

Lecture 22: Design of FIR / IIR Filters

Foundations of Digital Signal Processing

Outline

- Designing FIR Filters with Windows
- Designing FIR Filters with Frequency Selection
- Designing FIR Filters with Equi-ripples
- Designing IIR Filters with Discrete Differentiation
- Designing IIR Filters with Impulse Invariance
- Designing IIR Filters with the Bilinear Transform
- Related Analog Filters

■ Homework #9

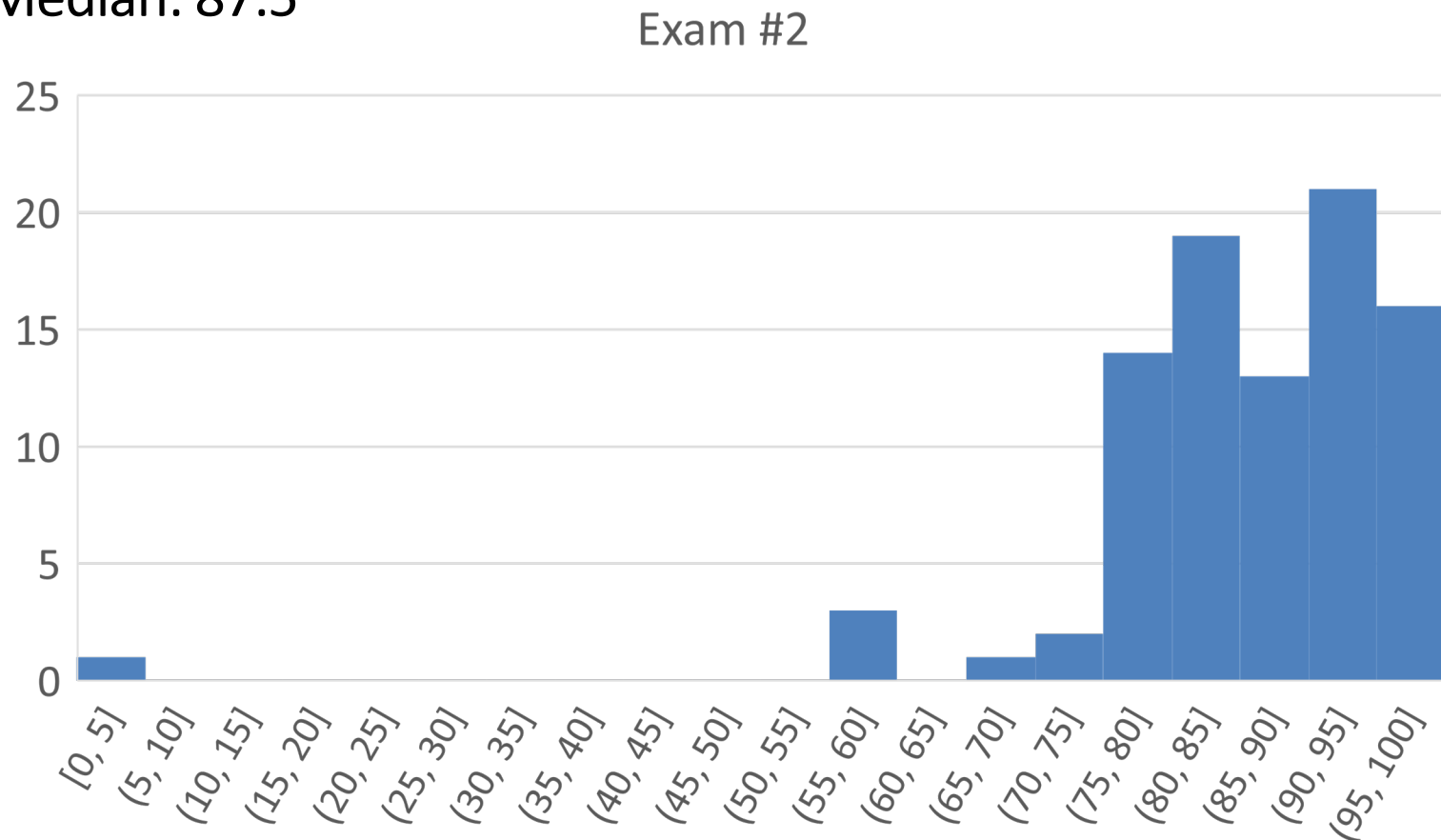
- Due on Thursday
- Submit via canvas

■ Coding Assignment #6

- Due on next Monday
- Submit via canvas

■ Exam #2 – Great Job!

- Mean: 86.3
- Median: 87.5



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Foundations of Digital Signal Processing

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- **Designing FIR Filters with Windows**
- Designing FIR Filters with Frequency Selection
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- Designing IIR Filters with Discrete Differentiation
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Causality and Linear Phase

■ **Question:** Consider a length- M symmetric, causal filter. What condition must be satisfied?

Causality and Linear Phase

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- $x[n] = \pm x[-n + (N - 1)] = \pm x[N - 1 - n]$
 - **Positive:** Even symmetry
 - **Negative:** Odd symmetry

Causality and Linear Phase

■ **Question:** Consider a length- M symmetric, causal filter. What is the phase response?

Causality and Linear Phase

■ **Question:** Consider a length-M symmetric, causal filter. What is the phase response? Assume M is even.

■ **Even Symmetry**

$$X(z) = a_0 + a_1z^{-1} + a_2z^{-2} + \cdots + a_1z^{-(M-2)} + a_0z^{-(M-1)}$$

■ **Odd Symmetry**

$$X(z) = a_0 + a_1z^{-1} + a_2z^{-2} + \cdots - a_1z^{-(M-2)} - a_0z^{-(M-1)}$$

Causality and Linear Phase

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$$\begin{aligned} X(z) &= a_0 + a_1 z^{-1} + a_2 z^{-2} + \cdots + a_1 z^{-(M-2)} + a_0 z^{-(M-1)} \\ &= z^{-\frac{(M-1)}{2}} \left[a_0 z^{\frac{(M-1)}{2}} + a_1 z^{\frac{(M-1)}{2}-1} + \cdots + a_1 z^{1-\frac{(M-1)}{2}} + a_0 z^{-\frac{(M-1)}{2}} \right] \end{aligned}$$

■ Odd Symmetry

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$$G(\omega) = |X(\omega)| e^{j\Theta(\omega)}$$

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$$G(\omega) = |X(\omega)| e^{j\Theta(\omega)} \quad \Theta(\omega) = \begin{cases} 0 & \text{for } G(\omega) > 0 \\ \pi & \text{for } G(\omega) < 0 \end{cases}$$

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Causality and Linear Phase

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■ Notice that

$$X(z) = z^{-(M-1)} X(z^{-1})$$

Causality and Linear Phase

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■ Pole-zero plot property?

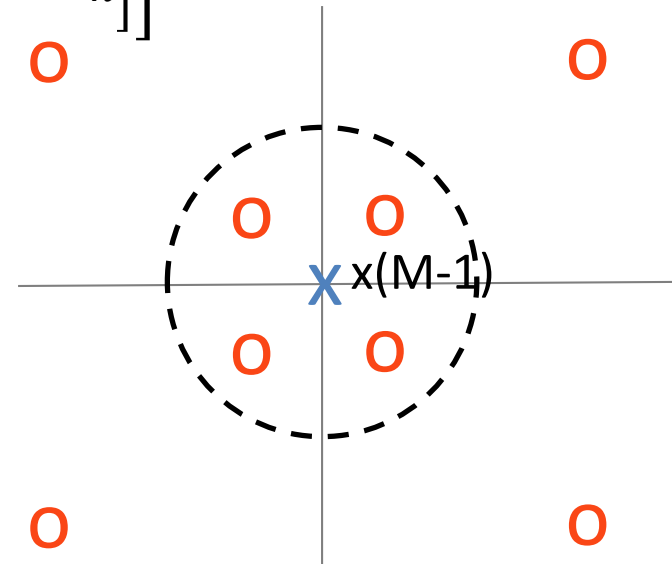
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Causality and Linear Phase

■ **Question:** Consider a length- M symmetric, causal filter. What is the phase response? Assume M is even.

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 \end{aligned}$$

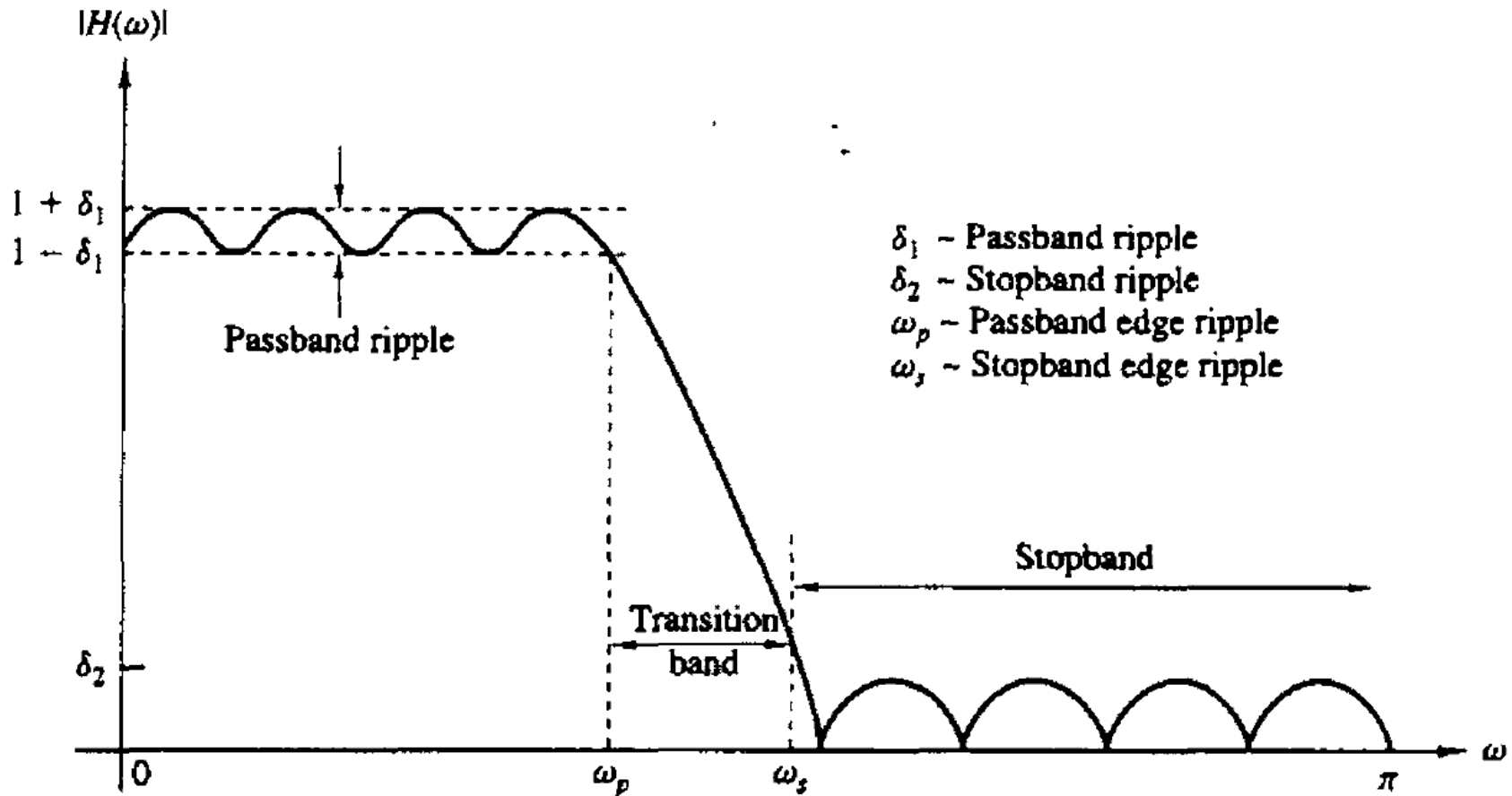


■ Pole-zero plot property?

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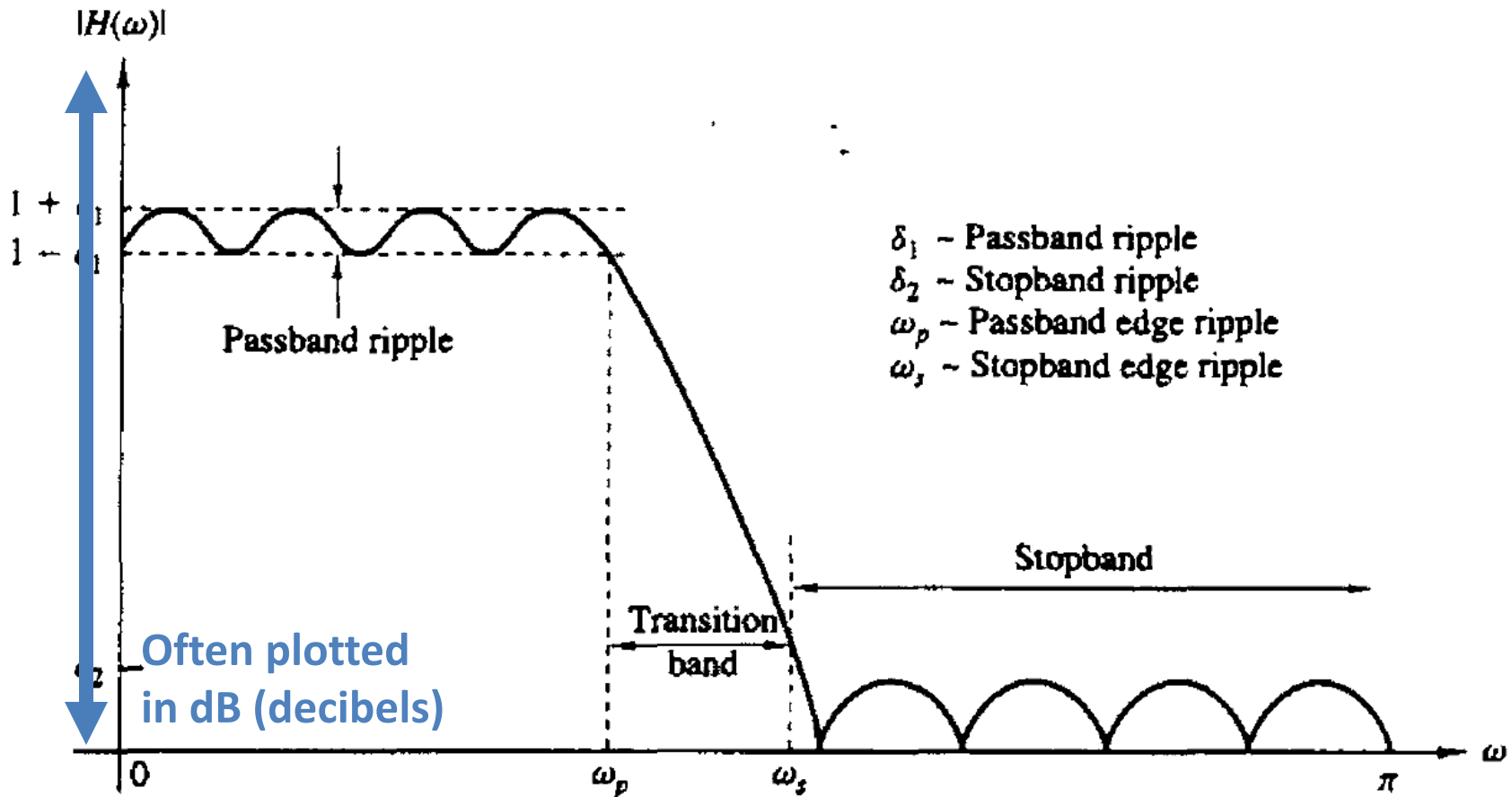
Causality

■ Question: How do we describe causal filter magnitude?



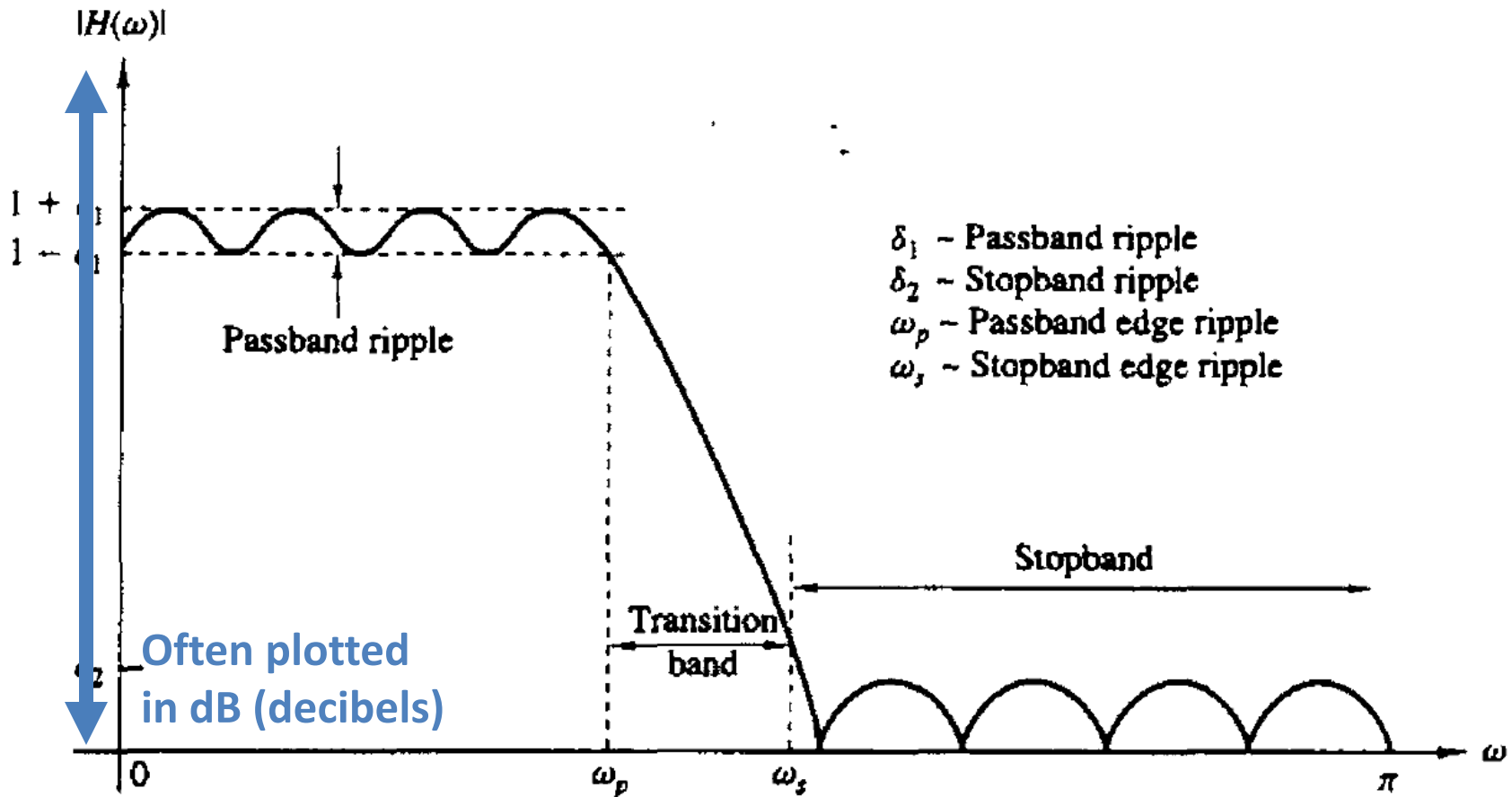
Causality

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Causality

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Lecture 21: Design of FIR Filters

