



university of
groningen

faculty of science
and engineering

Lecture 5: Frequency response of FIR filters

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Signals and Systems
1B - 2024/2025

Preliminaries

- ▶ Lab session 2 is on Brightspace. The deadline for reports and code is on **Friday, December 20, at 17:30.**
- ▶ Remember to submit your **code in Themis and documentation in Brightspace.**
- ▶ Lab 3 will be released next week.
- ▶ Office hours on Friday, December 20, are online –Google Meet.
- ▶ Tutorials next week are hybrid. More information will be provided in an announcement soon.

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Let us talk about the Lab 1:

- ▶ Thumbs up in clean code practices :D
- ▶ Chatbots and documentation not following the template or the page limit D:

You are responsible of your work!

Overview

1. Recap
2. Frequency Response
3. Properties of Frequency Response
4. Running Sum Filter
5. Image Smoothing
6. Closing Remarks

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Recap

The general form of causal FIR filters is

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Recap

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Quick questions!

- ▶ Does this equation seem familiar from other courses?
- ▶ How can we interpret $y[n]$?
- ▶ Indeed, how do you choose b_k ?

Recap

The convolution between a discrete signal x and filter h is

$$y[n] = x[n] \star h[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k]. \quad (2)$$

Indeed, the response of an **LTI filter** h is $y[n] = x[n] \star h[n]$.

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Indeed, the response of an **LTI filter** h is $y[n] = x[n] \star h[n]$.

Another quick question!

Why are LTI filters special?

Recap

We need to talk...

About convolutional neural networks and recurrent neural networks. They are not LTI systems, but...

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Sinusoidal response of FIR systems

Suppose the system faces a (discrete-time) complex exponential

$$x[n] = A e^{j\varphi} e^{j\hat{\omega}n} \quad (3)$$

Let us compute the FIR filter for such a signal.

Sinusoidal response of FIR systems

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

Sinusoidal response of FIR systems

$$\begin{aligned}y[n] &= \sum_{k=0}^M b_k x[n-k] \\&= \sum_{k=0}^M b_k A e^{j\varphi} e^{j\hat{\omega}(n-k)}\end{aligned}$$

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$$\begin{aligned}y[n] &= \sum_{k=0}^M b_k x[n-k] \\&= \sum_{k=0}^M \mathbf{b}_k A e^{j\varphi} e^{j\hat{\omega}(n-k)} \\&= \left(\sum_{k=0}^M \mathbf{b}_k e^{-j\hat{\omega}k} \right) A e^{j\varphi} e^{j\hat{\omega}n} \\&= H(e^{j\hat{\omega}}) A e^{j\varphi} e^{j\hat{\omega}n}\end{aligned}$$

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If $x[n]$ is complex exponential, $y[n]$ is complex exponential

- ▶ $x[n]$ and $y[n]$ have the same radian frequency $\hat{\omega}$

Sinusoidal response of FIR systems

Let us expand the $H(e^{j\hat{\omega}})$ component.

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Frequency response has the following components: a **magnitude** $|H(e^{j\hat{\omega}})|$ and an **angle** $e^{j\angle H(e^{j\hat{\omega}})}$. The response $H(e^{j\hat{\omega}})$ is only meaningful for sinusoidal input

Teacher's comment

Avoid writing $y[n] = H(e^{j\hat{\omega}})x[n]$. This expression mixes time and frequency domains!

Time domain and frequency domain

Impulse response $h[k]$ is the time-domain representation of frequency response $H(e^{j\hat{\omega}})$.

Example: Suppose $\{b_k\} = \{-1, 3, -1\}$, for $k = 0, 1, 2$, then,

$$h[n] = \sum_{k=0}^M b_k \delta[n - k]$$

$$H(e^{j\hat{\omega}}) = \sum_{k=0}^M b_k e^{-j\hat{\omega}k}$$

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$$h[n] = \sum_{k=0}^M b_k \delta[n - k] = -\delta[n] + 3\delta[n - 1] - \delta[n - 2]$$

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$$H(e^{j\hat{\omega}}) = \sum_{k=0}^M b_k e^{-j\hat{\omega}k} = -1 + 3e^{-j\hat{\omega}} - e^{-j2\hat{\omega}}$$

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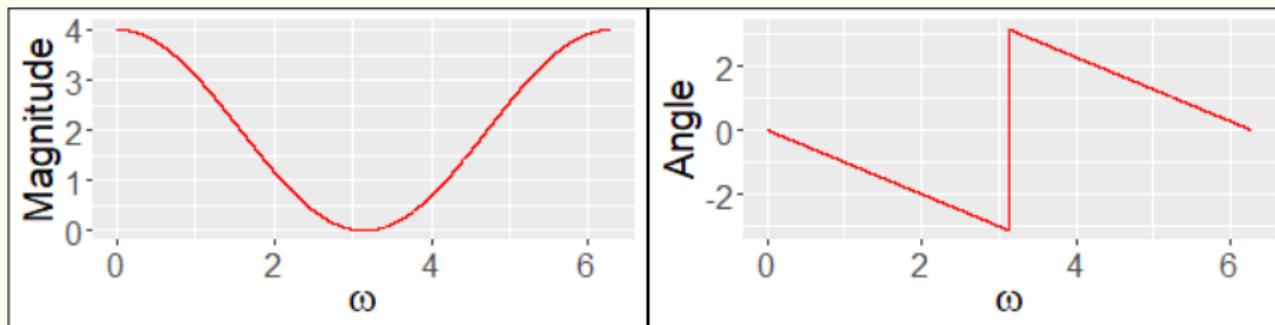


Figure: Magnitude and angle components in $H(e^{j\hat{\omega}})$ with $\{b_k\} = \{1, 2, 1\}$ for $k = 0, 1, 2$

Example - FIR complex exponential input

Consider the FIR filter $\{b_k\} = \{1, 2, 1\}$ for $k = 0, 1, 2$ and input signal

$$x[n] = 2e^{j\pi/4} e^{j(\pi/3)n} \quad (4)$$

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$$y[n] = 3e^{-j\pi/3} \cdot 2e^{j\pi/4} e^{j(\pi/3)n}$$

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$$\begin{aligned} y[n] &= 3e^{-j\pi/3} \cdot 2e^{j\pi/4} e^{j(\pi/3)n} \\ &= 6e^{j\pi/4-j\pi/3} e^{j(\pi/3)n} \\ &= 6e^{-j\pi/12} e^{j(\pi/3)n} \end{aligned}$$

Example - FIR sinusoidal input

Consider the FIR filter $\{b_k\} = \{1, 2, 1\}$ for $k = 0, 1, 2$ and input signal

$$x[n] = 2 \cos((\pi/3)n - \pi/2) = e^{-j((\pi/3)n - \pi/2)} + e^{j((\pi/3)n - \pi/2)} \quad (5)$$

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$$y[n] = 3e^{j\pi/3} e^{-j((\pi/3)n - \pi/2)} + 3e^{-j\pi/3} e^{j((\pi/3)n - \pi/2)}$$

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- ▶ The phase φ is increased by $\angle H(e^{j\hat{\omega}})$

That is,

$$y[n] = |H(e^{j\hat{\omega}})| \cdot A \cos\left(\hat{\omega}n + \varphi + \angle H(e^{j\hat{\omega}})\right) \quad (6)$$

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Comments

- ▶ **The gain of a filter cannot be negative!**
- ▶ It is called a gain even if $|H(e^{j\hat{\omega}})| < 1$
- ▶ The angle $\angle H(e^{j\hat{\omega}}) \in (-\pi, \pi]$ by convention.

First-difference filter

Consider the first-difference system $y[n] = x[n] - x[n - 1]$. Let us analize the FIR filter response.

