

6.003: Signal Processing

Convolution and Filtering

good morning! we will
start @ 8:05am
Eastern

time domain

$$y(t) = (h * x)(t) = \int h(\tau)x(t - \tau) d\tau$$

$$y[n] = (h * x)[n] = \sum_m h[m]x[n - m]$$

$$x[n] \rightarrow \boxed{h[n]} \rightarrow y[n]$$

$(x * h)[n]$

frequency domain

$$Y(\omega) = H(\omega)X(\omega)$$

$$Y(\Omega) = H(\Omega)X(\Omega)$$

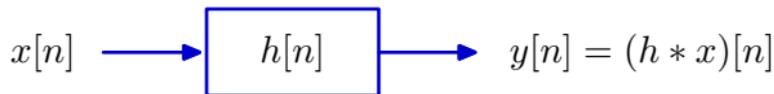
$$X(\Omega) \rightarrow \boxed{H(\Omega)} \rightarrow Y(\Omega)$$

$X(\Omega)H(\Omega)$

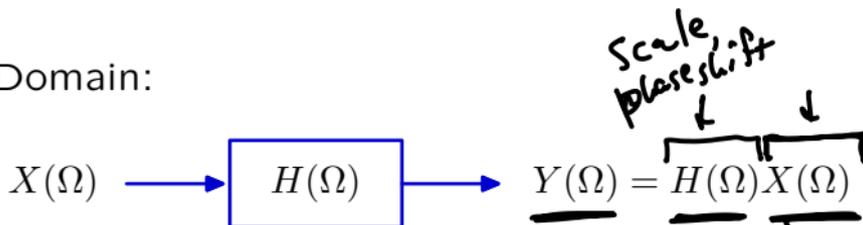
Filtering

We can view filtering in both the time and frequency domains:

Time Domain:



Frequency Domain:



Each frequency component $X(\Omega)$ is scaled by a factor $H(\Omega)$, which can be possibly complex.

The system is completely described by the set of scale factors $H(\cdot)$, which we refer to as the **frequency response** of the system.

$$y[n] = (x * h)[n]$$

$$Y(\Omega) = \sum_{n=-\infty}^{\infty} y[n] e^{-j\Omega n} = \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} x[m] h[n-m] e^{-j\Omega n}$$

swap order

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

pull this out of inner sum

now, let $p = n - m$

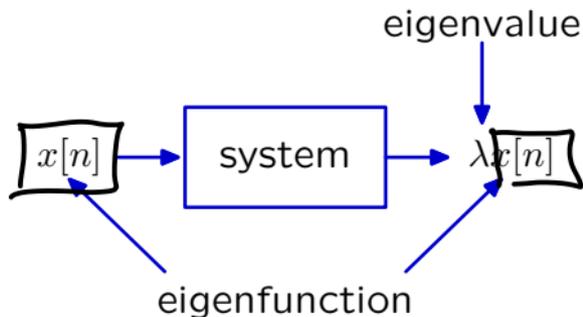
$$= \sum_{m=-\infty}^{\infty} x[m] \sum_{p=-\infty}^{\infty} h[p] e^{-j\Omega p} e^{-j\Omega m}$$

pull this out

$$= \sum_{m=-\infty}^{\infty} x[m] e^{-j\Omega m} \sum_{p=-\infty}^{\infty} h[p] e^{-j\Omega p} = X(\Omega) \cdot H(\Omega)$$

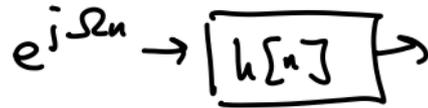
Eigenfunctions and Eigenvalues

If the output signal is a scalar multiple of the input signal, we refer to the signal as an **eigenfunction**, and the multiplier as the **eigenvalue**.