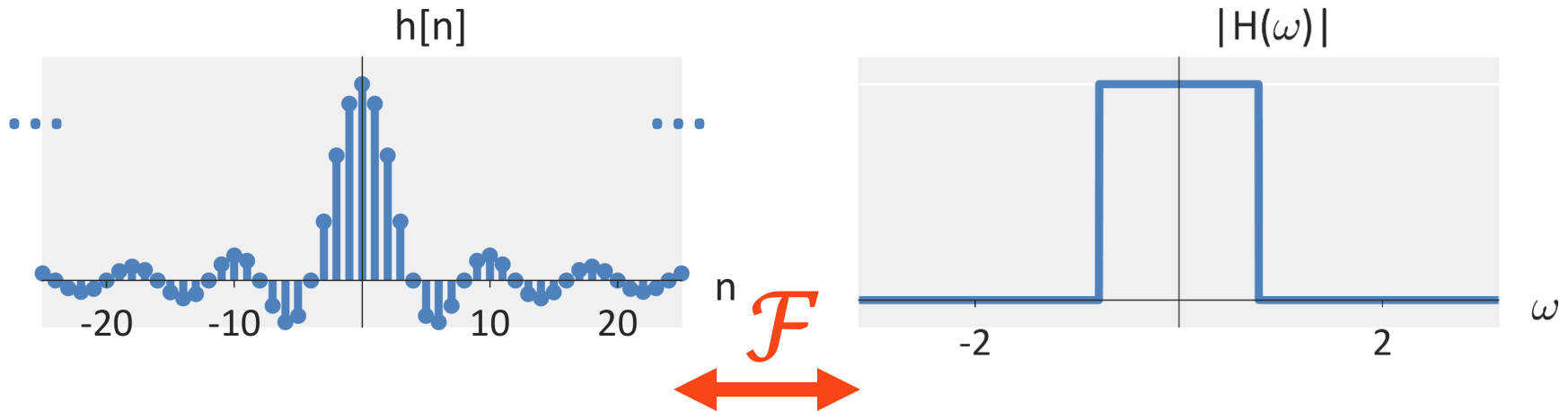
Foundations of Digital Signal Processing

Outline

- Review Downsampling & Upsampling
- Causality in Filters
- **Designing FIR Filters with Windows**
- Designing FIR Filters with Frequency Selection
- Designing FIR Filters with Equi-ripples

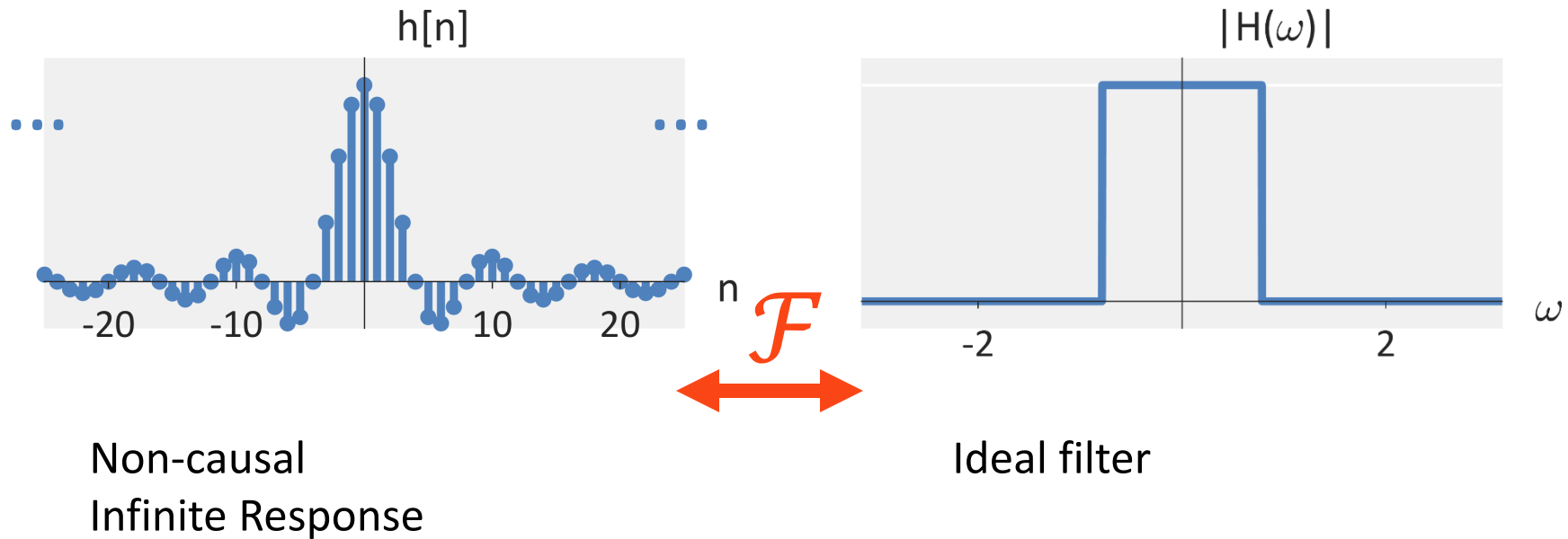
Designing with Windows

■ **Question:** How can I design an FIR filter from an ideal filter?



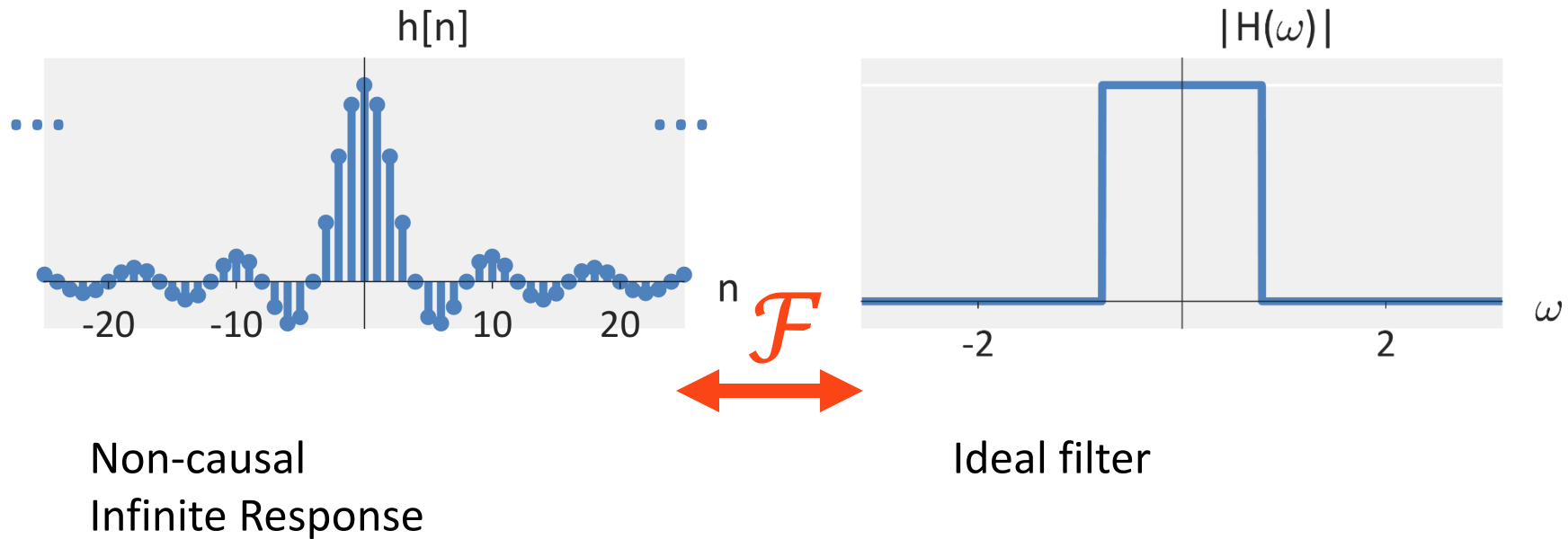
Designing with Windows

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Designing with Windows

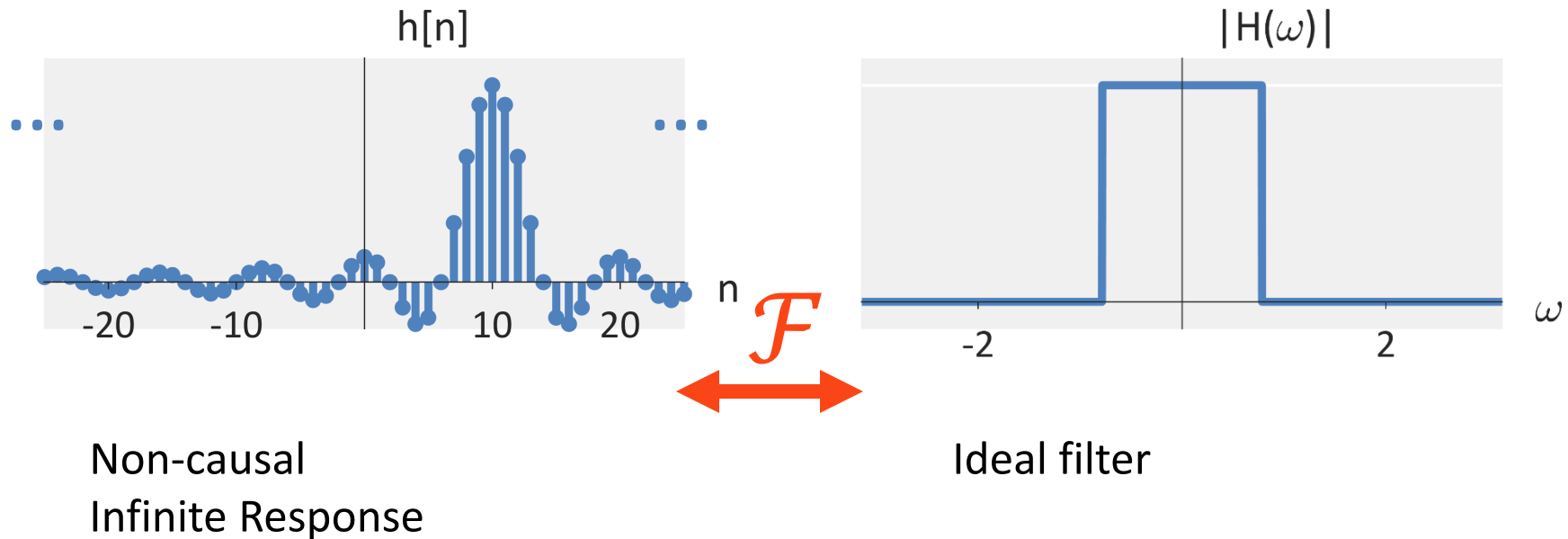
■ **Question:** How can I design an FIR filter from an ideal filter?



■ **Answer:** Window the response!

Designing with Windows

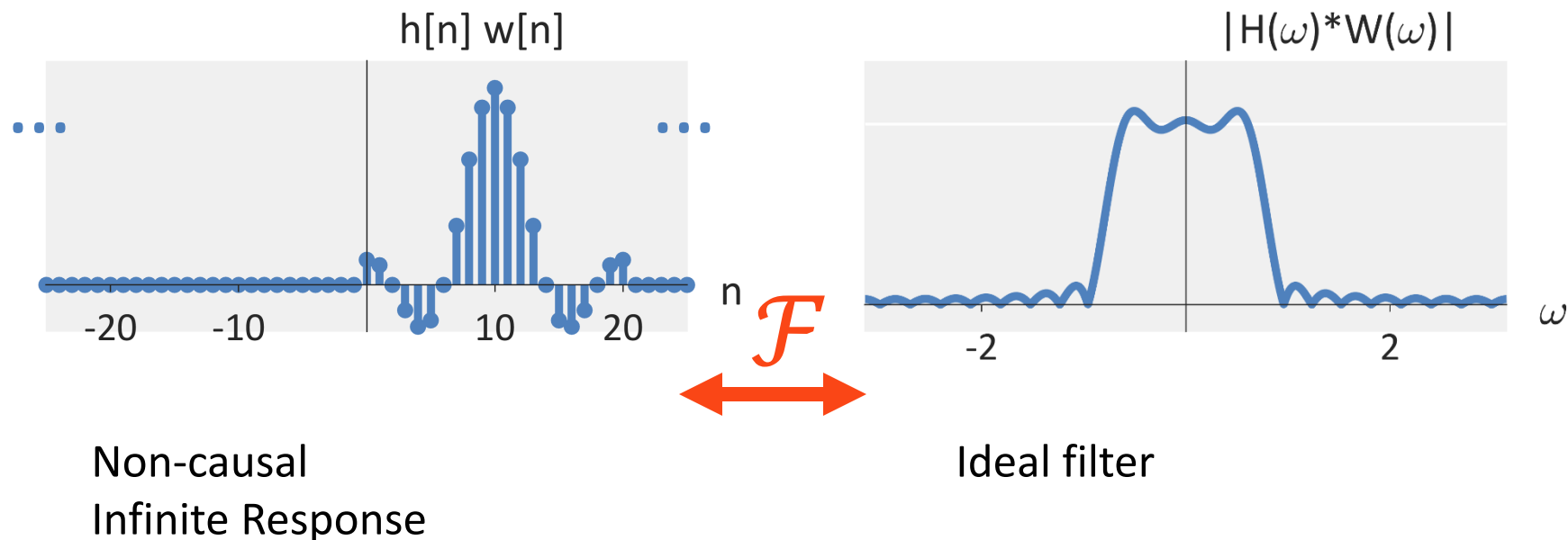
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Designing with Windows

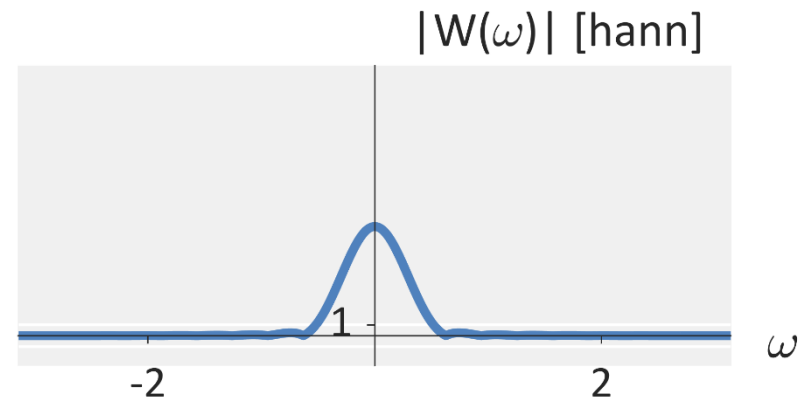
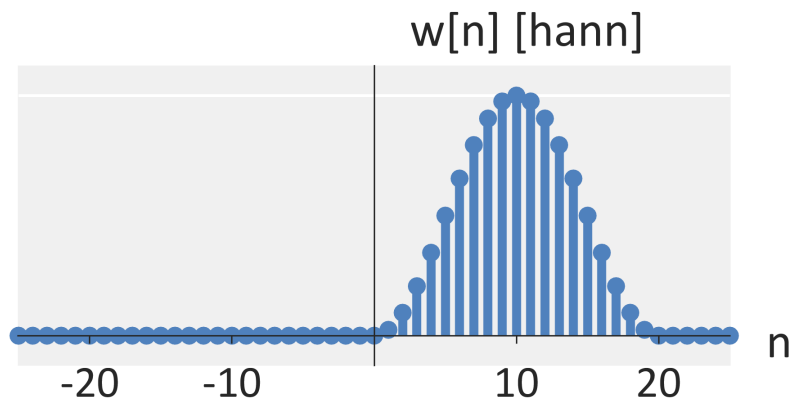
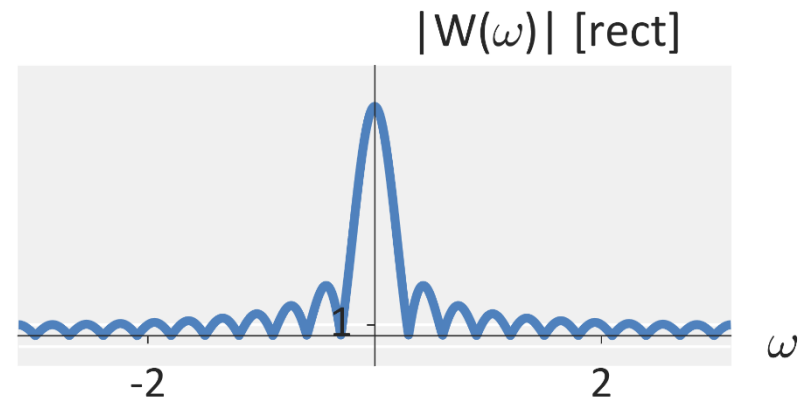
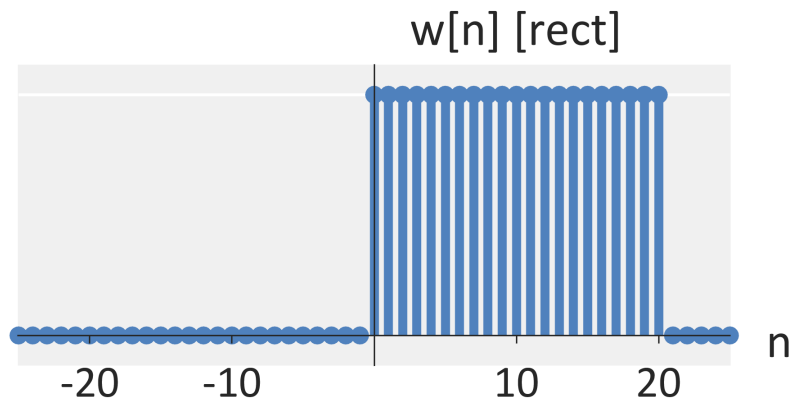
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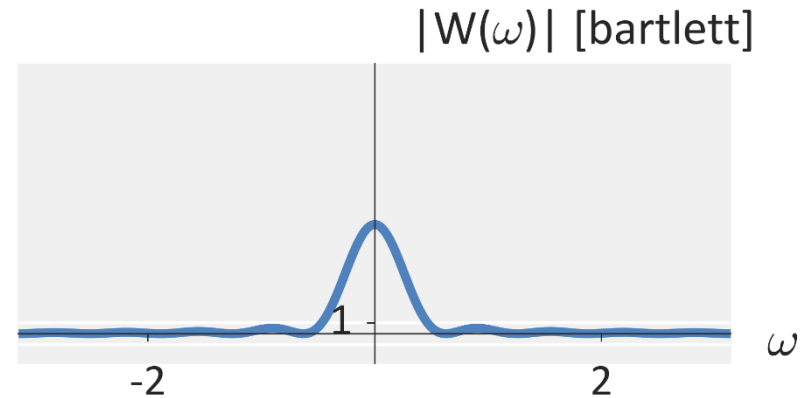
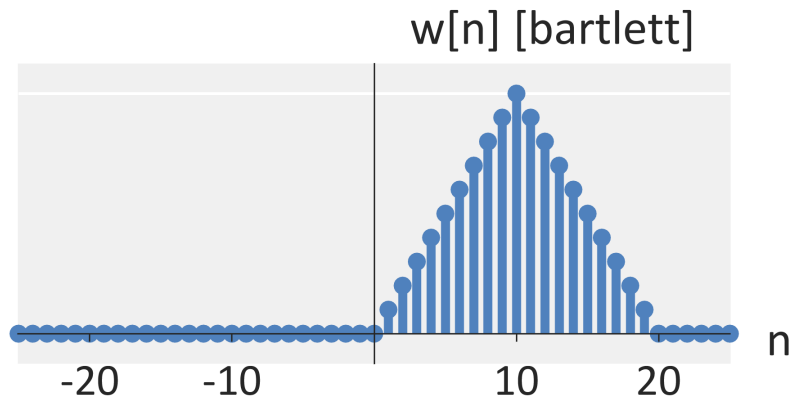
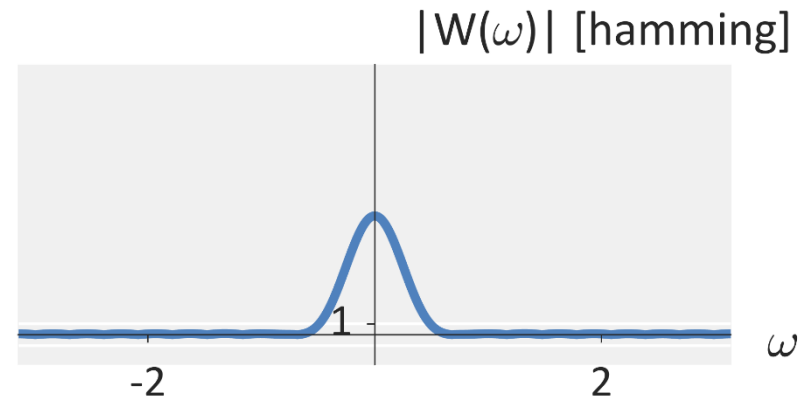
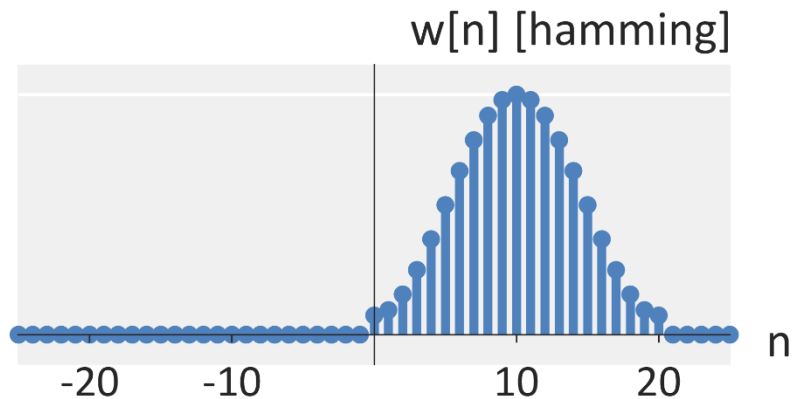
Designing with Windows