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The magnitude and angle are

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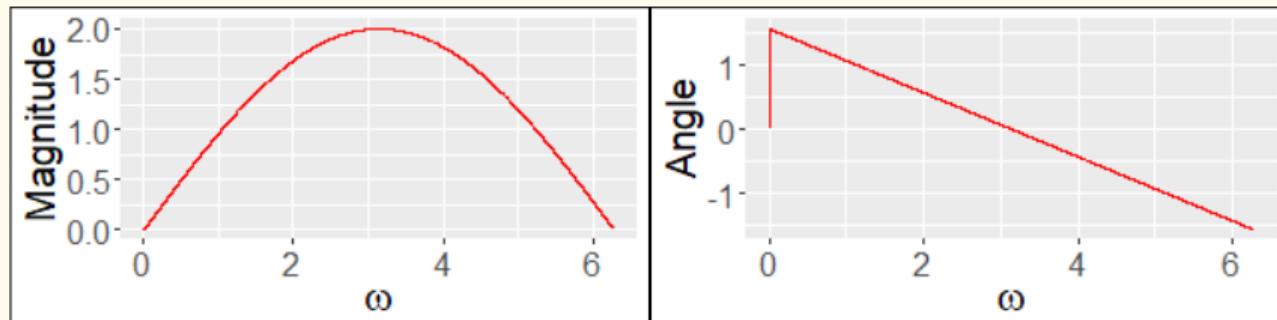


Figure: Magnitude and angle components in example system

First-difference filter

Consider again the first-difference system $y[n] = x[n] - x[n - 1]$. Let us compute the magnitude and angle of the FIR filter response in another way.

$$H(e^{j\hat{\omega}}) = 1 - e^{-j\hat{\omega}} = e^{-j\hat{\omega}/2}(e^{j\hat{\omega}/2} - e^{-j\hat{\omega}/2}) = e^{-j\hat{\omega}/2}(2 \sin(\hat{\omega}/2))$$

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$$H(e^{j\hat{\omega}}) = 1 - e^{-j\hat{\omega}} = e^{-j\hat{\omega}/2}(e^{j\hat{\omega}/2} - e^{-j\hat{\omega}/2}) = e^{-j\hat{\omega}/2}(2 \sin(\hat{\omega}/2))$$

The magnitude and angle are

$$\begin{aligned}|H(e^{j\hat{\omega}})| &= |2 \sin(\hat{\omega}/2)|\\\angle H(e^{j\hat{\omega}}) &= -\hat{\omega}/2 \quad \text{for } 0 < \hat{\omega} \leq 2\pi\end{aligned}$$

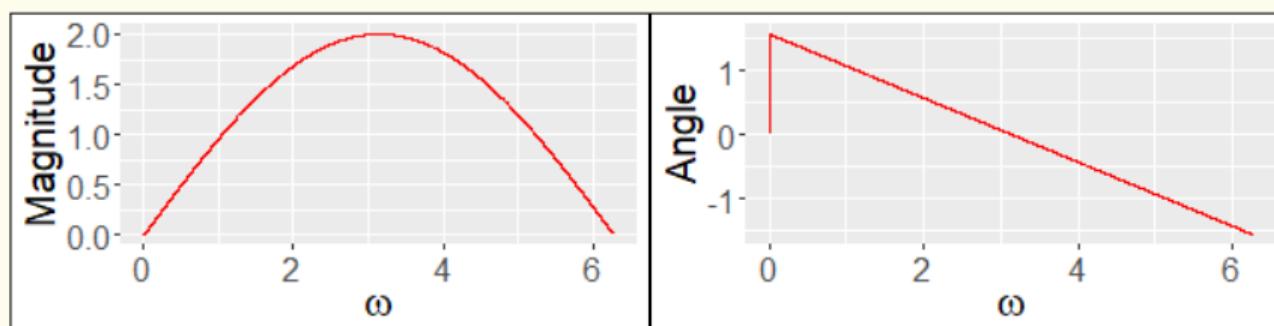


Figure: Magnitude and angle components in example system

First-difference filter

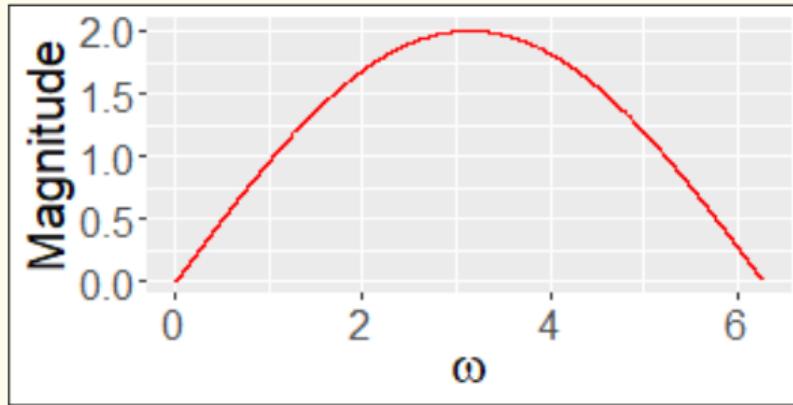


Figure: Magnitude components in example system

Note that the magnitude is zero for $\hat{\omega} = 0 + 2\pi k$.

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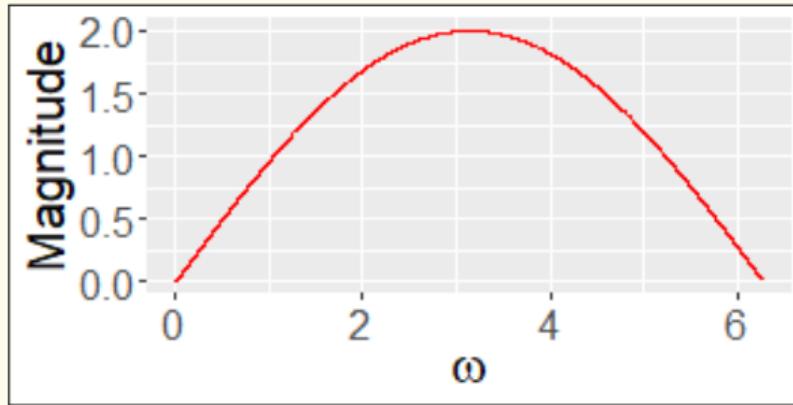


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- ▶ That is, any frequency $\hat{\omega}$ that is a multiple of 2π is eliminated.

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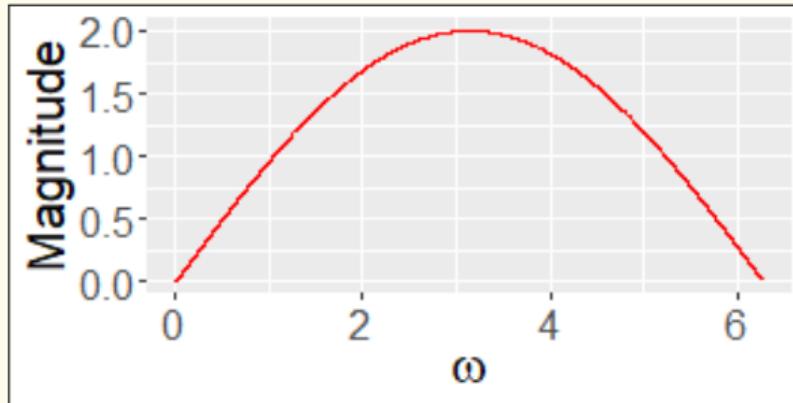


Figure: Magnitude components in example system

Note that the magnitude is zero for $\hat{\omega} = 0 + 2\pi k$.

- ▶ That is, any frequency $\hat{\omega}$ that is a multiple of 2π is eliminated.
- ▶ In particular, the DC component ($\hat{\omega} = 0$) is removed.

This behaviour makes sense for a first-difference filter...Why?

Eliminating a single frequency

Consider the FIR filter $\{b_k\} = \{1, \alpha, 1\}$ for $k = 0, 1, 2$.

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The magnitude in the FIR filter is

$$|H(e^{j\hat{\omega}})| = |\alpha + 2 \cos(\hat{\omega})|$$

By setting $|H(e^{j\hat{\omega}})| = 0$, we note this filter eliminates $\hat{\omega} = \arccos(-\alpha/2)$. E.g. $\alpha = -1$ eliminates $\hat{\omega} = 1/3\pi$.

