EE 110B Signals and Systems

Linear and Time-Invariant (LTI) Systems

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Why LTI systems?

- Linear and time-invariant systems are especially easy to analyze and design.
- In a lot of cases, they are good enough to do the "signal processing" job.
- Amenable to frequency analysis in the Fourier domain.

• An LTI system's response to an impulse input is called its **impulse response**.



• Because the system is LTI,

$$a\delta[n-n_0]$$
 \longrightarrow LTI SYSTEM \longrightarrow $ah[n-n_0]$

Not only that, but also

$$\begin{array}{c} a_1 \delta[n-n_1] \\ + a_2 \delta[n-n_2] \end{array} \longrightarrow \begin{array}{c} a_1 h[n-n_1] \\ + a_2 h[n-n_2] \end{array}$$

Now, extending this all the way,

$$\sum_{k=-\infty}^{\infty} a_k \delta[n-k] \longrightarrow \lim_{k=-\infty}^{\infty} a_k h[n-k]$$

• If only all signals came in the form

$$\sum_{k=-\infty}^{\infty} a_k \delta[n-k]$$

- Then we would figure out the output for any input in terms of the impulse response.
- But they DO come in that form!!! Recall that

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

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

which implies

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

- This sum is known as the convolution sum.
- <u>Summary</u>: If you know the impulse response of an LTI system, you know everything there is to know!

The convolution sum

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

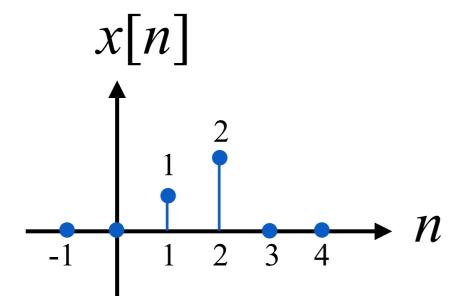
This operation is also denoted as

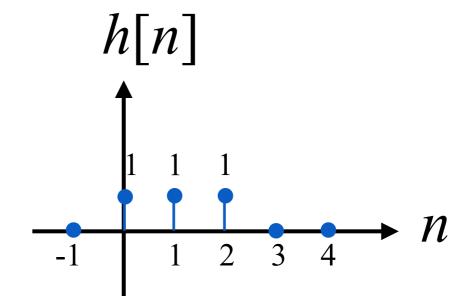
$$y[n] = x[n] \star h[n]$$

- As before, two ways to interpret this formula:
 - An infinite summation of shifted impulse responses h[n-k] each scaled with x[k].
 - For every n, an infinite sum of the samples of the product signal x[k] h[n-k].

