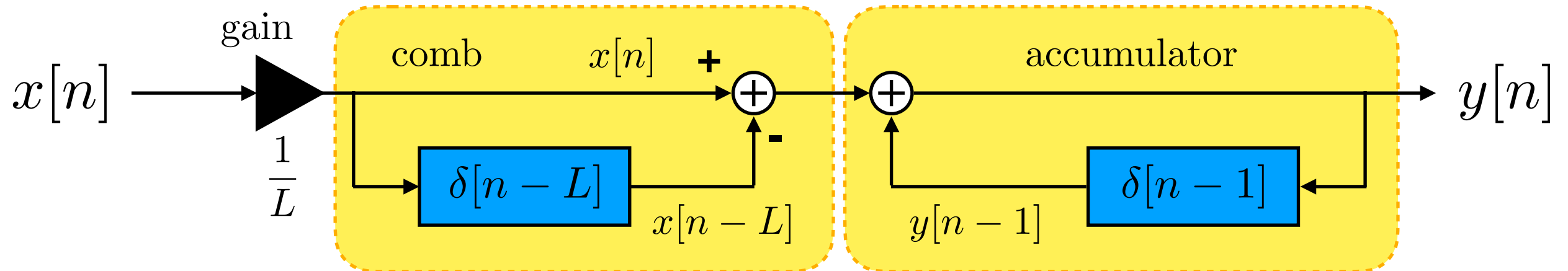
(What is the difference?)

Example: Moving Average

$$b[n] = \begin{cases} \frac{1}{L}, & n = 0, 1, \dots, L-1 \\ 0, & \text{otherwise} \end{cases}$$

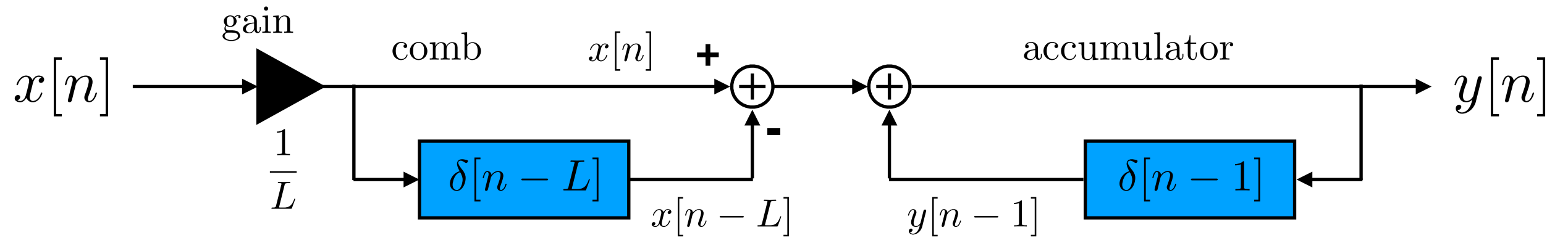
$$= \left(\frac{1}{L} \right) (u[n] - u[n-L])$$

$$= \left(\frac{1}{L} \right) (\delta[n] - \delta[n-L]) * u[n]$$



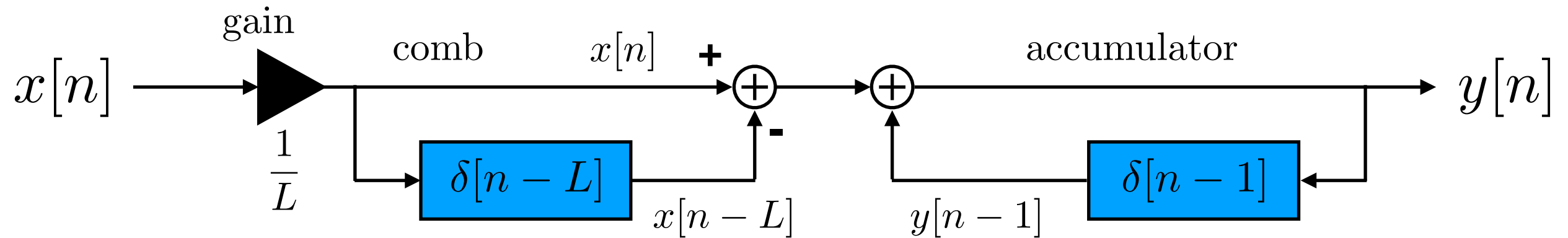
This representation uses less arithmetic: 2 adds.