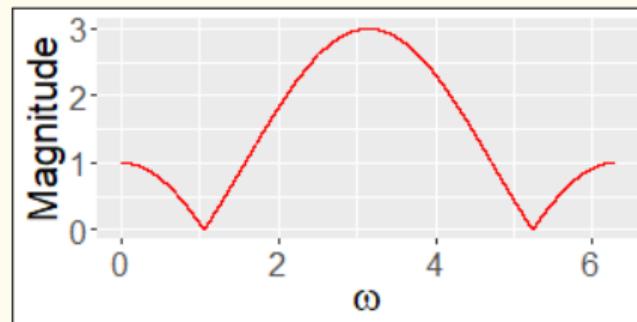


Figure: Magnitude and angle components in example system

Example - Eliminating a single frequency

Consider the FIR filter $\{b_k\} = \{1, -1, 1\}$ for $k = 0, 1, 2$, and the input signal

$$x[n] = 4 + 3 \cos\left(\frac{1}{3}\pi n - \frac{\pi}{2}\right) + 3 \cos\left(\frac{7}{8}\pi n\right) \quad (7)$$

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The output is

$$y[n] = 1 \cdot 4 + 0 \cdot 3 \cos\left(\frac{1}{3}\pi n - \frac{\pi}{2}\right) + 2.8477 \cdot 3 \cos\left(\frac{7}{8}\pi n + \frac{1}{8}\pi\right)$$

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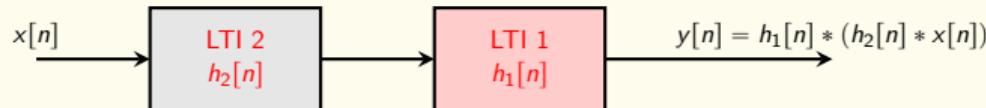
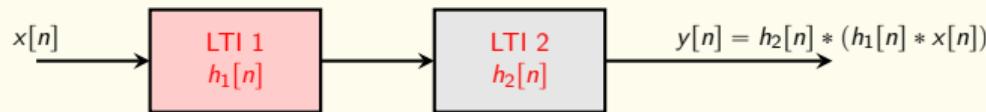
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$$\begin{aligned} y[n] &= 1 \cdot 4 + 0 \cdot 3 \cos\left(\frac{1}{3}\pi n - \frac{\pi}{2}\right) + 2.8477 \cdot 3 \cos\left(\frac{7}{8}\pi n + \frac{1}{8}\pi\right) \\ &= 4 + 8.5433 \cos\left(\frac{7}{8}\pi n + \frac{1}{8}\pi\right) \end{aligned}$$

Cascaded LTI systems

LTI cascades connect multiple LTI systems together

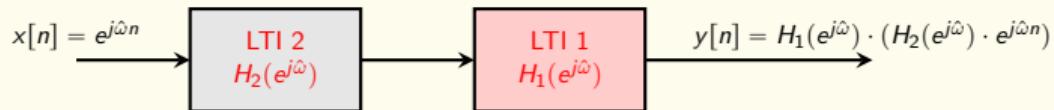
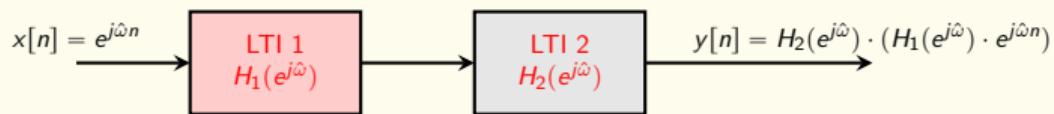
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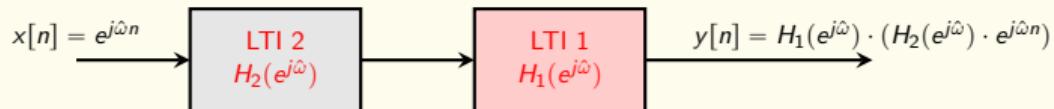
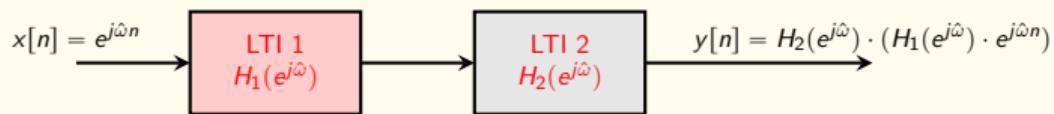
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Convolution in the time domain is multiplication in the frequency domain.

Cascaded LTI systems

Convolving FIR filters equals multiplying their frequency responses. Consider the following FIR filters in time-domain,

$$h_1[n] = \{2, 1, 2\},$$

$$h_2[n] = \{0, 3, 0, -3\}$$

The respective frequency-domain response is

$$H_1(e^{j\hat{\omega}}) = 2 + e^{-j\hat{\omega}} + 2e^{-j2\hat{\omega}}$$

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Break!

See you at _____

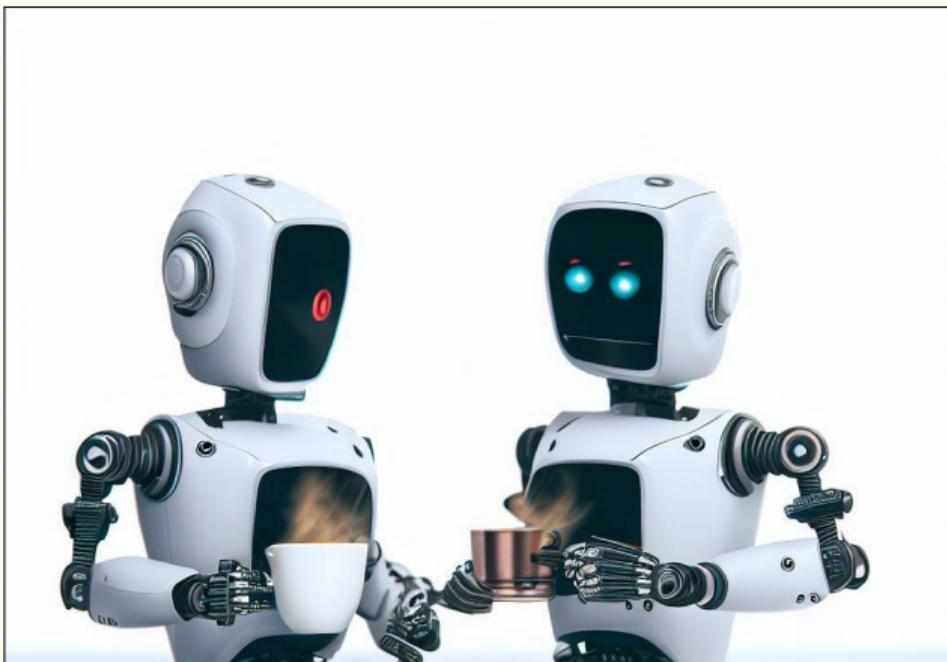


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1. Recap
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3. Properties of Frequency Response
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5. Image Smoothing
6. Closing Remarks

Running sum

The L -point running sum filter is a common LTI system

$$y[n] = \sum_{k=0}^{L-1} x[n-k] = x[n] + x[n-1] + \cdots + x[n-L+1] \quad (8)$$

What is the frequency response?

