



<http://elec3004.com>

Digital Filters (Part II)

ELEC 3004: Systems: Signals & Controls
Dr. Surya Singh

Lecture 12
(with material from Lathi and Cannon (Discrete Systems))

elec3004@itee.uq.edu.au

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<http://robotics.itee.uq.edu.au/~elec3004/>

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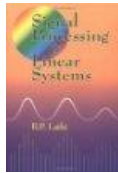


Lecture Schedule:

Week	Date	Lecture Title
1	29-Feb	Introduction
	3-Mar	Systems Overview
2	7-Mar	Systems as Maps & Signals as Vectors
	10-Mar	Data Acquisition & Sampling
3	14-Mar	Sampling Theory
	17-Mar	Antialiasing Filters
4	21-Mar	Discrete System Analysis
	24-Mar	Convolution Review
	28-Mar	Holiday
	31-Mar	Holiday
5	4-Apr	Frequency Response & Filter Analysis
	7-Apr	Filters
6	11-Apr	Digital Filters
	14-Apr	Digital Filters
7	18-Apr	Digital Windows
	21-Apr	FFT
8	25-Apr	Holiday
	28-Apr	Feedback
9	3-May	Introduction to Feedback Control
	5-May	Servoregulation/PID
10	9-May	Introduction to (Digital) Control
	12-May	Digital Control
11	16-May	Digital Control Design
	19-May	Stability
12	23-May	Digital Control Systems: Shaping the Dynamic Response & Estimation
	26-May	Applications in Industry
13	30-May	System Identification & Information Theory
	2-Jun	Summary and Course Review



Follow Along Reading:



B. P. Lathi
*Signal processing
and linear systems*
1998
[TK5102.9.L38 1998](#)

Today

- Chapter 10
**(Discrete-Time System Analysis
Using the z -Transform)**
 - § 10.3 Properties of DTFT
 - § 10.5 Discrete-Time Linear System
analysis by DTFT
 - § 10.7 Generalization of DTFT
to the \mathcal{Z} -Transform

- Chapter 12
(Frequency Response and Digital Filters)
 - § 12.1 Frequency Response of Discrete-Time Systems
 - § 12.3 Digital Filters
 - § 12.4 Filter Design Criteria
 - § 12.7 Nonrecursive Filters

Next Time

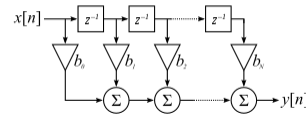


FIR!

** FIR Filter Design **

- How to get all these coefficients?

$$H(\omega) = h_0 + h_1 e^{-i\omega} + \dots + h_{n-1} e^{-i(n-1)\omega}$$



FIR Design Methods:

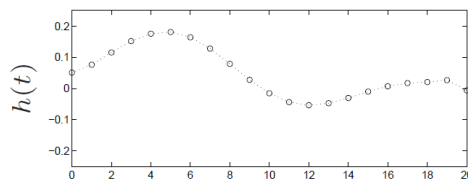
1. Impulse Response Truncation
 - + Simplest
 - Undesirable frequency domain-characteristics, not very useful
2. Windowing Design Method
 - + Simple
 - Not optimal (not minimum order for a given performance level)
3. Optimal filter design methods
 - + "More optimal"
 - Less simple...



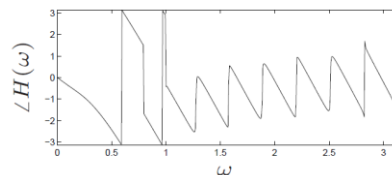
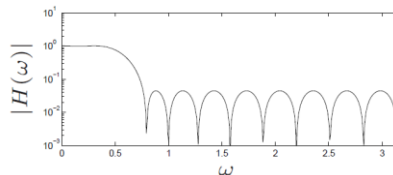
FIR Filter Design & Operation

Ex: Lowpass FIR filter

- Set Impulse response (order $n = 21$)
- "Determine" $h(t)$
 - $h(t)$ is a 20 element vector that we'll use to as a weighted sum



- FFT ("Magic") gives $H(\omega)$ Frequency Response & Phase



Why is this “hard”? Looking at the Low-Pass Example

$$H_d(\omega) = \begin{cases} 1 & \text{if } |\omega| \leq \omega_c \\ 0 & \text{if } \omega_c < |\omega| < \pi \end{cases}$$

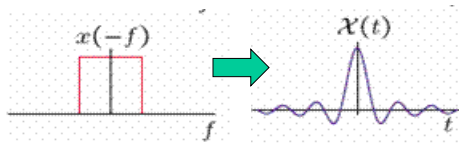
- Why is this hard?
 - Shouldn't it be “easy” ??
 - ... just hit it with some FFT “magic” and then keep the bands we want and then hit it with some Inverse-FFT “supermagic”???
 - Remember we need a “system” that does this “rectangle function” in frequency
 - Let's consider what that means...
 - It basically suggests we need an **Inverse FFT** of a **“rectangle function”**