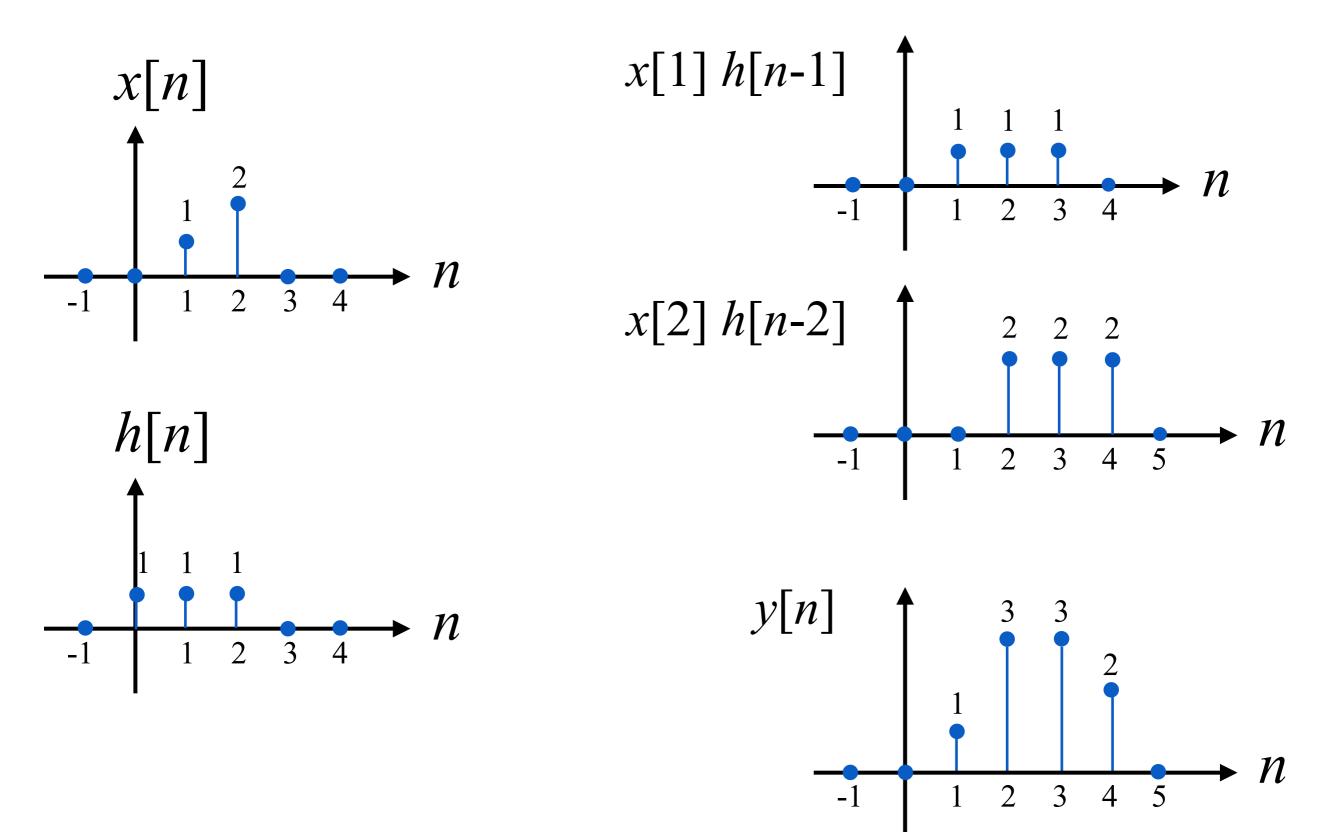
The convolution sum

• Example: Find $y[n] = x[n] \star h[n]$ if

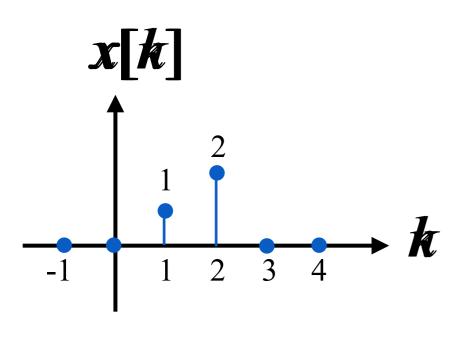


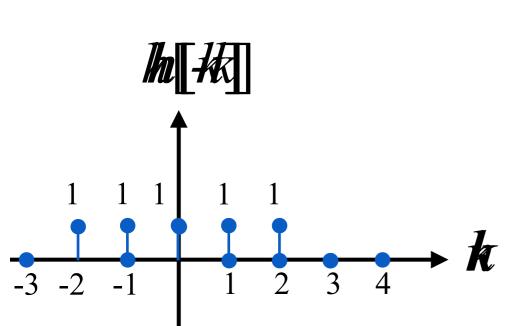


• Method 1: Accumulate x[k] h[n-k]'s.

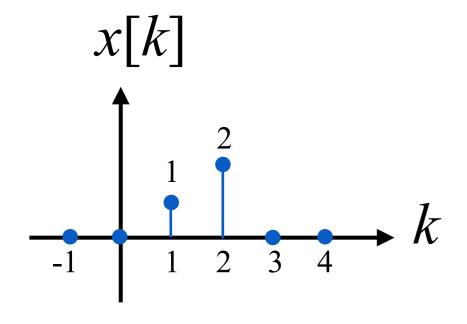


• Method 2: Calculate sum for each *n*.





- Change the running variable to *k*
- To plot h[-k], flip h[k] around the y-axis.
- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].

