

EE16B

Designing Information Devices and Systems II

Lecture 12B

Discrete Signals and Systems
LTI Systems, Convolution sum

Intro

- Last time:
 - Sampling Theorem
 - Aliasing
 - Discrete Signals
- Today
 - Discrete systems

Complex Frequencies

- Sinusoids are sums of left and right rotating complex exponentials

$$2 \cos(\omega t) = e^{j\omega t} + e^{-j\omega t}$$

“Positive” and “Negative” frequencies

Discrete frequencies with period N:

$$y[n] = e^{j2\pi n/N}$$

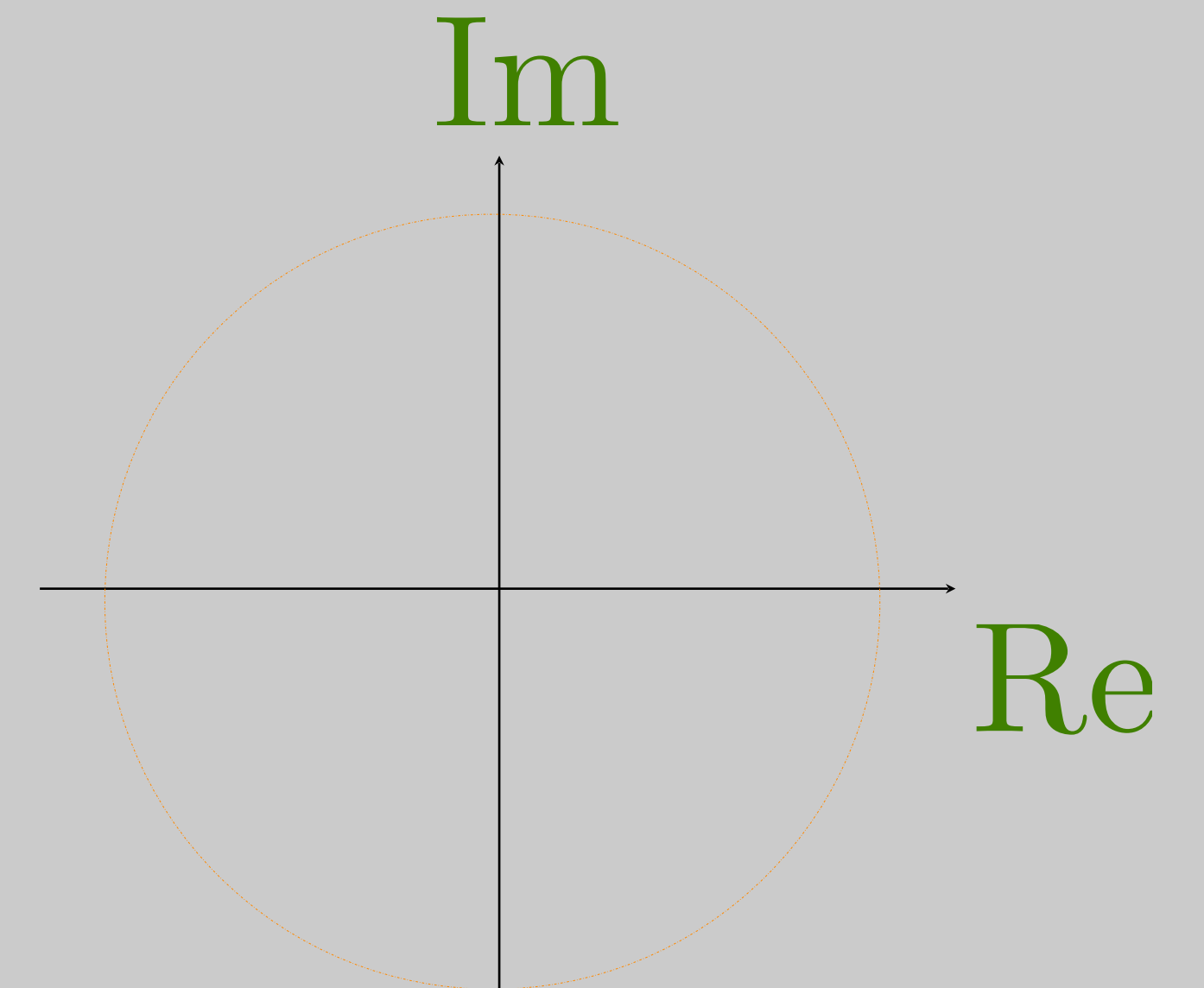
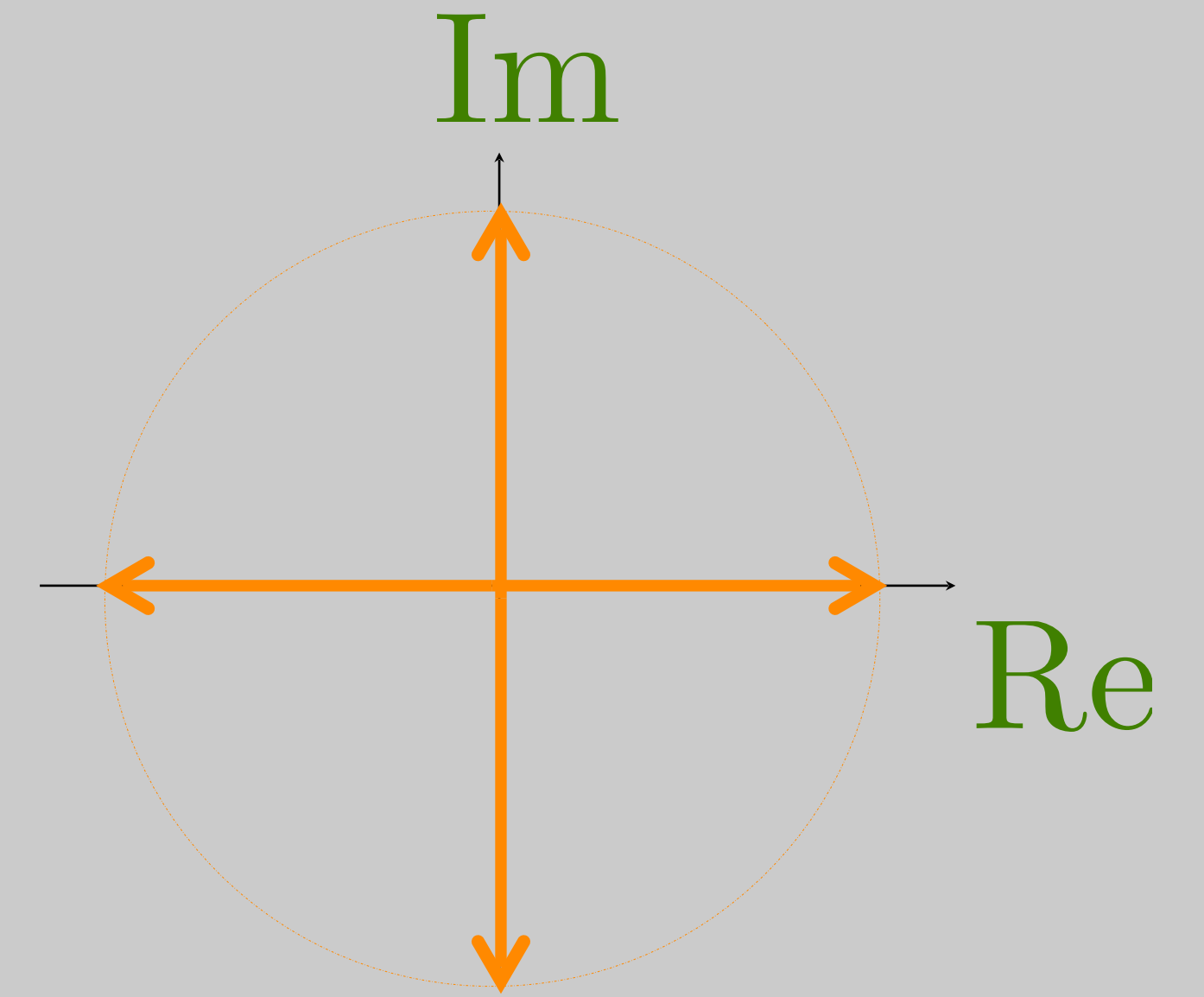
$$W_N \triangleq e^{j2\pi/N} \Rightarrow y[n] = W_N^n$$

Complex Frequencies

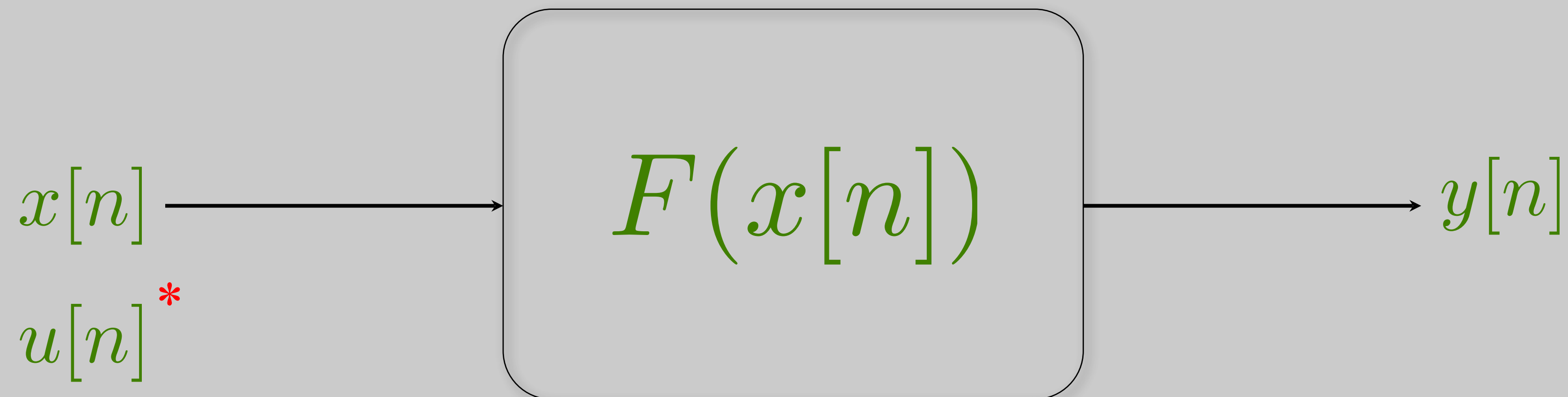
$$W_N \triangleq e^{j2\pi/N} \Rightarrow y[n] = W_N^n$$

- $N = 4$ $y[n] = W_4^n$

- $N = 6$, neg. freq. $y[n] = W_6^{-n}$



Discrete Time Systems



- What Properties?
 - Causality
 - Linearity
 - Stability
 - Time/shift invariance

*WARNING: Going to interchange $x[n]$ and $u[n]$ as inputs

$\vec{x}[n]$ will be a state, not input

$U[n]$ is unit step, not to be confused with $u[n]$

Properties of D.T. Systems

- Causality:
 - $y[n_0]$ depends only on $x[n]$ for $\infty \leq n \leq n_0$

Causal?

$$\vec{x}[n + 1] = A\vec{x}[n] + Bu[n]$$

$$y[n] = C\vec{x}[n]$$

$$\vec{x}[n] = A^n \vec{x}[0] + \sum_{k=0}^{n-1} A^{n-1-k} Bu[k]$$

$$y[n] = CA^n \vec{x}[0] + \sum_{k=0}^{n-1} CA^{n-1-k} Bu[k]$$

Properties of D.T. Systems

$$y[n] = F\{x[n]\}$$

- **Linearity**

- Homogeneity: scaling the input, scales the output

$$F\{ax[n]\} = aF\{x[n]\} = ay[n]$$

Properties of D.T. Systems

$$y[n] = F\{x[n]\}$$

- Linearity

- Homogeneity / scaling: scaling the input, scales the output

$$F\{ax[n]\} = aF\{x[n]\} = ay[n]$$

- Additivity: sum of inputs \Rightarrow sum of outputs

$$F\{x_1[n] + x_2[n]\} = F\{x_1[n]\} + F\{x_2[n]\} = y_1[n] + y_2[n]$$

- Super position: Both additivity and homogeneity

Example:

$$\vec{x}[n + 1] = A\vec{x}[n] + Bu[n]$$

$$y[n] = C\vec{x}[n]$$

Linear?

$$y[n] = CA^n \vec{x}[0] + \sum_{k=0}^{n-1} CA^{n-1-k} Bu[k]$$

Properties of D.T. Systems

$$y[n] = F\{x[n]\}$$

- **BIBO Stability**

- If $x[n]$ is bounded, then $y[n]$ is bounded

$$|x[n]| < M < \infty \quad \forall n \Rightarrow |y[n]| < P < \infty \quad \forall n$$

BIBO stable?

$$y[n] = CA^n \vec{x}[0] + \sum_{k=0}^{n-1} CA^{n-1-k} Bu[k]$$

Properties of D.T. Systems

$$y[n] = F\{x[n]\}$$

- Time Invariance: Shifted input \Rightarrow shifted output

$$y[n - n_0] = F\{x[n - n_0]\}$$

Time Invariant? $\vec{x}[n + 1] = A\vec{x}[n] + Bu[n]$

$$y[n] = C\vec{x}[n]$$

$$y[n] = CA^n \vec{x}[0] + CBu[n - 1] + CABu[n - 2] + \cdots + CA^{n-1} Bu[0]$$

Linear Time Invariant Systems

- Linear Time/Shift Invariant (LTI/LSI) systems are completely characterized by their impulse response $h[n]$



$h[n]$ is the “DNA” of an LTI system

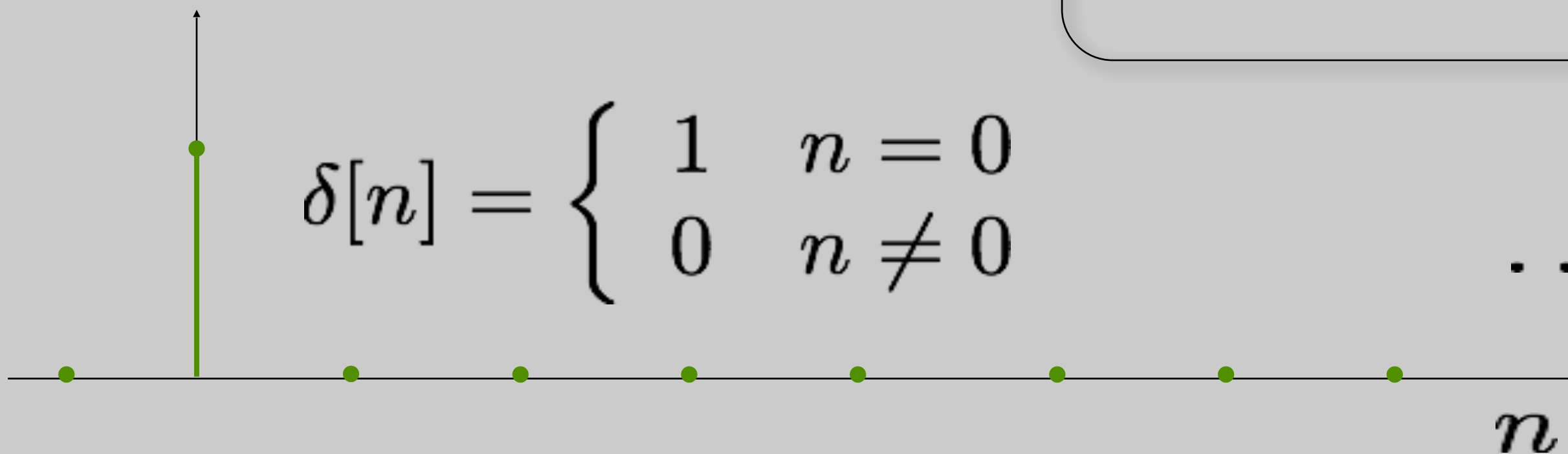
Knowing $h[n]$ is enough to find $y[n]$ for ANY $x[n]$!