■ Different Filters



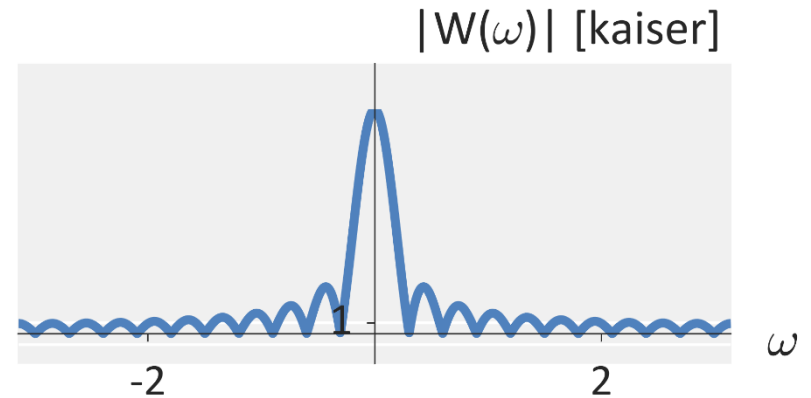
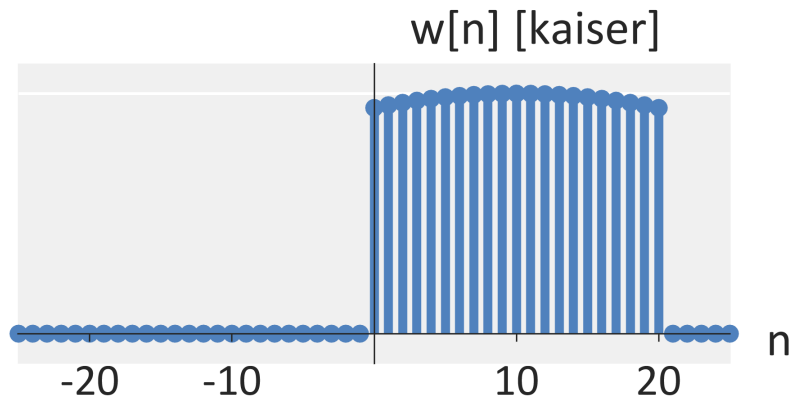
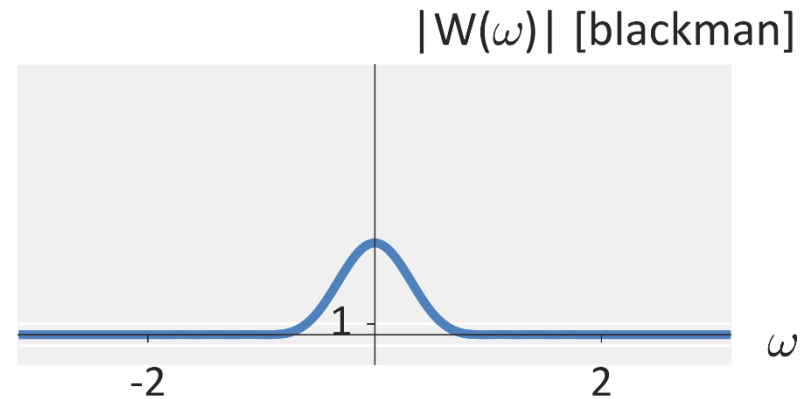
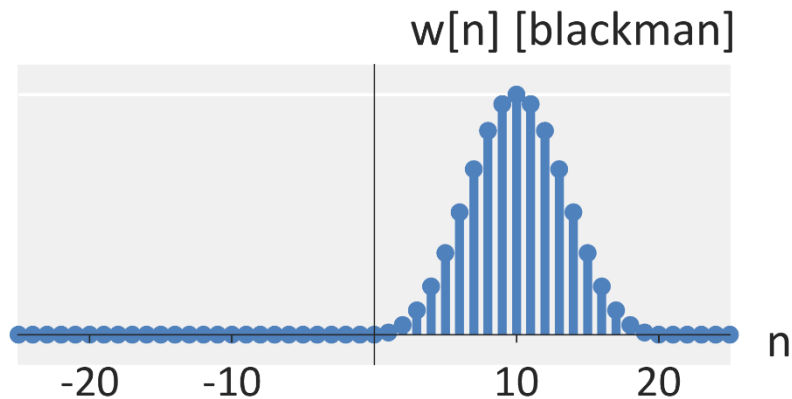
Designing with Windows

■ Different Filters



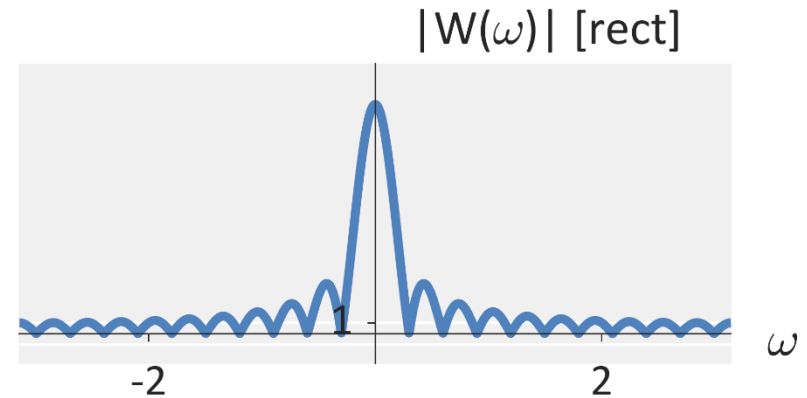
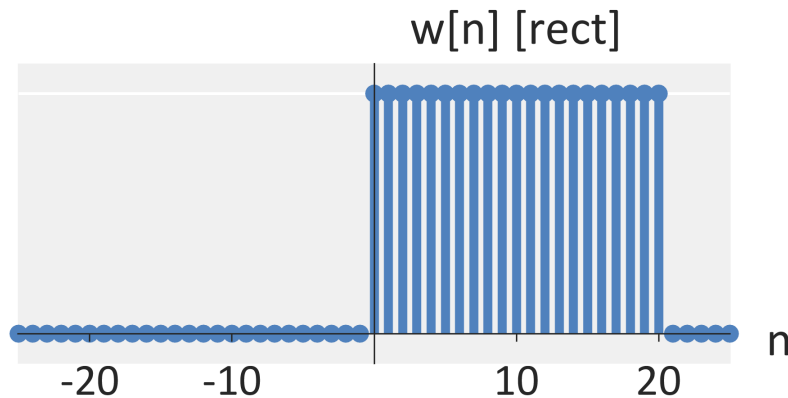
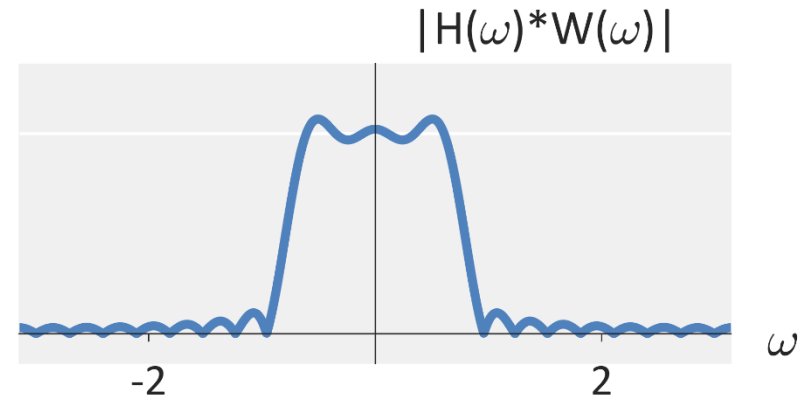
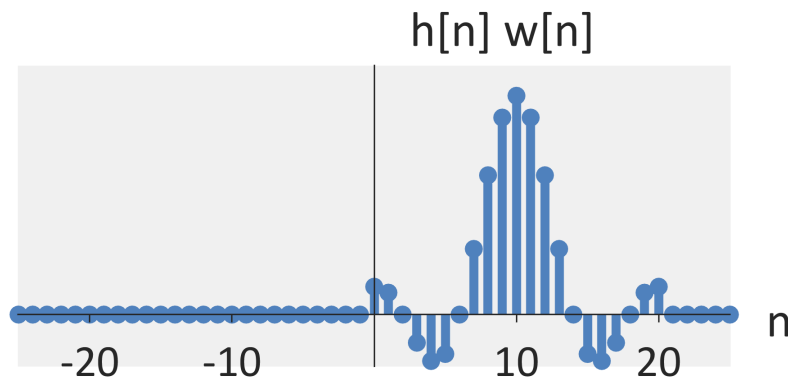
Designing with Windows

■ Different Filters



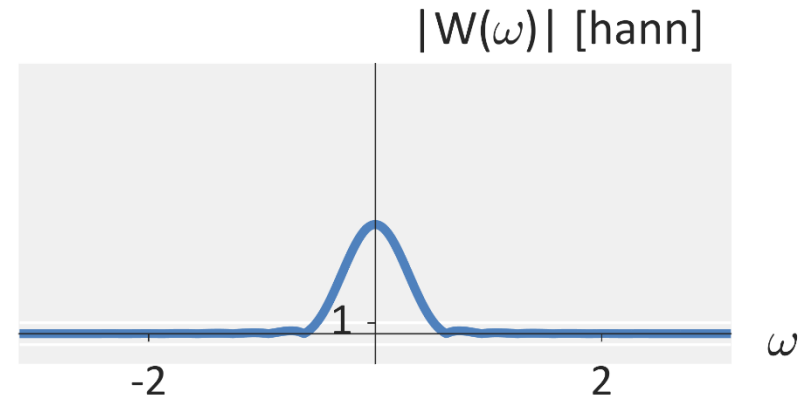
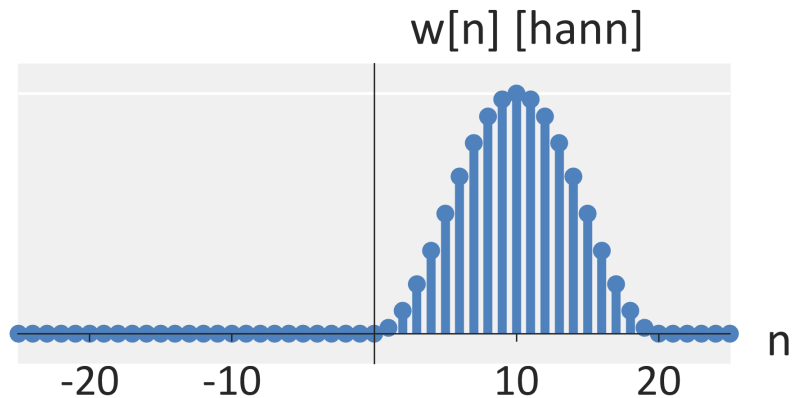
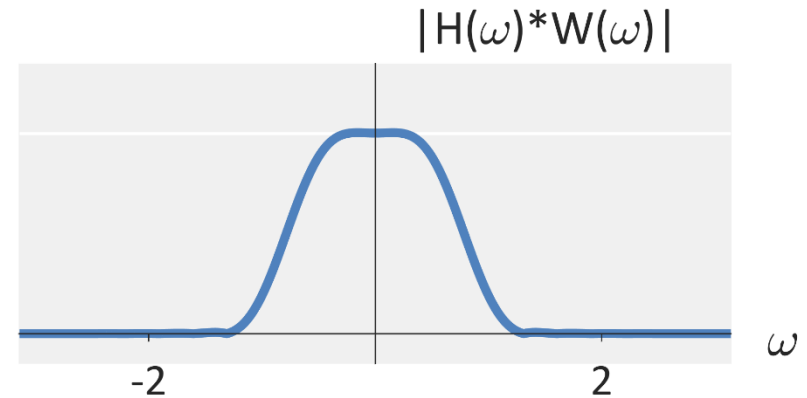
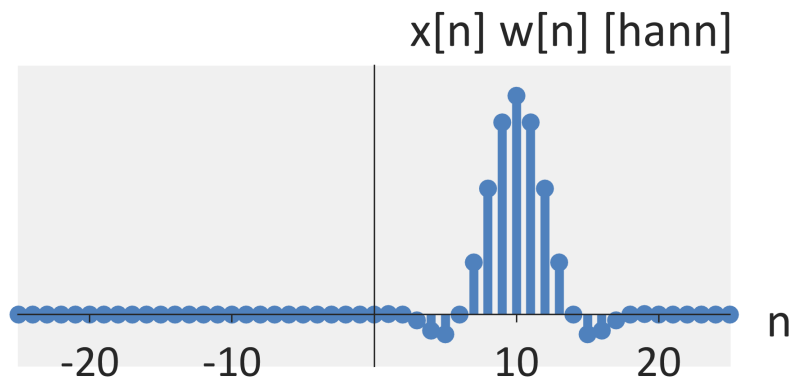
Designing with Windows

■ Windowing the sinc impulse response



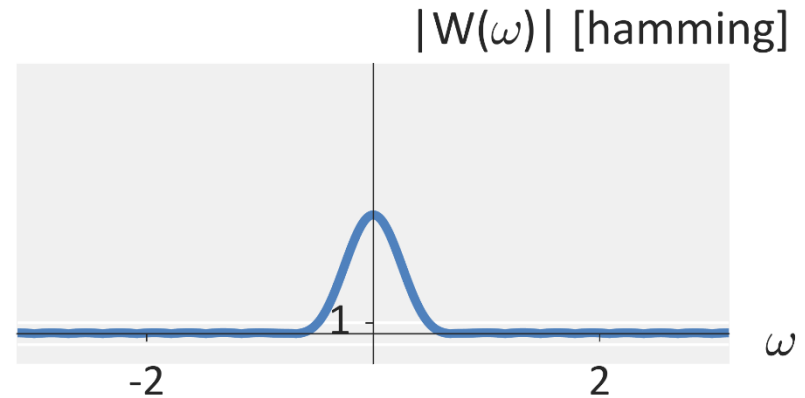
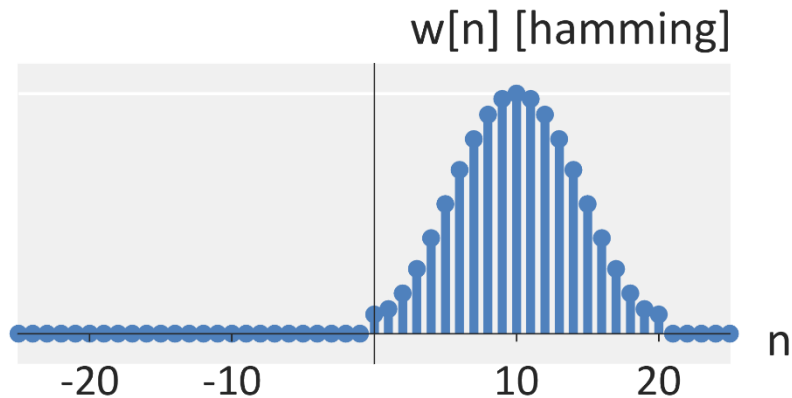
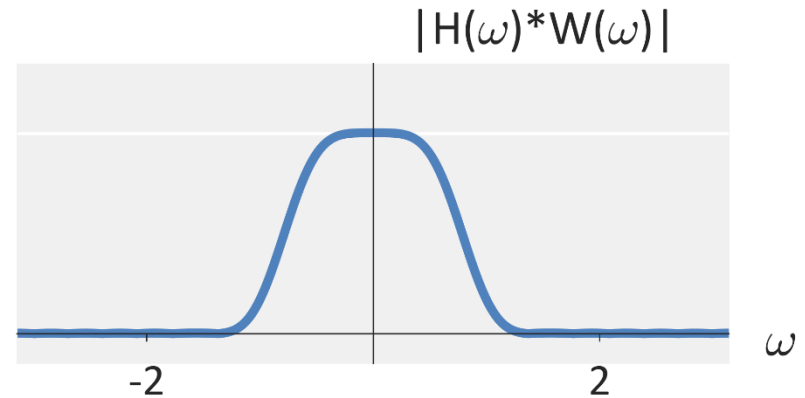
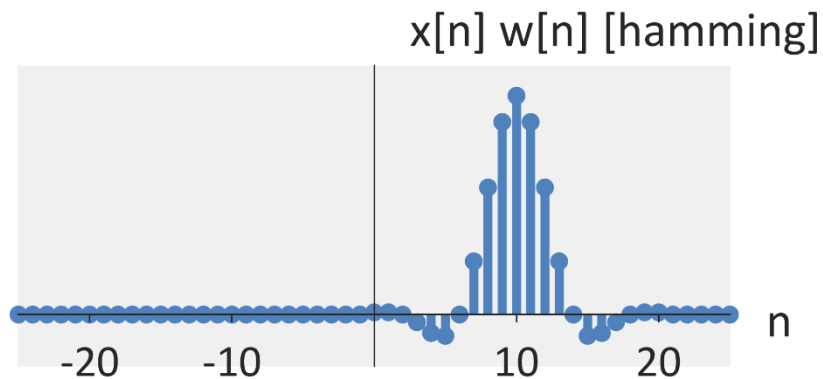
Designing with Windows