Complex Exponentials

Complex exponentials are eigenfunctions of LTI systems. If $h[\cdot]$ is a system's unit sample response, and $x[n] = e^{j\Omega n}$, then the system's output is:



$$y[n] = \sum_{m=-\infty}^{\infty} h[m] e^{j\Omega(n-m)}$$

The equation shows the convolution sum for the output $y[n]$. The input signal $e^{j\Omega(n-m)}$ is highlighted with a red bracket and a blue circle around the $e^{j\Omega n}$ term, indicating the separation of the input from the system response.

$$= e^{j\Omega n} \sum_{m=-\infty}^{\infty} h[m] e^{-j\Omega m}$$

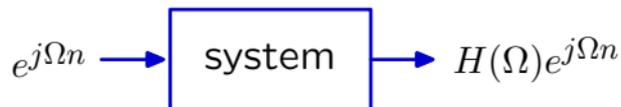
The equation is simplified by factoring out the input signal $e^{j\Omega n}$. The remaining sum $\sum_{m=-\infty}^{\infty} h[m] e^{-j\Omega m}$ is highlighted with a pink bracket, representing the system's frequency response $H(\Omega)$.

$$= e^{j\Omega n} H(\Omega) = |H(\Omega)| e^{j(\Omega n + \angle H(\Omega))}$$

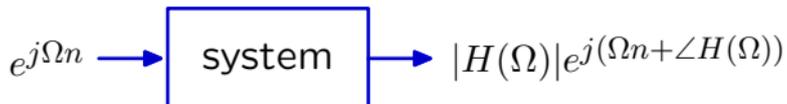
The final result shows the output signal as the product of the input signal and the system's frequency response $H(\Omega)$. The magnitude $|H(\Omega)|$ and the phase $\angle H(\Omega)$ are highlighted with pink underlines, showing that the output is a complex exponential with the same frequency as the input, but with a magnitude and phase shift determined by the system's response.

Complex Exponentials

Complex exponentials are eigenfunctions of LTI systems.



The eigenvalues $H(\Omega)$ are generally complex-valued, and so they affect both the amplitude and phase of the output:



Response to Eternal Sinusoids

If $h[n]$ is purely real, we have $H(-\Omega) = H^*(\Omega)$.

Consider $x[n] = \cos(\Omega n)$ (for all n), which can be written as:

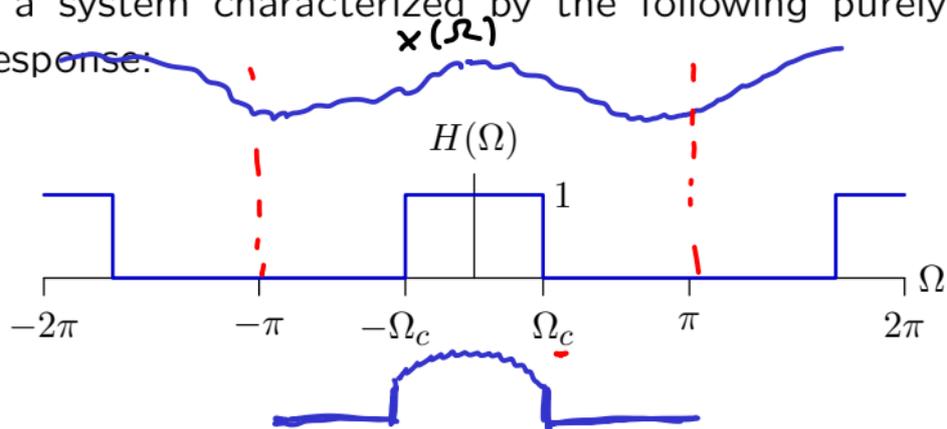
$$x[n] = \frac{1}{2} \left(e^{j\Omega n} + e^{-j\Omega n} \right) \quad \cos(\Omega n) \rightarrow \boxed{h[n]} \rightarrow$$

Then:

$$\begin{aligned} y[n] &= \frac{1}{2} \left(e^{j\Omega n} H(\Omega) + e^{-j\Omega n} H^*(\Omega) \right) \\ &= \operatorname{Re} \left(e^{j\Omega n} H(\Omega) \right) \\ &= \operatorname{Re} \left(|H(\Omega)| e^{j(\Omega n + \angle H(\Omega))} \right) \\ &= |H(\Omega)| \cos(\Omega n + \angle H(\Omega)) \end{aligned}$$

The "Ideal" Low-Pass Filter

Consider a system characterized by the following purely real frequency response:

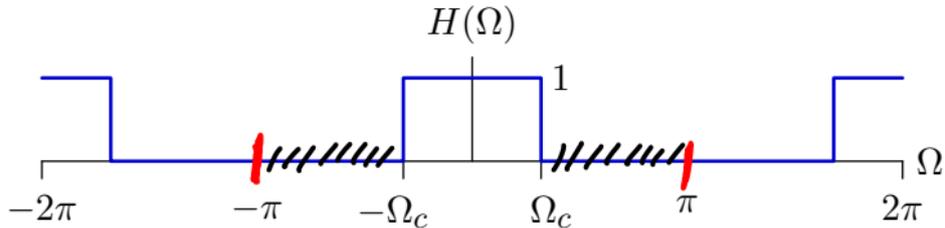


Such a system is called a low-pass filter, because it allows low frequencies to pass through unmodified, while attenuating high frequencies.

We could apply this filter to a signal by multiplying the DTFT of that signal by the values above. But we could also apply the filter by operating in the time domain.

$$(x * h)[n]$$

The "Ideal" Low-Pass Filter



We can apply this filter to a signal by convolving with its unit sample response. What is the unit sample response of the system whose frequency response is shown above?

$$(x * h)[n]$$

unit sample resp \longleftrightarrow freq resp
 $h \longleftrightarrow H$
 DTFT

$$h[n] = \frac{1}{2\pi} \int_{-2\pi}^{2\pi} H(\Omega) e^{j\Omega n} d\Omega = \frac{1}{2\pi} \int_{-\Omega_c}^{\Omega_c} e^{j\Omega n} d\Omega$$

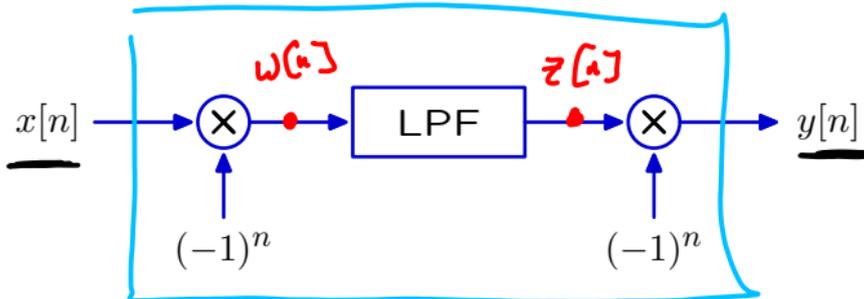
$$= \frac{1}{2\pi} \frac{1}{jn} e^{j\Omega n} \Big|_{\Omega=-\Omega_c}^{\Omega_c}$$

$$= \frac{1}{2\pi j n} \left(e^{j\Omega_c n} - e^{-j\Omega_c n} \right)$$

$\frac{2j \sin(\Omega_c n)}{2\pi n}$

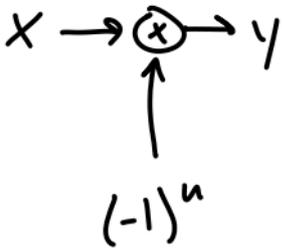
Cascaded System

Consider this system, where LPF represents a lowpass filter of the form discussed on the previous slides.



How many of the following statements are true?

- The transformation from $x[n]$ to $y[n]$ is linear.
- The transformation from $x[n]$ to $y[n]$ is time invariant.
- The transformation from $x[n]$ to $y[n]$ is a high-pass filter.



linear? ✓

$$x[n] = a x_1[n] + b x_2[n]$$

$$y[n] \stackrel{?}{=} a y_1[n] + b y_2[n]$$

$$y[n] = (-1)^n (a x_1[n] + b x_2[n])$$

$$= a \underbrace{(-1)^n x_1[n]}_{y_1[n]} + b \underbrace{(-1)^n x_2[n]}_{y_2[n]}$$

time-invariant? ✗

$$x[n] = x_1[n - n_0]$$

$$y[n] \stackrel{?}{=} y_1[n - n_0] = (-1)^{n - n_0} x_1[n - n_0]$$

$$= (-1)^n x_1[n - n_0]$$

$$X \rightarrow \boxed{\text{LPF}} \rightarrow Y$$

$$Y(\Omega) = X(\Omega) H(\Omega)$$

\Downarrow

$$y[n] = (x * h)[n]$$

[convolution is linear
time-invariant

(proof on next 2 pages)