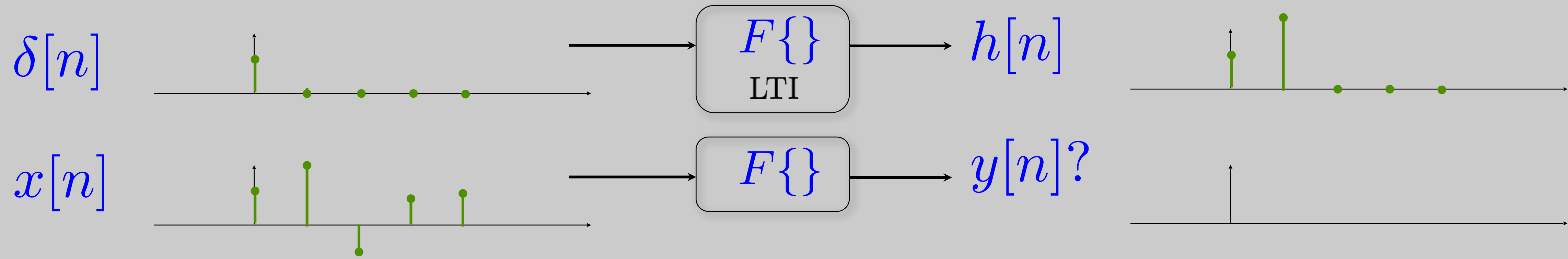
Linear Time Invariant Systems

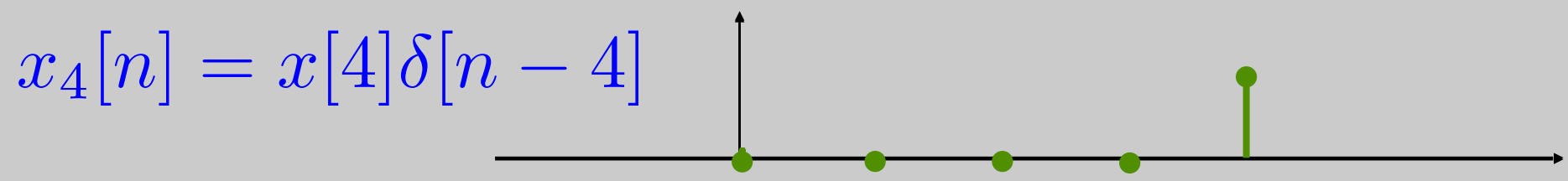
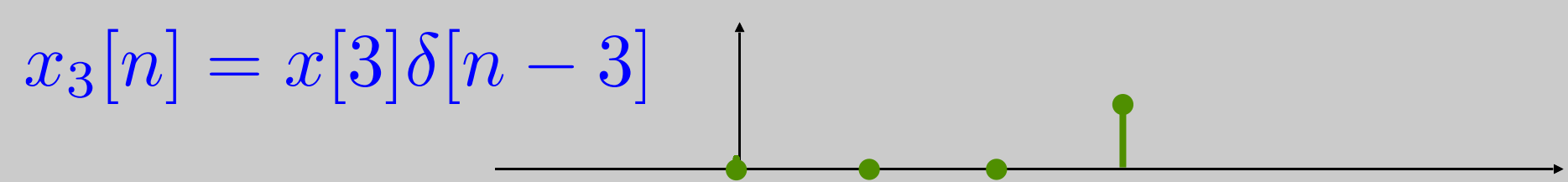
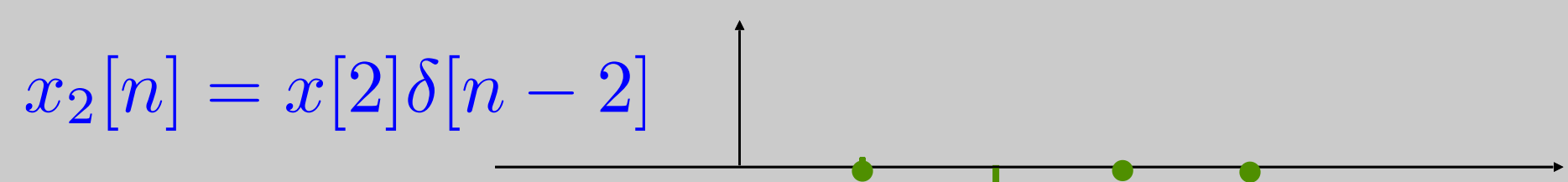
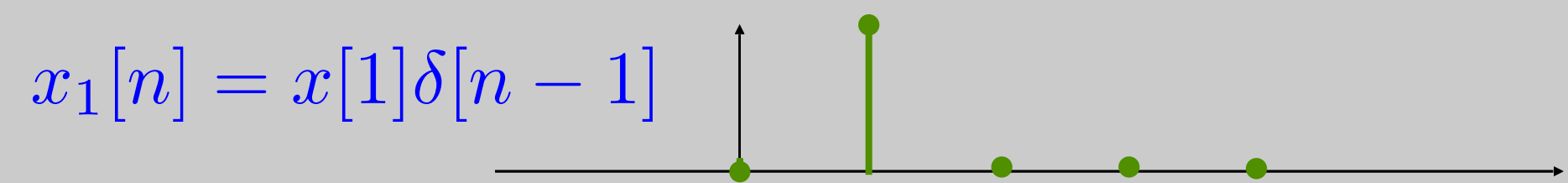
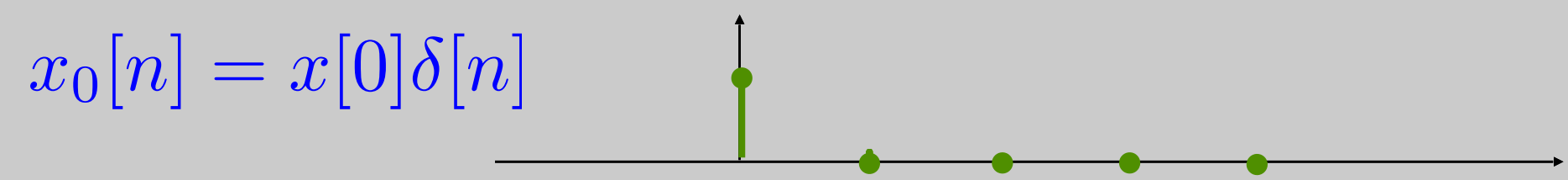
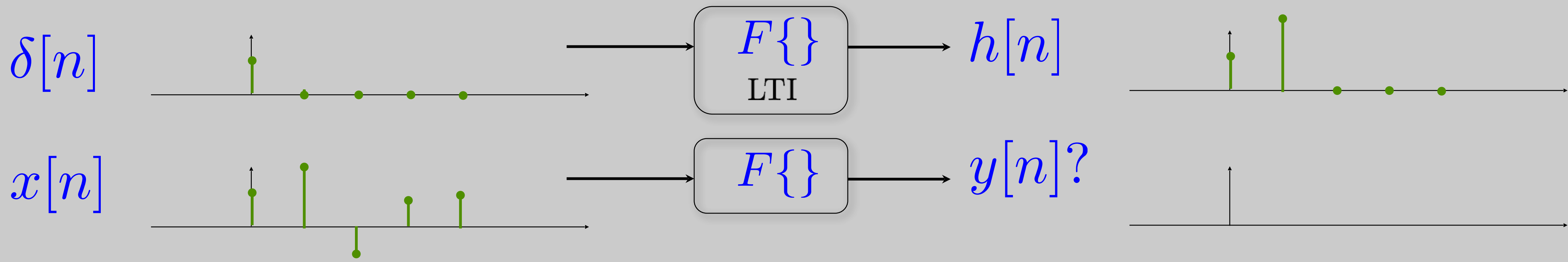
- Linear Time/Shift Invariant (LTI/LSI) systems are completely characterized by their impulse response $h[n]$



$$\delta[n] = \begin{cases} 1 & n = 0 \\ 0 & n \neq 0 \end{cases} \quad \dots$$

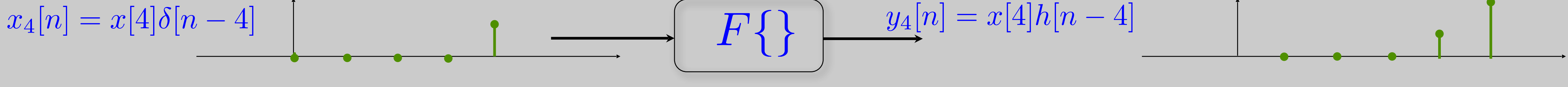
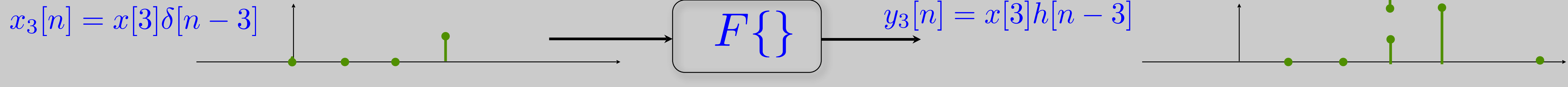
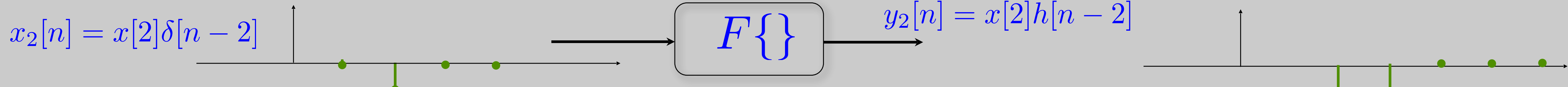
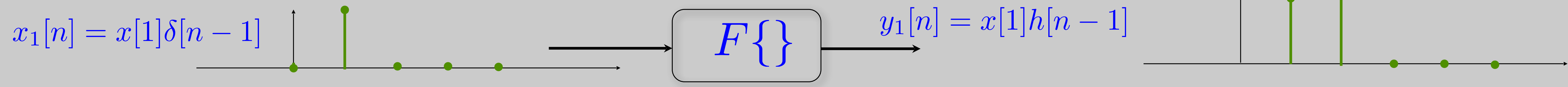
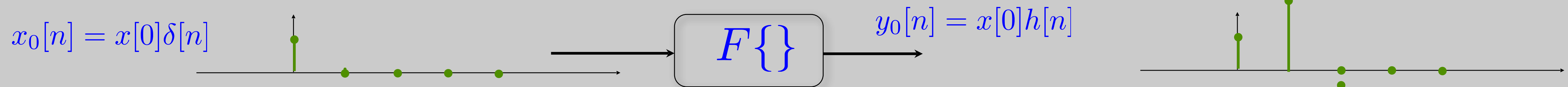
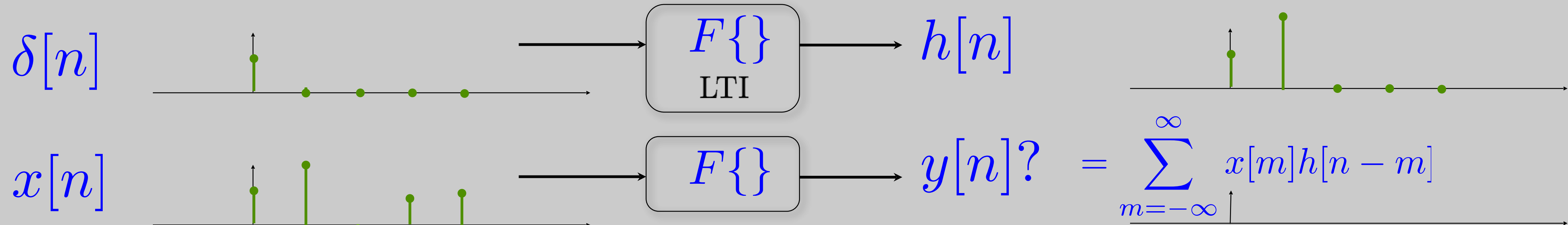






$$x[n] = \sum_{m=-\infty}^{\infty} x_m[n]$$

$$= \sum_{m=-\infty}^{\infty} x[m]\delta[n-m]$$



Linear Time Invariant Systems



- Decompose $x[n]$:

$$x[n] = \sum_{m=-\infty}^{\infty} x[m] \delta[n - m] = \begin{cases} 1 & n = m \\ 0 & n \neq m \end{cases}$$

- Compute output:

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

Convolution
sum \downarrow

Sum of weighted, delayed impulse responses!

Example:

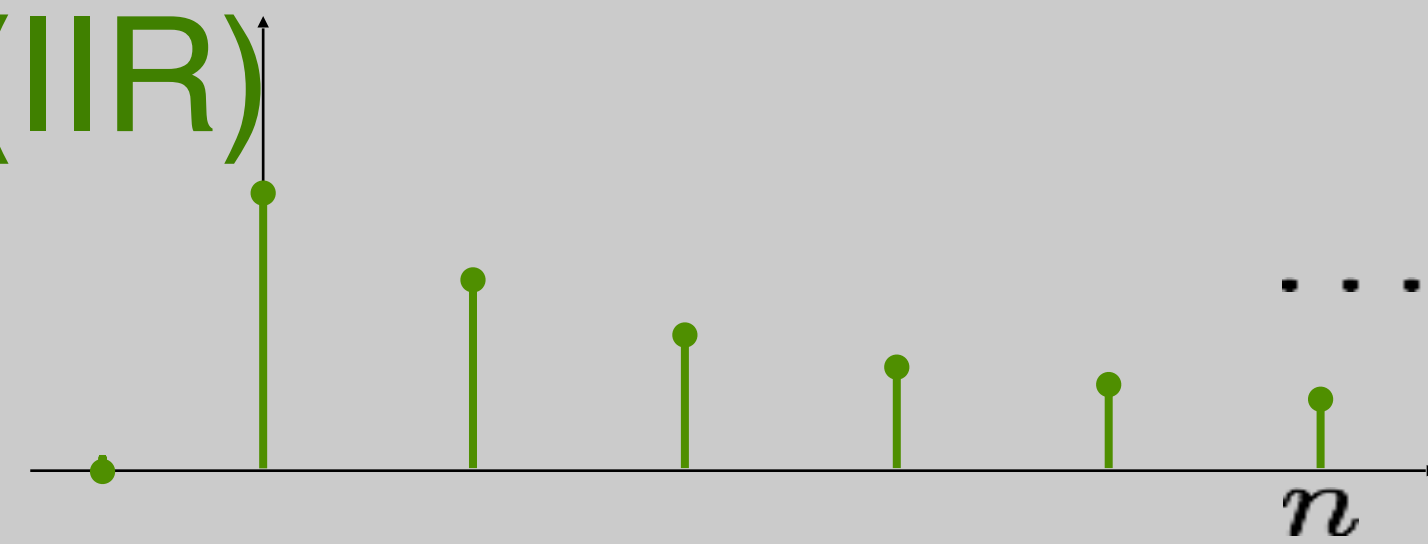
$$y[n] = ay[n - 1] + x[n]$$

$$h[n] = \begin{cases} a^n & n \geq 0 \\ 0 & n < 0 \end{cases}$$

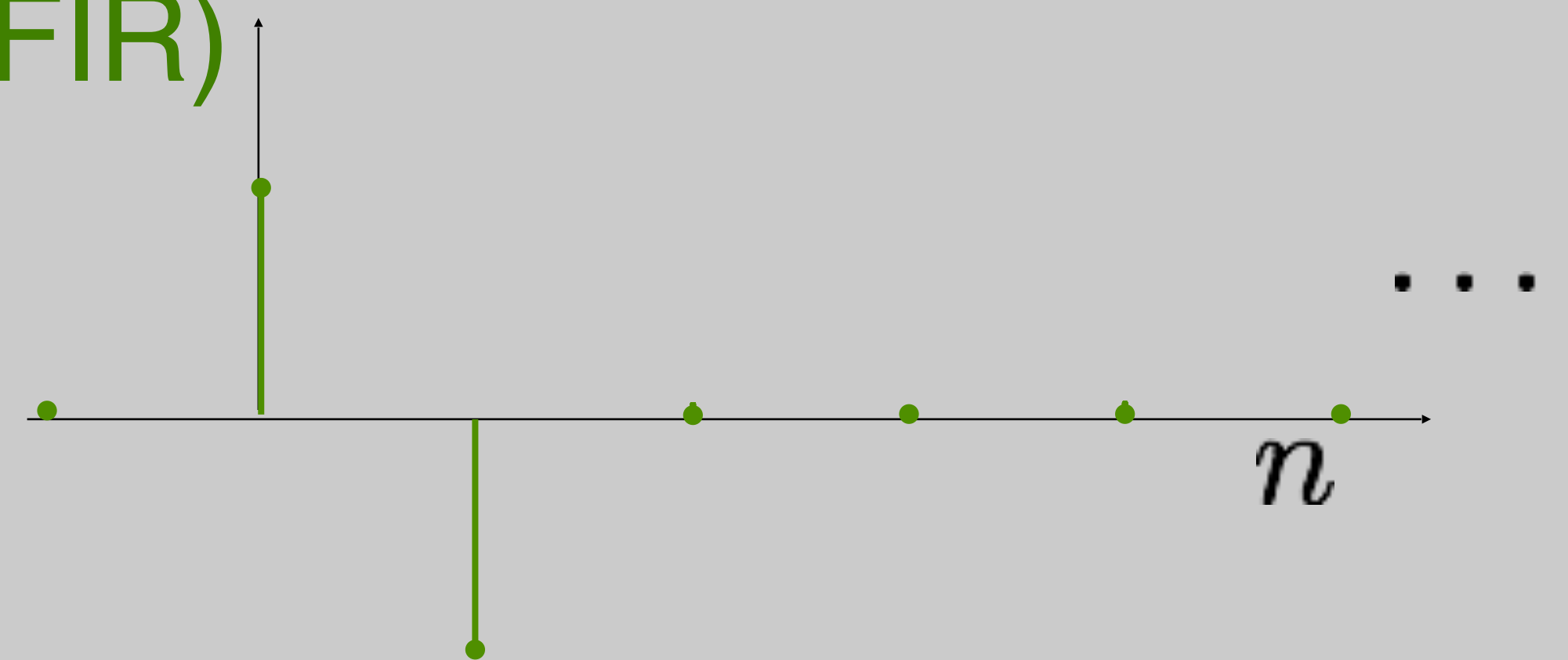
$$y[n] = x[n] - x[n - 1]$$

$$h[n] = \delta[n] - \delta[n - 1]$$

Infinite impulse response
(IIR)