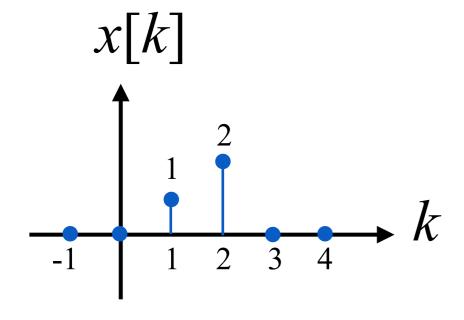


- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 0, the sum yields

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

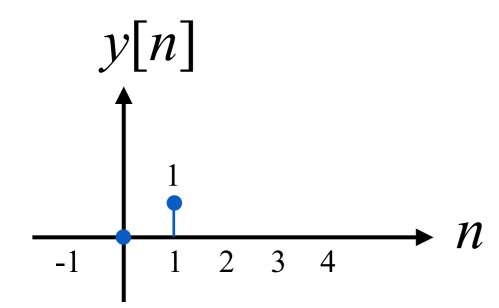
$$y[n]$$

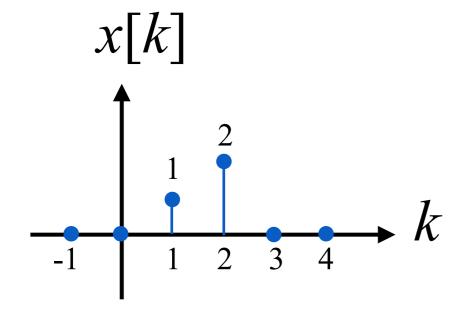
$$1 \quad 2 \quad 3 \quad 4$$



- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 1, the sum yields

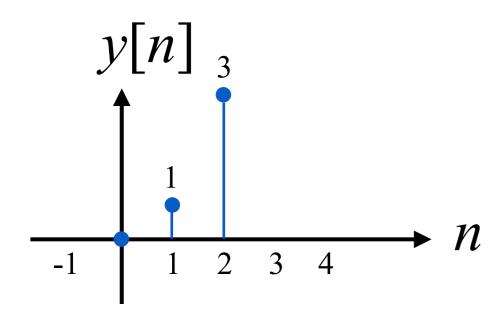
$$y[1] = \sum_{k=-\infty}^{\infty} x[k]h[1-k]$$
$$= 1 \times 1$$

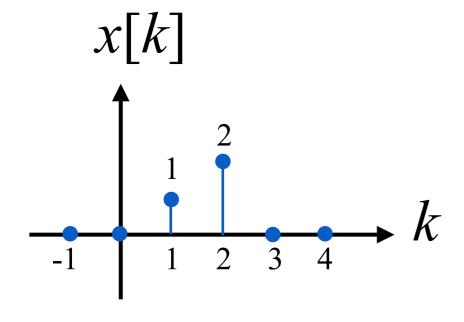




- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 2, the sum yields

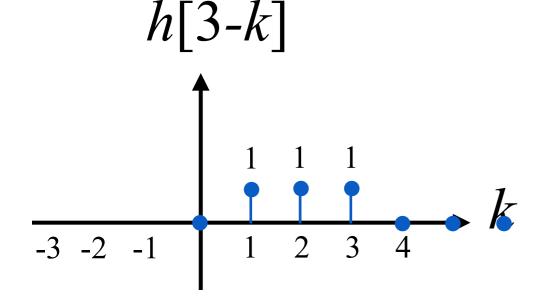
$$y[2] = \sum_{k=-\infty}^{\infty} x[k]h[2-k]$$
$$= 1 \times 1 + 2 \times 1$$
$$= 3$$

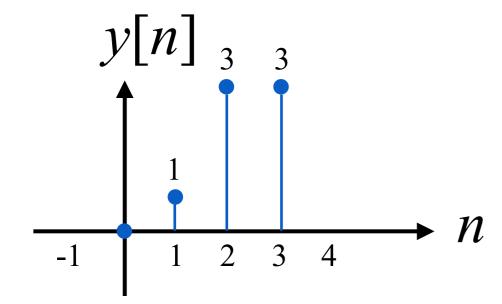


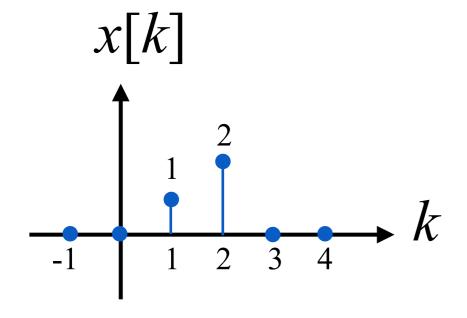


- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 3, the sum yields

$$y[3] = \sum_{k=-\infty}^{\infty} x[k]h[3 - k]$$
$$= 1 \times 1 + 2 \times 1$$
$$= 3$$