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What is the frequency response?

$$\begin{aligned} H(e^{j\hat{\omega}}) &= 1 + e^{-j\hat{\omega}} + e^{-j2\hat{\omega}} + \cdots + e^{-j(L-1)\hat{\omega}} \\ -e^{-j\hat{\omega}}H(e^{j\hat{\omega}}) &= -e^{-j\hat{\omega}} - e^{-j2\hat{\omega}} - \cdots - e^{-j(L-1)\hat{\omega}} - e^{-jL\hat{\omega}} \end{aligned}$$

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$$\frac{H(e^{j\hat{\omega}})}{-e^{-j\hat{\omega}}H(e^{j\hat{\omega}})} = \frac{1 + e^{-j\hat{\omega}} + e^{-j2\hat{\omega}} + \cdots + e^{-j(L-1)\hat{\omega}}}{(1 - e^{-j\hat{\omega}})H(e^{j\hat{\omega}})} = \frac{1 + 0 + 0 + \cdots + 0}{-e^{-j\hat{\omega}} - e^{-j2\hat{\omega}} - \cdots - e^{-j(L-1)\hat{\omega}} - e^{-jL\hat{\omega}}}$$

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By solving the equation above, we have

$$H(e^{j\hat{\omega}}) = \frac{1 - e^{-jL\hat{\omega}}}{1 - e^{-j\hat{\omega}}} \quad (9)$$

Running sum

$$H(e^{j\hat{\omega}}) = \frac{1 - e^{-jL\hat{\omega}}}{1 - e^{-j\hat{\omega}}}$$

Running sum

$$\begin{aligned} H(e^{j\hat{\omega}}) &= \frac{1 - e^{-jL\hat{\omega}}}{1 - e^{-j\hat{\omega}}} \\ &= \frac{e^{-j\hat{\omega}L/2}}{e^{-j\hat{\omega}/2}} \cdot \left(\frac{e^{j\hat{\omega}L/2} - e^{-j\hat{\omega}L/2}}{e^{j\hat{\omega}/2} - e^{-j\hat{\omega}/2}} \right) \end{aligned}$$

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Running sum

$$\begin{aligned} H(e^{j\hat{\omega}}) &= \frac{1 - e^{-jL\hat{\omega}}}{1 - e^{-j\hat{\omega}}} \\ &= \frac{e^{-j\hat{\omega}L/2}}{e^{-j\hat{\omega}/2}} \cdot \left(\frac{e^{j\hat{\omega}L/2} - e^{-j\hat{\omega}L/2}}{e^{j\hat{\omega}/2} - e^{-j\hat{\omega}/2}} \right) \\ &= e^{-j\hat{\omega}(L-1)/2} \cdot \left(\frac{\sin(\hat{\omega}L/2)}{\sin(\hat{\omega}/2)} \right) \\ &= e^{-j\hat{\omega}(L-1)/2} \cdot D_L(\hat{\omega}) \end{aligned}$$

$$D_L(\hat{\omega}) = \frac{\sin(\hat{\omega}L/2)}{\sin(\hat{\omega}/2)}$$

The function $D_L(\hat{\omega})$ is called the Dirichlet form

Dirichlet form

Consider an L -point running sum filter with $L = 11$.

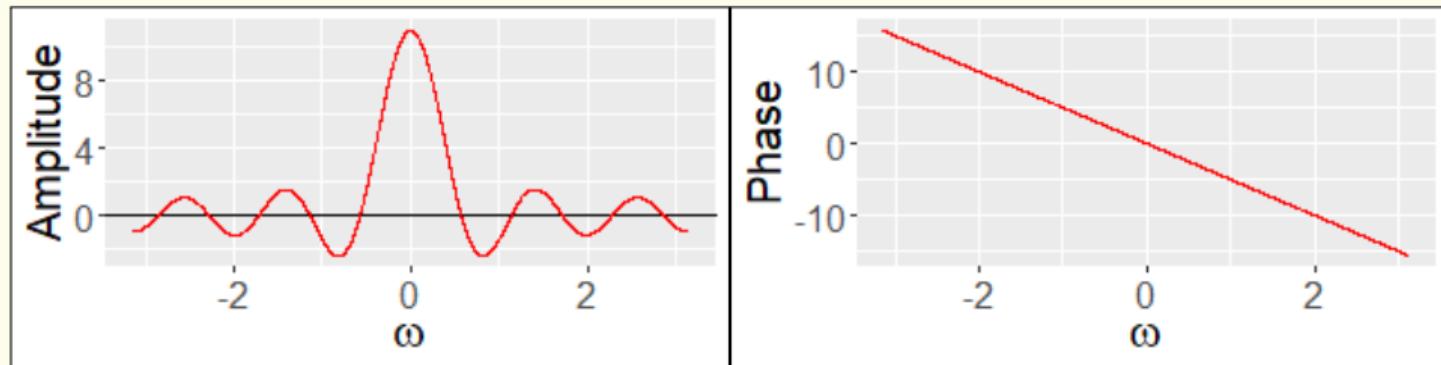


Figure: Amplitude and phase for a running sum filter with $L = 11$.

- ▶ $D_{11}(\hat{\omega})$ is symmetric around the vertical axis
- ▶ $D_{11}(\hat{\omega})$ has period 2π
- ▶ $D_{11}(\hat{\omega})$ has maximum 11 for $\hat{\omega} = 0$
- ▶ $D_{11}(\hat{\omega})$ is zero for $\hat{\omega} = \frac{2\pi k}{11}$ for integers $k \neq 0$

Dirichlet form

Quick question!

Do you notice something strange in the amplitude and phase?

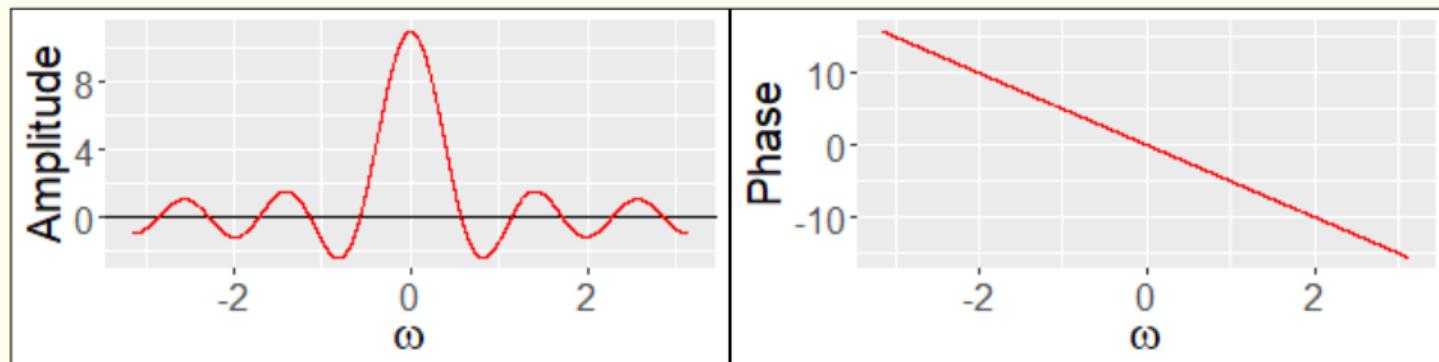


Figure: Amplitude and phase for a running sum filter with $L = 11$.

Running sum

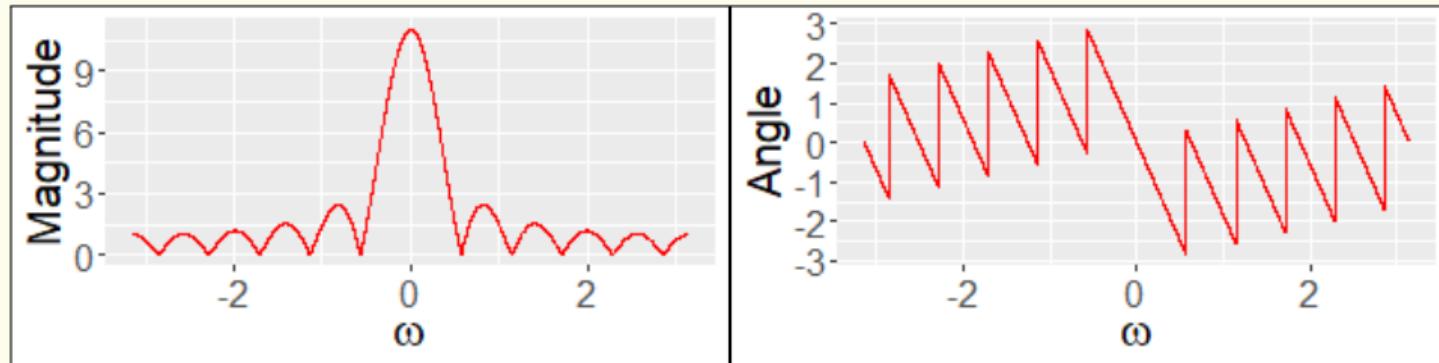


Figure: Magnitude and angle for a running sum filter with $L = 11$.