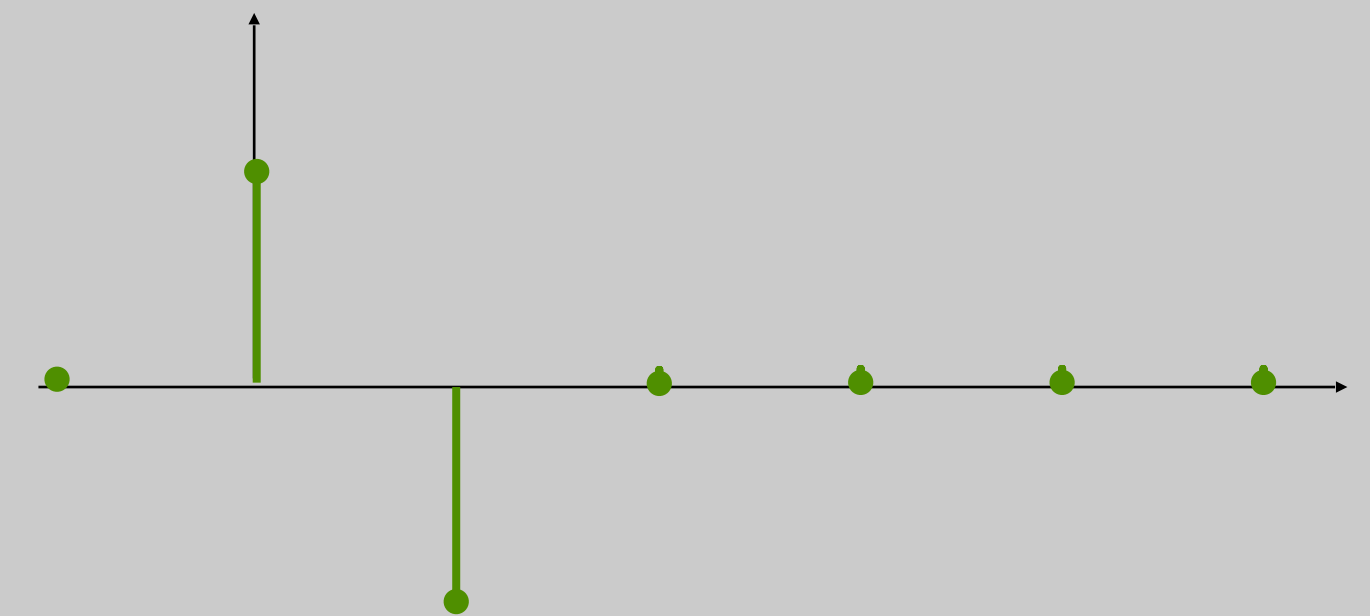
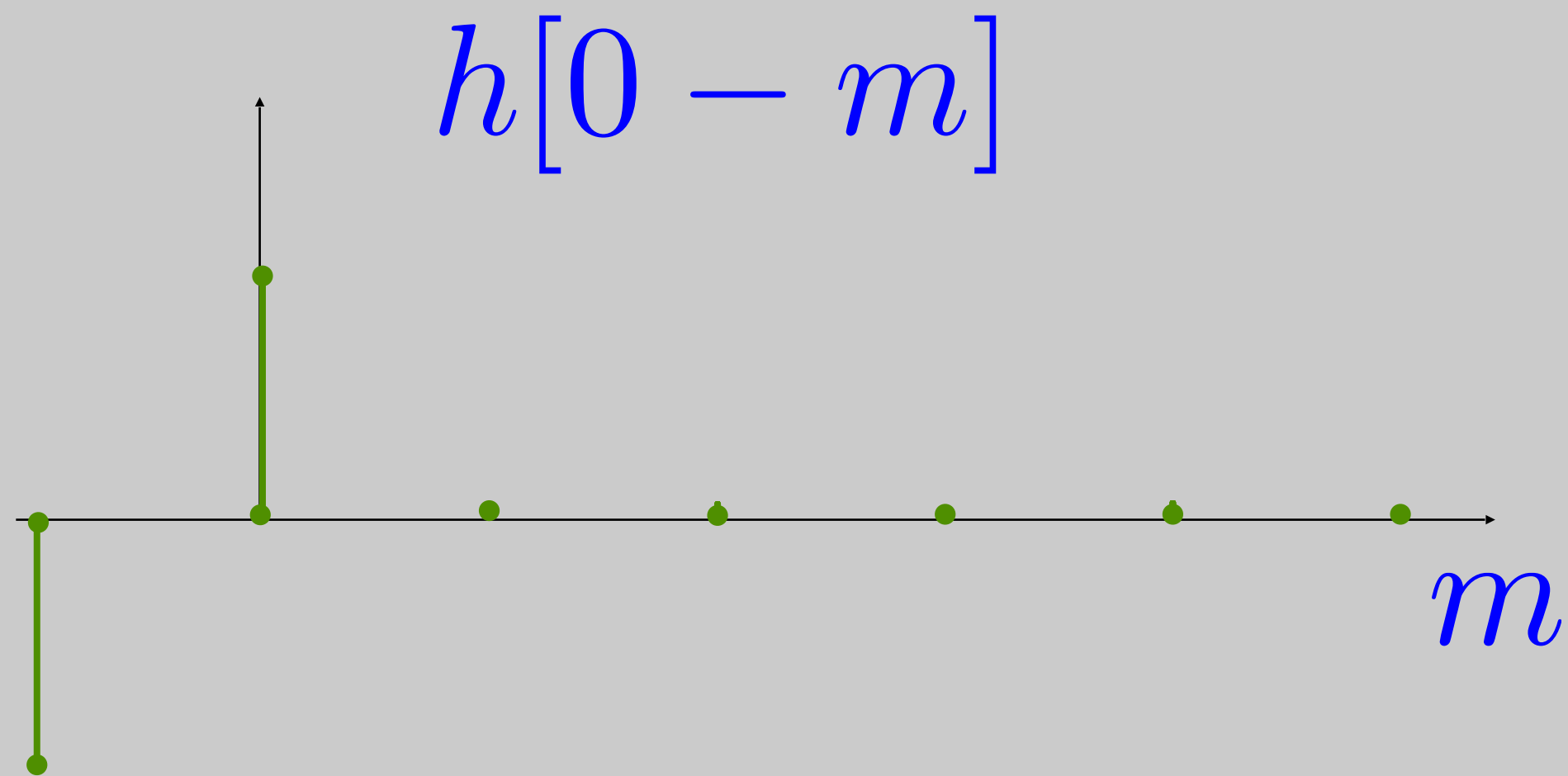
finite impulse response
(FIR)



Convolution Sum

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

- What is $h[n-m]$ for different n 's?

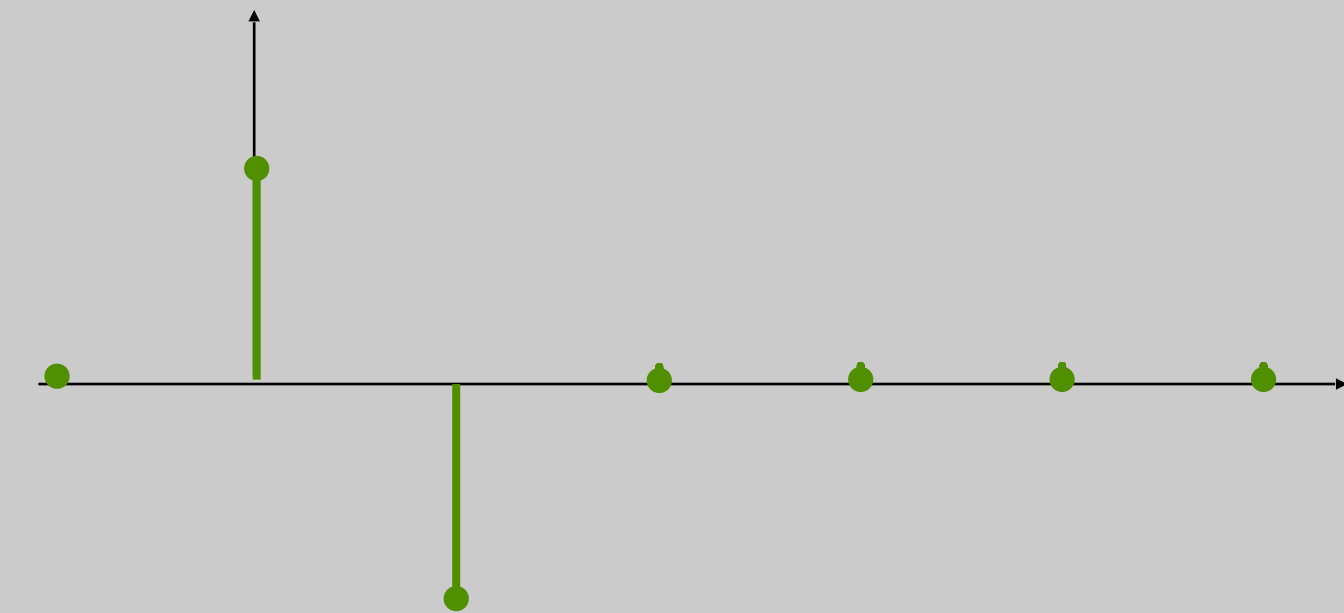
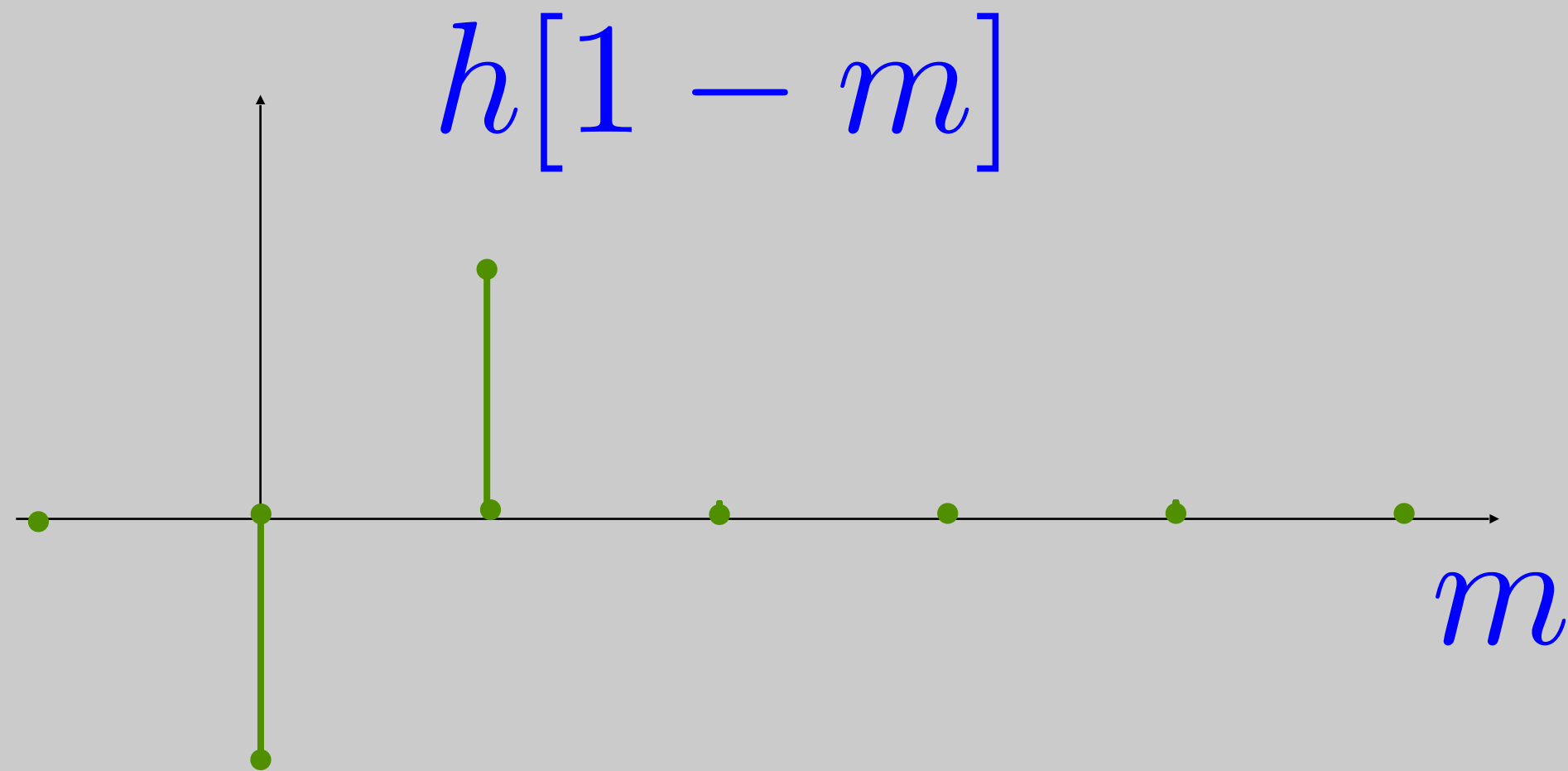


$$h[n] = \delta[n] - \delta[n-1]$$

Convolution Sum

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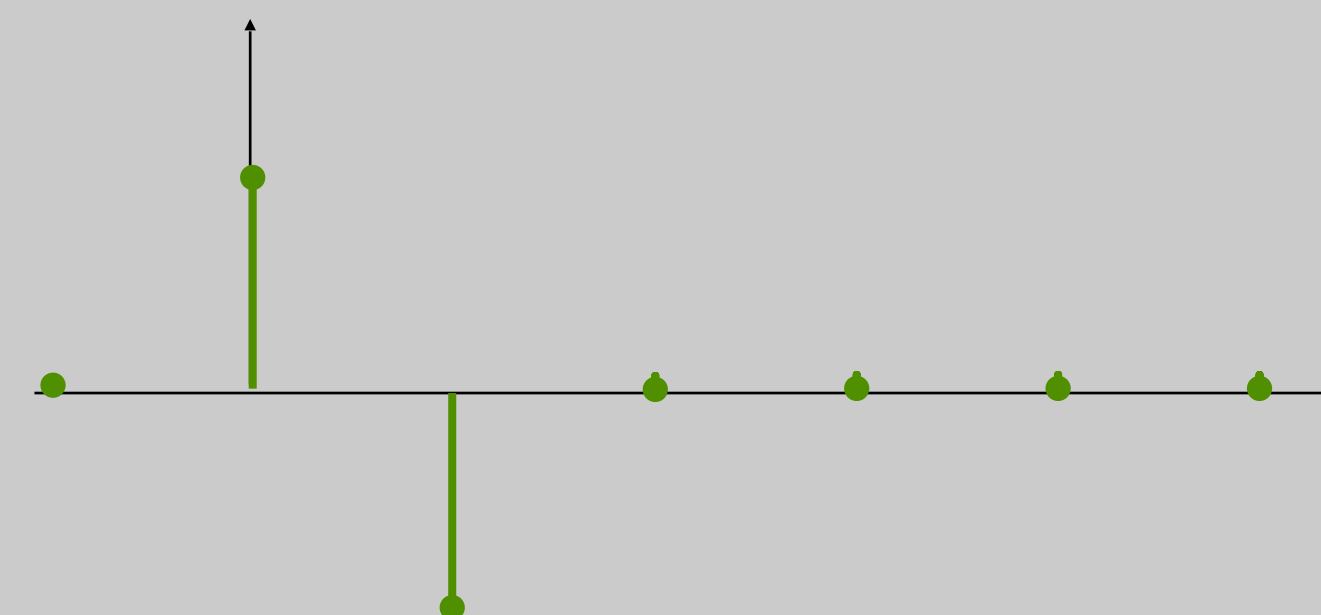
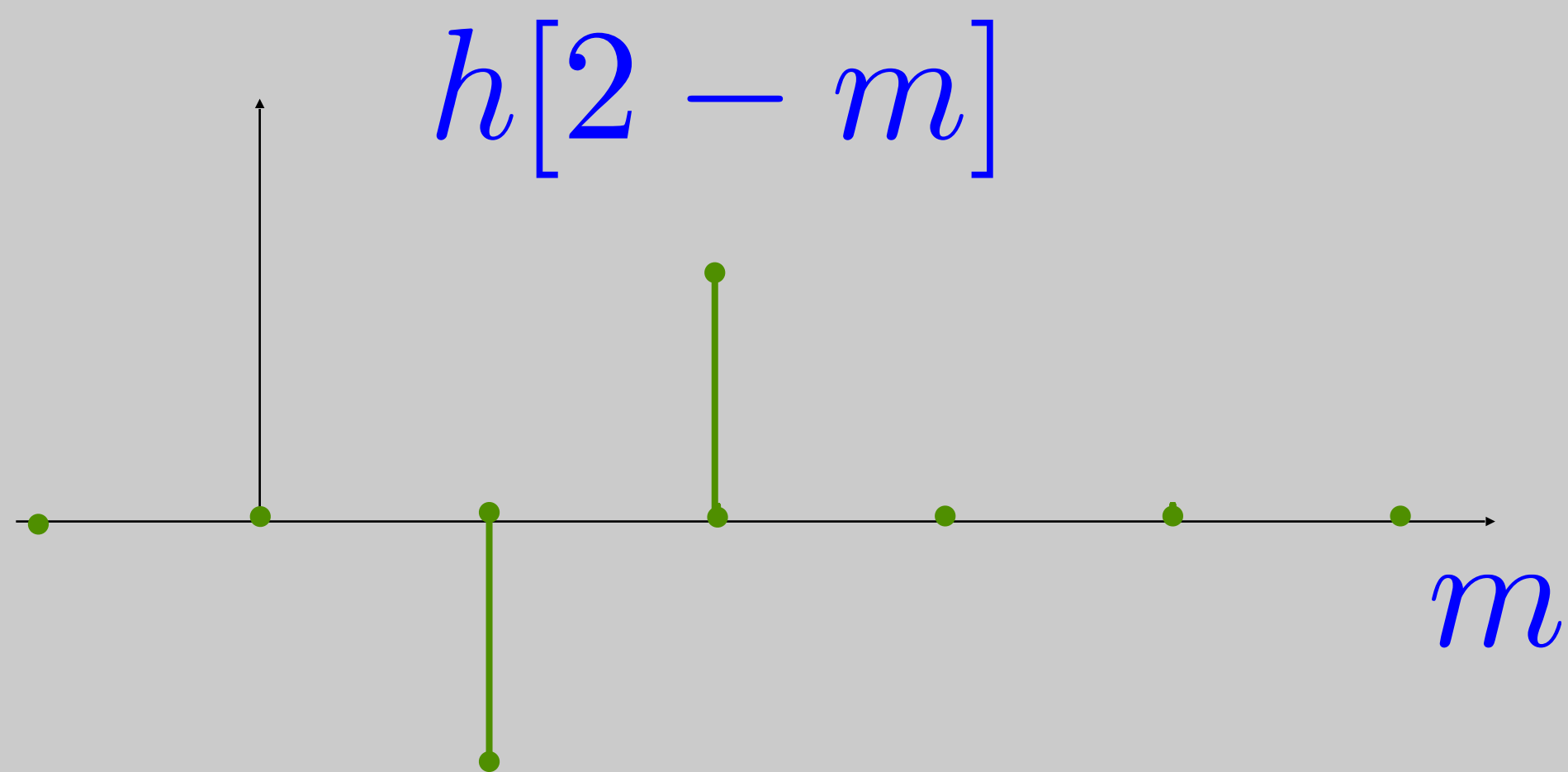