Flashback: Fourier Series & Rectangular Functions

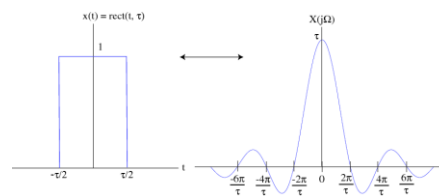
\mathfrak{F} : Fourier Transform

$$\mathfrak{F}^{-1} \left\{ \text{rect} \left(\frac{\omega}{2} \right) \right\} = \frac{\text{sinc}(t)}{\pi}$$



Ref: <http://cnx.org/content/m26719/1.1/>
<http://www.wolframalpha.com/input/?=IFFT%28sinc%28%29%29>

$$\mathfrak{F} \{ \text{rect}(t) \} = \text{sinc} \left(\frac{\omega}{2} \right)$$



Ref: <http://cnx.org/content/m32899/1.8/>
<http://www.thefouriertransform.com/pairs/box.php>

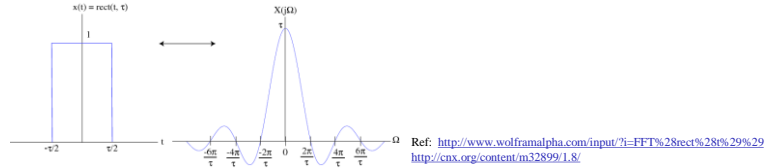
See:

- Table 7.1 (p. 702) Entry 17
& Table 9.1 (p. 852) Entry 7



Flashback: Fourier Series & Rectangular Functions [2]

- The sinc function might look familiar
 - This is the frequency content of a square wave (box)



- This also applies to **signal reconstruction!**
 - **Whittaker–Shannon interpolation formula**
 - This says that the “better way” to go from Discrete to Continuous (i.e. D to A) is not ZOH, but rather via the sinc!

$$x(t) = \sum_{n=-\infty}^{\infty} x[n] \cdot \text{sinc}\left(\frac{t-nT}{T}\right)$$

∴ FIR and Low Pass Filters...

$$\therefore H_d(\omega) = \begin{cases} 1 & \text{if } |\omega| \leq \omega_c \\ 0 & \text{if } \omega_c < |\omega| < \pi \end{cases}$$

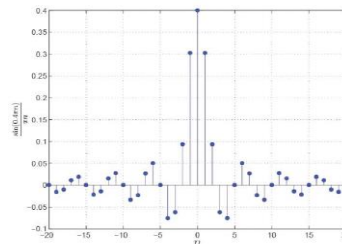
Has impulse response:

$$h_d(n) = \frac{\omega_c}{\pi} \frac{\sin \omega_c n}{\omega_c n}$$

Thus, to filter an impulse train with an ideal **low-pass filter** use:

$$x(t) = \left(\sum_{n=-\infty}^{\infty} x[n] \cdot \delta(t - nT) \right) * \text{sinc}\left(\frac{t}{T}\right)$$

- However!!**
a sinc is non-causal and infinite in duration



And, this **cannot be implemented in practice** ☹

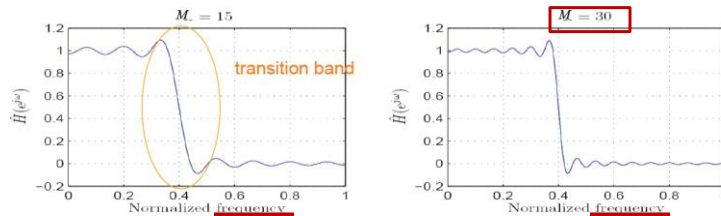
- ∴ we need to know all samples of the input, both in the **past** and in the **future**

Plan 0: Impulse Response Truncation

Maybe we saw this coming...

∴ Clip off the sinc at some large n

$$\hat{h}(n) = \frac{\sin(n\omega_c)}{\pi n} \text{ for } |n| \leq M \text{ and } 0 \text{ otherwise}$$



- *Ripples* in both passband/stopband and the transition not abrupt (i.e., a *transition band*).
- As $M \rightarrow \infty$, transition band $\rightarrow 0$ (as expected!)