Example: Moving Average



$$y[n] = y[n-1] + \frac{1}{L} (x[n] - x[n-L])$$

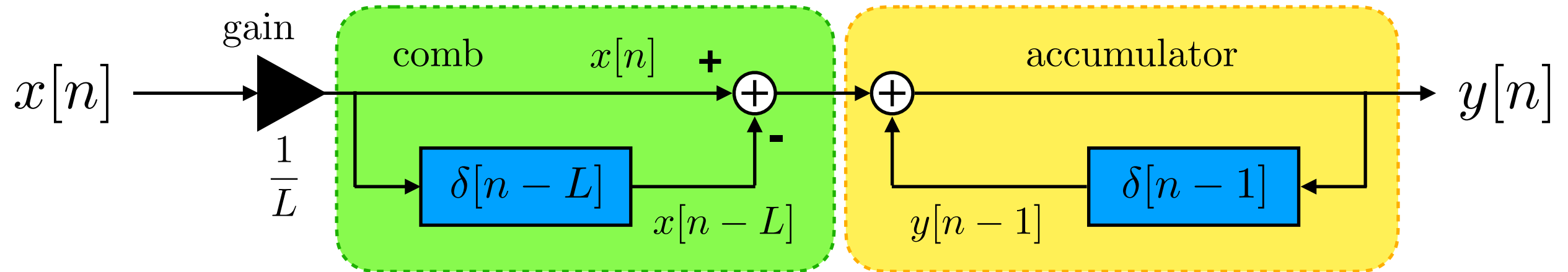
Example: Moving Average



$$y[n] = y[n-1] + \frac{1}{L} (x[n] - x[n-L])$$

$$y[n] - y[n-1] = \frac{1}{L} (x[n] - x[n-L])$$

Example: Moving Average



$$y[n] = y[n-1] + \frac{1}{L} (x[n] - x[n-L])$$

$$y[n] - y[n-1] = \frac{1}{L} (x[n] - x[n-L])$$