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Convolution Sum

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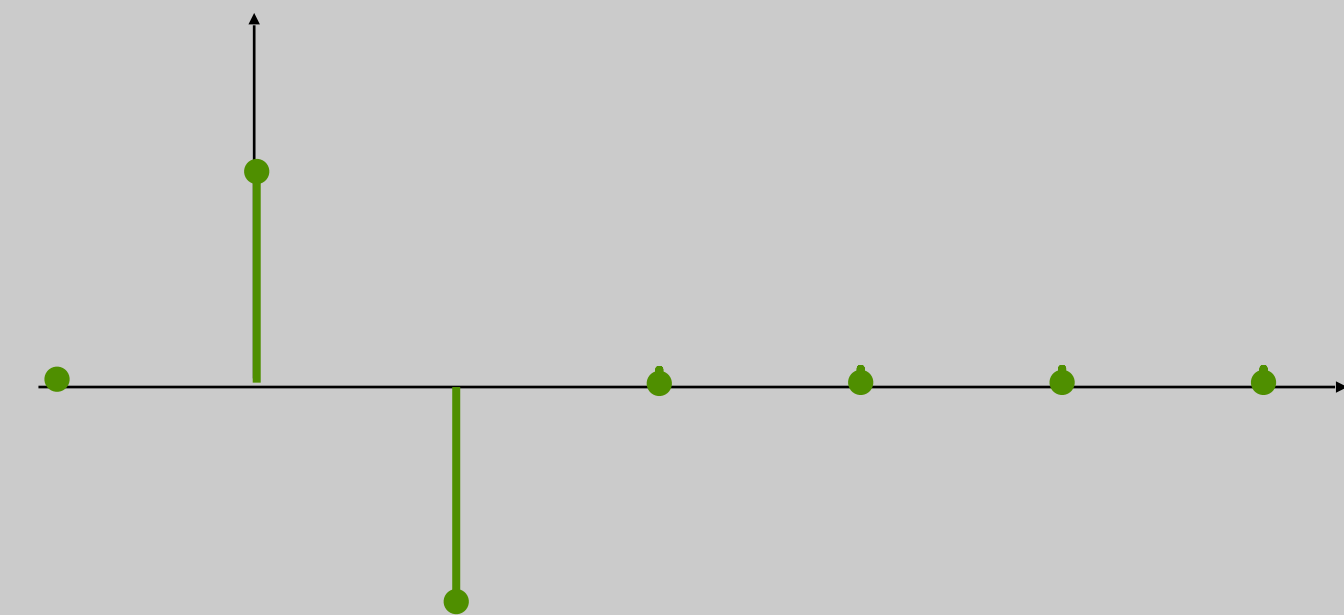
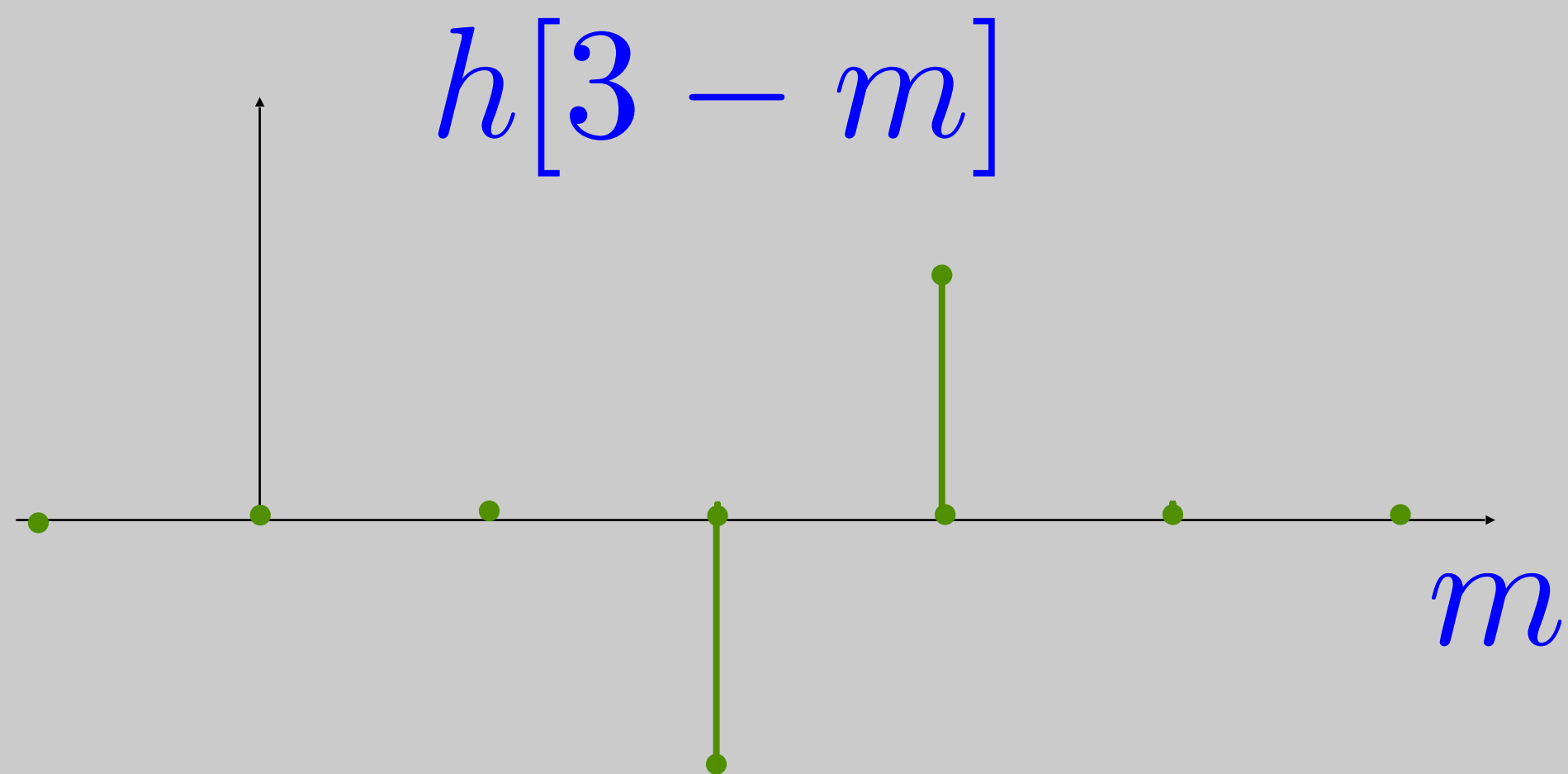


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Convolution Sum

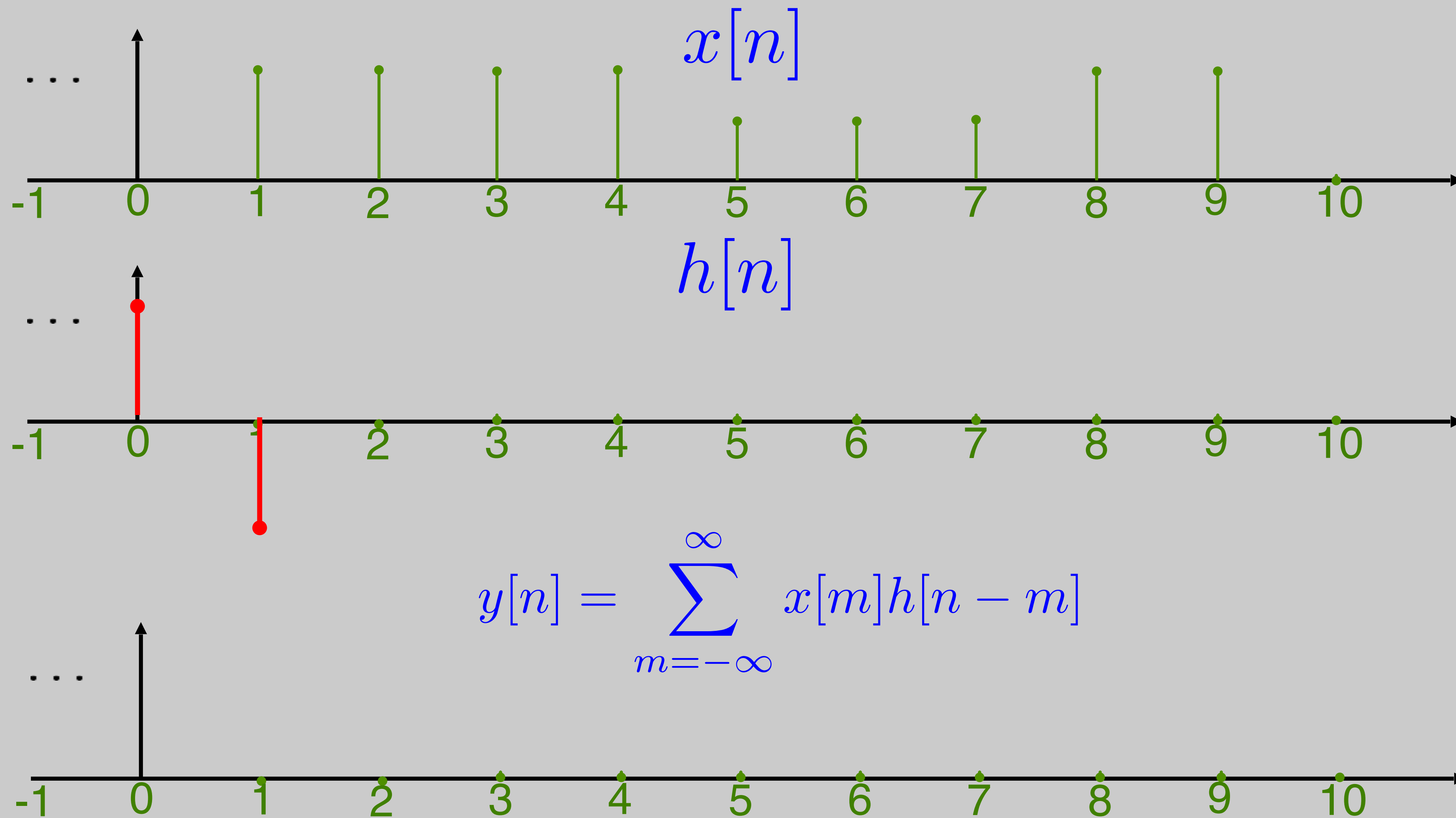
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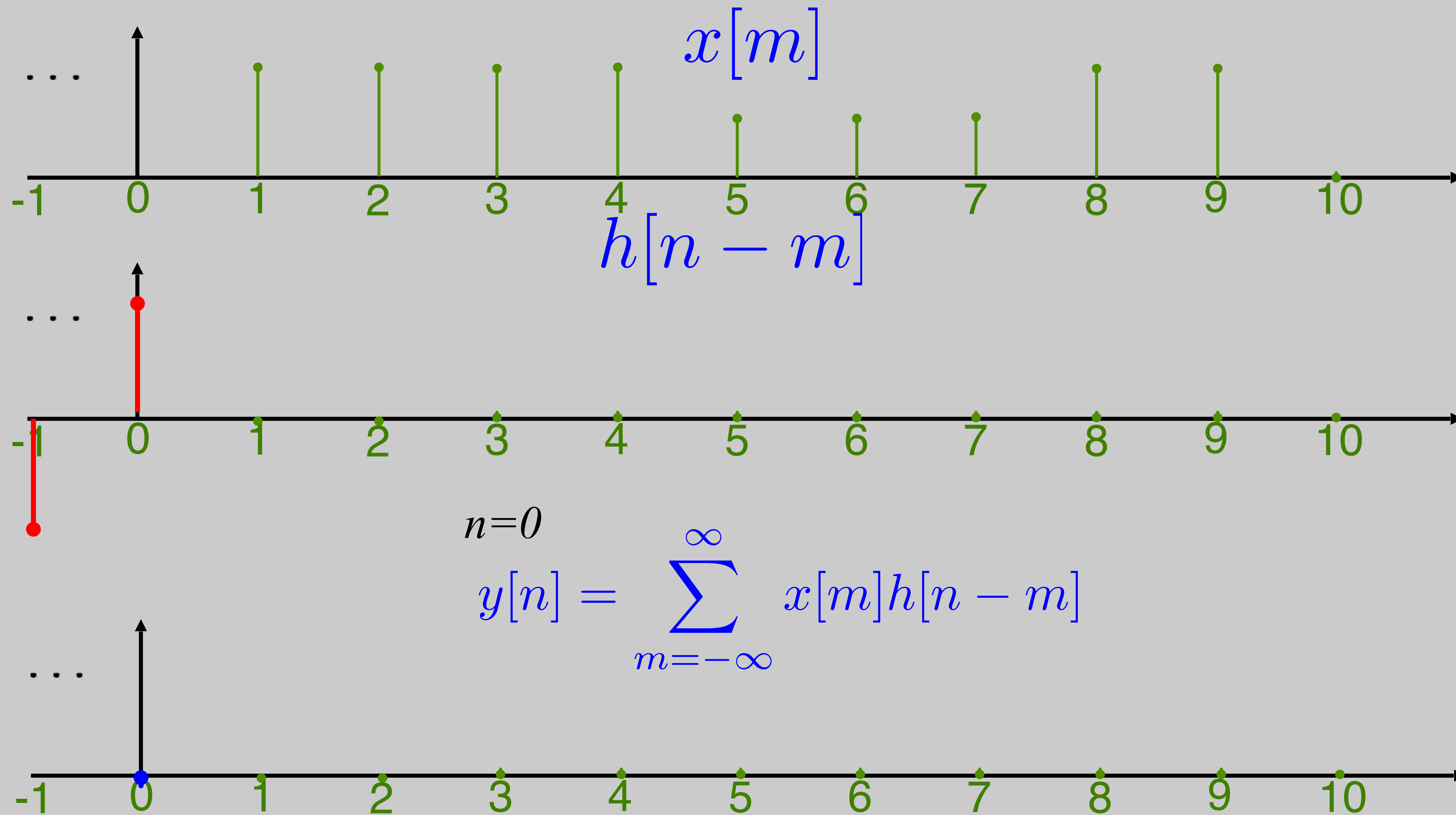