- ▶ The magnitude $|H(e^{j\hat{\omega}})| = |D_{11}(\hat{\omega})|$ cannot be negative
- ▶ Instead, $-1 = e^{j\pi}$ is a phase rotation of π radians
 - ▶ The angle is depicted in principal form $-\pi < \angle H(e^{j\hat{\omega}}) \leq \pi$

Running sum

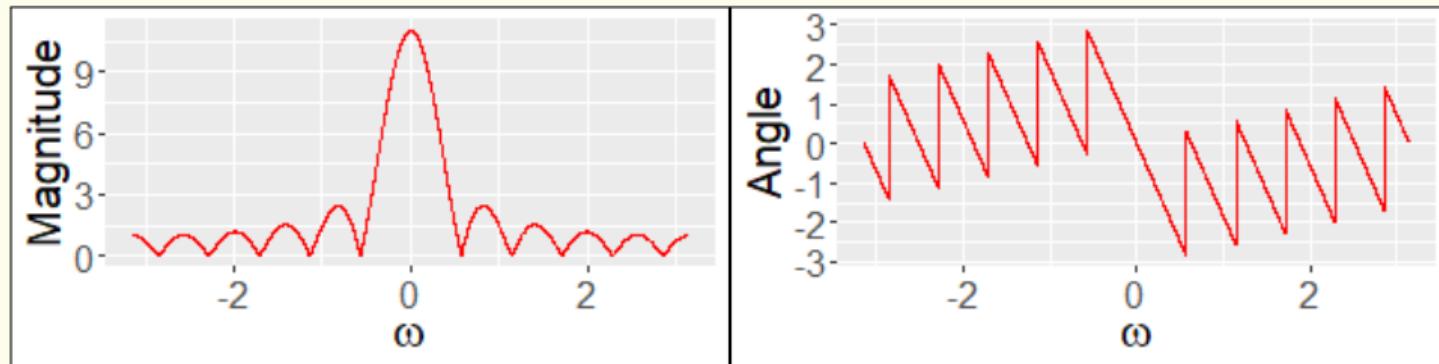
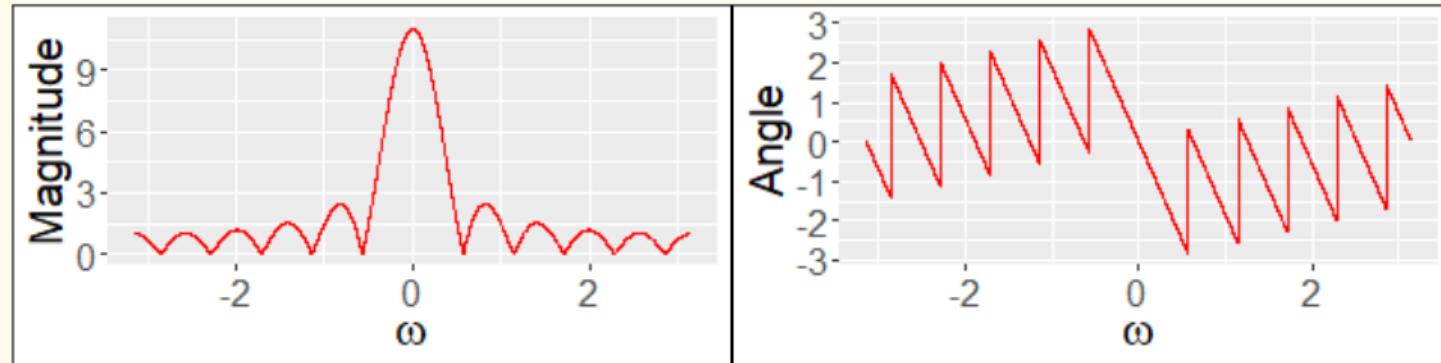


Figure: Magnitude and angle for a running sum filter with $L = 11$.

- ▶ The magnitude $|H(e^{j\hat{\omega}})| = |D_{11}(\hat{\omega})|$ cannot be negative
- ▶ Instead, $-1 = e^{j\pi}$ is a phase rotation of π radians
 - ▶ The angle is depicted in principal form $-\pi < \angle H(e^{j\hat{\omega}}) \leq \pi$
- ▶ The running sum behaves as a **low-pass filter**
 - ▶ Low frequencies are multiplied with a high magnitude
 - ▶ High frequencies are multiplied with a low magnitude

Continuous and discrete signals



- ▶ The frequency response $H(e^{j\hat{\omega}})$ is defined for any frequency $\hat{\omega}$
- ▶ In practice, frequencies outside the range $-\pi < \hat{\omega} \leq \pi$ cannot be reconstructed by an ideal D-to-C converter

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$.

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) =$$

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(25)}{1000}} \left(\frac{\sin(11\pi(25)/1000)}{\sin(\pi(25)/1000)} \right)$$

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(25)}{1000}} \left(\frac{\sin(11\pi(25)/1000)}{\sin(\pi(25)/1000)} \right) \approx 0.8811 e^{-j\pi/4}$$

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(25)}{1000}} \left(\frac{\sin(11\pi(25)/1000)}{\sin(\pi(25)/1000)} \right) \approx 0.8811 e^{-j\pi/4}$$

$$H(e^{j\frac{2\pi(250)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(250)}{1000}} \left(\frac{\sin(11\pi(250)/1000)}{\sin(\pi(250)/1000)} \right)$$

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(25)}{1000}} \left(\frac{\sin(11\pi(25)/1000)}{\sin(\pi(25)/1000)} \right) \approx 0.8811e^{-j\pi/4}$$

$$H(e^{j\frac{2\pi(250)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(250)}{1000}} \left(\frac{\sin(11\pi(250)/1000)}{\sin(\pi(250)/1000)} \right) \approx 0.0909e^{-j\pi/2}$$

Continuous and discrete signals



Consider the signal $x(t) = \cos(2\pi(25)t) + \sin(2\pi(250)t)$ passing through an **11-point averager** (not sum!) with sampling frequency is $f_s = 1000$.

There are two frequency components $\hat{\omega}_1 = \frac{2\pi(25)}{1000}$ and $\hat{\omega}_2 = \frac{2\pi(250)}{1000}$. The 11-point average filter response is

$$H(e^{j\frac{2\pi(25)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(25)}{1000}} \left(\frac{\sin(11\pi(25)/1000)}{\sin(\pi(25)/1000)} \right) \approx \mathbf{0.8811} e^{-j\pi/4}$$

$$H(e^{j\frac{2\pi(250)}{1000}}) = \frac{1}{11} e^{-j5\frac{2\pi(250)}{1000}} \left(\frac{\sin(11\pi(250)/1000)}{\sin(\pi(250)/1000)} \right) \approx \mathbf{0.0909} e^{-j\pi/2}$$

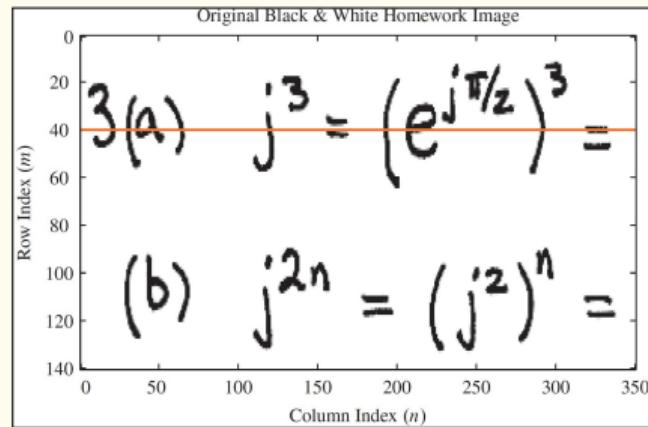
And the output signal is

$$y(t) = \mathbf{0.8811} \cos(2\pi(25)t - \pi/4) + \mathbf{0.0909} \sin(2\pi(250)t - \pi/2) \quad (10)$$