$$N = 1, \quad \mathbf{a} = [1, -1]$$

$$M = L - 1,$$

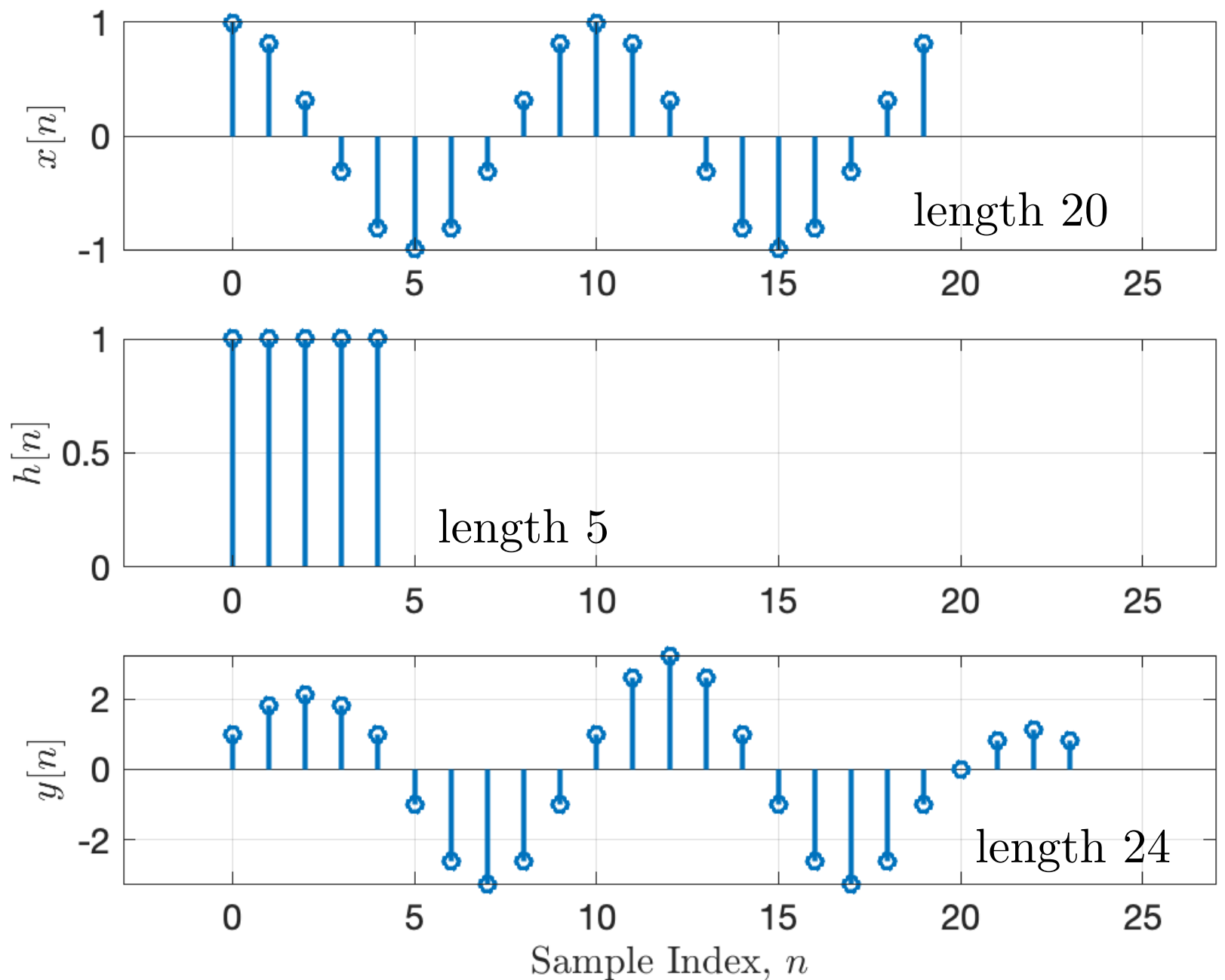
$$\mathbf{b} = \left[\frac{1}{L}, 0, \dots, 0, -\frac{1}{L} \right]$$

$L - 1$ zeros
 $L + 1$ length

`y = filter(b,a,x)` in Matlab

Convolution and Filtering in Matlab

```
x = cos(2*pi*0.1*[0:19]);  
h = ones(1,5);  
y = conv(h,x);
```