$$h[n] = \delta[n] - \delta[n-1]$$

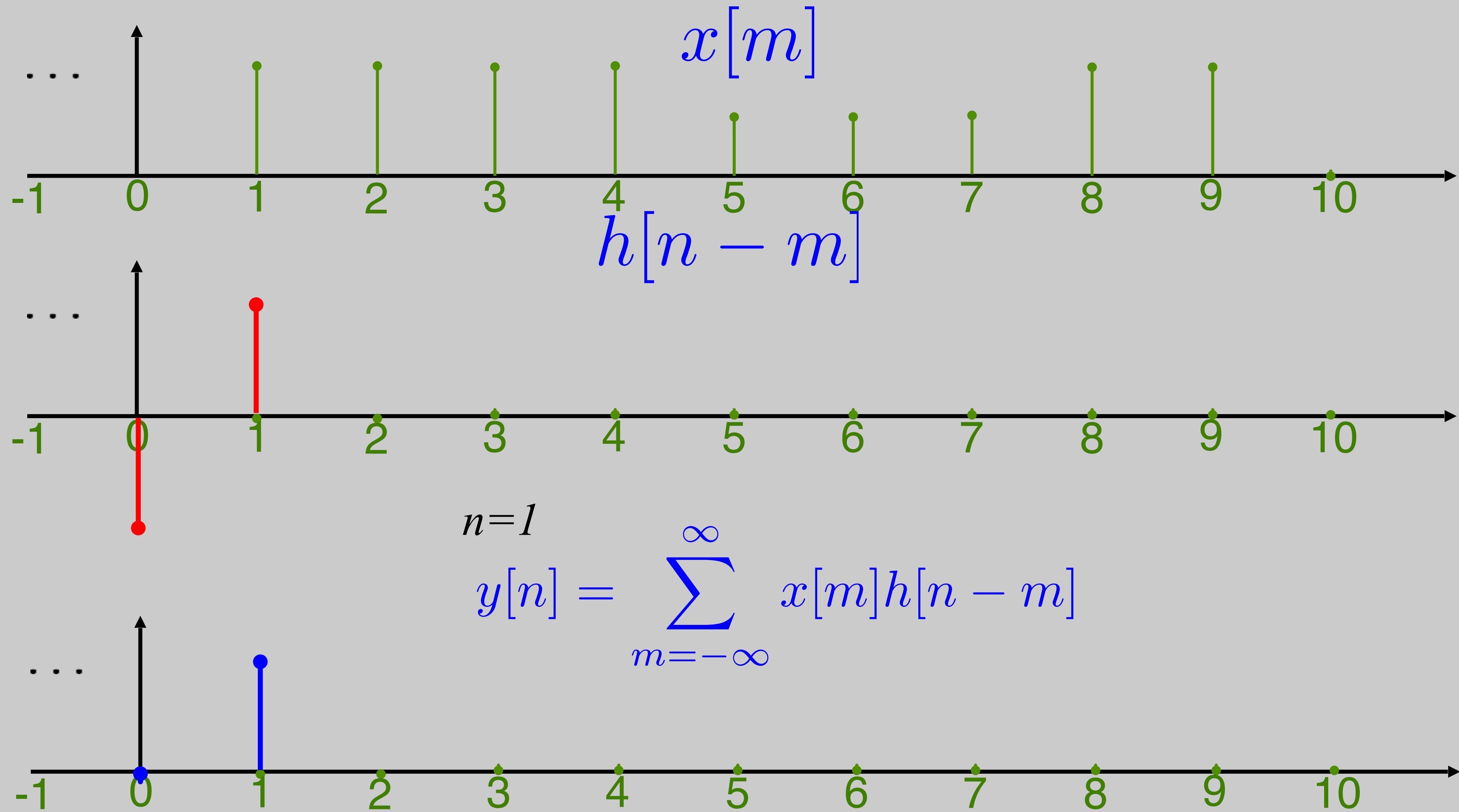
Graphical Example of Convolution



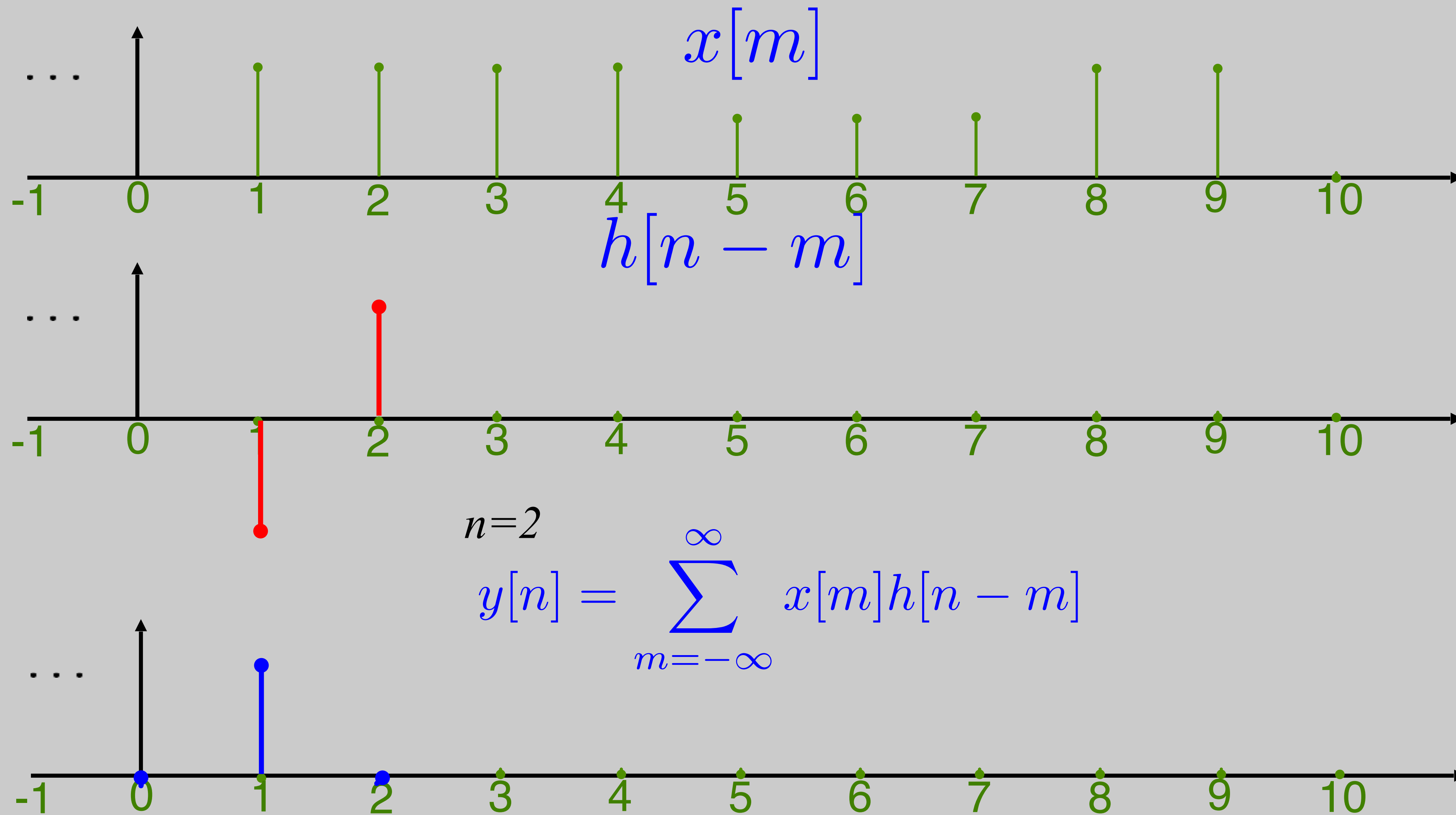
Graphical Example of Convolution



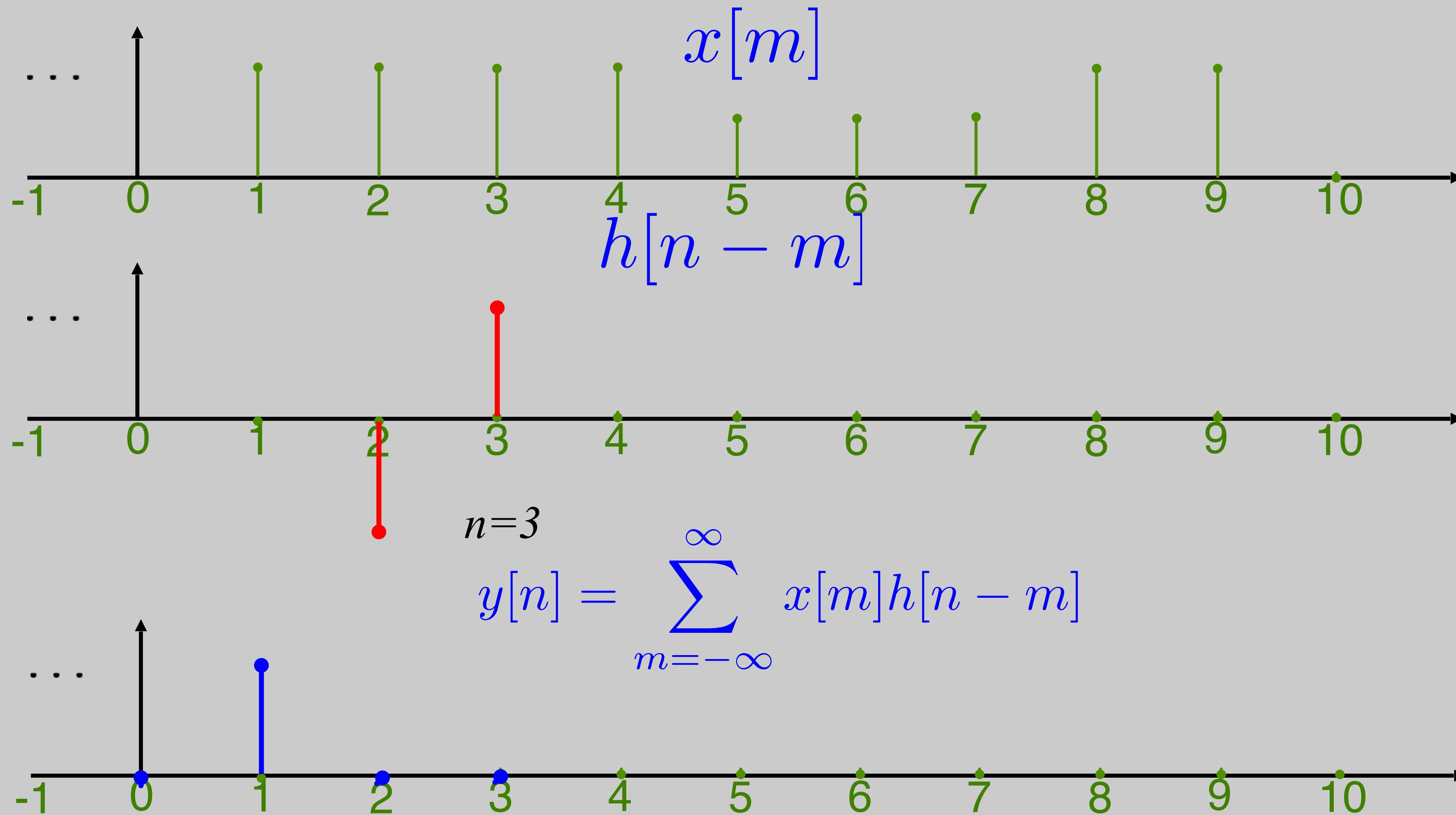
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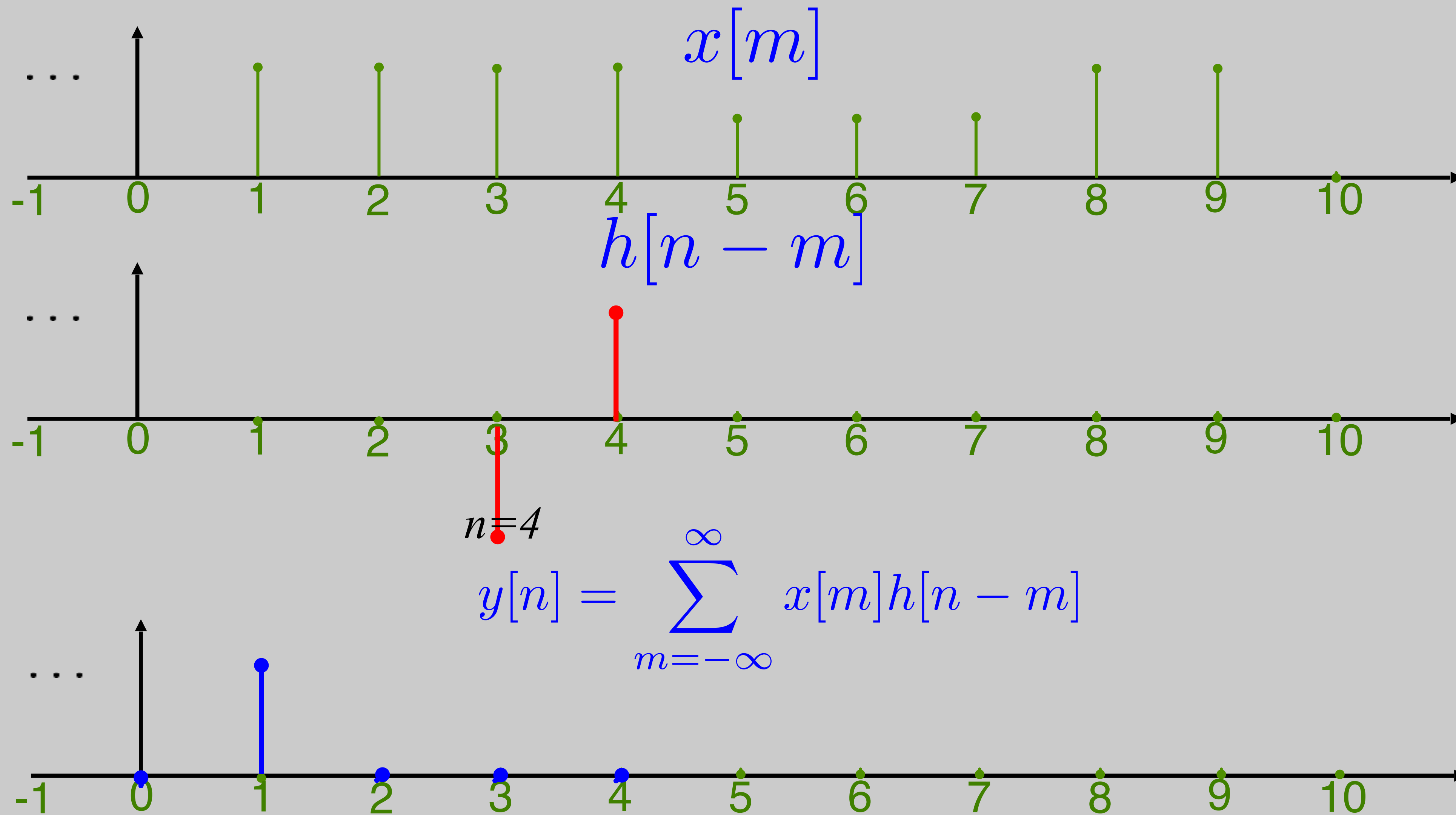
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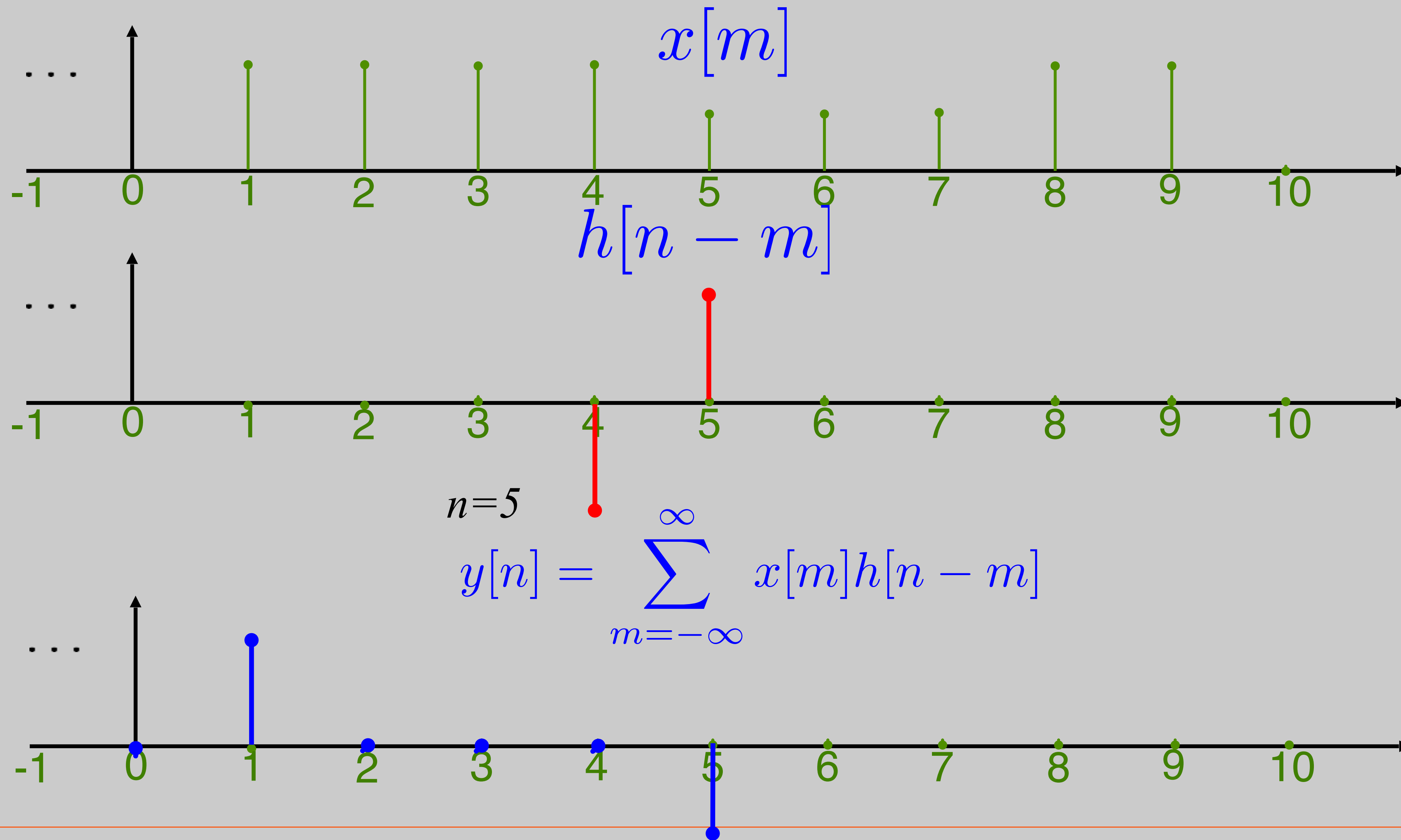
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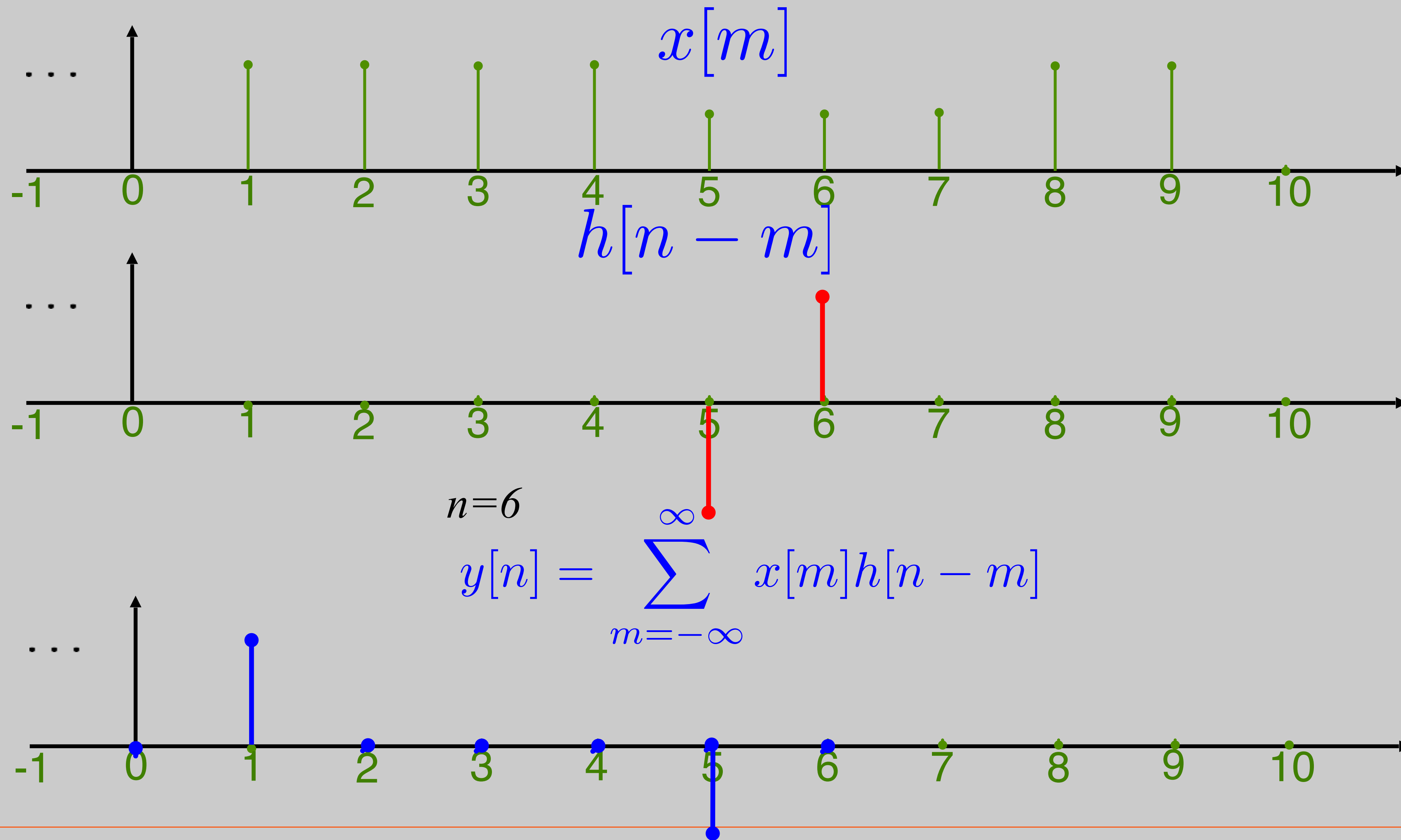
Graphical Example of Convolution