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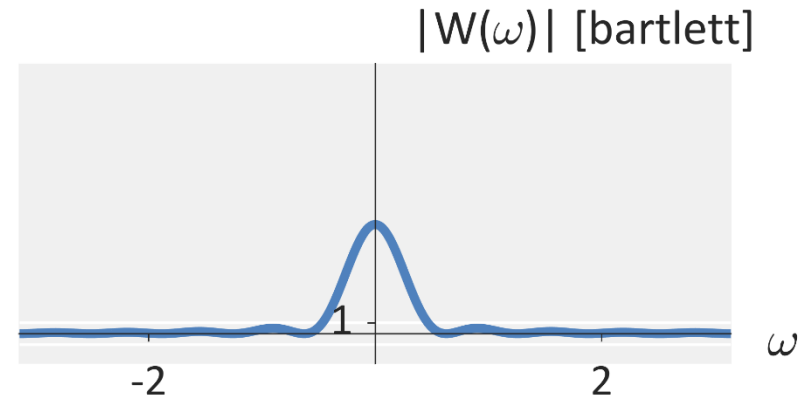
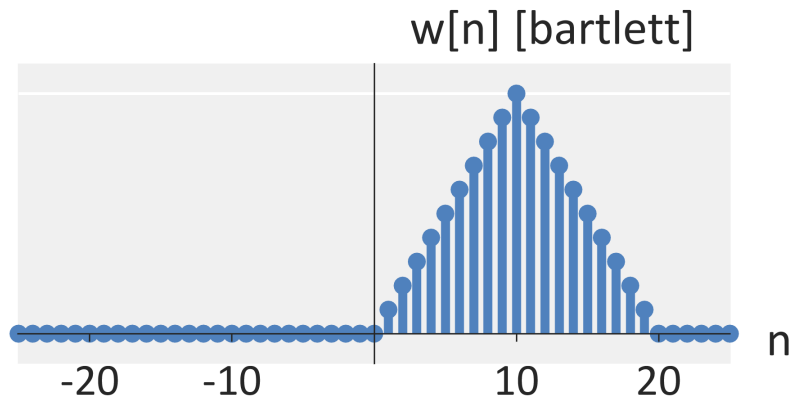
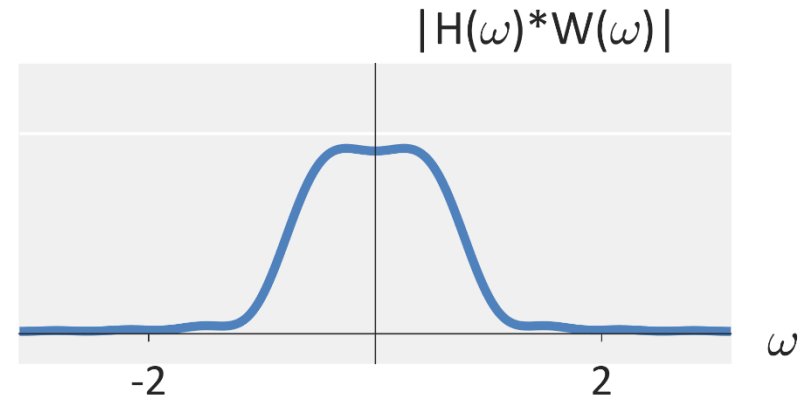
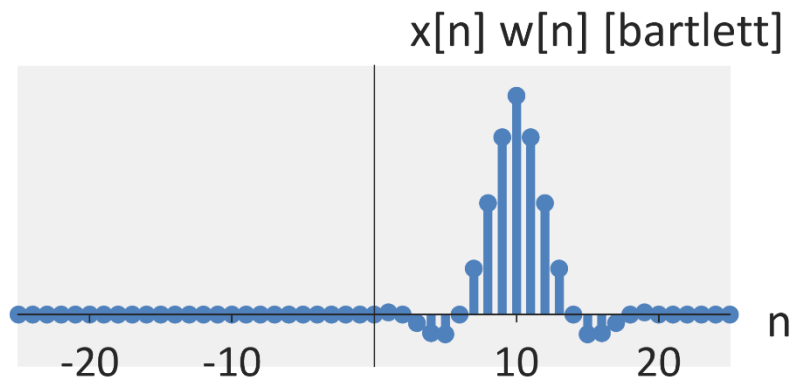
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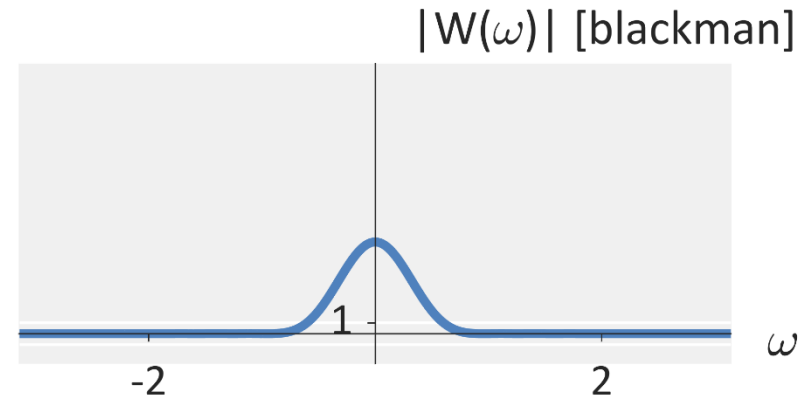
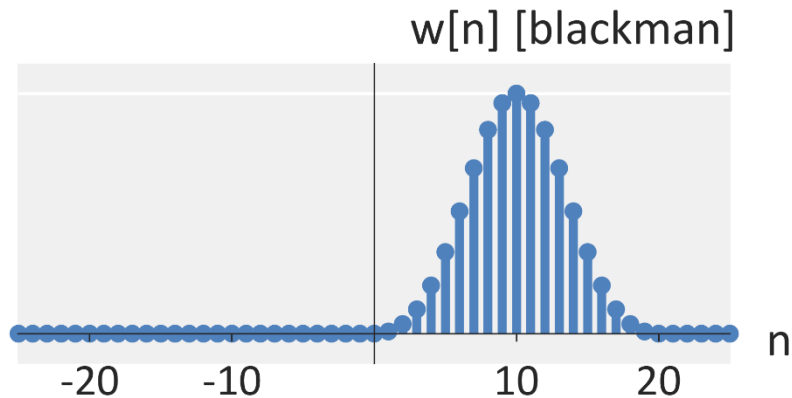
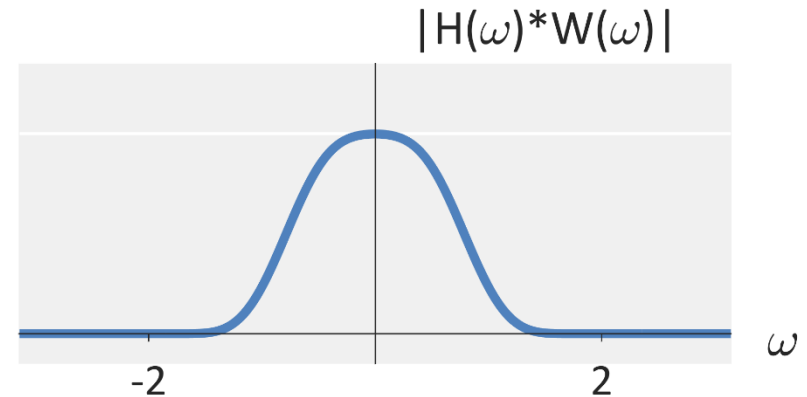
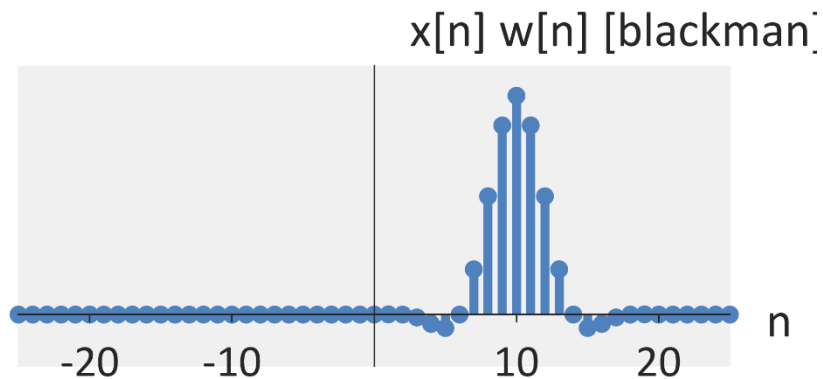
Designing with Windows

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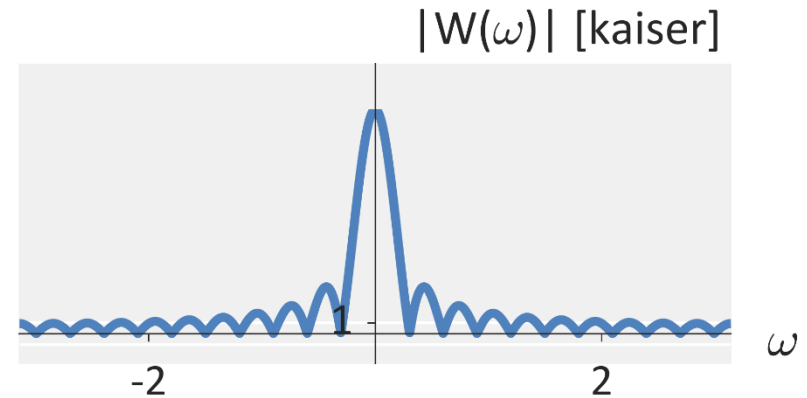
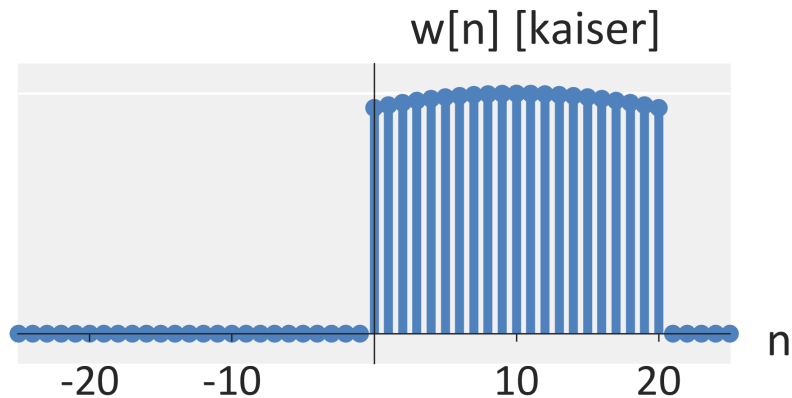
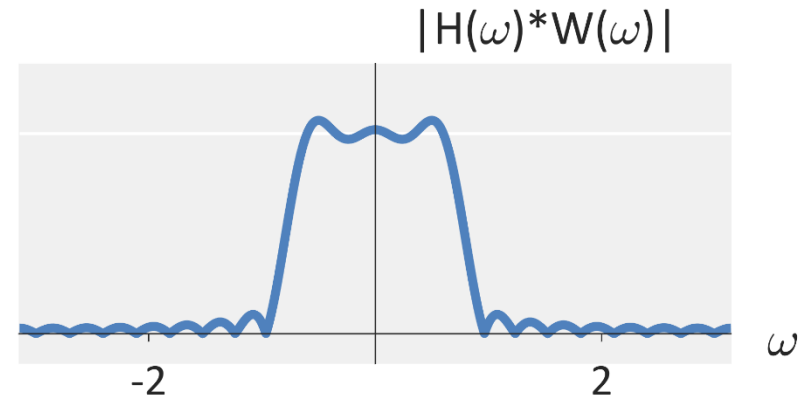
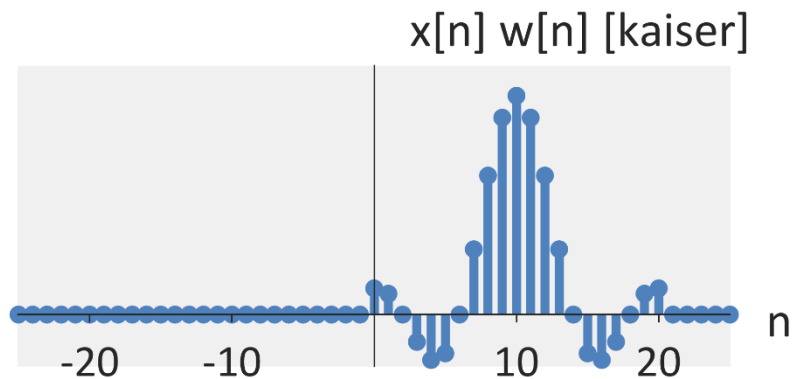
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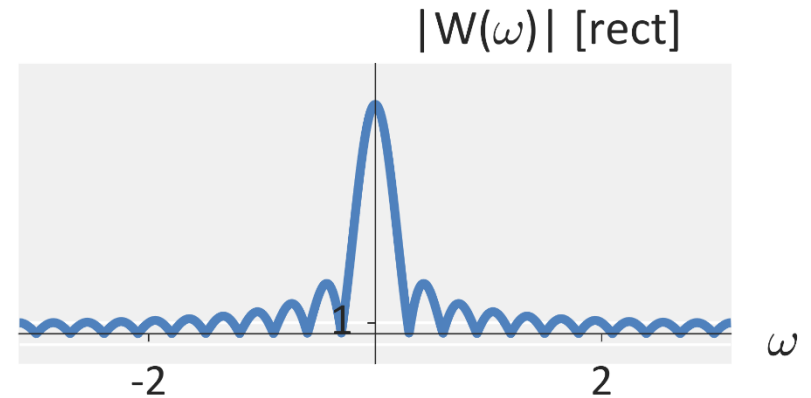
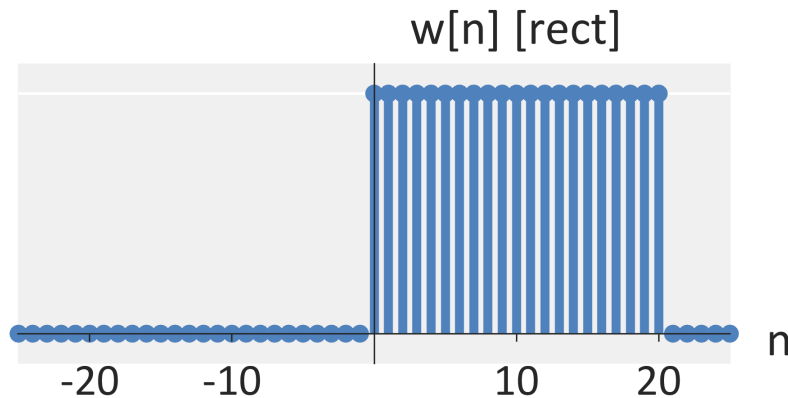
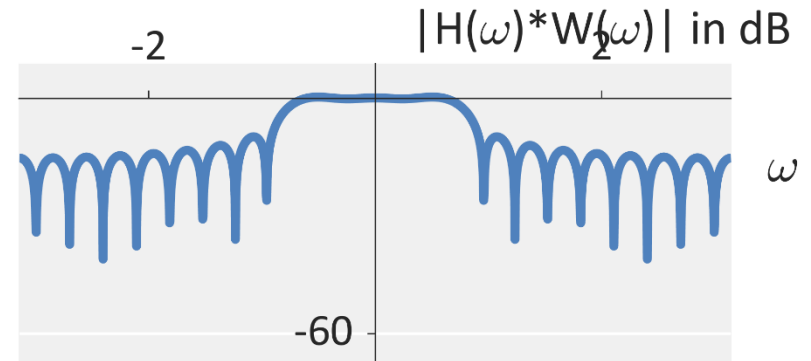
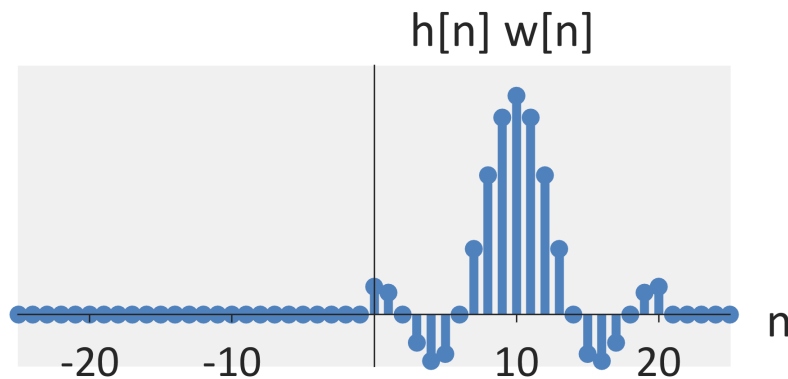
Designing with Windows

■ Windowing the sinc impulse response



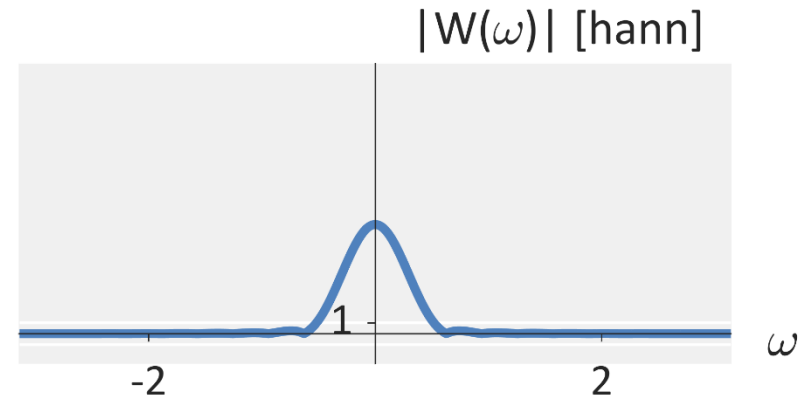
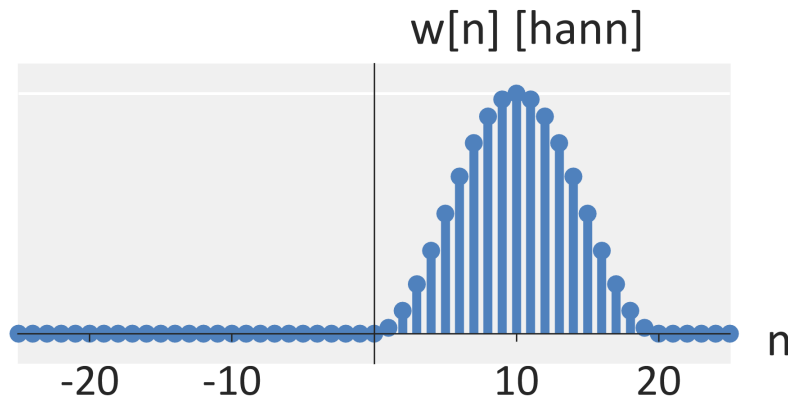
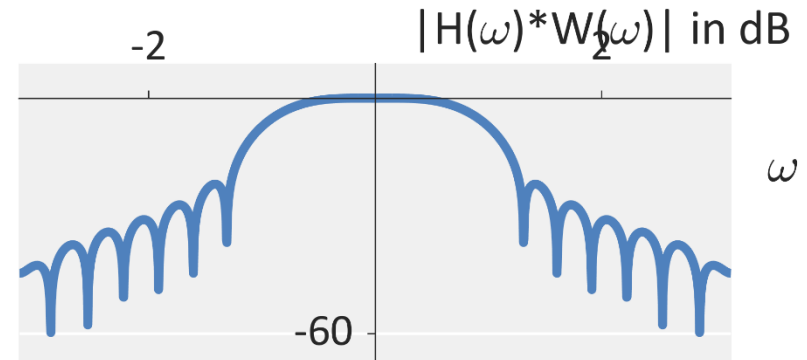
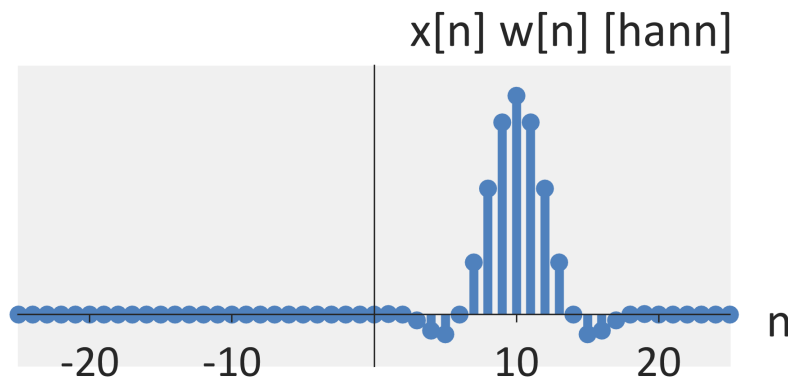
Designing with Windows