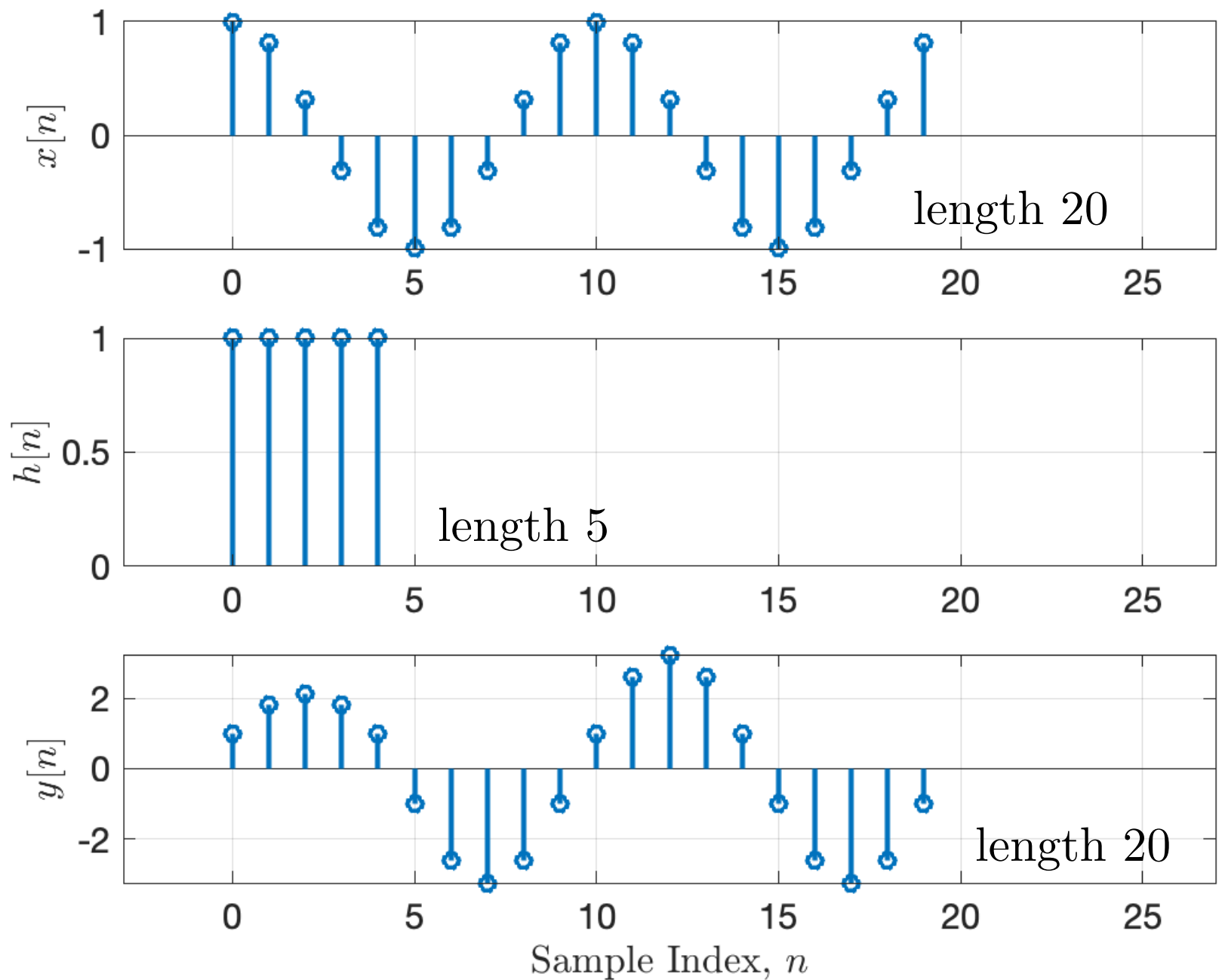


Length of convolution
result:

$$L_y = L_x + L_h - 1$$

Convolution and Filtering in Matlab

```
x = cos(2*pi*0.1*[0:19]);  
h = ones(1,5);  
y = filter(h,1,x);
```



Length of filtering
result:

$$L_y = L_x$$

Convolution and Filtering in Matlab

```
x = [cos(2*pi*0.1*[0:19]), zeros(1,4)];
```

```
h = ones(1,5);
```