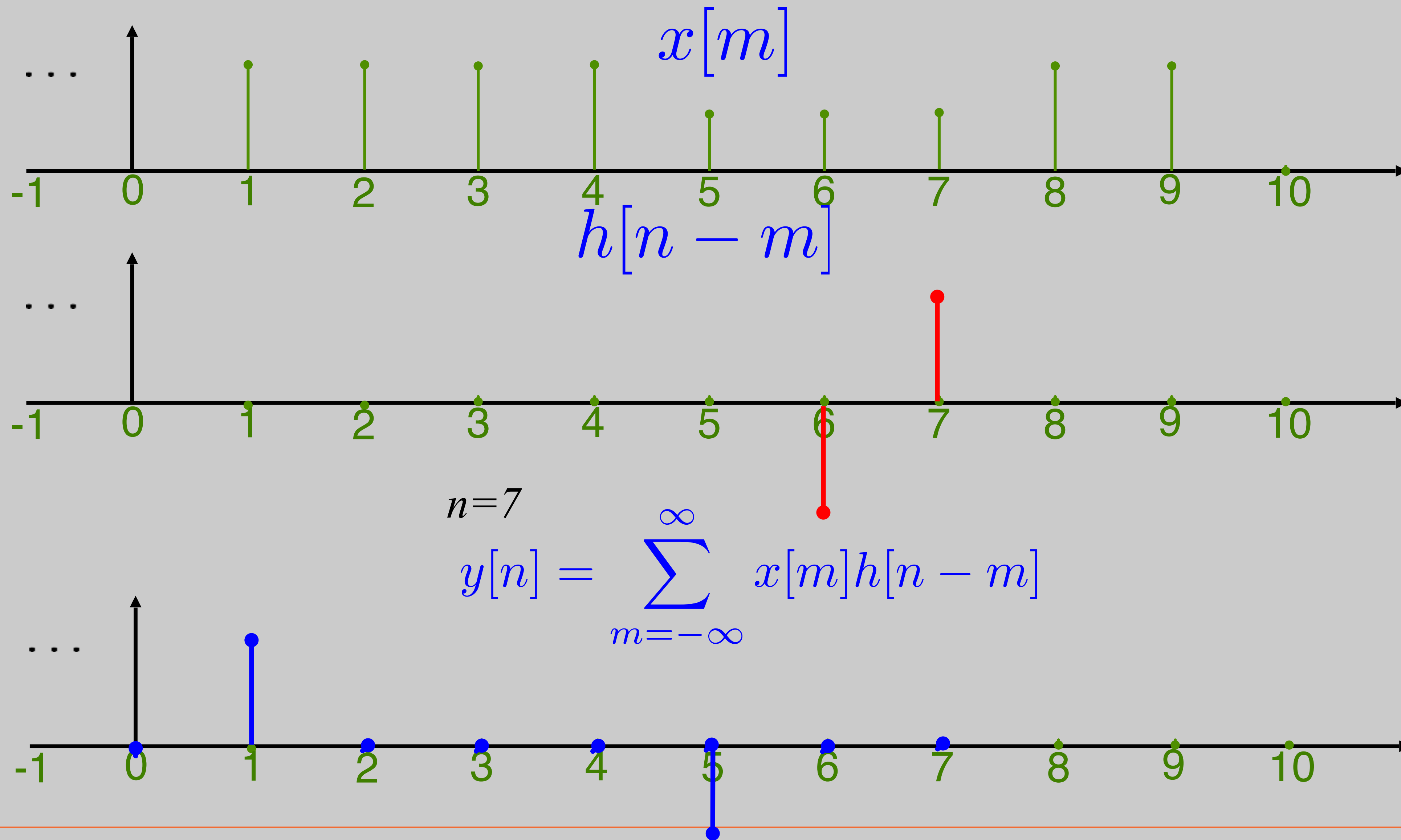
Graphical Example of Convolution



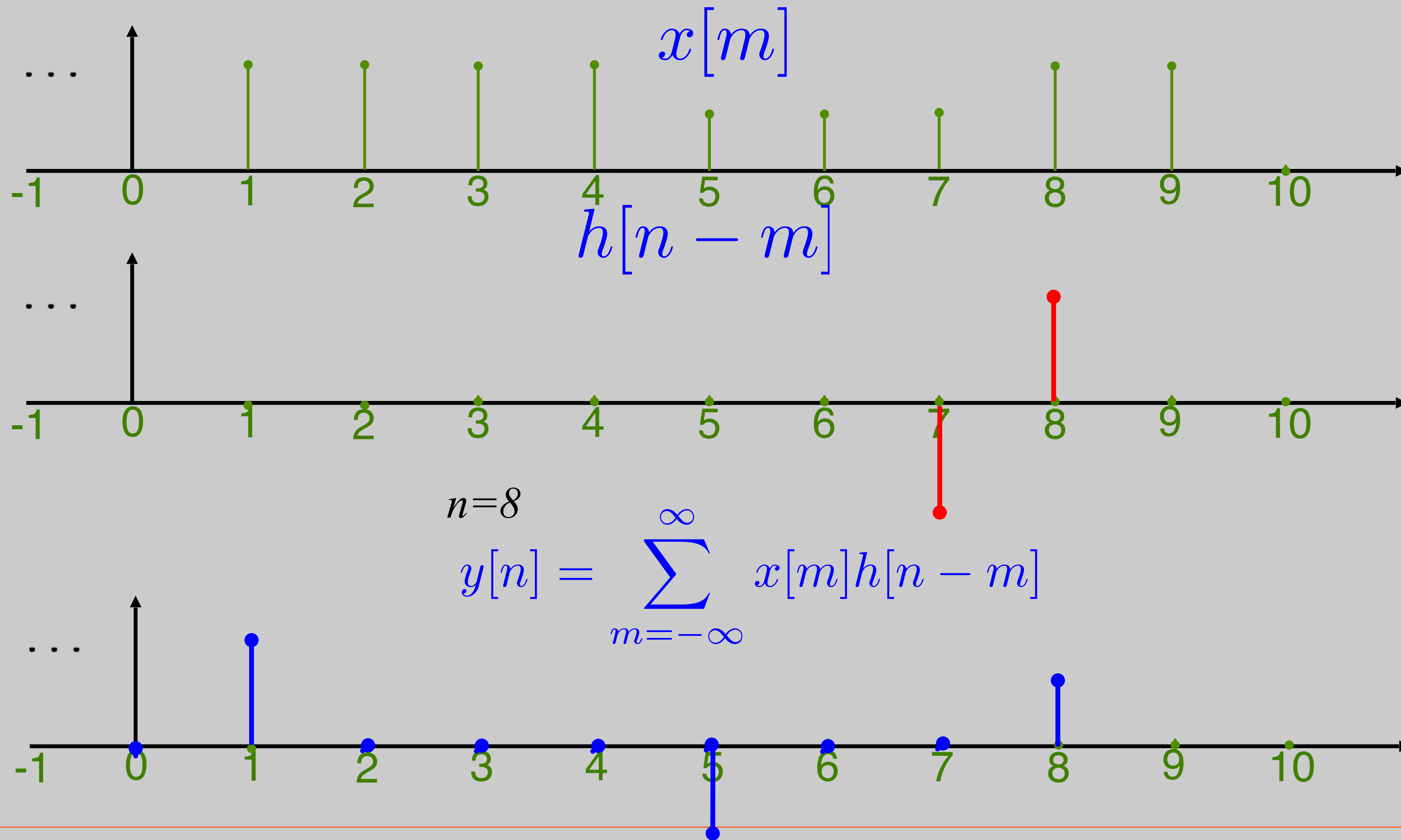
Graphical Example of Convolution



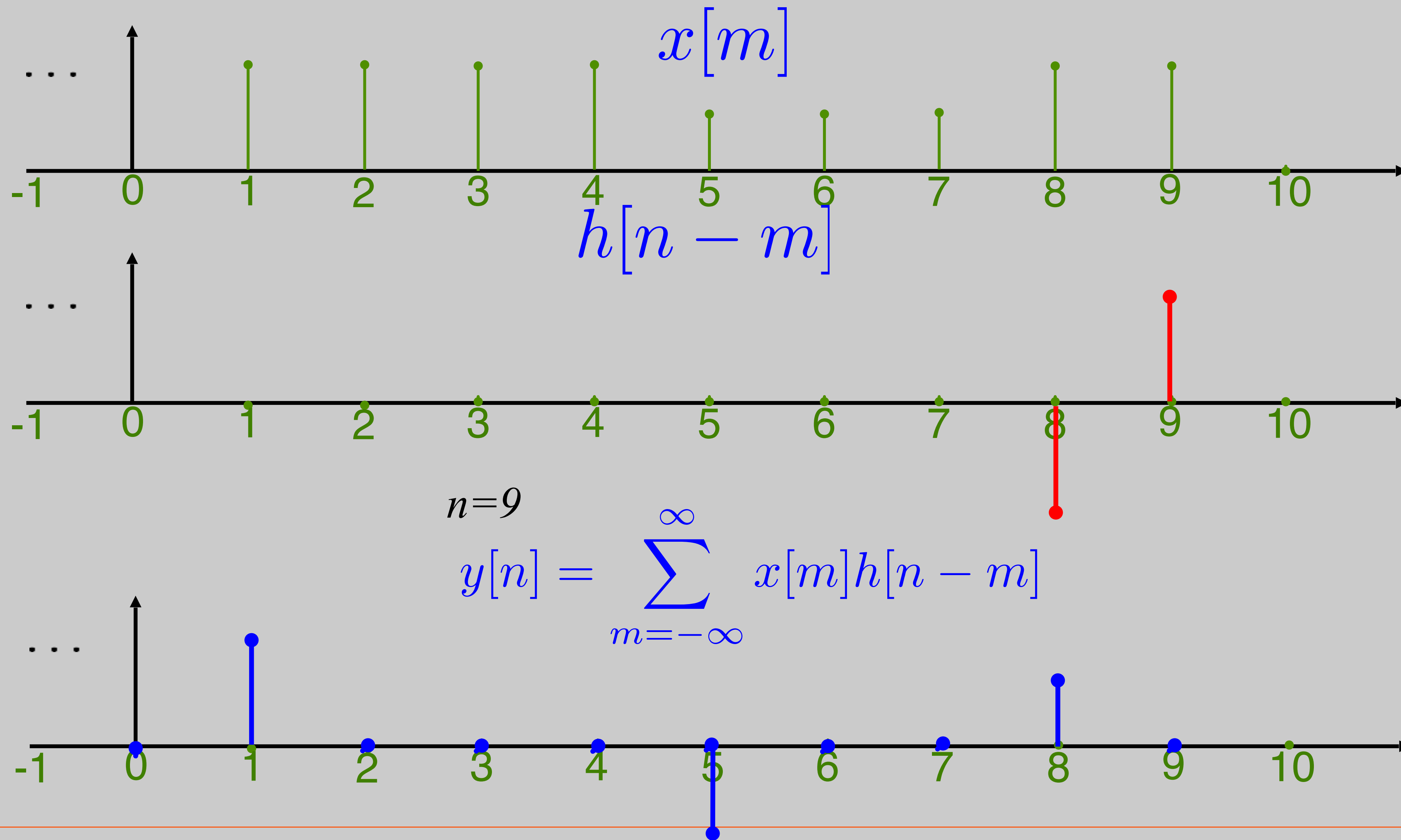
Graphical Example of Convolution



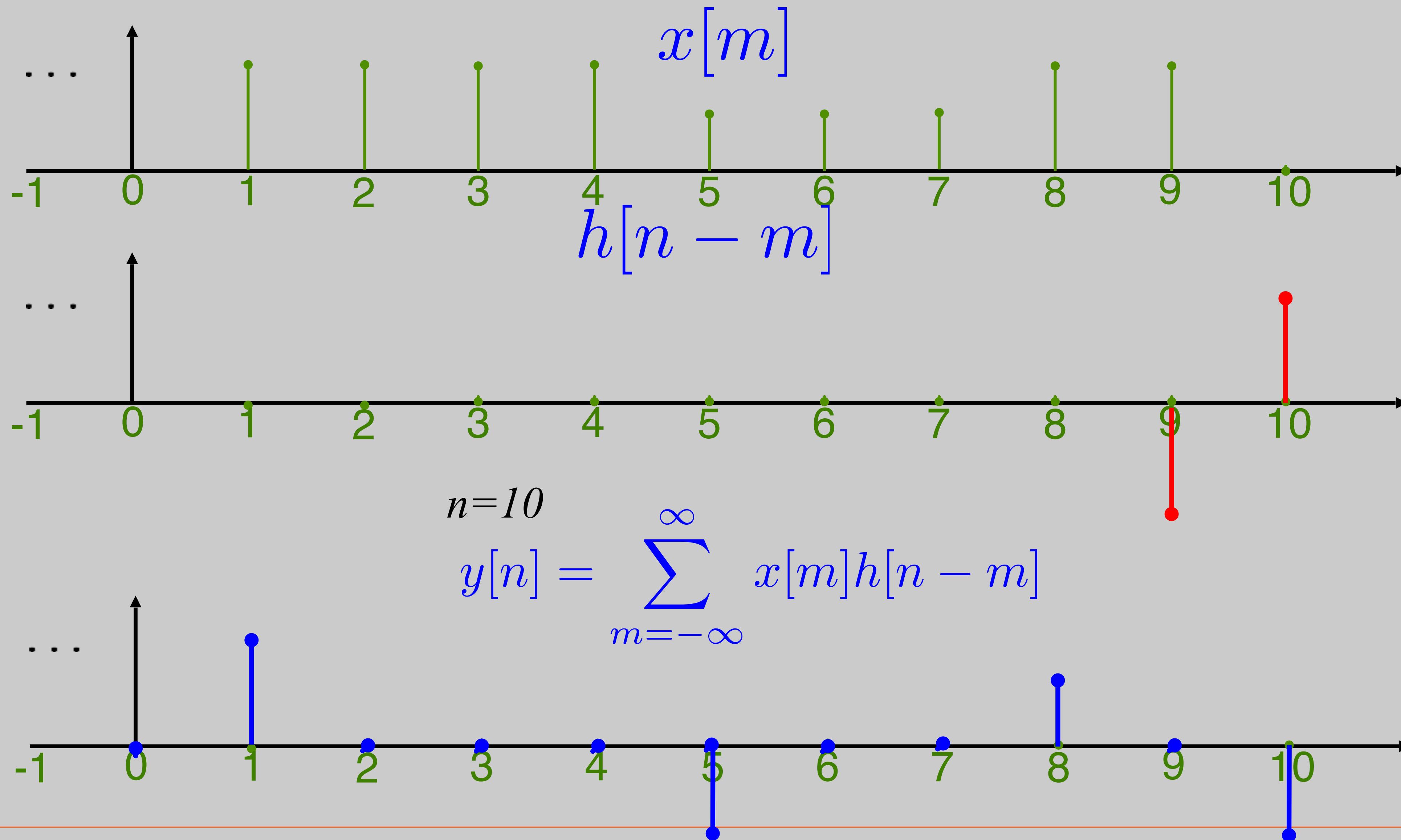
Graphical Example of Convolution



Graphical Example of Convolution



Graphical Example of Convolution

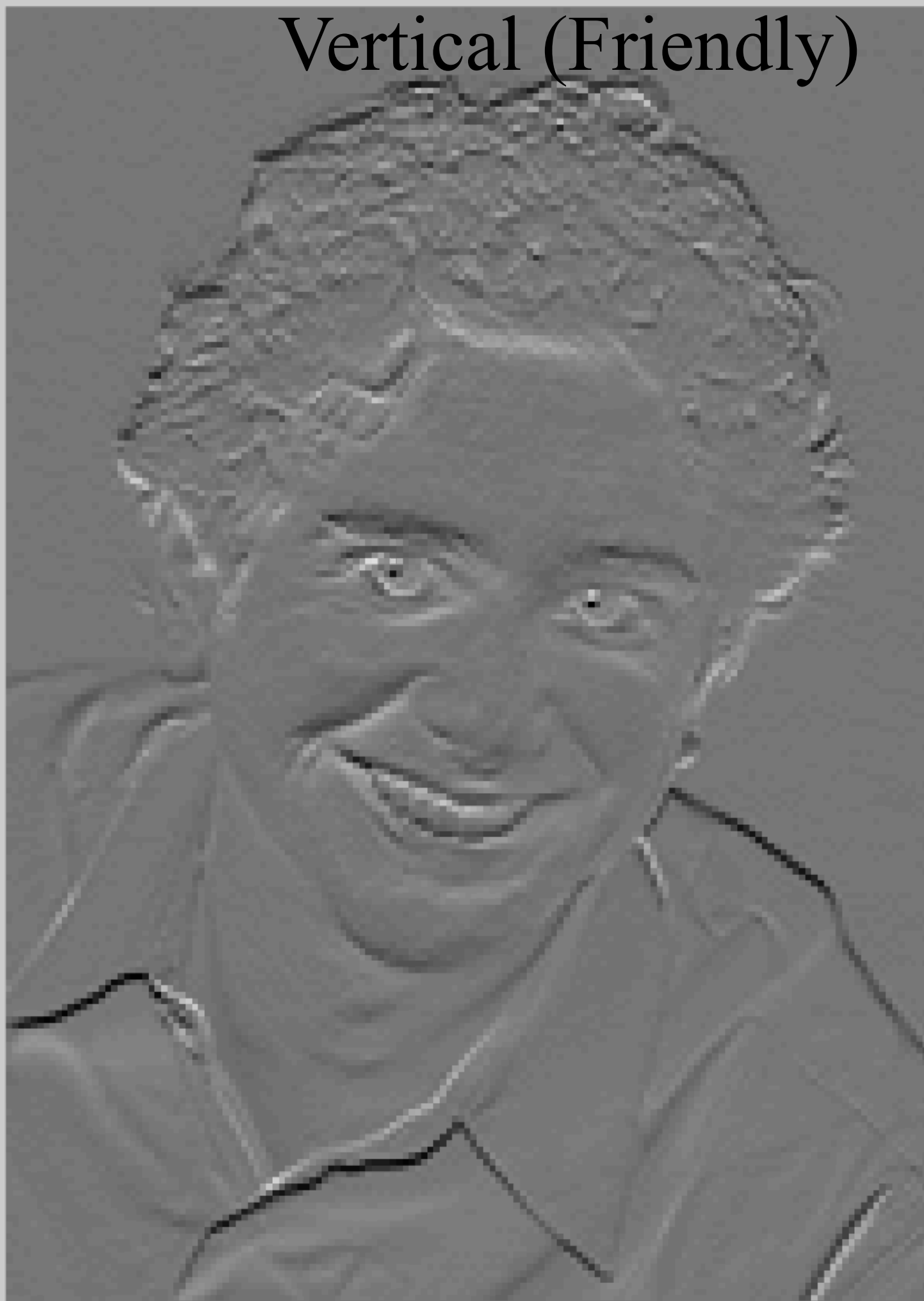


Example

Awekward



Vertical (Friendly)



Horizontal Mean



BIBO Stability of LTI systems

- LTI system is BIBO stable if, and only if $h[n]$ is absolutely summable.

$$\sum_{n=-\infty}^{\infty} |h[n]| < \infty$$

Proof (if):

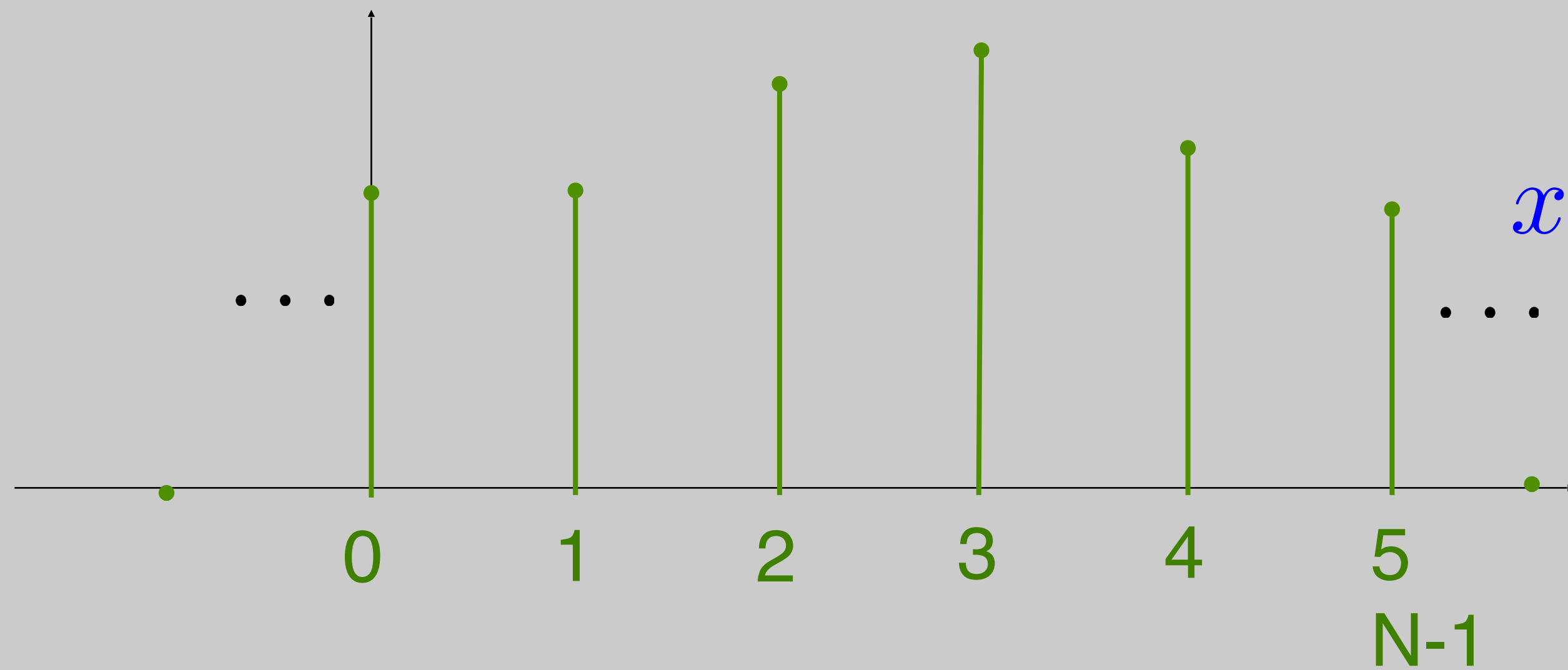
$$\begin{aligned} |y[n]| &= \left| \sum_{m=-\infty}^{\infty} x[m]h[n-m] \right| \leq \sum_{m=-\infty}^{\infty} |x[m]| \cdot |h[n-m]| \\ &\leq M \sum_{m=-\infty}^{\infty} |h[n-m]| = M \sum_{n=-\infty}^{\infty} |h[n]| < \infty \end{aligned}$$

$|x[m]| < M$

Only if in EE123

Finite Sequences

- Consider a finite sequence of length N



$$x[n] = \begin{cases} \text{something} & 0 \leq n < N \\ 0 & \text{otherwise} \end{cases}$$

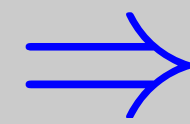
- Can also be written as a vector

$$\vec{x} = \begin{bmatrix} x(0) \\ x(1) \\ \vdots \\ x(N-1) \end{bmatrix}$$

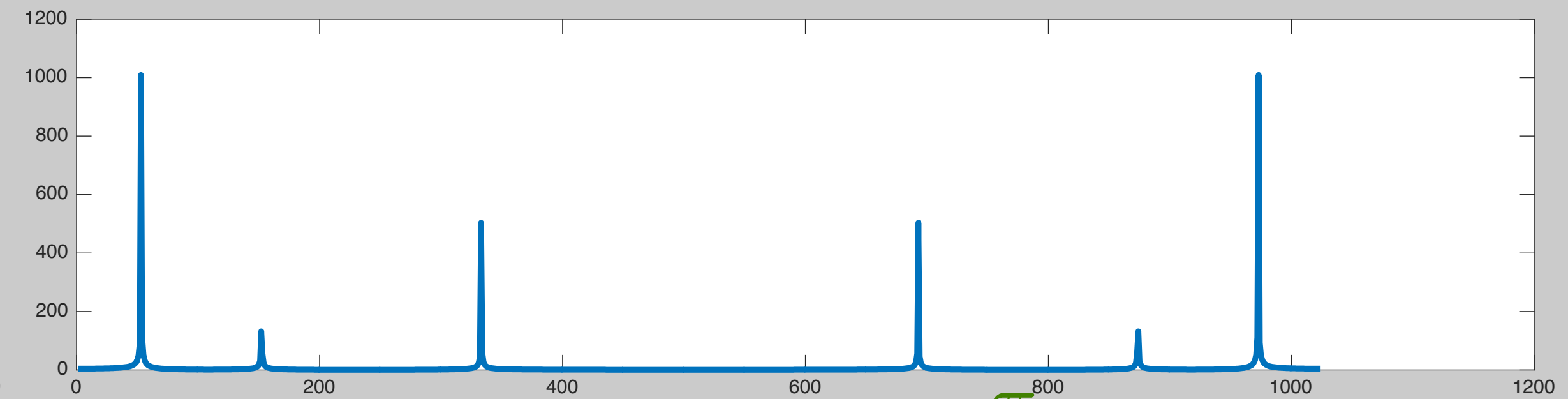