■ Windowing the sinc impulse response



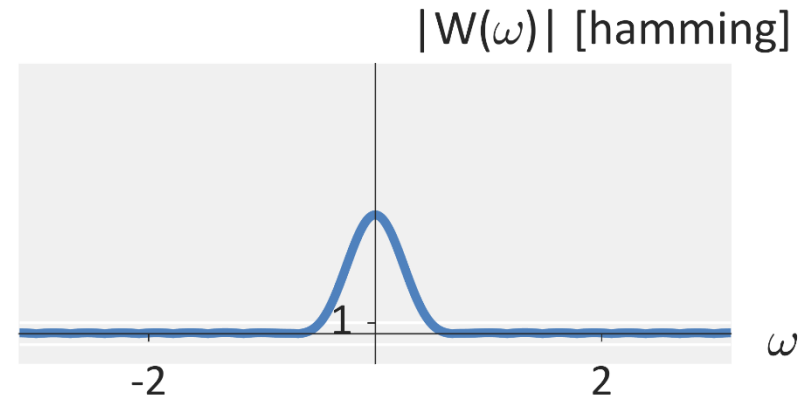
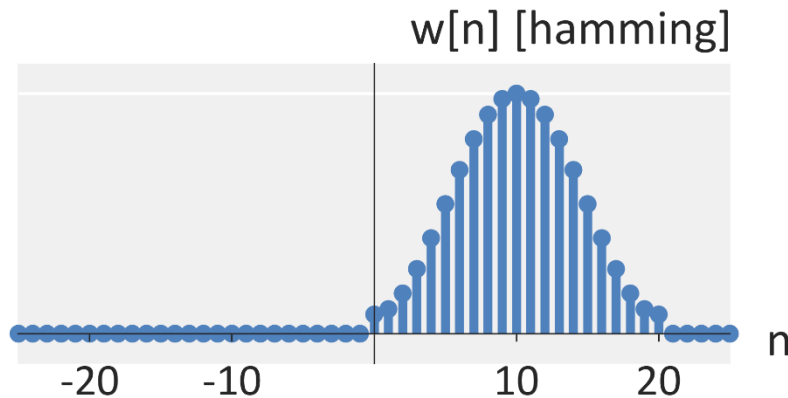
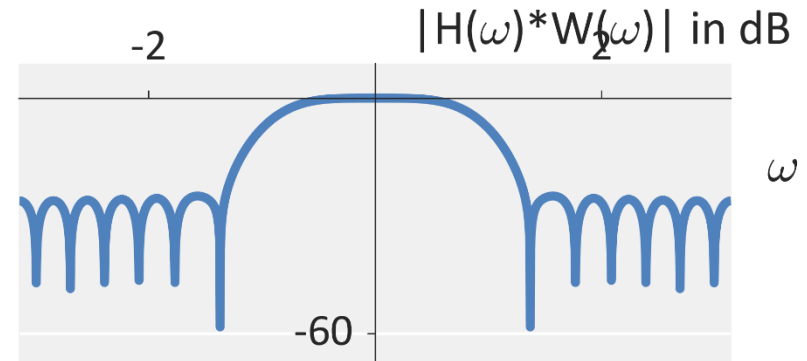
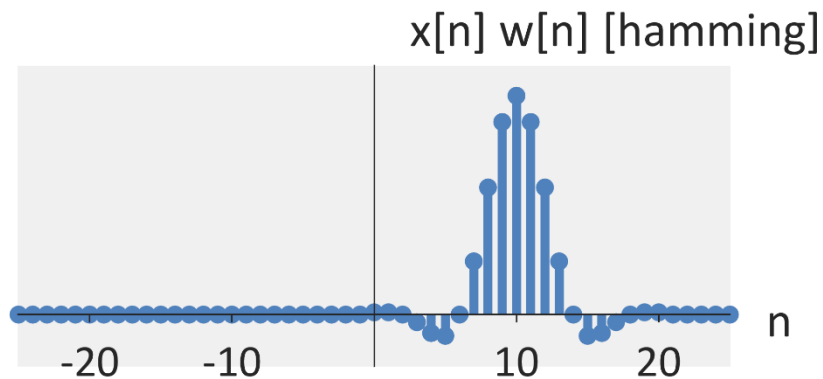
Designing with Windows

■ Windowing the sinc impulse response



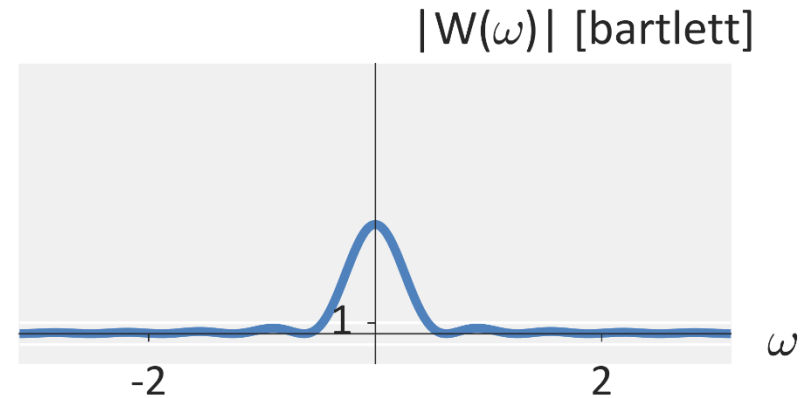
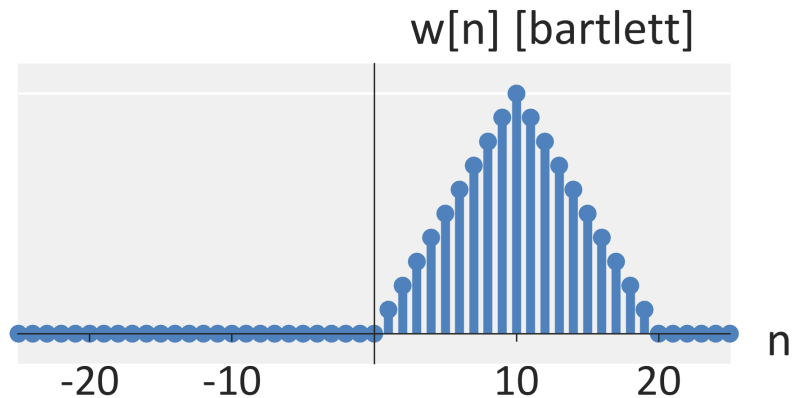
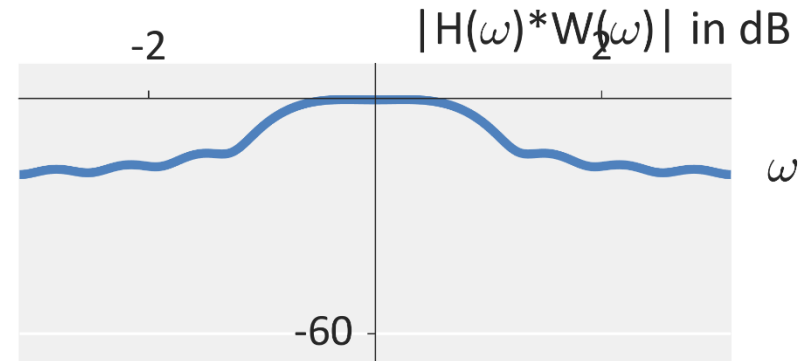
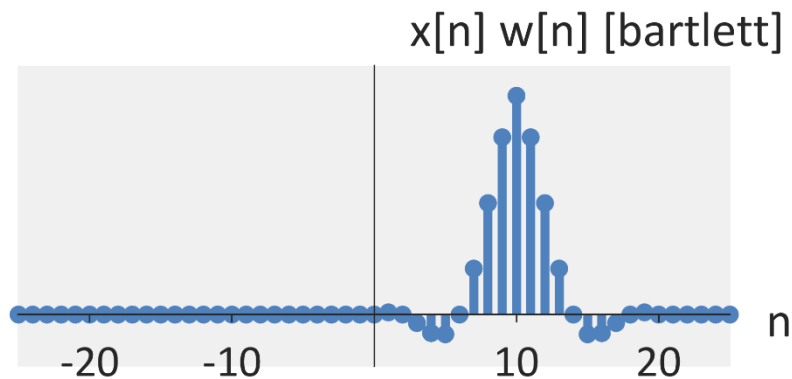
Designing with Windows

■ Windowing the sinc impulse response



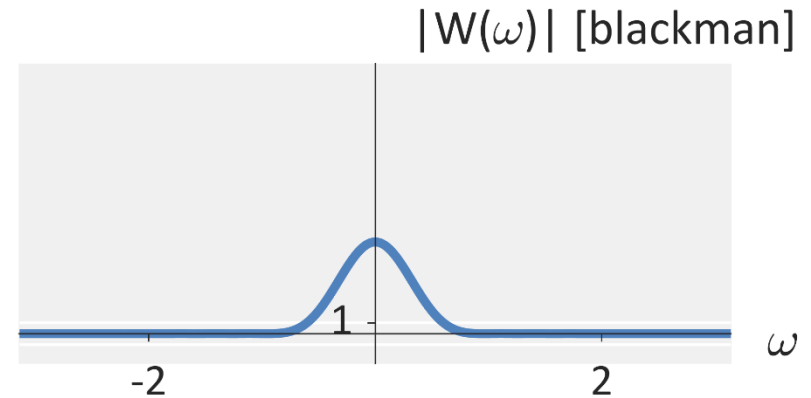
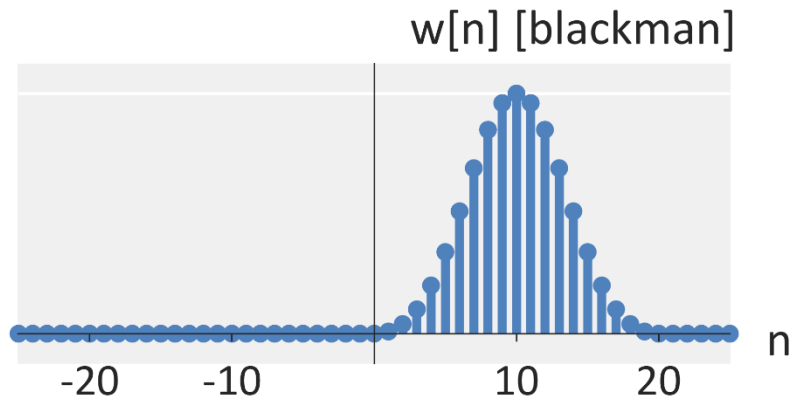
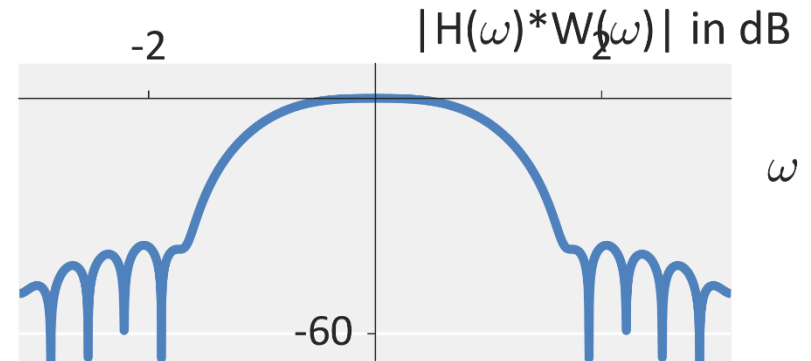
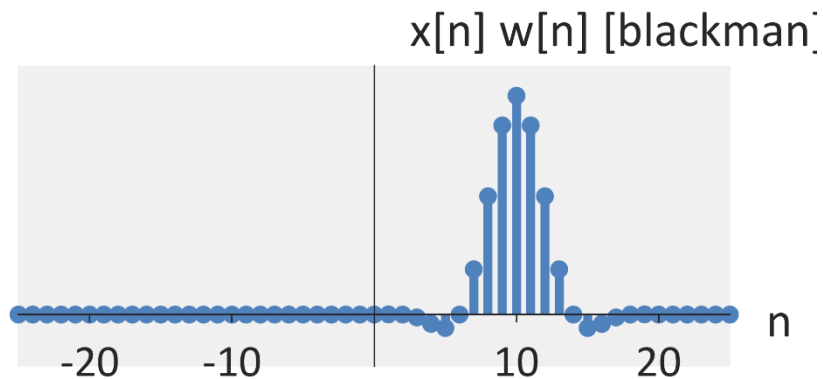
Designing with Windows

■ Windowing the sinc impulse response



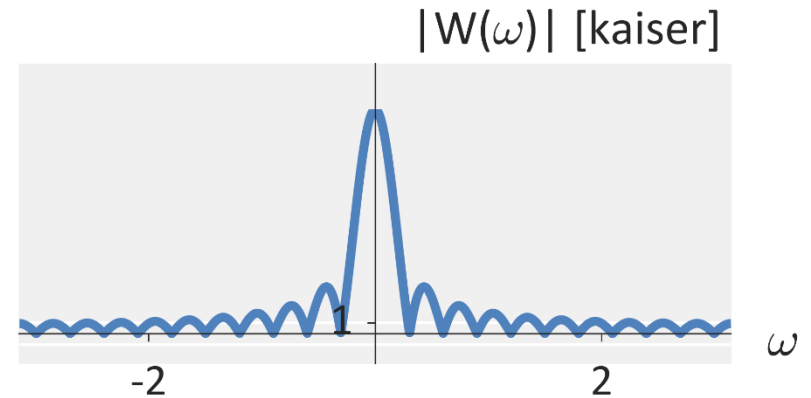
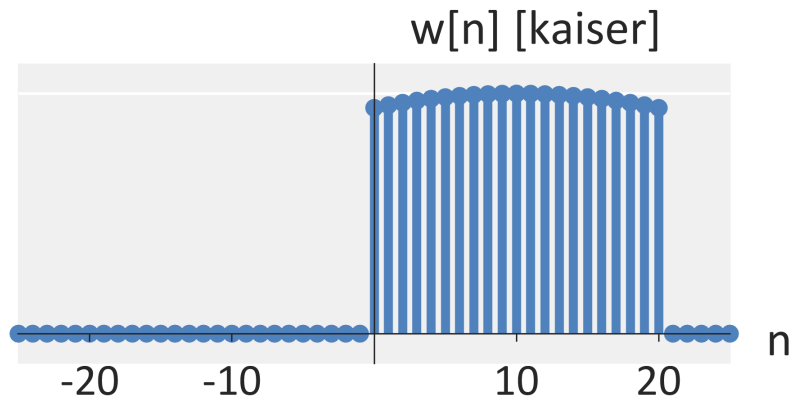
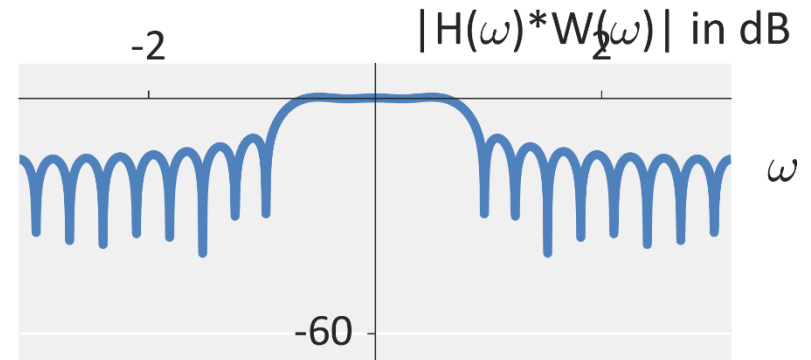
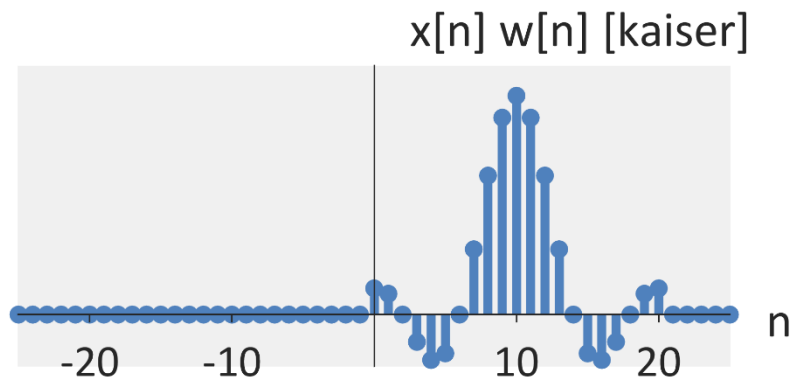
Designing with Windows

■ Windowing the sinc impulse response



Designing with Windows

■ Windowing the sinc impulse response



Lecture 22: Design of FIR / IIR Filters

Foundations of Digital Signal Processing

Outline

- Designing FIR Filters with Windows
- **Designing FIR Filters with Frequency Sampling**
- Designing FIR Filters with Equi-ripples
- Designing IIR Filters with Discrete Differentiation
- Designing IIR Filters with Impulse Invariance
- Designing IIR Filters with the Bilinear Transform
- Related Analog Filters

Design with Frequency Sampling

■ Option 2: Work backwards with constraints

Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

Design with Frequency Sampling

■ Option 2: Work backwards with constraints

Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=(N+1)/2}^{N-1} H[k] e^{j\frac{2\pi}{N}nk}$$

Design with Frequency Sampling

■ Option 2: Work backwards with constraints

Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

Design with Frequency Sampling

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Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

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$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

Design with Frequency Sampling

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Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + H[k] e^{-j\frac{2\pi}{N}nk} e^{j2\pi n}$$

Design with Frequency Sampling

■ Option 2: Work backwards with constraints

Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + H[k] e^{-j\frac{2\pi}{N}nk}$$

Design with Frequency Sampling

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Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] \left(e^{j\frac{2\pi}{N}nk} + e^{-j\frac{2\pi}{N}nk} \right)$$

Design with Frequency Sampling

■ Option 2: Work backwards with constraints

Consider the DFT:

$$h[n] = \sum_{k=0}^{N-1} H[k] e^{j\frac{2\pi}{N}nk} \quad \text{such that} \quad H[k] = H[N - k]$$

$$h[n] = H[0] + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}nk} + \sum_{k=1}^{(N-1)/2} H[k] e^{j\frac{2\pi}{N}n(N-k)}$$

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos\left(\frac{2\pi}{N}nk\right)$$

Design with Frequency Sampling

- **An inverse DFT that forces time-symmetry**

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos\left(\frac{2\pi}{N} nk\right)$$

Design with Frequency Sampling

- **An inverse DFT that forces time-symmetry**

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos\left(\frac{2\pi}{N} nk\right)$$

- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$
 - Compute the frequency sampled filter

Design with Frequency Sampling

- **An inverse DFT that forces time-symmetry**

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos\left(\frac{2\pi}{N} nk\right)$$

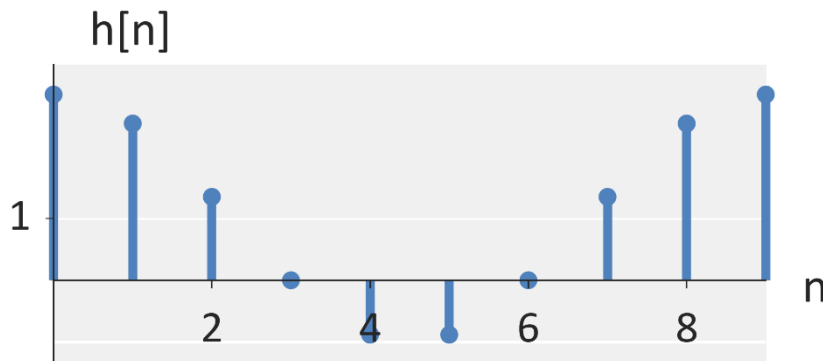
- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$
 - Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos((2\pi/9)n)$$

Design with Frequency Sampling

- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$
- Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos((2\pi/9)n)$$

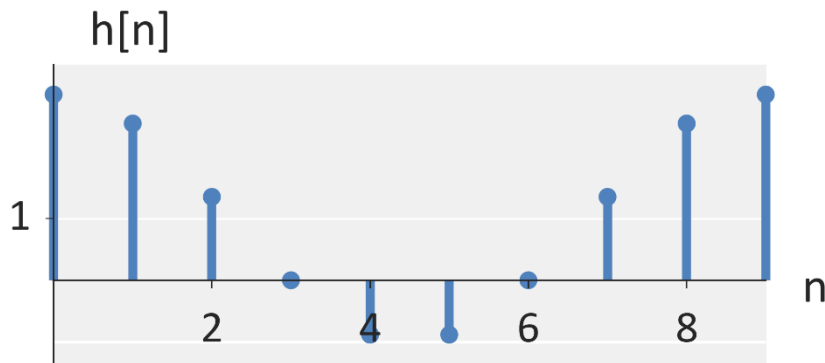


Design with Frequency Sampling

- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$
- Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos((2\pi/9)n)$$

- In practice, this should be circularly shifted so that the maximum is centered.