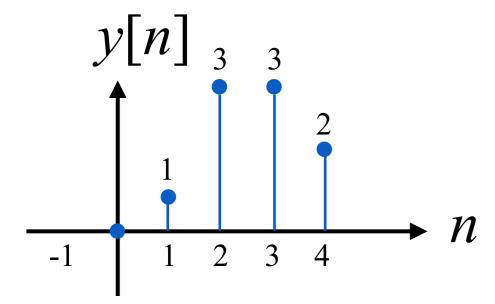


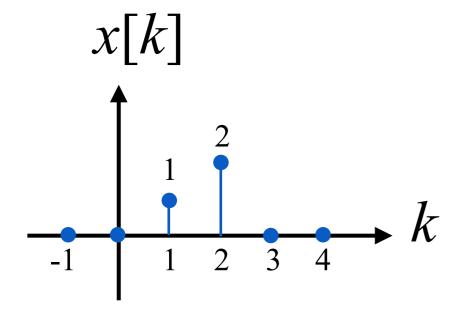




- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 4, the sum yields

$$y[4] = \sum_{k=-\infty}^{\infty} x[k]h[4-k]$$
$$= 2 \times 1$$
$$= 2$$





- To plot h[n-k], shift h[-k] to the right by n units.
- For each n, sum up all samples of the product signal x[k]h[n-k].
- For n = 5, the sum yields

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

$$y[n]_{3} \xrightarrow{3}$$

$$1$$

$$1$$

$$1$$

$$2$$

$$-1$$

$$1$$

$$2$$

$$3$$

$$4$$

$$5$$

The convolution sum

• Example: Find $y[n] = x[n] \star h[n]$ if

$$x[n] = u[n]$$
$$h[n] = 0.5^n u[n]$$

- Now, since x[n] has infinitely many non-zero samples, Method 1 won't work.
- For Method 2, the infinite sum becomes

$$y[n] = \sum_{k=-\infty}^{\infty} 0.5^{n-k} u[k] u[n-k]$$

$$y[n] = \sum_{k=-\infty}^{\infty} 0.5^{n-k} u[k] u[n-k]$$

$$= \sum_{k=0}^{\infty} 0.5^{n-k} u[n-k]$$

$$= 0.5^{n} \sum_{k=0}^{\infty} 0.5^{-k} u[n-k]$$

- If n < 0, the above sum is zero since it contains u[n], u[n-1], u[n-2], ..., all of which is zero.
- Otherwise, u[n-k] = 0 only when k > n. Thus,

$$y[n] = 0.5^n u[n] \sum_{k=0}^{\infty} 0.5^{-k}$$

Digression: Power sums

• How do we compute $S = \sum_{k=0}^{\infty} \alpha^k$?

• Here is the trick:

$$S = 1 + \alpha + \alpha^2 + \dots + \alpha^n$$

$$\alpha S = \alpha + \alpha^2 + \alpha^3 + \dots + \alpha^{n+1}$$

$$\alpha S + 1 - \alpha^{n+1} = 1 + \alpha + \alpha^2 + \dots + \alpha^n$$

$$\alpha S + 1 - \alpha^{n+1} = S$$

$$\alpha S + 1 - \alpha^{n+1} = S$$

In other words,

$$S = \frac{\alpha^{n+1} - 1}{\alpha - 1}$$

• What about the infinite sum $S = \sum_{k=0}^{\infty} \alpha^k$?