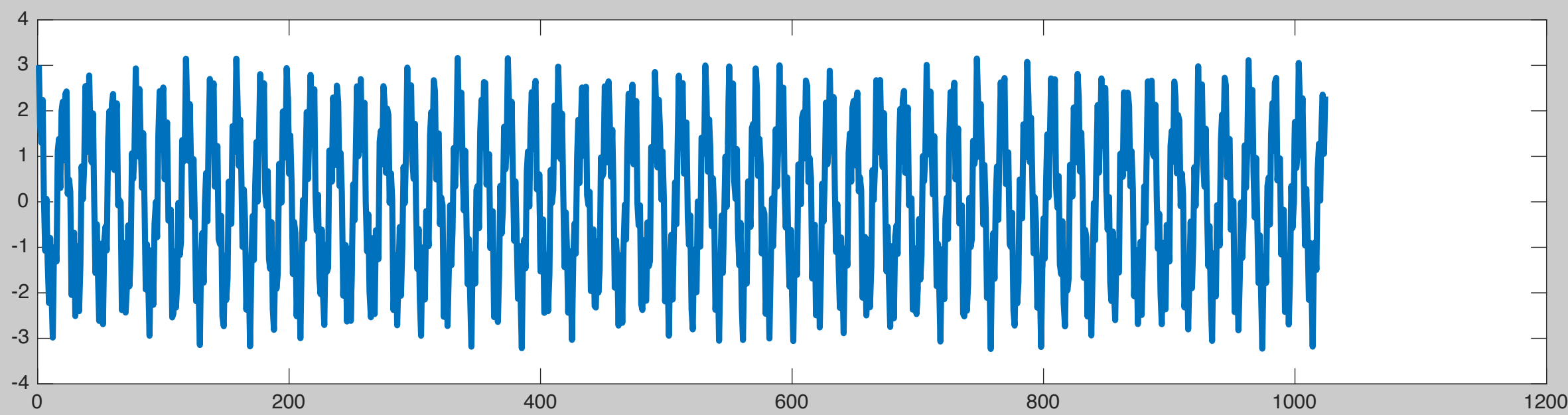
Why?

- To compute this:

$x[n]$



$X[k]$



$\frac{\pi}{2}$

π

$-\frac{\pi}{2}$

Finite Sequences as Vectors

- Define an inner-product (for \mathbb{R}^N):

$$\begin{aligned}\langle \vec{x}, \vec{y} \rangle &= \vec{x} \cdot \vec{y} = \sum_{n=0}^{N-1} x[n]y[n] = \\ &= \vec{x}^T \vec{y}\end{aligned}$$

So,

$$\begin{aligned}\langle \vec{x}, \vec{x} \rangle &= \sum_{n=0}^{N-1} x[n]x[n] = \sum_{n=0}^{N-1} x^2[n] = \|\vec{x}\|^2 \\ &\Rightarrow \vec{x}^T \vec{x} = \|\vec{x}\|^2\end{aligned}$$

Finite Sequences as Vectors

- What about complex?

$$x \cdot x = x^2 = (x_r + jx_i)(x_r + jx_i) = x_r^2 - x_i^2 + 2jx_r x_i \neq \|x\|^2$$

but,

$$x^* \cdot x = (x_r - jx_i)(x_r + jx_i) = x_r^2 + x_i^2 = \|x\|^2$$

- Transpose vs Transpost conjugate

$$\vec{x} = \begin{bmatrix} 1 \\ j \\ 1 + j \end{bmatrix} \quad \vec{x}^T = \begin{bmatrix} 1 & j & 1 + j \end{bmatrix}$$
$$\vec{x}^* = \begin{bmatrix} 1 & -j & 1 - j \end{bmatrix}$$

Finite Sequences as Vectors

- Define Complex inner product

$$\langle \vec{x}, \vec{y} \rangle = \overline{\vec{x}} \cdot \vec{y} = \sum_{n=0}^{N-1} x[n]^* y[n] = \vec{x}^* \vec{y} = \vec{x}^H \vec{y}$$

$$\vec{x} = \begin{bmatrix} 1 \\ j \end{bmatrix} \Rightarrow \vec{x}^* \vec{x} = \begin{bmatrix} 1 & -j \end{bmatrix} \begin{bmatrix} 1 \\ j \end{bmatrix} = 2$$