Design with Frequency Sampling

- **An inverse DFT that forces time-symmetry**

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos\left(\frac{2\pi}{N}\left(n - \frac{N-1}{2}\right)k\right)$$

- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$

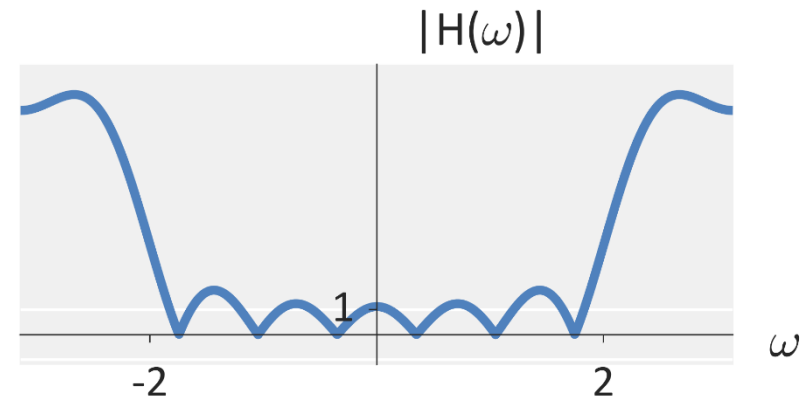
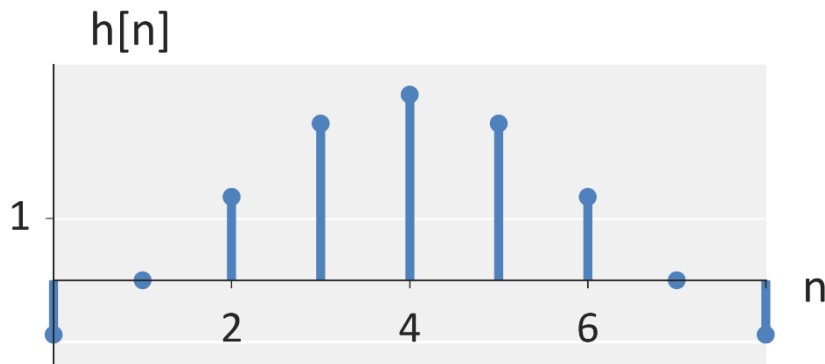
- Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos((2\pi/9)n)$$

Design with Frequency Sampling

- **Example:** Consider the desired 9-sample frequency response with the first half defined by $[1 \ 1 \ 0 \ 0]$
- Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos\left((2\pi/9)(n - 8/2)\right)$$



Design with Frequency Sampling

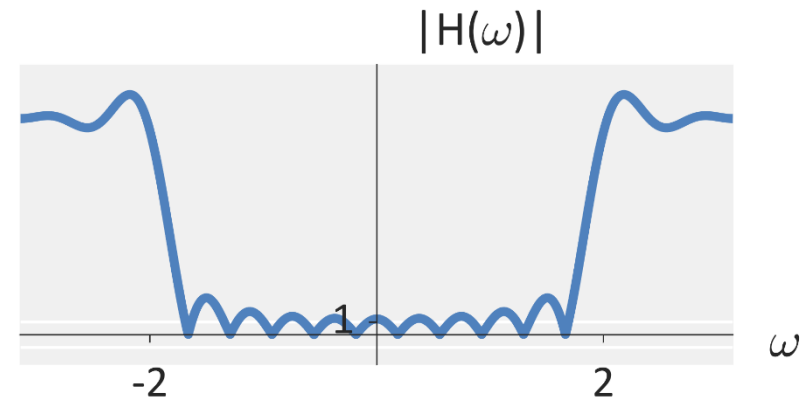
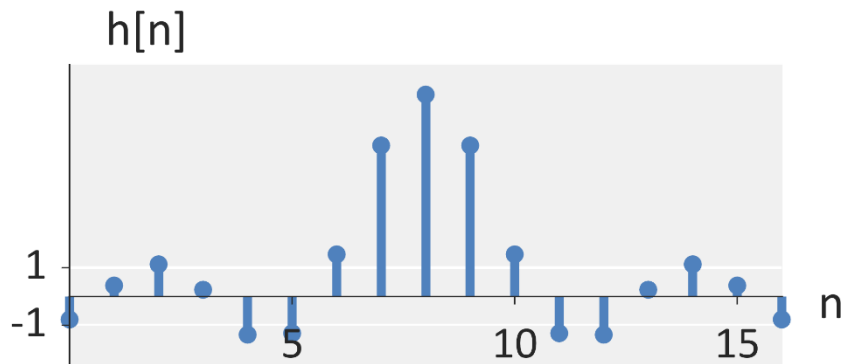
- **Example:** Consider the desired 17-sample frequency response with the first half defined by $[1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 0 \ 0]$
 - Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos((2\pi/19)n_c) + 2 \cos((4\pi/19)n_c) + 2 \cos((6\pi/19)n_c)$$
$$n_c = n - \frac{16}{2}$$

Design with Frequency Sampling

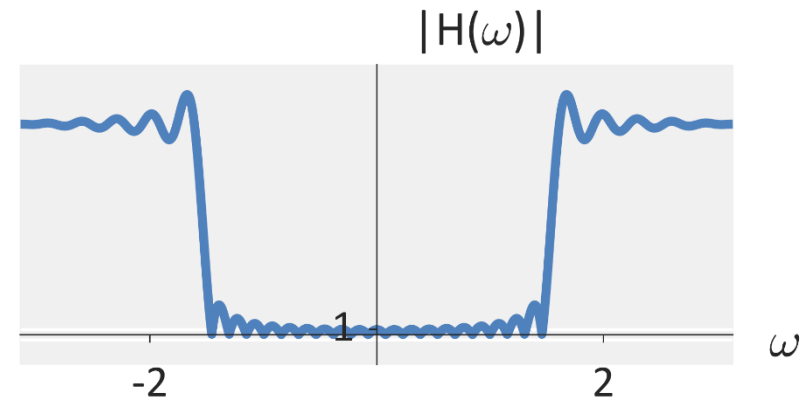
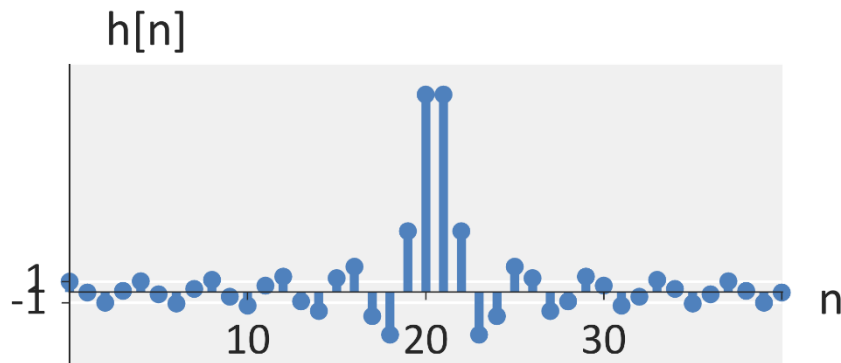
- **Example:** Consider the desired 17-sample frequency response with the first half defined by $[1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 0 \ 0]$
 - Compute the frequency sampled filter

$$h[n] = 1 + 2 \cos\left(\left(2\pi/19\right)n_c\right) + 2 \cos\left(\left(4\pi/19\right)n_c\right) + 2 \cos\left(\left(6\pi/19\right)n_c\right)$$
$$n_c = n - \frac{16}{2}$$



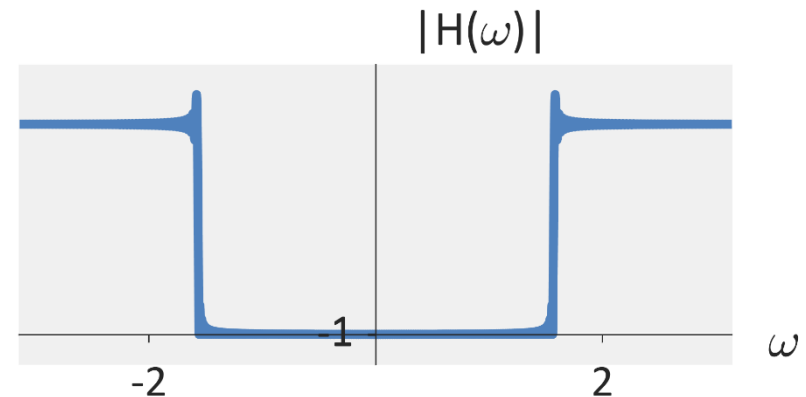
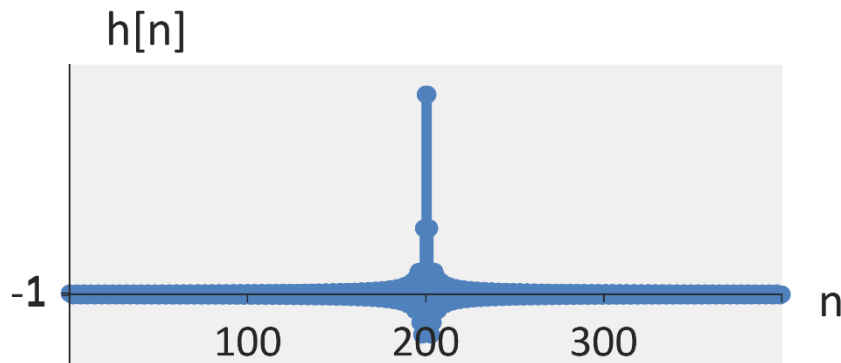
Design with Frequency Sampling

- **Example:** Consider the desired 41-sample frequency response with the first 10 values defined by 1
- Compute the frequency sampled filter



Design with Frequency Sampling

- **Example:** Consider the desired 401-sample frequency response with the first 100 values defined by 1
- Compute the frequency sampled filter
- Note that in practice, this needs to be circularly shifted to the center



Design with Frequency Sampling

■ **Note:** The definition can be slightly modified

- Our definition:

$$\begin{aligned} h[n] &= H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos \left(\frac{2\pi}{N} \left(n - \frac{N-1}{2} \right) k \right) \\ &= H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos \left(\frac{2\pi}{N} \left(n - \frac{N}{2} + \frac{1}{2} \right) k \right) \\ &= H[0] + 2 \sum_{k=1}^{(N-1)/2} H[k] \cos \left(\frac{2\pi}{N} \left(n + \frac{1}{2} \right) k - \pi k \right) \\ &= H[0] + 2 \sum_{k=1}^{(N-1)/2} (-1)^k H[k] \cos \left(\frac{2\pi}{N} \left(n + \frac{1}{2} \right) k \right) \end{aligned}$$

Design with Frequency Sampling

■ Final Definition

$$h[n] = H[0] + 2 \sum_{k=1}^{(N-1)/2} (-1)^k H[k] \cos\left(\frac{2\pi}{N} \left(n + \frac{1}{2}\right) k\right)$$

Side note: This is very closely related to the discrete cosine transform

Lecture 22: Design of FIR / IIR Filters

Foundations of Digital Signal Processing

Outline

- Designing FIR Filters with Windows
- Designing FIR Filters with Frequency Sampling
- **Designing FIR Filters with Equi-ripples**
- Designing IIR Filters with Discrete Differentiation
- Designing IIR Filters with Impulse Invariance
- Designing IIR Filters with the Bilinear Transform
- Related Analog Filters

Design with Equi-ripples

Previously derived:

$$X(z) = z^{-\frac{(M-1)}{2}} \sum_{k=0}^{M/2-1} a_k \left[z^{\frac{(M-1)}{2}-k} + z^{-\left[\frac{(M-1)}{2}-k\right]} \right]$$

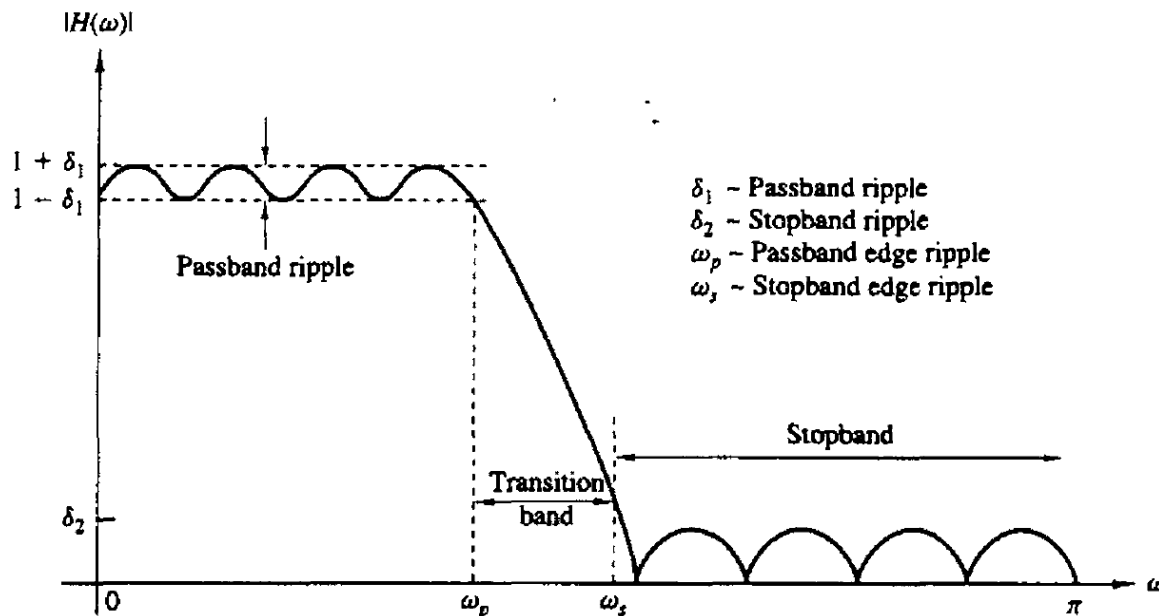
$$\begin{aligned} X(\omega) &= e^{-j\omega \frac{(M-1)}{2}} \sum_{k=0}^{M/2-1} a_k \left[e^{j\omega \left[\frac{(M-1)}{2}-k\right]} + e^{-j\omega \left[\frac{(M-1)}{2}-k\right]} \right] \\ &= 2e^{-j\omega \frac{(M-1)}{2}} \sum_{k=0}^{M/2-1} a_k \cos \left(\omega \left[\frac{M-1}{2} - k \right] \right) \end{aligned}$$

Design with Equi-ripples

■ Equi-ripple design

$$X(\omega) = 2e^{-j\omega \frac{(M-1)}{2}} \sum_{k=0}^{M/2-1} a_k \cos \left(\omega \left[\frac{M-1}{2} - k \right] \right)$$

- **Goal:** Find the optimal a_k s that satisfies **passband / stopband** ripple constraints.



Design with Equi-ripples

■ Equi-ripple design

$$\min_{a_k} W(\omega) \left[H_d(\omega) - 2e^{-j\omega \frac{(M-1)}{2}} \sum_{k=0}^{M/2-1} a_k \cos \left(\omega \left[\frac{M-1}{2} - k \right] \right) \right]$$

Diagram illustrating the equi-ripple design equation. A blue arrow points from the text "Desired frequency response" to the term $H_d(\omega)$ in the equation. Another blue arrow points from the text "Equals:" to the term \min_{a_k} in the equation.

Equals:

$$\frac{\delta_2}{\delta_1} \quad \text{for } \omega \text{ in pass band}$$
$$1 \quad \text{for } \omega \text{ in stop band}$$

δ_2 = stopband ripple
 δ_1 = passband ripple

Lecture 22: Design of FIR / IIR Filters

Foundations of Digital Signal Processing

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IIR Filter Design from Derivatives

■ Designing IIR Filters

- No easy ways to design digital IIR filters
- So let us start from analog filters

IIR Filter Design from Derivatives

■ Designing IIR Filters

- No easy ways to design digital IIR filters
- So let us start from analog filters

■ **Option 1:** Preserve the difference equation!

IIR Filter Design from Derivatives

■ **Question:** What is a derivative in discrete-time?

- In continuous-time

$$\frac{dx(t)}{dt} \rightarrow sX(s)$$

- In discrete-time

IIR Filter Design from Derivatives

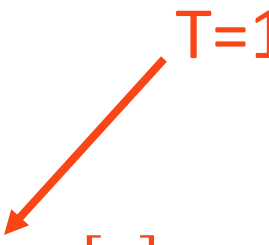
■ Question: What is a derivative in discrete-time?

- In continuous-time

$$\frac{dx(t)}{dt} \rightarrow sX(s)$$

- In discrete-time

$$\begin{aligned} \frac{dx(t)}{dt} &= \lim_{\Delta T \rightarrow 0} \frac{x(t) - x(t - \Delta T)}{\Delta T} \\ \left. \frac{dx(t)}{dt} \right|_{t=nT} &= \frac{x(nT) - x(nT - T)}{T} = x[n] - x[n - 1] \end{aligned}$$

 $T=1$

IIR Filter Design from Derivatives

■ Question: What is a derivative in discrete-time?