linearity of convolution

$$x[n] = ax_1[n] + bx_2[n]$$

$$(x * h)[n] \stackrel{?}{=} a(x_1 * h)[n] + b(x_2 * h)[n]$$

$$(x * h)[n] = \sum_{m=-\infty}^{\infty} (ax_1[m] + bx_2[m])h[n-m]$$

distribute

$$= \sum_{m=-\infty}^{\infty} (ax_1[m]h[n-m] + bx_2[m]h[n-m])$$

break apart sum

$$= a \sum_{m=-\infty}^{\infty} x_1[m]h[n-m] + b \sum_{m=-\infty}^{\infty} x_2[m]h[n-m]$$

$$= a(x_1 * h)[n] + b(x_2 * h)[n] \quad \checkmark \quad \text{convolution is linear}$$

time-invariance of convolution

$$X[n] = x_1[n-n_0]$$

$$(x * h)[n] \stackrel{?}{=} (x_1 * h)[n-n_0]$$

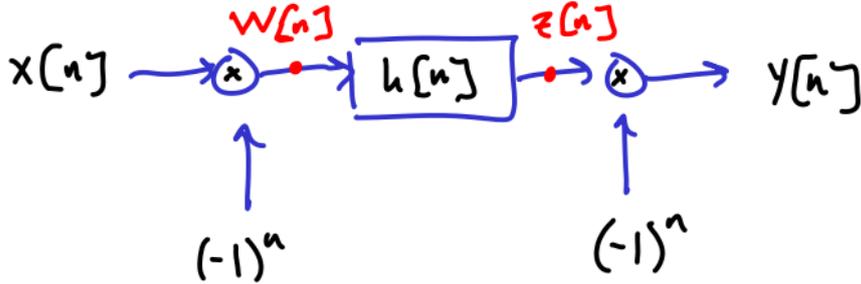
$$(x * h)[n] = \sum_{m=-\infty}^{\infty} x_1[n-n_0] h[n-m]$$

$$\text{let } p = m - n_0$$

$$= \sum_{p=-\infty}^{\infty} x_1[p] h[(n-n_0)-p]$$

$$= (x_1 * h)[n-n_0] \quad \checkmark \text{ convolution is time-invariant}$$

now back to the big system
on the next few slides



$$(-1)^{n+m} = (-1)^{n-m}$$

$$w[n] = (-1)^n x[n]$$

$$z[n] = (w * h)[n] = \sum_{m=-\infty}^{\infty} (-1)^m x[m] h[n-m]$$

$$y[n] = (-1)^n z[n]$$

$$= (-1)^n \sum_{m=-\infty}^{\infty} (-1)^m x[m] h[n-m]$$

$$= \sum_{m=-\infty}^{\infty} (-1)^{n+m} x[m] h[n-m]$$

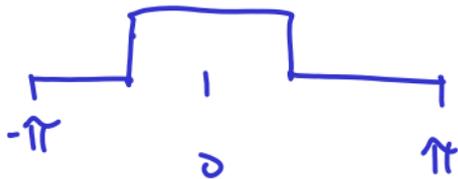
$$g[n] = (-1)^n h[n]$$

$$= \sum_{m=-\infty}^{\infty} x[m] \left((-1)^{n-m} h[n-m] \right) = (x * g)[n]$$

$$y[n] = (x * g)[n]$$

$$Y(\omega) = X(\omega)G(\omega)$$

$H(\omega)$



$G(\omega)$



high-pass filter

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

$\Downarrow \quad \underline{e^{j\pi n}}$

$$G(\omega) = \sum_{n=-\infty}^{\infty} (-1)^n h[n] e^{-j\omega n}$$

$$= \sum_{n=-\infty}^{\infty} e^{j\pi n} h[n] e^{-j\omega n}$$

$$= \sum_{n=-\infty}^{\infty} h[n] e^{-j(\omega - \pi)n}$$

$$= \underline{H(\omega - \pi)}$$