Projections

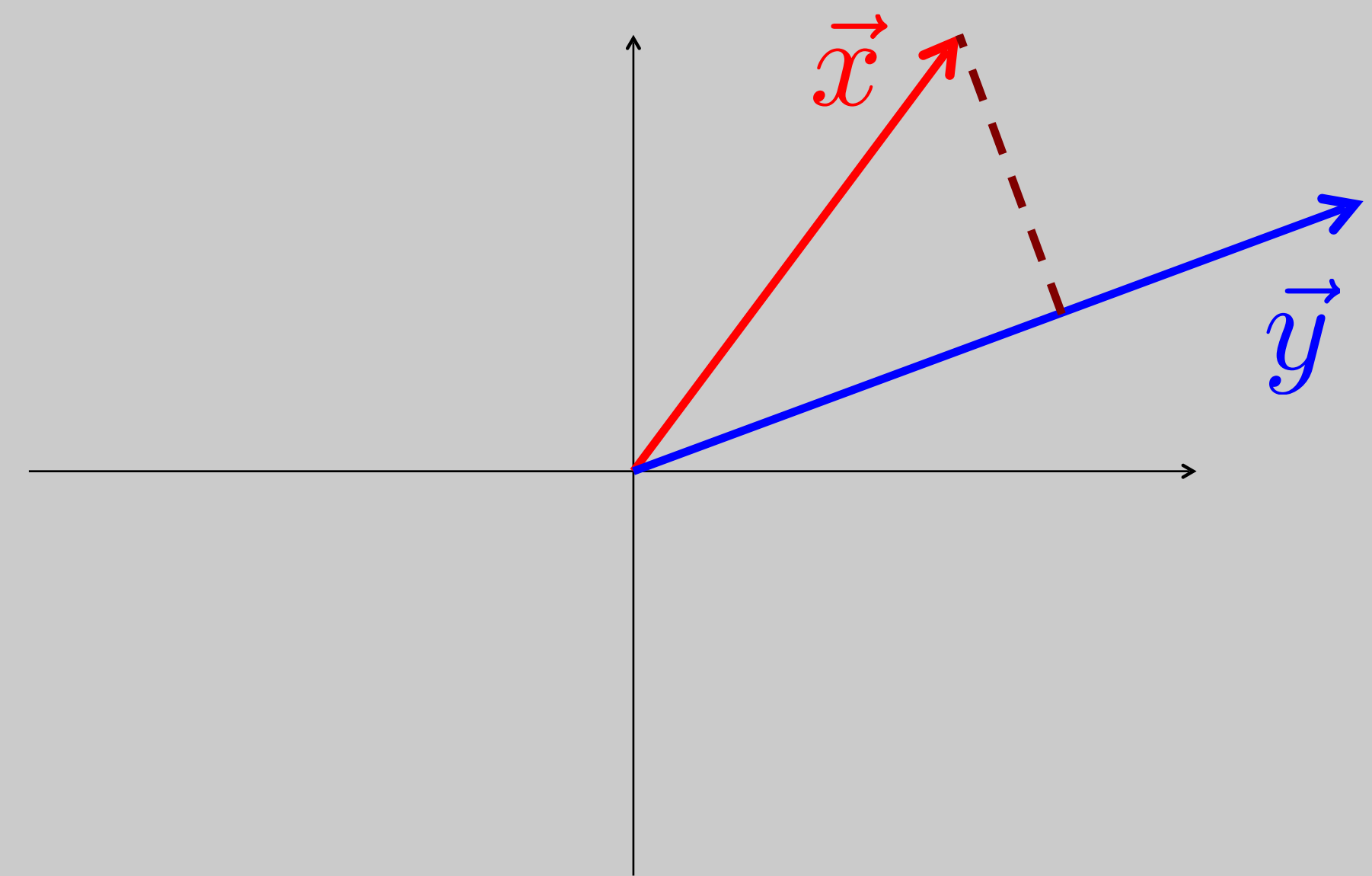
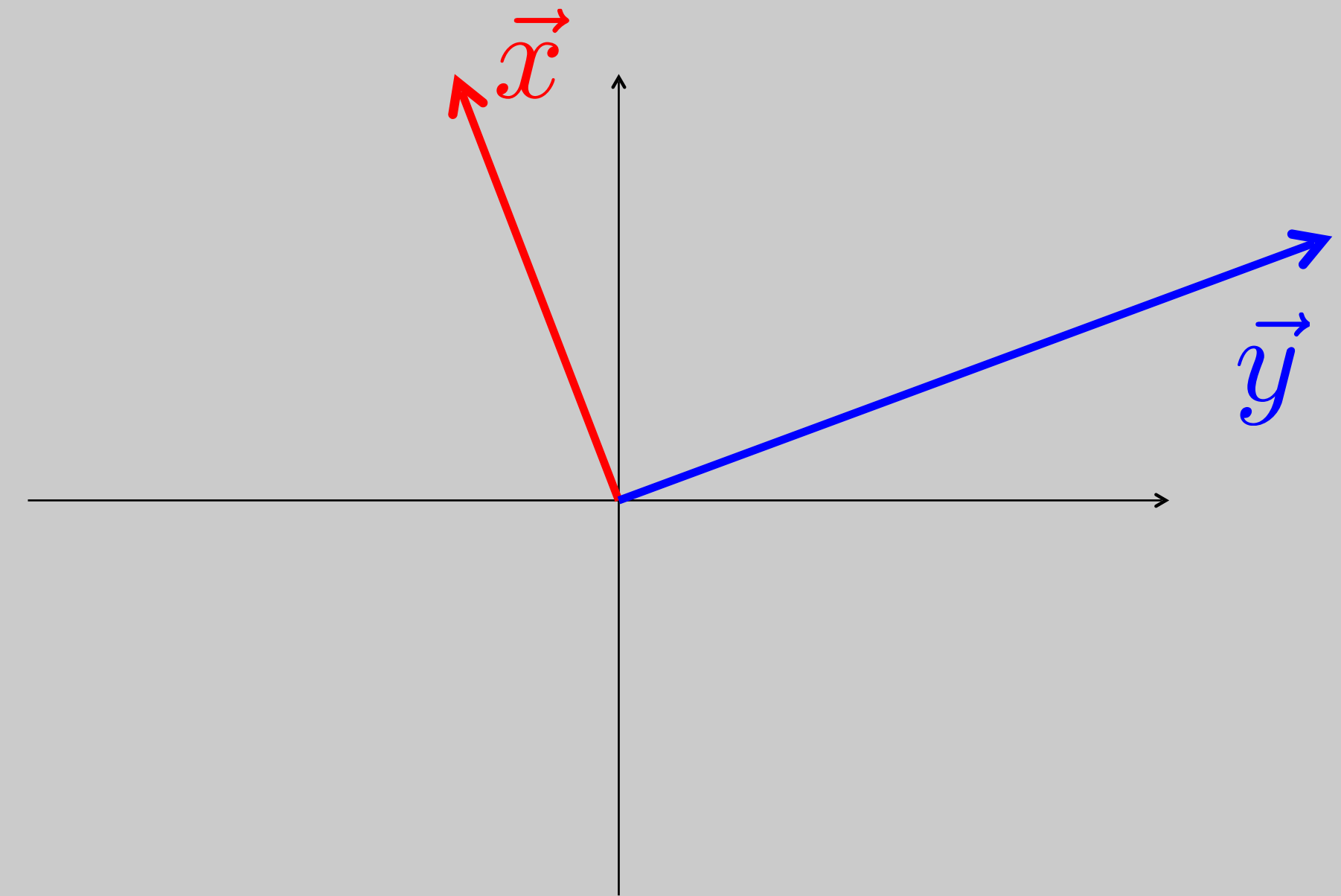
- Orthogonality:

$$\vec{x}^* \vec{y} = \sum_{n=0}^{N-1} x[n]^* y[n] = 0$$

- Unit vector: $||\hat{x}|| = 1$

$$\hat{x} = \frac{\vec{x}}{||\vec{x}||}$$

- Define projection as: $\frac{\vec{y}^* x}{||\vec{y}||}$

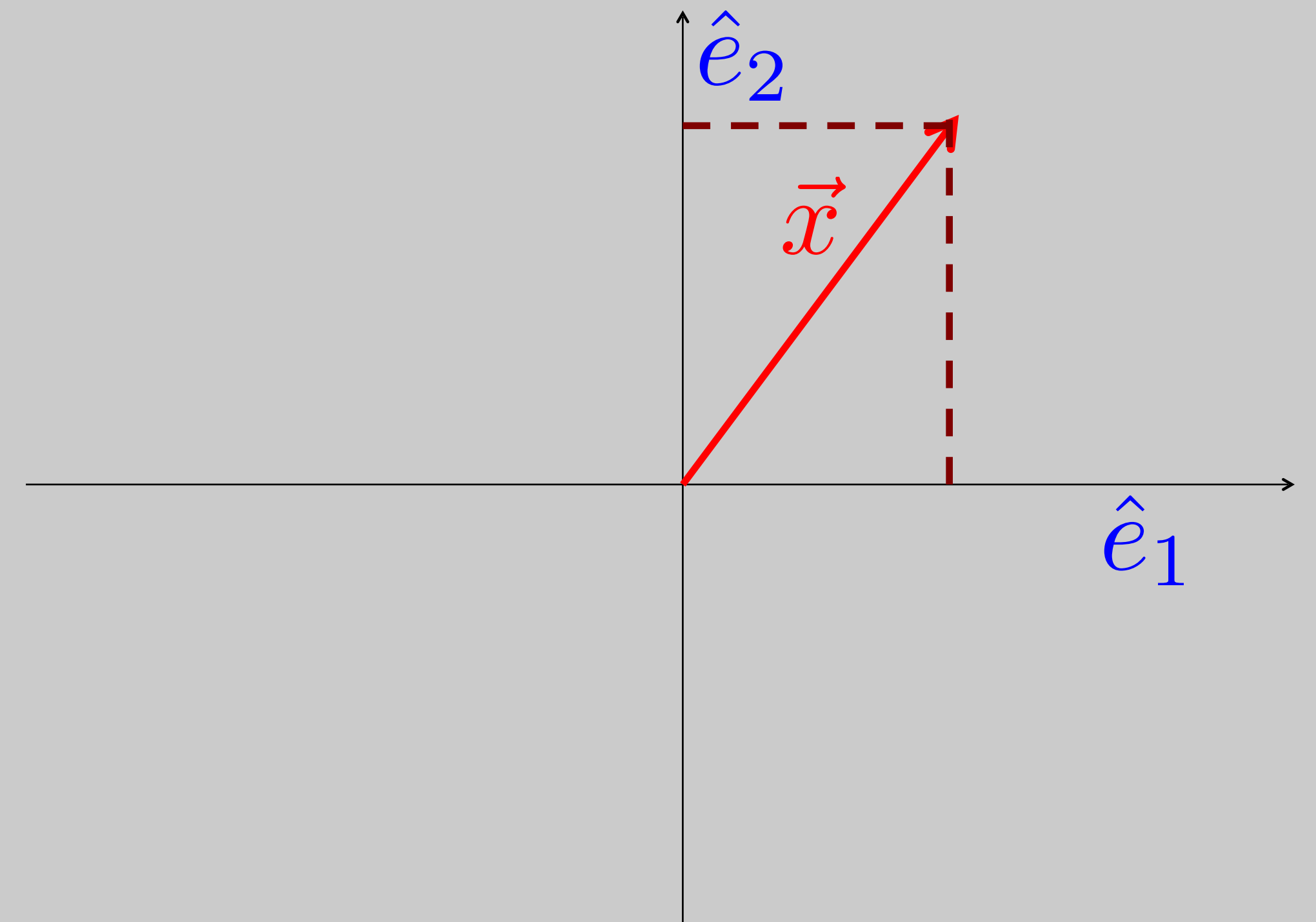


Change of Coordinates (Basis)

- We can compute new coordinates by projections onto orthonormal basis vectors

$$\hat{e}_1^* \vec{x} = [1 \ 0] \vec{x} = x_1$$

$$\hat{e}_2^* \vec{x} = [0 \ 1] \vec{x} = x_2$$



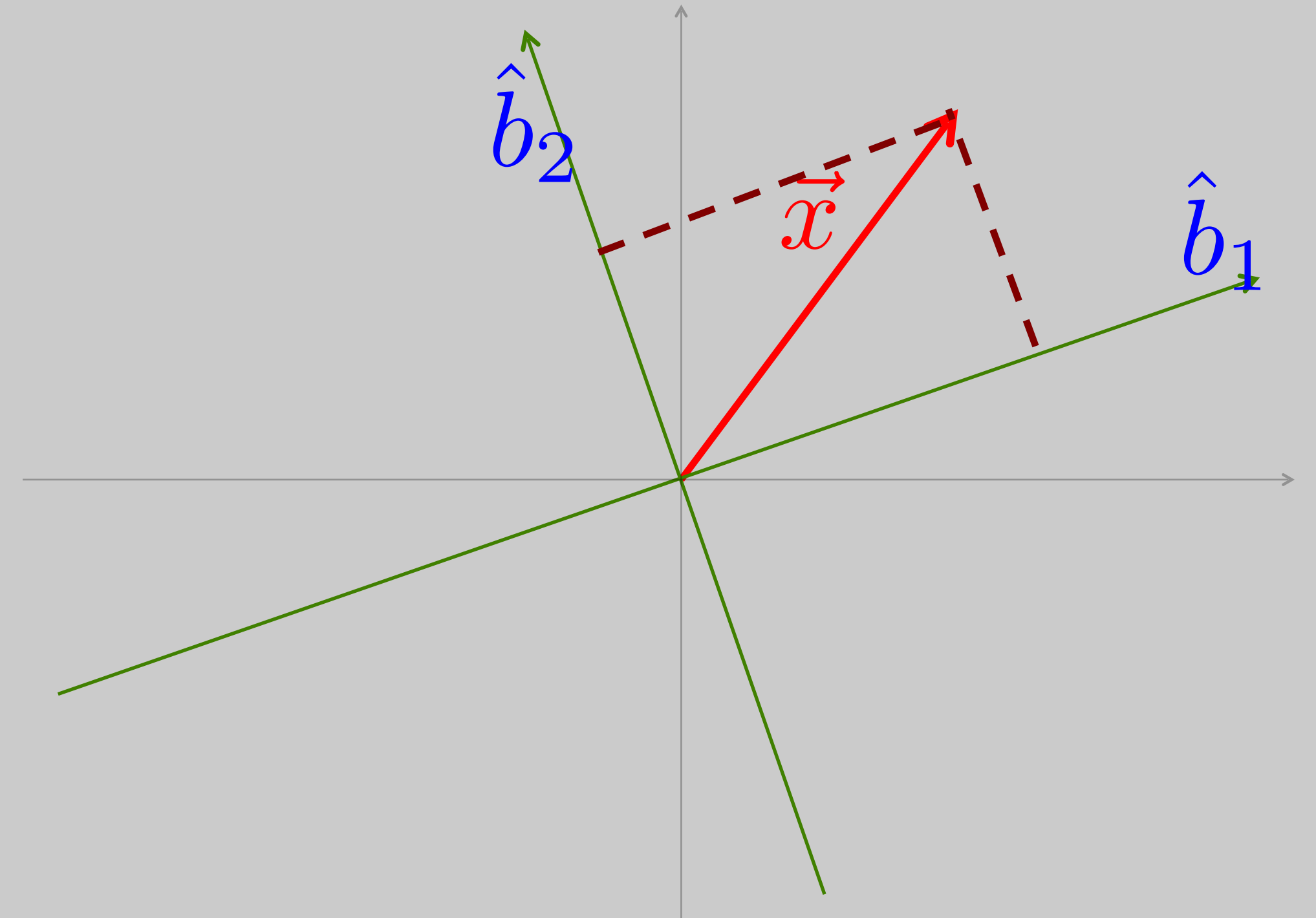
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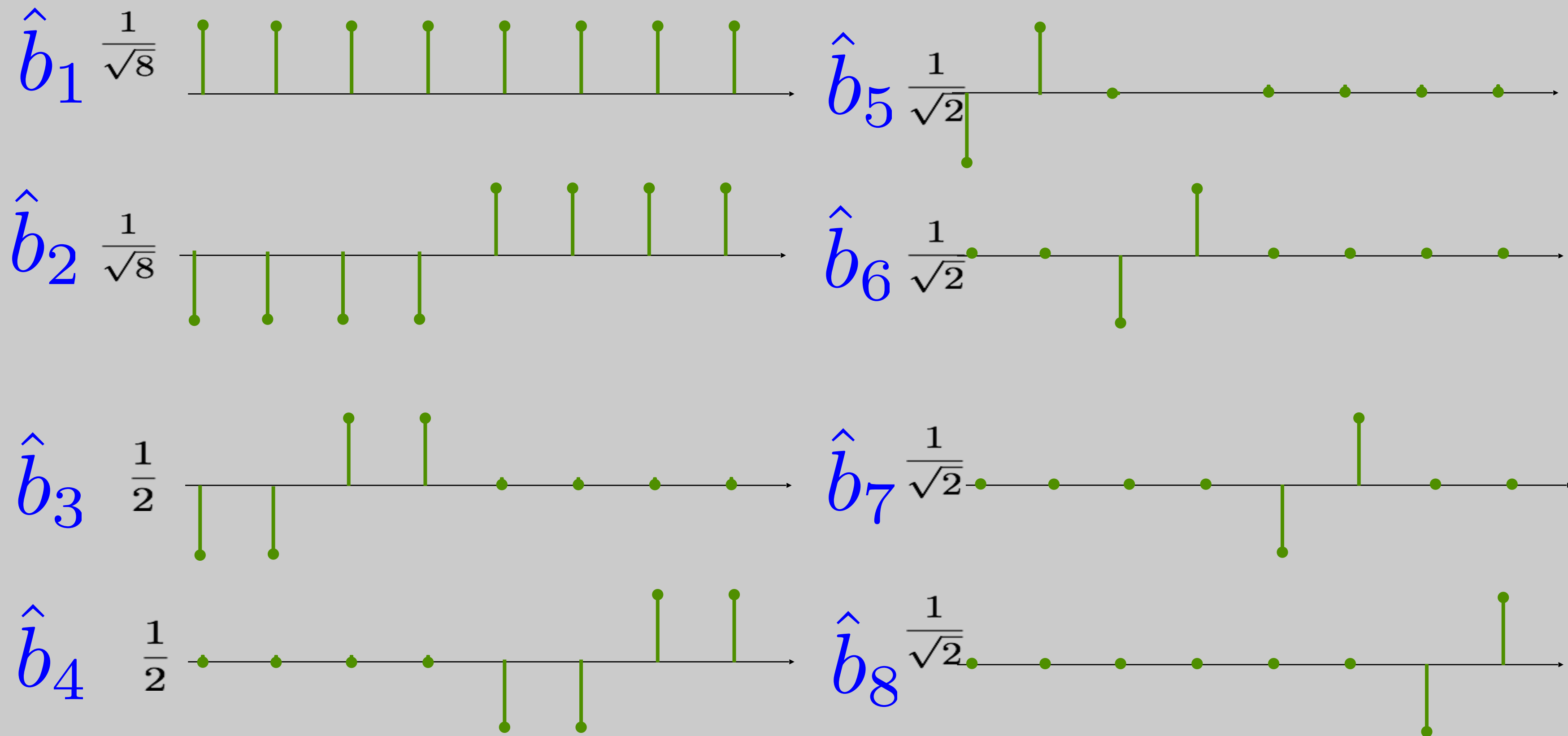
New coordinates:

$$\begin{bmatrix} \hat{b}_1^* \vec{x} \\ \hat{b}_2^* \vec{x} \end{bmatrix} = \begin{bmatrix} \hat{b}_1 & \hat{b}_2 \end{bmatrix}^* \vec{x}$$

$$\Rightarrow \vec{x} = (\hat{b}_1^* \vec{x}) \hat{b}_1 + (\hat{b}_2^* \vec{x}) \hat{b}_2$$



Change of basis



$$\begin{array}{c} 1 \\ \uparrow \\ \bullet \bullet \bullet \bullet \bullet \bullet \bullet \bullet \end{array} = \hat{b}_1 + \hat{b}_2 + \hat{b}_3 + \hat{b}_4 + \hat{b}_5 + \hat{b}_6 + \hat{b}_7 + \hat{b}_8$$