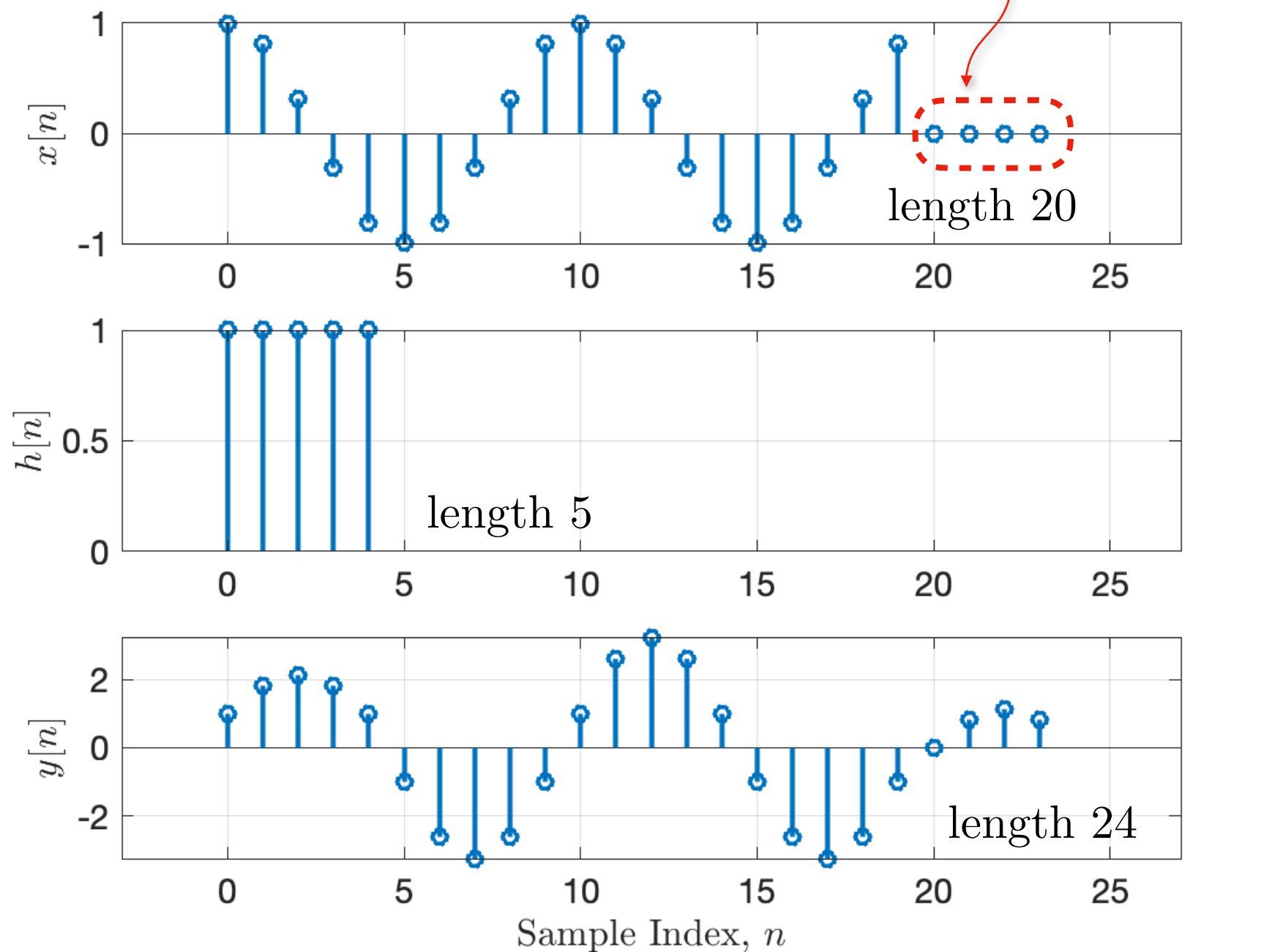
```
y = filter(h,1,x);
```

Make filtering perform
convolution by zero
padding the input.

Zero pad by $L_h - 1$.