- In continuous-time

$$\frac{dx(t)}{dt} \rightarrow sX(s)$$

- In discrete-time

$$\frac{dx(t)}{dt} = \lim_{\Delta T \rightarrow 0} \frac{x(t) - x(t - \Delta T)}{\Delta T}$$

$$\left. \frac{dx(t)}{dt} \right|_{t=nT} = \frac{x(nT) - x(nT - T)}{T} = \frac{1}{T} (x[n] - x[n - 1])$$

IIR Filter Design from Derivatives

■ Question: What is a derivative in discrete-time?

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$$\left. \frac{dx(t)}{dt} \right|_{t=nT} = \frac{x(nT) - x(nT - T)}{T} = \frac{1}{T} (x[n] - x[n - 1])$$

$$\left. \frac{dx(t)}{dt} \right|_{t=nT} \rightarrow \frac{1}{T} (1 - z^{-1})X(z)$$

IIR Filter Design from Derivatives

■ **Question:** What is a second-derivative in discrete-time?

- In continuous-time

$$\frac{d^2x(t)}{dt^2} \rightarrow s^2X(s)$$

- In discrete-time

$$\frac{d^2x(t)}{dt^2} = \frac{dx(t)}{dt} \left[\frac{dx(t)}{dt} \right]$$

$$\left. \frac{dx(t)}{dt} \right|_{t=nT} = \frac{x(nT) - x(nT - T)}{T}$$

IIR Filter Design from Derivatives

■ **Question:** What is a second-derivative in discrete-time?

- In continuous-time

$$\frac{d^2x(t)}{dt^2} \rightarrow s^2X(s)$$

- In discrete-time

$$\frac{d^2x(t)}{dt^2} = \frac{dx(t)}{dt} \left[\frac{dx(t)}{dt} \right]$$

$$\left. \frac{d^2x(t)}{dt^2} \right|_{t=nT} = \frac{[x(nT) - x(nT - T)]/T - [x(nT - T) - x(nT - 2T)]/T}{T}$$

IIR Filter Design from Derivatives

■ **Question:** What is a second-derivative in discrete-time?

- In continuous-time

$$\frac{d^2x(t)}{dt^2} \rightarrow s^2X(s)$$

- In discrete-time

$$\frac{d^2x(t)}{dt^2} = \frac{dx(t)}{dt} \left[\frac{dx(t)}{dt} \right]$$

$$\left. \frac{d^2x(t)}{dt^2} \right|_{t=nT} = \frac{x(nT) - 2x(nT - T) + x(nT - 2T)}{T^2} \rightarrow \frac{x[n] - 2x[n - 1] + x[n - 2]}{T^2}$$

IIR Filter Design from Derivatives

■ Question: What is a second-derivative in discrete-time?

- In continuous-time

$$\frac{d^2x(t)}{dt^2} \rightarrow s^2X(s)$$

- In discrete-time

$$\frac{d^2x(t)}{dt^2} = \frac{dx(t)}{dt} \left[\frac{dx(t)}{dt} \right]$$

$$\left. \frac{d^2x(t)}{dt^2} \right|_{t=nT} = \frac{x(nT) - 2x(nT - T) + x(nT - 2T)}{T^2} \rightarrow \frac{x[n] - 2x[n - 1] + x[n - 2]}{T^2}$$

$$\left. \frac{dx(t)}{dt} \right|_{t=nT} \rightarrow \frac{1}{T^2} (1 - 2z^{-1} + z^{-2})X(z) = \frac{1}{T^2} (1 - z^{-1})^2 X(z)$$

IIR Filter Design from Derivatives

■ **Question:** What is a derivative in discrete-time?

- Translate continuous-time to discrete-time

$$\frac{d^k x(t)}{dt^k} \rightarrow s^k X(s)$$

$$\left. \frac{d^k x(t)}{dt^k} \right|_{t=nT} \rightarrow \frac{1}{T} (1 - z^{-1})^k X(z)$$

IIR Filter Design from Derivatives

■ **Question:** What is a derivative in discrete-time?

- Translate continuous-time to discrete-time

$$\frac{d^k x(t)}{dt^k} \rightarrow s^k X(s)$$

$$\left. \frac{d^k x(t)}{dt^k} \right|_{t=nT} \rightarrow \frac{1}{T} (1 - z^{-1})^k X(z)$$

$$s \rightarrow \frac{1}{T} (1 - z^{-1})$$

IIR Filter Design from Derivatives

■ **Example:** $s \rightarrow \frac{1}{T} (1 - z^{-1})$

- Use the derivative conversion to transform the following biquad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

IIR Filter Design from Derivatives

■ **Example:** $s \rightarrow \frac{1}{T} (1 - z^{-1})$

- Use the derivative conversion to transform the following bi-quad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

$$\begin{aligned} H(z) &= \frac{1}{\left(\frac{1}{T} (1 - z^{-1}) + 0.1\right)^2 + 9} = \frac{T^2}{T^2 \left[\left(\frac{1}{T} (1 - z^{-1}) + 0.1\right)^2 + 9\right]} \\ &= \frac{\sqrt{T}}{\left((1 - z^{-1}) + 0.1T\right)^2 + 9T^2} = \frac{\sqrt{T}}{\left((1 + 0.1T) - z^{-1}\right)^2 + 9T^2} \end{aligned}$$

IIR Filter Design from Derivatives

■ **Example:** $s \rightarrow \frac{1}{T} (1 - z^{-1})$

- Use the derivative conversion to transform the following bi-quad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

$$H(z) = \frac{T^2}{((1 + 0.1T) - z^{-1})^2 + 9T^2}$$

$$((1 + 0.1T) - z^{-1})^2 + 9T^2 = 0$$

$$((1 + 0.1T) - z^{-1})^2 = -9T^2$$

$$(1 + 0.1T) - z^{-1} = \pm 3Tj$$

$$z^{-1} = (1 + 0.1T) \mp 3Tj$$

$$z = \frac{10}{1 + (0.1 + 3j)T}$$

Finding poles



IIR Filter Design from Derivatives

■ **Example:** $s \rightarrow \frac{1}{T} (1 - z^{-1})$

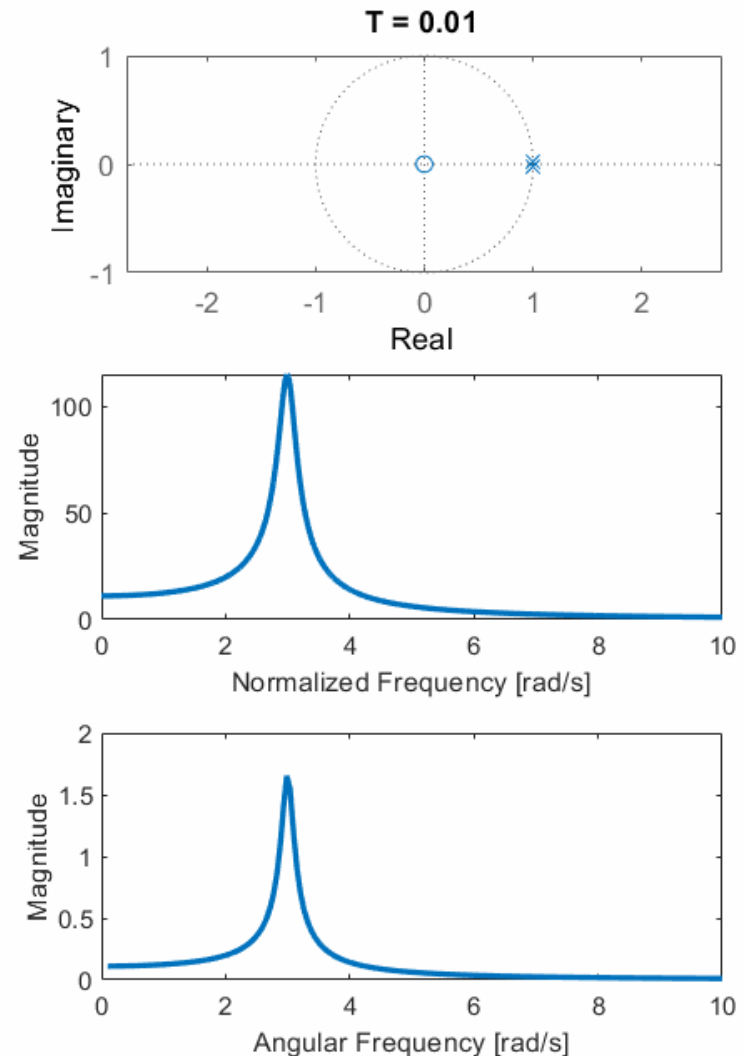
- Use the derivative conversion to transform the following bi-quad filter into the discrete-time domain.

$$z = \frac{10}{1 + (0.1 + 3j)T} \quad \text{Poles}$$

IIR Filter Design from Derivatives

■ **Example:** $s \rightarrow \frac{1}{T} (1 - z^{-1})$

$$Z = \frac{10}{1 + (0.1 + 3j)T}$$



IIR Filter Design from Derivatives

■ Question: What is a derivative in discrete-time?

- Translate continuous-time to discrete-time

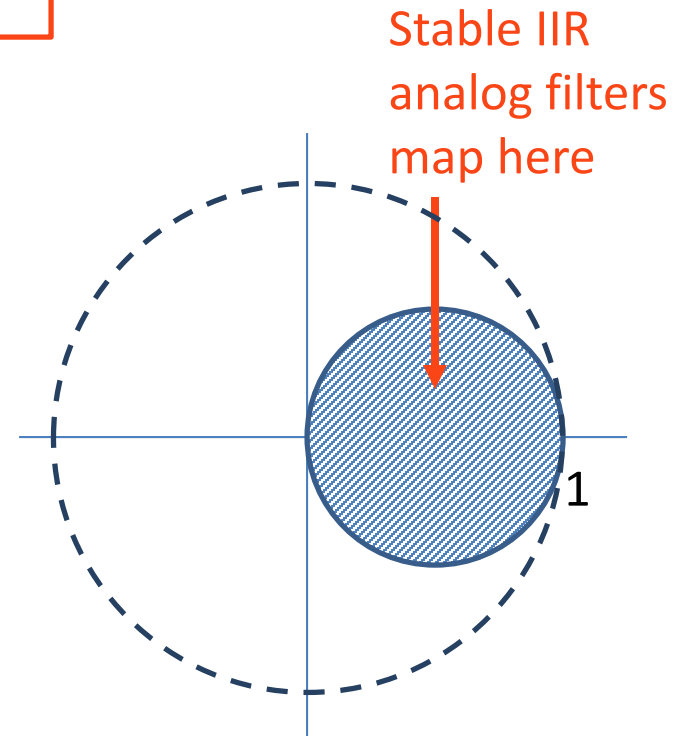
$$s \rightarrow \frac{1}{T} (1 - z^{-1})$$

■ Pros:

- Relatively simple

■ Cons:

- Very limiting
- Stable continuous-time poles can only be mapped to low frequencies



Lecture 22: Design of FIR / IIR Filters

Foundations of Digital Signal Processing

Outline

- Designing FIR Filters with Windows
- Designing FIR Filters with Frequency Sampling
- Designing FIR Filters with Equi-ripples
- Designing IIR Filters with Discrete Differentiation
- **Designing IIR Filters with Impulse Invariance**
- Designing IIR Filters with the Bilinear Transform
- Related Analog Filters

IIR Filter Design by Impulse Invariance

■ Designing IIR Filters

- No easy ways to design digital IIR filters
- So let us start from analog filters

■ Option 2: Preserve the impulse response!

IIR Filter Design by Impulse Invariance

■ **Question:** How else can I represent my transfer function?

$$H(s) = \prod_{k=1}^K \frac{1}{s - p_k}$$

IIR Filter Design by Impulse Invariance

■ **Question:** How else can I represent my transfer function?

$$H(s) = \prod_{k=1}^K \frac{1}{s - p_k} = \sum_{k=1}^K c_k e^{p_k t}$$

IIR Filter Design by Impulse Invariance

■ **Question:** How else can I represent my transfer function?

$$H(s) = \prod_{k=1}^K \frac{1}{s - p_k} = \sum_{k=1}^K c_k e^{p_k t}$$

$$h(t) = \sum_{k=1}^K c_k e^{p_k t}$$

IIR Filter Design by Impulse Invariance

■ **Question:** How else can I represent my transfer function?

$$H(s) = \prod_{k=1}^K \frac{1}{s - p_k} = \sum_{k=1}^K c_k e^{p_k t}$$

$$h(t) = \sum_{k=1}^K c_k e^{p_k t}$$

$$h(nT) = h[n] = \sum_{k=1}^K c_k e^{p_k nT} = \sum_{k=1}^K c_k [e^{p_k T}]^n$$

IIR Filter Design by Impulse Invariance

■ **Question:** How else can I represent my transfer function?

$$H(s) = \prod_{k=1}^K \frac{1}{s - p_k} = \sum_{k=1}^K c_k e^{p_k t}$$

$$h(t) = \sum_{k=1}^K c_k e^{p_k t}$$

$$h(nT) = h[n] = \sum_{k=1}^K c_k e^{p_k nT} = \sum_{k=1}^K c_k [e^{p_k T}]^n$$

$$H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$$

IIR Filter Design from Derivatives

■ **Example:** $H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$

- Use impulse invariance to transform the following biquad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

IIR Filter Design from Derivatives

■ **Example:** $H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$

- Use impulse invariance to transform the following biquad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

Poles:

$$(s + 0.1)^2 + 9 = 0$$

$$s = \pm 3j - 0.1$$

IIR Filter Design from Derivatives

■ **Example:** $H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$

- Use impulse invariance to transform the following biquad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9} = \frac{1/2}{s + 3j + 0.1} + \frac{1/2}{s - 3j + 0.1}$$

Poles:

$$(s + 0.1)^2 + 9 = 0$$

$$s = \pm 3j - 0.1$$

IIR Filter Design from Derivatives

■ **Example:** $H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$

- Use impulse invariance to transform the following biquad filter into the discrete-time domain.

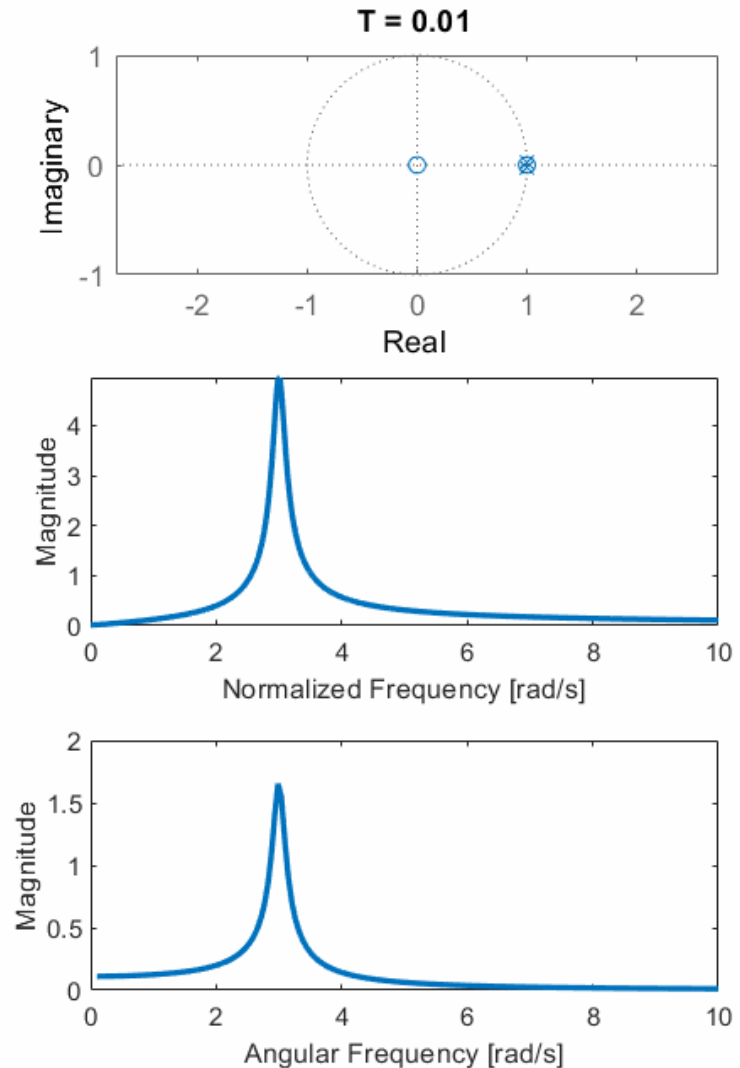
$$H(s) = \frac{1}{(s + 0.1)^2 + 9} = \frac{1/2}{s + 3j + 0.1} + \frac{1/2}{s - 3j + 0.1}$$

$$H(z) = \frac{1/2}{1 - e^{(-3j-0.1)T} z^{-1}} + \frac{1/2}{1 - e^{(3j-0.1)T} z^{-1}}$$

IIR Filter Design from Derivatives

■ **Example:** $H(z) = \sum_{k=1}^K \frac{c_k}{1 - e^{p_k T} z^{-1}}$

$$H(z) = \frac{1/2}{1 - e^{(-3j-0.1)T} z^{-1}} + \frac{1/2}{1 - e^{(3j-0.1)T} z^{-1}}$$



Lecture 22: Design of FIR / IIR Filters

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- Designing IIR Filters with Impulse Invariance
- **Designing IIR Filters with the Bilinear Transform**
- Related Analog Filters

IIR Filter Design by Bilinear Transform

■ Designing IIR Filters

- No easy ways to design digital IIR filters
- So let us start from analog filters

■ Option 3: Preserve the definition of z !

IIR Filter Design by Bilinear Transform

■ Question: How are z and s related?

- From continuous-time to discrete-time

$$s = j\Omega$$

$$e^{st} = e^{snT} = z^n$$

$$z = e^{sT}$$

Taylor Series
Expansion / Approximation



- Building an approximation ($e^x \approx 1 + x$)

$$z = \frac{e^{\frac{sT}{2}}}{e^{-\frac{sT}{2}}} \approx \frac{1 + sT/2}{1 - sT/2}$$

IIR Filter Design by Bilinear Transform

■ Question: How are z and s related?

- From continuous-time to discrete-time

$$s = j\Omega$$

$$e^{st} = e^{snT} = z^n$$

$$z = e^{sT} \quad s = \frac{1}{T} \ln(z)$$

**Bilinear
Expansion / Approximation**



- Building an approximation

$$s \approx \frac{2}{T} \frac{z - 1}{z + 1}$$

IIR Filter Design by Bilinear Transform

■ The Bilinear Transform

Continuous-time to discrete-time

$$s \rightarrow \frac{2}{T} \frac{1 - z^{-1}}{1 + z^{-1}}$$

Discrete-time to continuous-time

$$z \rightarrow \frac{1 + sT/2}{1 - sT/2}$$

IIR Filter Design by Bilinear Transform

■ **Example:** $s \rightarrow \frac{2}{T} \frac{1-z^{-1}}{1+z^{-1}}, \quad z \rightarrow \frac{1+sT/2}{1-sT/2}$

- Use the bilinear transform to transform the following biquad filter into the discrete-time domain.

$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

IIR Filter Design by Bilinear Transform

■ **Example:** $s \rightarrow \frac{2}{T} \frac{1-z^{-1}}{1+z^{-1}}, \quad z \rightarrow \frac{1+sT/2}{1-sT/2}$

- Use the bilinear transform to transform the following biquad filter into the discrete-time domain.