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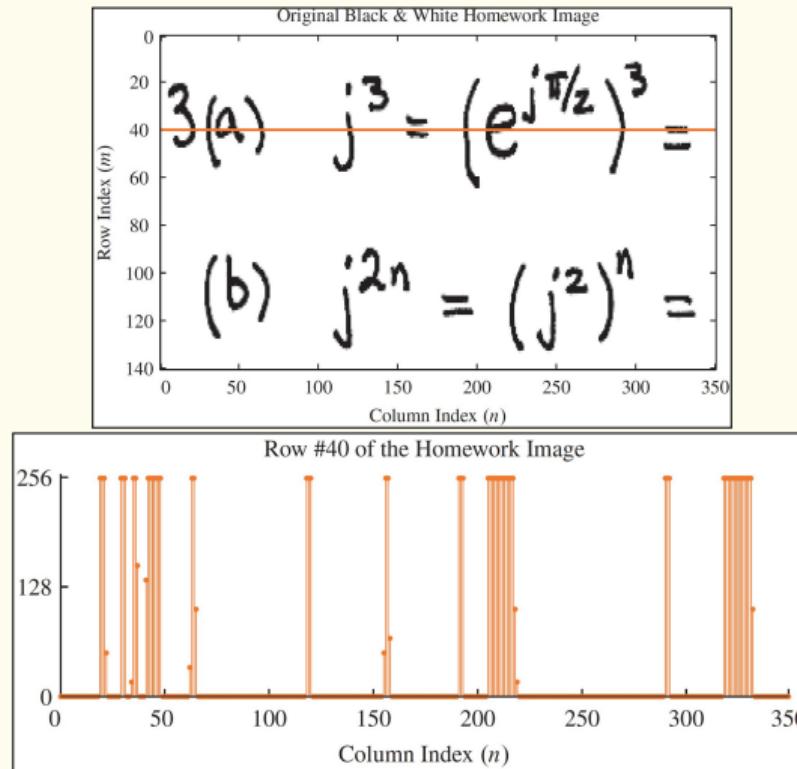
1. Recap
2. Frequency Response
3. Properties of Frequency Response
4. Running Sum Filter
5. Image Smoothing
6. Closing Remarks

Images as signals



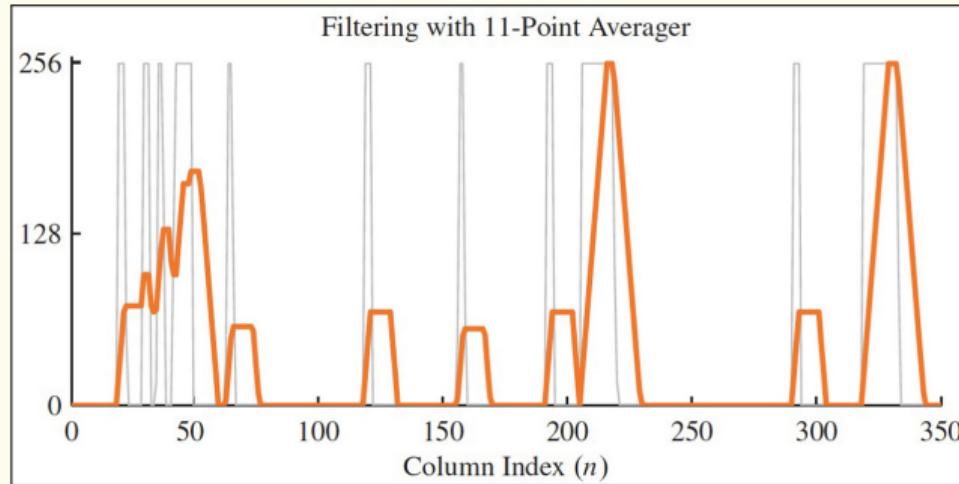
- ▶ Each row in the image can be interpreted as a signal

Images as signals



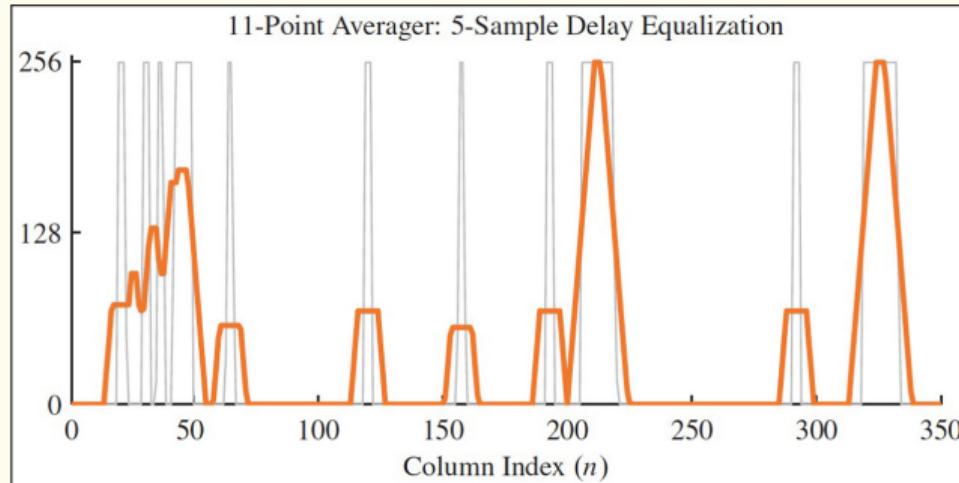
Images from McClellan, Shafer, & Yoder (2015) *DSP First*

Blurring



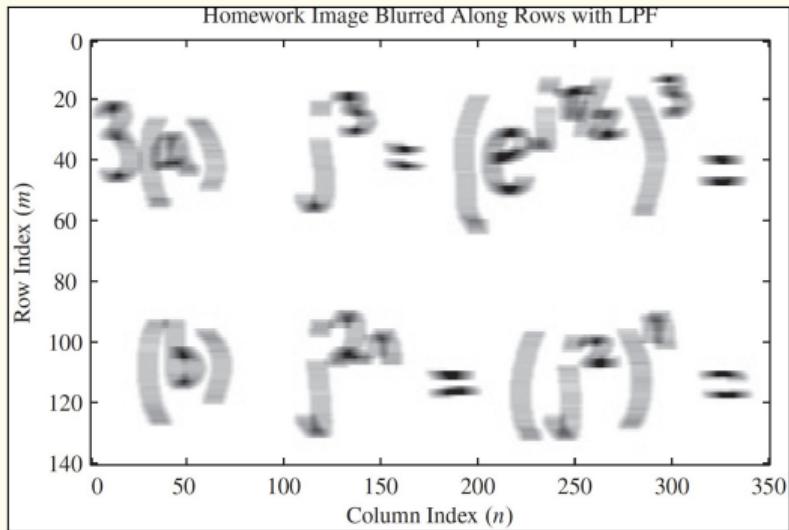
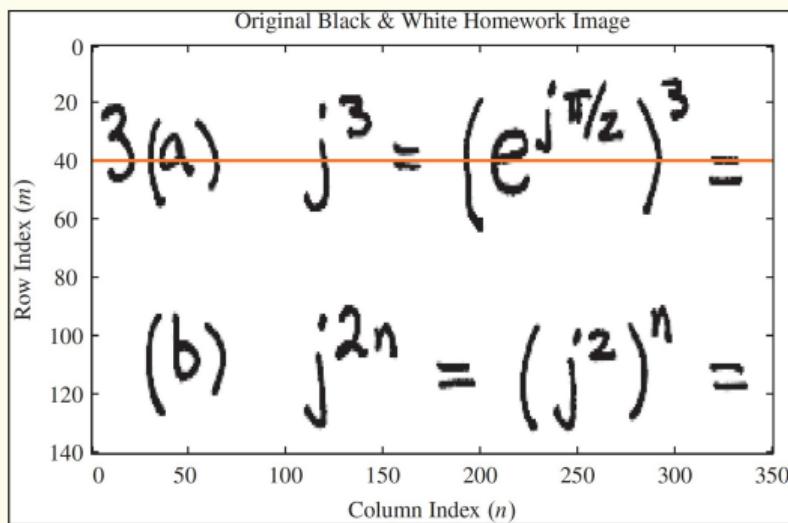
- ▶ The 11-point running average filter blurs the signal horizontally
 - ▶ Note that causal filters cause a shift to the right