Length of convolution
result:

$$L_y = L_x + L_h - 1$$



Q: Given a difference equation, how do you compute the impulse response?

$$a[0]y[n] = \sum_{m=0}^M b[m]x[n-m] - \sum_{k=1}^N a[k]y[n-k]$$

A: Put in an impulse and see what comes out.

$$\mathbf{a} = [a_0, a_1, \dots, a_N]$$

$$\mathbf{b} = [b_0, b_1, \dots, b_M]$$

$$\boldsymbol{\delta} = [1, 0, 0, \dots, 0] \quad (\text{Kronecker delta sequence})$$

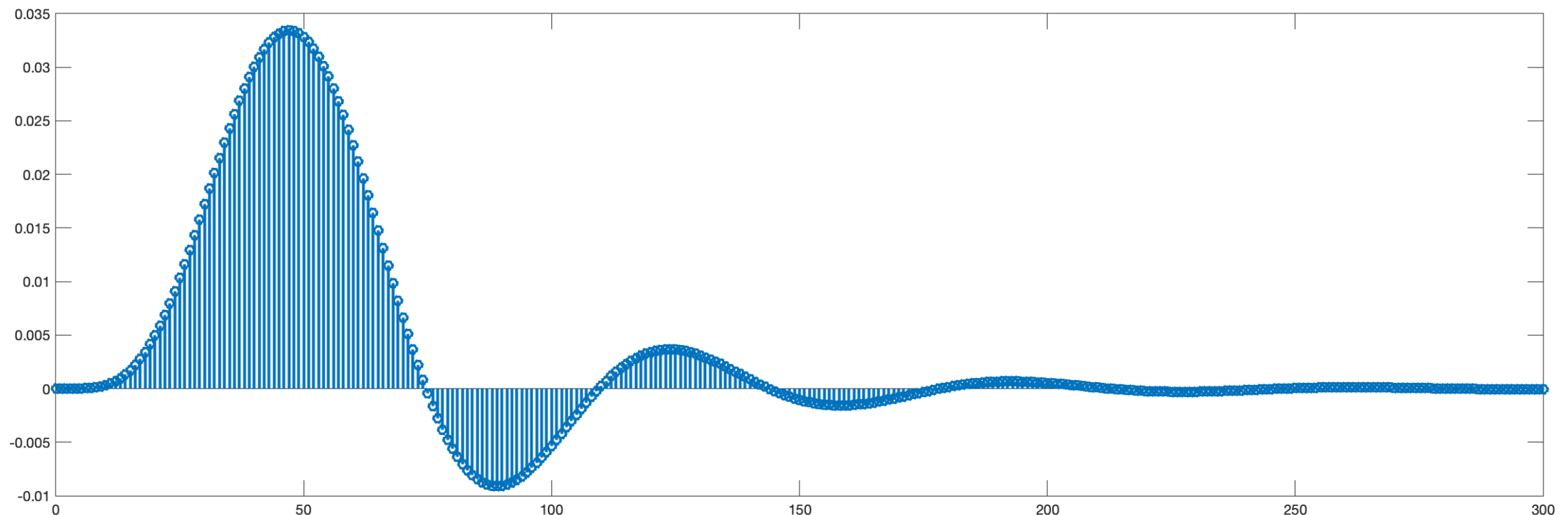
$$\mathbf{h} = [h_0, h_1, \dots, h_L] \quad (\text{Impulse response})$$

$$\mathbf{h} = \text{filter}(\mathbf{b}, \mathbf{a}, \boldsymbol{\delta})$$

Note: This only computes the first L samples of the impulse response. To compute the entire impulse response sequence, we will need the z -transform.

Butterworth Filter Impulse Response

```
[b,a]=butter(6,0.03); % LPF  
h = filter(b,a,[1,zeros(1,300)]);  
stem([0:300],h,'LineWidth',2);
```



- This gives the first 301 samples of the impulse response, which continues forever.
- At some point the impulse response amplitude becomes negligible.
- Truncation gives an approximate FIR filter.
- Difference equation realization more computationally efficient than convolution with truncated impulse response.