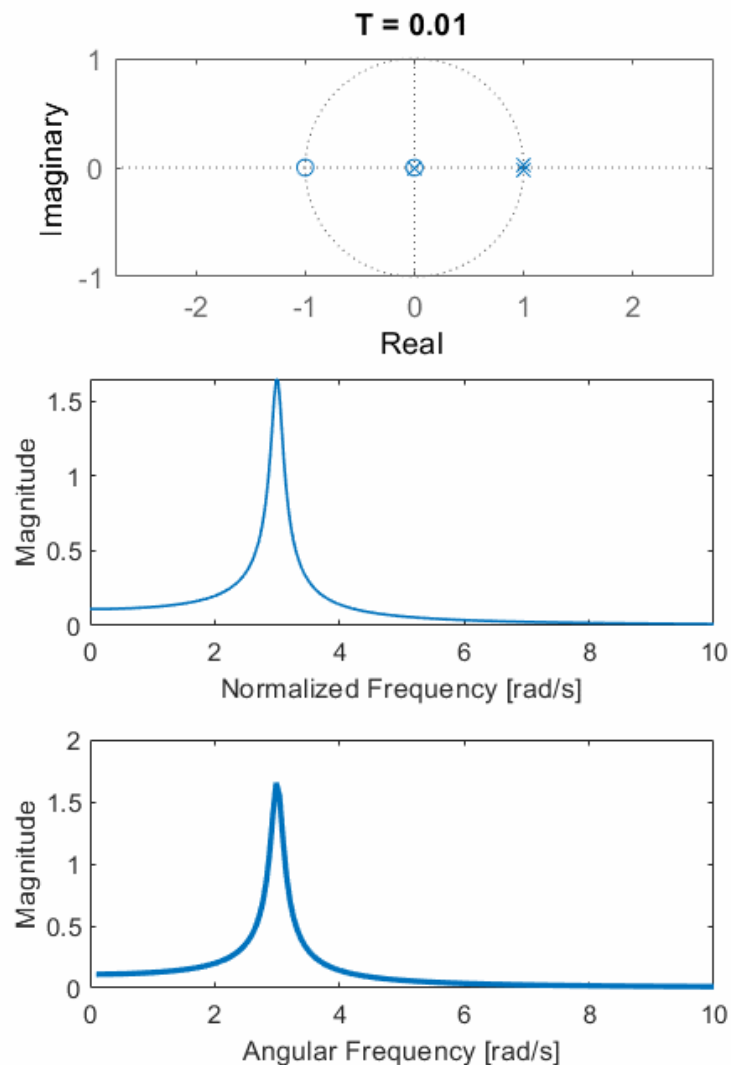
$$H(s) = \frac{1}{(s + 0.1)^2 + 9}$$

$$\begin{aligned} H(z) &= \frac{1}{\left(\frac{2}{T} \frac{1-z^{-1}}{1+z^{-1}} + 0.1\right)^2 + 9} \\ &= \frac{(1+z^{-1})^2}{\left(\frac{2}{T}(1-z^{-1}) + 0.1\right)^2 + 9(1+z^{-1})^2} \end{aligned}$$

IIR Filter Design by Bilinear Transform

■ **Example:** $s \rightarrow \frac{2}{T} \frac{1-z^{-1}}{1+z^{-1}}$, $Z \rightarrow \frac{1+sT/2}{1-sT/2}$

$$H(z) = \frac{(1+z^{-1})^2}{\left(\frac{2}{T}(1-z^{-1}) + 0.1\right)^2 + 9(1+z^{-1})^2}$$



Lecture 22: Design of FIR / IIR Filters

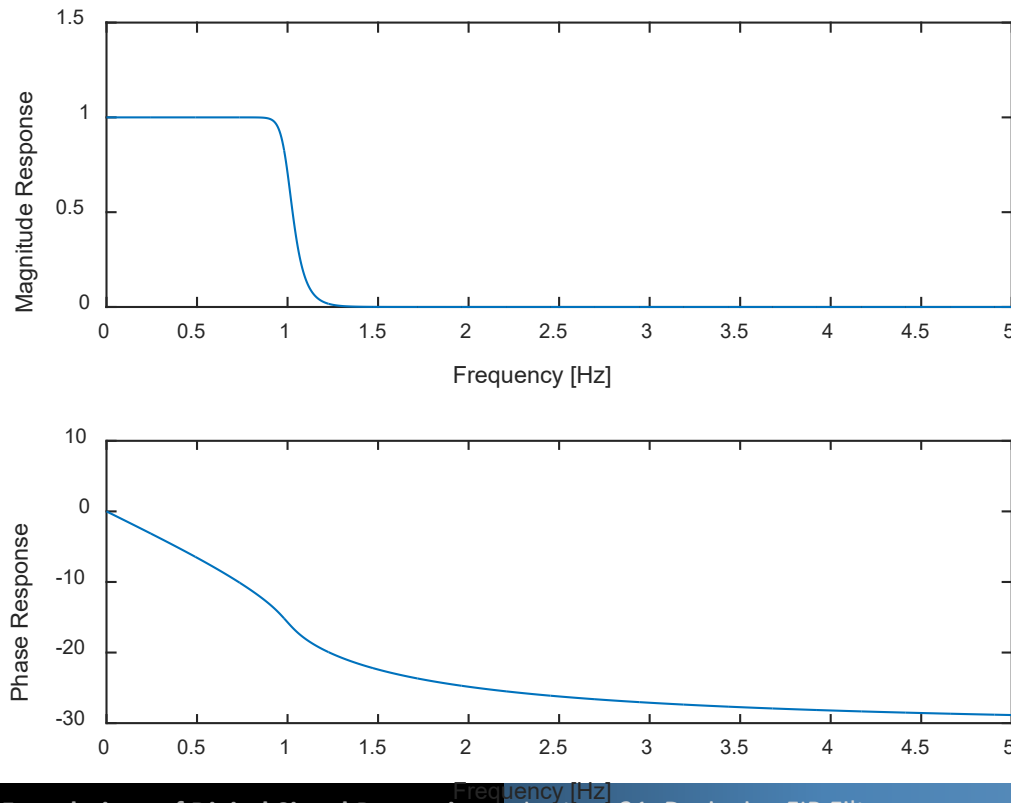
Foundations of Digital Signal Processing

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Multi-pole Filters

- **Butterworth:** Maximally flat passband
- **Chebyshev:** Faster cutoff with passband ripple
- **Elliptic:** Fastest cutoff with passband and stopband ripple



Butterworth Filter

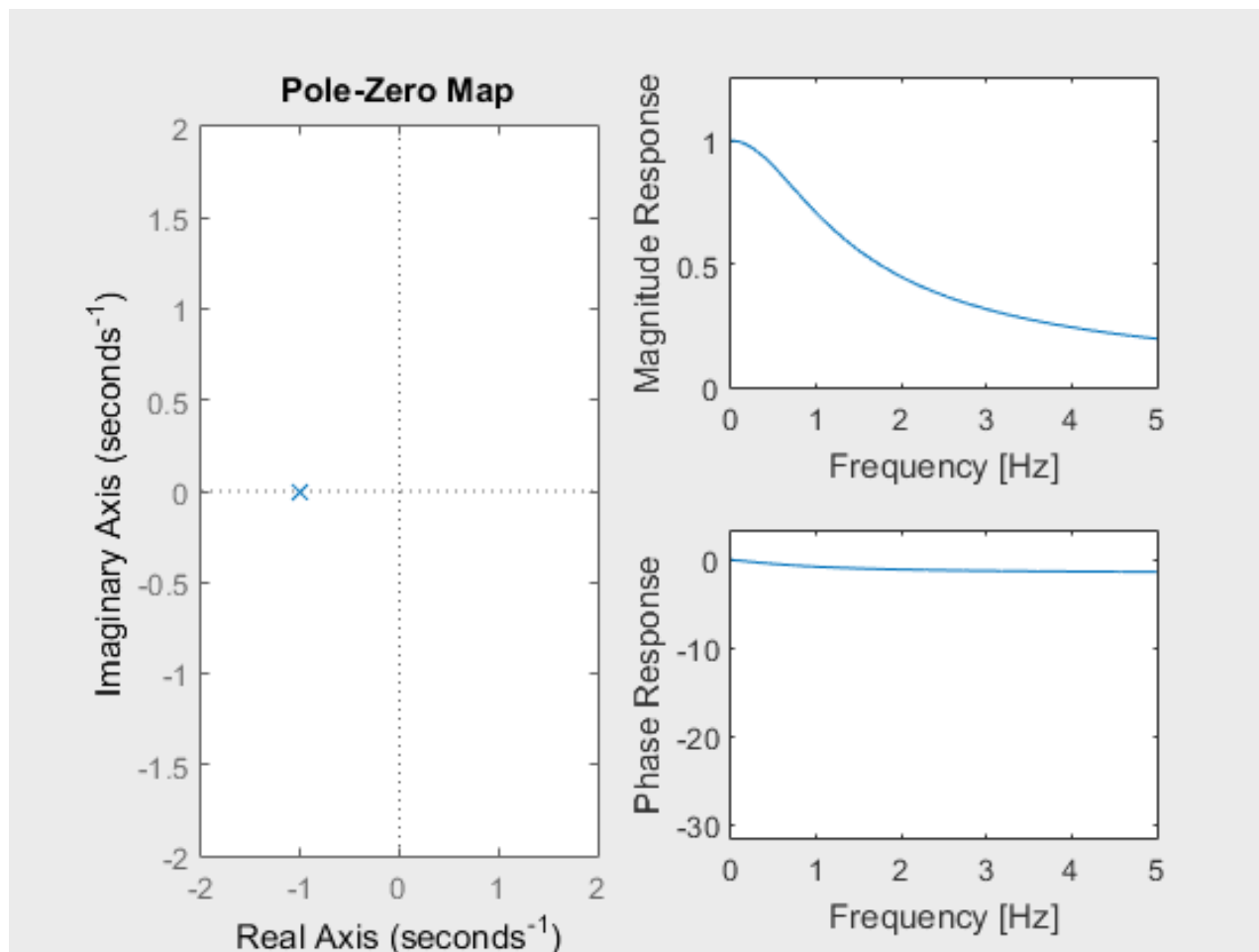
■ Butterworth Filter of order N

$$|H(j\omega)| = \frac{1}{\sum_{k=1}^N (s - s_k)} \quad s_k = e^{\frac{j(2k+N-1)\pi}{2N}}$$

Butterworth Filter

■ Butterworth Filter of order N

- N equally spaced poles on a circle on the left-hand-side of the s-plane



Butterworth Filter

■ Properties of the Butterworth Filter

- It is maximally flat at $\omega = 0$
- It has a cutoff frequency $|H(\omega)| = \frac{1}{\sqrt{2}}$ at $\omega = \omega_c$
- For large n , it becomes an ideal filter

Chebyshev Filter

■ Chebyshev Filter of order N

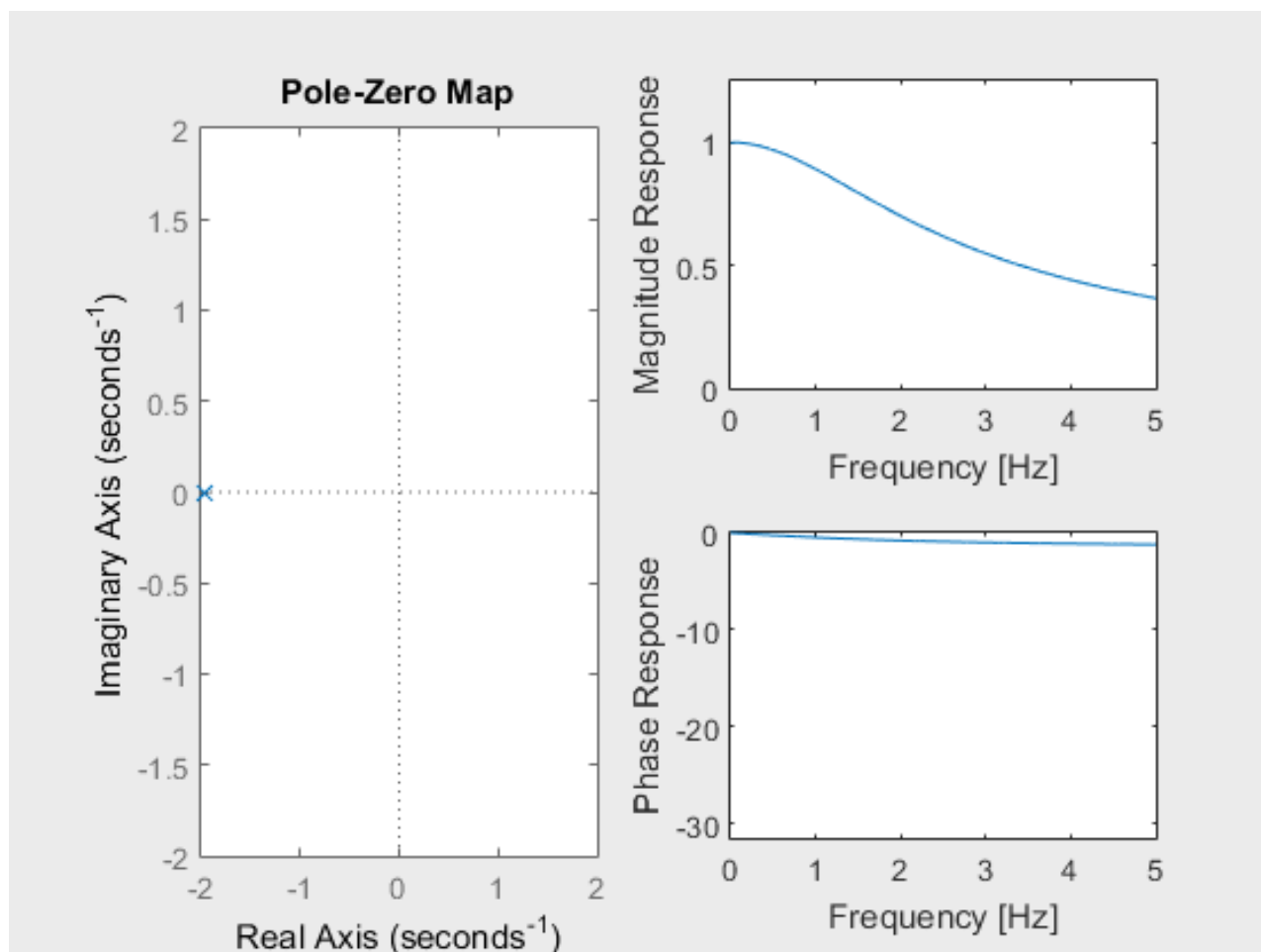
$$|H(j\omega)| = \frac{1}{\sqrt{1 + \epsilon^2 C_n^2(\omega)}}$$

$C_n^2(\omega)$ is an nth-order Chebyshev Polynomial

ϵ^2 controls ripple

Chebyshev Filter

■ Chebyshev Filter of order N



Chebyshev Filter

■ Properties of the Chebyshev Filter

- It has ripples in the passband and is smooth in the stopband.
- The ratio between the maximum and minimum ripples in the passband is

$$(1 + \epsilon^2)^{-1/2}$$

- If ϵ is reduced (i.e., the ripple size is reduced), then the stopband attenuation is reduced.
- It has a sharper cut-off than a Butterworth filter, but at the expense of passband rippling

Elliptic Filter

■ Elliptic Filter of order N

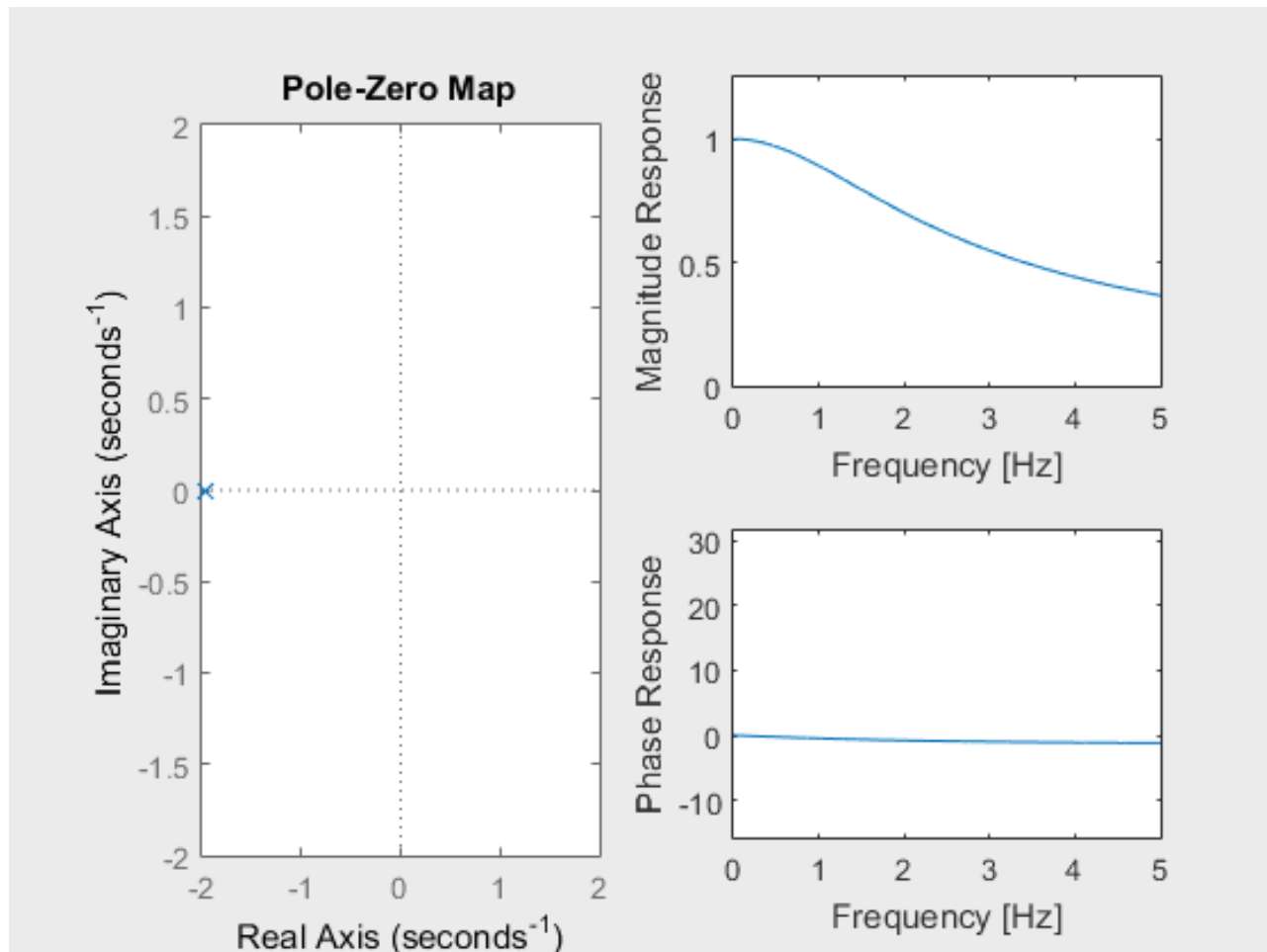
$$|H(j\omega)| = \frac{1}{\sqrt{1 + \epsilon^2 R_n^2(\omega)}}$$

$R_n^2(\omega)$ is an nth-order elliptic function

ϵ^2 controls ripple

Elliptic Filter

■ Elliptic Filter of order N



Elliptic Filter

■ Properties of the Elliptic Filter

- It has ripples in the passband and the stopband
- The ratio between the maximum and minimum ripples is larger than the Chebyshev filter, but it has an even quicker transition from passband to stopband
- It has poles and zeros, but they are much more difficult to compute compared with the Butterworth and Chebyshev filters