Blurring

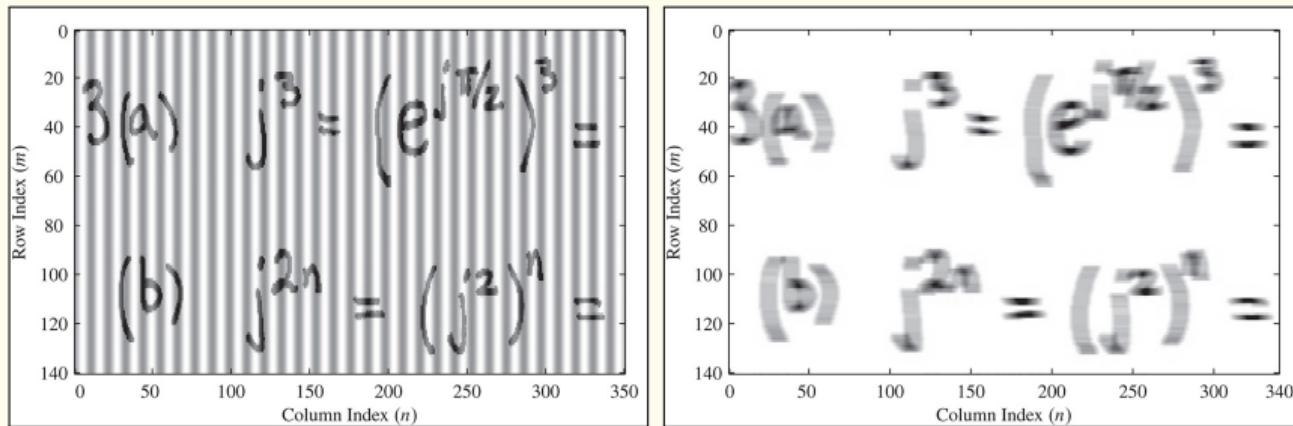


- ▶ The 11-point running average filter blurs the signal horizontally
 - ▶ Note that causal filters cause a shift to the right

Horizontal blurring on images

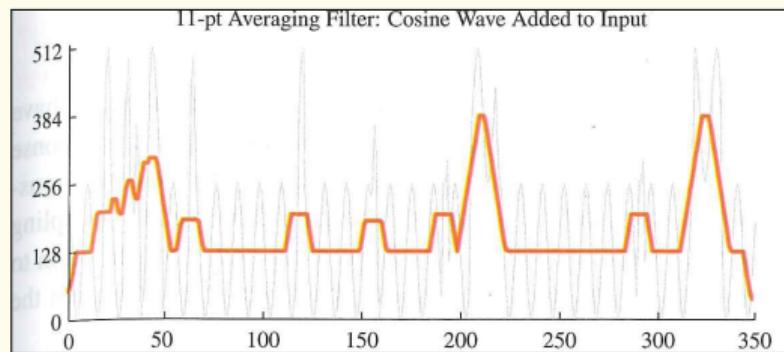
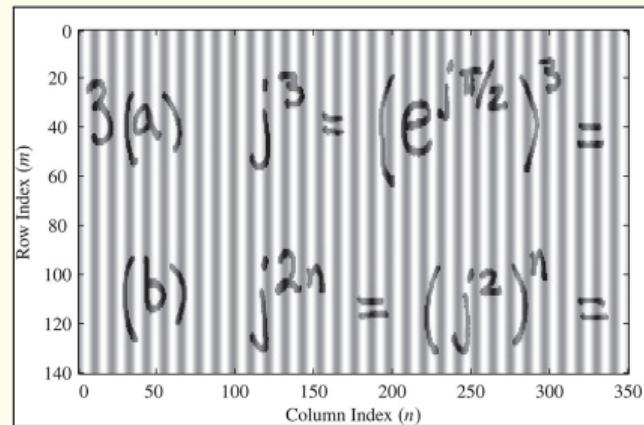


Eliminating frequencies



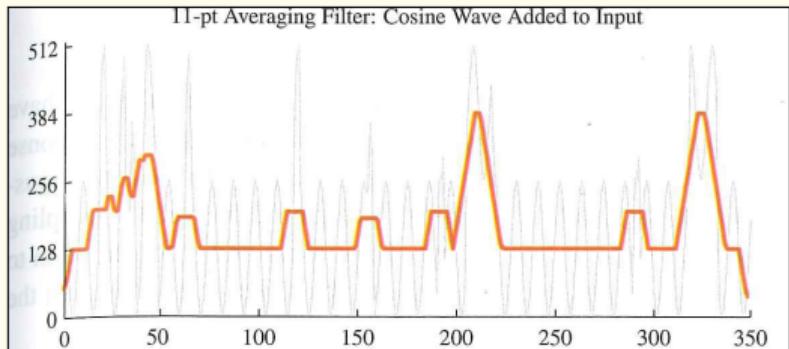
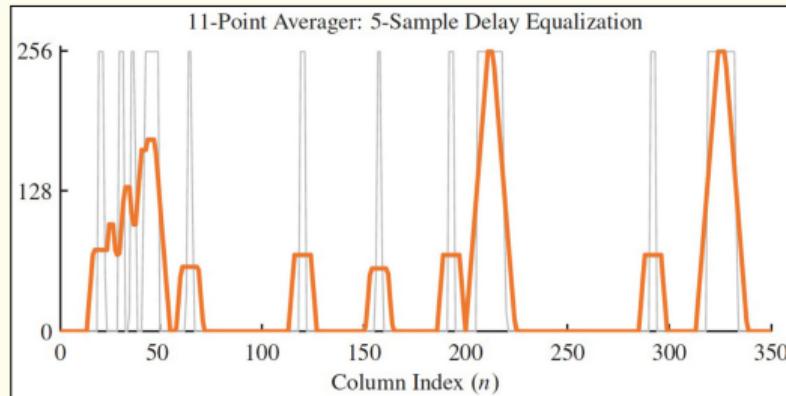
- ▶ The 11-point running average filter eliminates frequency multiples of 11
 - ▶ Cosine signal $128 \cos(2\pi n/11) + 128$ is added to the signal

Eliminating frequencies



- ▶ The 11-point running average filter eliminates frequency multiples of 11
 - ▶ Cosine signal $128 \cos(2\pi n/11) + 128$ is added to the signal
 - ▶ The running average divides the cosine across all samples