• Converges only if $|\alpha| < 1$, and to

$$S = \frac{-1}{\alpha - 1} = \frac{1}{1 - \alpha}$$

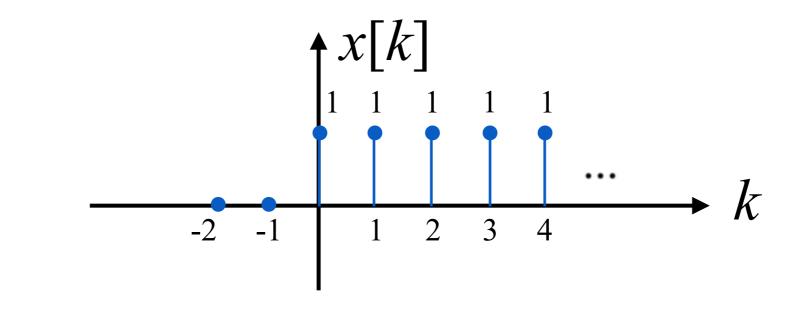
Back to the example

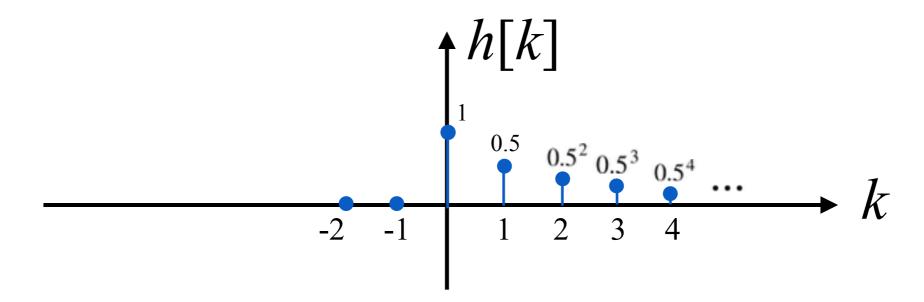
• Example: Find $y[n] = x[n] \star h[n]$ if

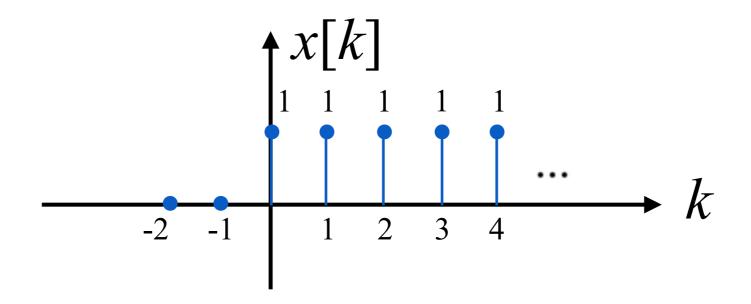
$$x[n] = u[n]$$
$$h[n] = 0.5^n u[n]$$

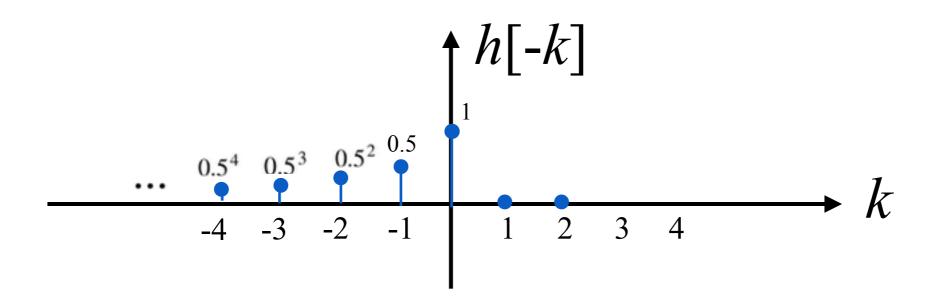
We had found

$$y[n] = 0.5^{n} u[n] \sum_{k=0}^{n} 0.5^{-k} = 0.5^{n} u[n] \frac{2^{n+1} - 1}{2 - 1}$$
$$= u[n] (2^{n+1} 0.5^{n} - 0.5^{n})$$
$$= (2 - 0.5^{n}) u[n]$$

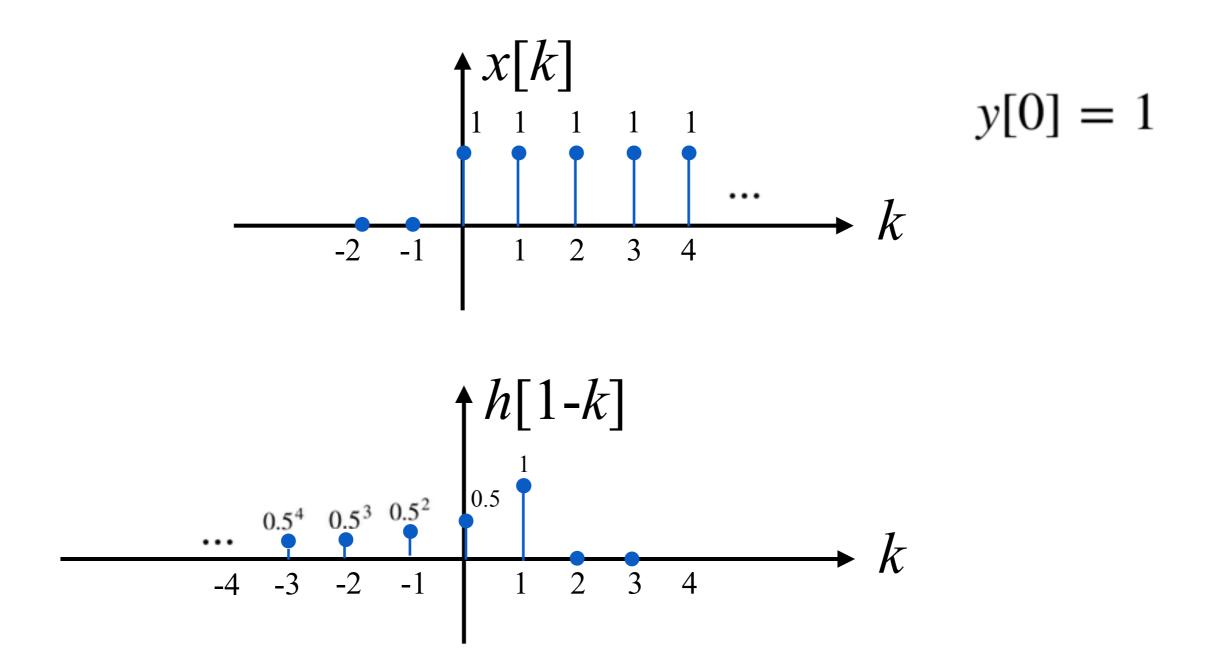




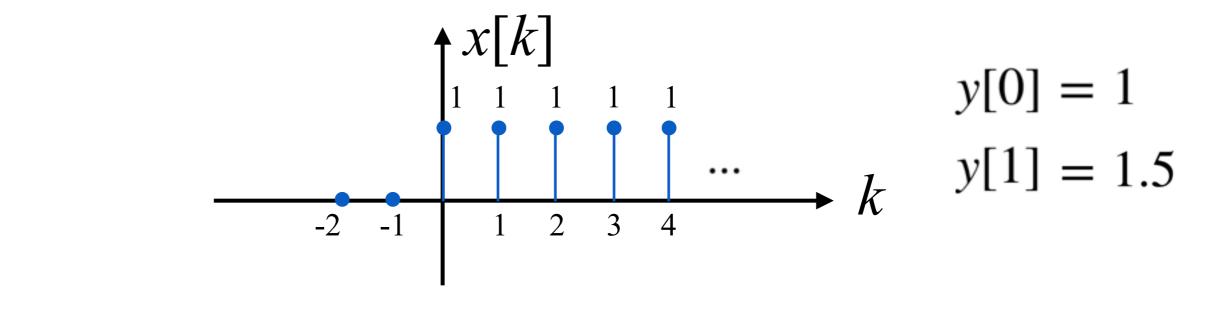


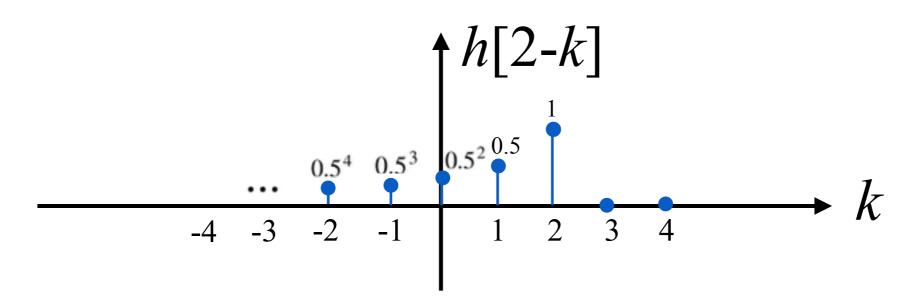


$$y[0] = 1$$

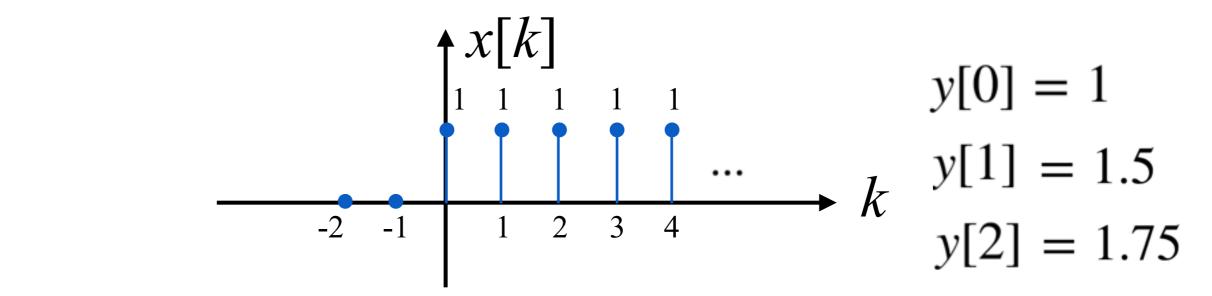


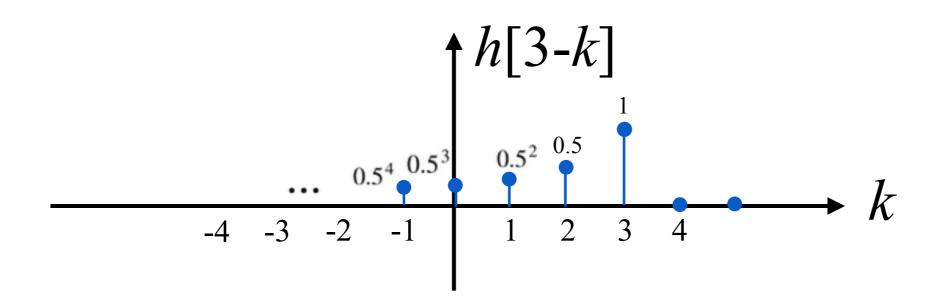
$$y[1] = 1 + 0.5 = 1.5$$





$$y[2] = 1 + 0.5 + 0.5^2 = 1.75$$





$$y[3] = 1 + 0.5 + 0.5^2 + 0.5^3 = 1.875$$

• Commutativity:

$$x[n] \star h[n] = h[n] \star x[n]$$

• Proof:

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

 $m=-\infty$

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

$$= \sum_{n=0}^{\infty} h[m]x[n-m] = h[n] \star x[n]$$

• Associativity:

$$x[n] \star (y[n] \star z[n]) = (x[n] \star y[n]) \star z[n]$$

• Proof:

$$x[n] \star (y[n] \star z[n]) = x[n] \star \sum_{k=-\infty}^{\infty} y[k]z[n-k]$$

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

$$= \sum_{l=-\infty}^{\infty} x[l] \sum_{m=-\infty}^{\infty} y[m-l] z[n-m]$$

$$x[n] \star (y[n] \star z[n]) \stackrel{(m=k+l)}{=} \sum_{l=-\infty}^{\infty} x[l] \sum_{m=-\infty}^{\infty} y[m-l] z[n-m]$$

$$= \sum_{m=-\infty}^{\infty} z[n-m] \sum_{l=-\infty}^{\infty} x[l]y[m-l]$$

$$= \sum_{m=-\infty}^{\infty} z[n-m](x[m] \star y[m])$$

$$= (x[n] \star y[n]) \star z[n]$$

• Linearity:

$$x_1[n] \star h[n] = y_1[n]$$

 $x_2[n] \star h[n] = y_2[n]$

implies

$$(ax_1[n] + bx_2[n]) \star h[n] = ay_1[n] + by_2[n]$$

• <u>Proof</u>: Follows from the fact that convolution of the input with the impulse response yields the output for **linear** and time-invariant systems.

• The same logic leads to Time-invariance:

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

implies

$$x[n - n_0] \star h[n] = y[n - n_0]$$

• Thanks to commutativity, we also have

$$x[n-n_1] \star h[n-n_2] = y[n-(n_1+n_2)]$$

Time-reversal:

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

implies

$$x[-n] \star h[-n] = y[-n]$$

• Proof:

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

$$\stackrel{\stackrel{l=-k)}{=}}{=} \sum_{l=\infty}^{\infty} x[l]h[-l-n] = y[-n]$$

• Identity element:

$$x[n] \star \delta[n] = x[n]$$

• Proof:

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

• All the terms in the above sum is zero, except at k = 0, where it is equal to x[n].

- For an LTI system, we can tell whether the system is **memoryless**, **causal**, **stable**, or **invertible** just by analyzing the impulse response.
- It may be more convenient to write the convolution sum as

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

• In general, this indicates that y[n] depends on all samples of x[n].