Eliminating frequencies

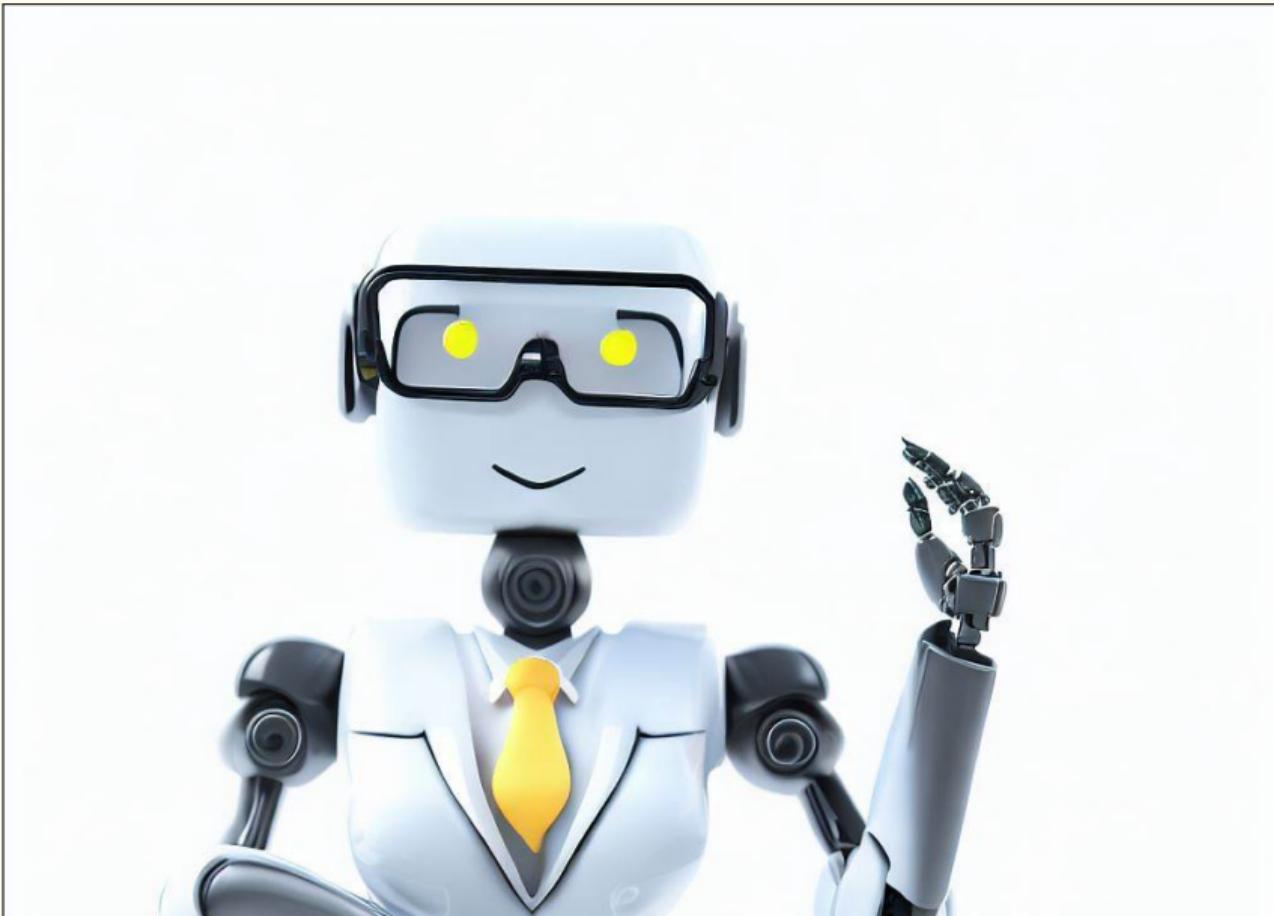


- ▶ The 11-point running average filter eliminates frequency multiples of 11
 - ▶ Cosine signal $128 \cos(2\pi n/11) + 128$ is added to the signal
 - ▶ The running average divides the cosine across all samples
 - ▶ Note the cosine causes transient effects at the start and end of the signal

Table of Contents

1. Recap
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Let us wrap up the lecture!



Take-home Messages

LTI filters respond to complex exponentials in a particular way

$$\begin{aligned}x[n] &= Ae^{j\varphi} e^{j\hat{\omega}n} \\y[n] &= \left(\sum_{k=0}^M b_k e^{-j\hat{\omega}k} \right) Ae^{j\varphi} e^{j\hat{\omega}n} \\&= H(e^{j\hat{\omega}})Ae^{j\varphi} e^{j\hat{\omega}n}\end{aligned}$$

- ▶ An LTI system is characterized by impulse response $h[n]$
- ▶ An LTI system is characterized by frequency response $H(e^{j\hat{\omega}})$

Take-home Messages

$$H(e^{j\hat{\omega}}) = \sum_{k=0}^M b_k e^{-j\hat{\omega}k} = |H(e^{j\hat{\omega}})|e^{j\angle H(e^{j\hat{\omega}})}$$

- ▶ The frequency response consists of
 - ▶ Magnitude $|H(e^{j\hat{\omega}})| \geq 0$
 - ▶ Phase $-\pi < \angle H(e^{j\hat{\omega}}) \leq \pi$
- ▶ The input frequency is not changed by an LTI system
- ▶ Each frequency component in the input may have a different frequency response

And Finally.

A convolution in the time domain is a multiplication in the frequency domain.

Practice Questions

The following questions might appear in the final exam:

- ▶ Let us suppose that $y[n]$ is the response of a FIR filter $h[n]$ and the signal $x[n]$ whose (radian) frequency is $\hat{\omega}$. What is the (radian) frequency of $y[n]$?
- ▶ What is the magnitude and angle in a sinusoidal response of an FIR system?
- ▶ Why would you eliminate a single frequency in a signal?
- ▶ What is advantage of computing $h[n] \star x[n]$ in the frequency domain?

Tutorial exercises

During the tutorial, the exercises below will be discussed in class

- ▶ Attempt to complete the exercises **before** class starts
- ▶ As the weeks progress, more time is needed for an explanation

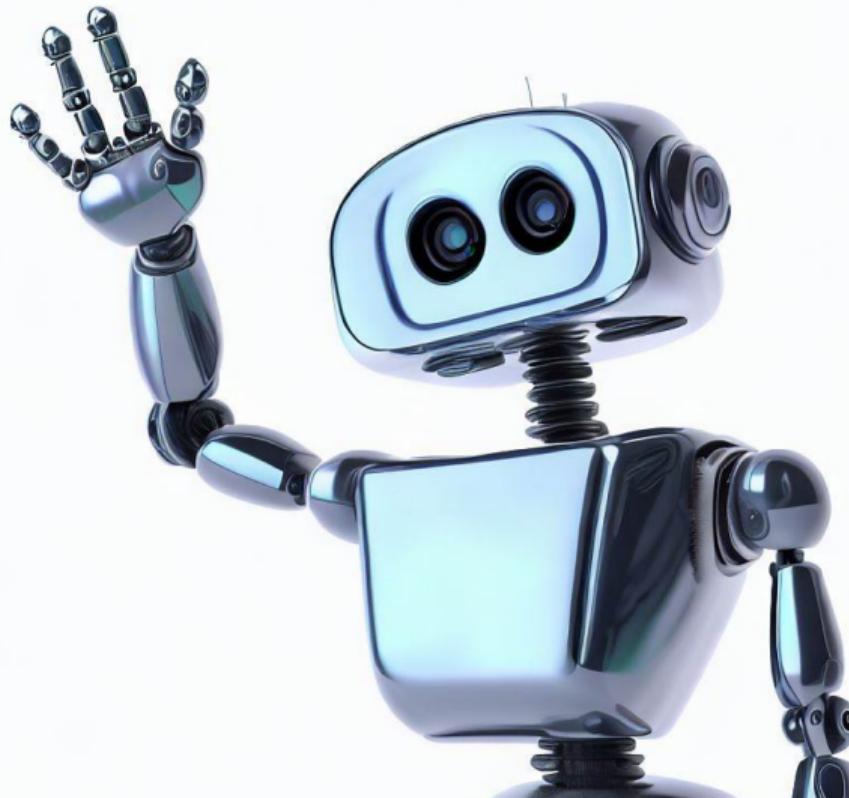
SPF	DSPF
P 6.1 (p. 157)	P 6.1 (p. 250)
P 6.4 (p. 157)	P 6.4 (p. 251)
P 6.5 (p. 158)	P 6.5 (p. 252)
P 6.7 (p. 158)	P 6.7 (p. 252)
P 6.16 (p. 160)	P 6.17 (p. 256)

Closing Remarks: Next Lecture

Let us discuss how we can transform a periodic signal from time to frequency domain.

Fourier transform

Have a nice day!



Acknowledgements

The material for this lecture series was developed by dr. Arnold Meijster and dr. Harmen de Weerd and modified by Juan Diego Cardenas-Cartagena.

Disclaimer

- ▶ Grammar was checked with Grammarly and Grammar checker GPT.
- ▶ Images without source were created with the assistance of DALL·E.