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

Write this more openly as

$$y[n] = ... + h[2]x[n - 2]$$

+ $h[1]x[n - 1]$ PAST
+ $h[0]x[n]$ PRESENT
+ $h[-1]x[n + 1]$ FUTURE
+ $h[-2]x[n + 2] + ...$

- For the system to be **memoryless**, the present value of y[n] must depend only on the **present** value of x[n].
- That is the same as

$$h[n] = 0 \quad \forall n \neq 0$$

• In other words, the impulse response must be of the form $h[n] = c\delta[n]$ for some c.

- For the system to be **causal**, the present value of y[n] must depend only on the **present** and **past** values of x[n].
- That is the same as

$$h[n] = 0 \quad \forall n < 0$$

• In other words, the impulse response must be of the form h[n] = g[n]u[n] for some g[n].

• For **stability**, let us analyze |y[n]|:

$$|y[n]| = \left| \sum_{k=-\infty}^{\infty} h[k]x[n-k] \right|$$

$$\leq \sum_{k=-\infty}^{\infty} |h[k]| \cdot |x[n-k]|$$

• Now, if |x[n]| is bounded by B for all n,

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

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

• Therefore, a sufficient condition for stability is

$$\sum_{k=-\infty}^{\infty} \left| h[k] \right| < \infty$$

• It is also **necessary** because otherwise, we could just select x[n] = sign(h[-n]) to obtain

$$y[0] = \sum_{k=-\infty}^{\infty} h[k]x[-k] = \sum_{k=-\infty}^{\infty} h[k]\operatorname{sign}(h[k]) = \sum_{k=-\infty}^{\infty} |h[k]|$$

• The system has an LTI **inverse** if and only if there exists a signal g[n] such that

$$x[n] \star h[n] \star g[n] = x[n]$$
 for all $x[n]$.

• This is equivalent to

$$h[n] \star g[n] = \delta[n]$$

• If such g[n] exists, it is the impulse response of the inverse system.

• Example: Determine if the system is memoryless, causal, stable, or invertible if its impulse response is given by

$$h[n] = 0.5^n u[n]$$

• Memory: h[n] is not of the form $c\delta[n]$

HAS
MEMORY
CAUSAL

• Causality: h[n] is of the form g[n]u[n]



Stability:

$$\sum_{k=-\infty}^{\infty} |h[k]| = \sum_{k=0}^{\infty} 0.5^k = \frac{1}{1-0.5} = 2$$
 STABLE

• Example: Determine if the system is memoryless, causal, stable, or invertible if its impulse response is given by

$$h[n] = 0.5^n u[n]$$

• Invertibility: Observe that

$$h[n] - 0.5h[n - 1] = 0.5^{n}u[n] - 0.5 \cdot 0.5^{n-1}u[n - 1]$$

$$= 0.5^{n} (u[n] - u[n - 1])$$

$$= 0.5^{n} \delta[n]$$

$$= \delta[n]$$

• Invertibility: Observe that

$$h[n] - 0.5h[n - 1] = \delta[n]$$

- Now, can we rewrite this as $h[n] \star g[n] = \delta[n]$ for some g[n]?
- Yes. Take $g[n] = \delta[n] 0.5\delta[n-1]$:

$$h[n] \star (\delta[n] - 0.5\delta[n-1])$$

$$= h[n] \star \delta[n] - 0.5h[n] \star \delta[n-1]$$

$$= h[n] - 0.5h[n - 1]$$



• Example: Determine if the system is memoryless, causal, stable, or invertible if its impulse response is given by

$$h[n] = (-1)^n$$

• Memory: h[n] is not of the form $c\delta[n]$

HAS MEMORY

- Causality: h[n] is not of the form $g[n]u[n] \frac{NON}{CAUSAL}$
- Stability:

$$\sum_{k=-\infty}^{\infty} \left| h[k] \right| = \sum_{k=-\infty}^{\infty} \left| (-1)^k \right| = \sum_{k=-\infty}^{\infty} 1 = \infty \quad \text{UNSTABLE}$$

• Invertibility: If g[n] exists such that

$$h[n] \star g[n] = \delta[n]$$

what would be the result of $h[n-2] \star g[n]$?

- Due to time invariance, it must be $\delta[n-2]$
- Due to the fact that h[n-2] = h[n], it must be $\delta[n]$
- Contradiction!!!!
- No such g[n] can exist.

NOT INVERTIBLE