→ FIR Filters: Window Function Design Method

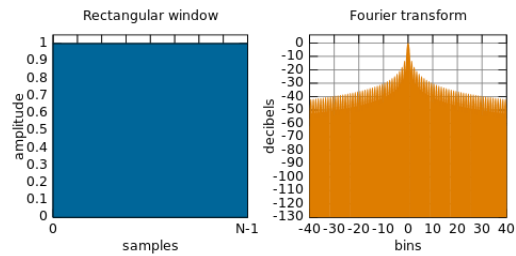
- Windowing: a generalization of the truncation idea
- There many, many “window” functions:
 - Rectangular
 - Triangular
 - Hanning
 - Hamming
 - Blackman
 - Kaiser
 - Lanczos
 - Many More ... (see: http://en.wikipedia.org/wiki/Window_function)



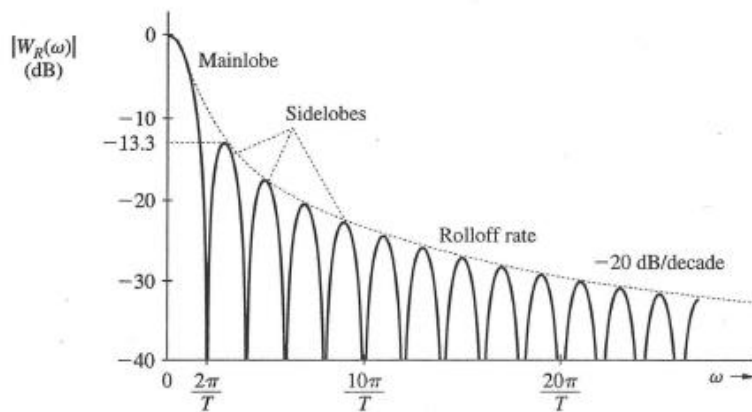
Some Window Functions [1]

1. Rectangular

$$w(n) = 1$$



Windowing and its effects/terminology



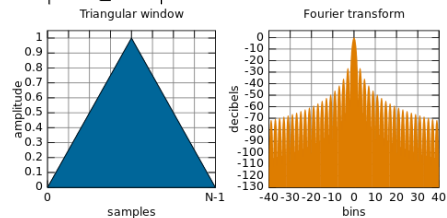
Lathi, Fig. 7.45



Some More Window Functions ...

2. Triangular window

$$w(n) = 1 - \left| \frac{n - \frac{N-1}{2}}{\frac{N+1}{2}} \right|$$



- And Bartlett Windows
 - A slightly narrower variant with zero weight at both ends:

$$w(n) = 1 - \left| \frac{n - \frac{N-1}{2}}{\frac{N-1}{2}} \right|$$



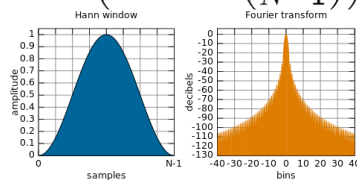
Some More Window Functions...

3. Generalized Hamming Windows

$$w(n) = \alpha - \beta \cos\left(\frac{2\pi n}{N-1}\right)$$

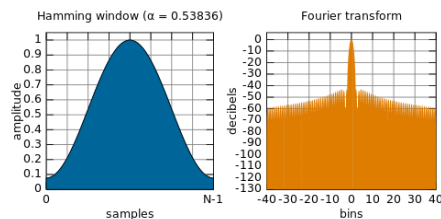
→ Hanning Window

$$\rightarrow w(n) = 0.5 \left(1 - \cos\left(\frac{2\pi n}{N-1}\right) \right)$$



→ Hamming's Window

$$\rightarrow \alpha = 0.54, \beta = 1 - \alpha = 0.46$$

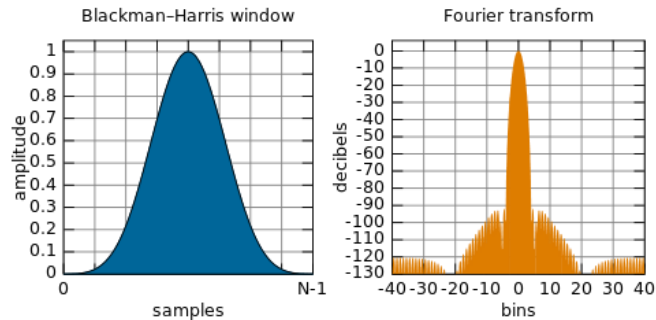


Some More Window Functions...

4. Blackman–Harris Windows

- A generalization of the Hamming family,
- Adds more shifted sinc functions for less side-lobe levels

$$w(n) = a_0 - a_1 \cos\left(\frac{2\pi n}{N-1}\right) + a_2 \cos\left(\frac{4\pi n}{N-1}\right) - a_3 \cos\left(\frac{6\pi n}{N-1}\right)$$



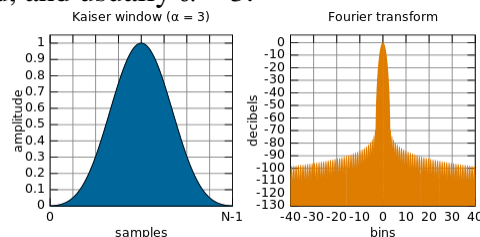
Some More Window Functions...

5. Kaiser window

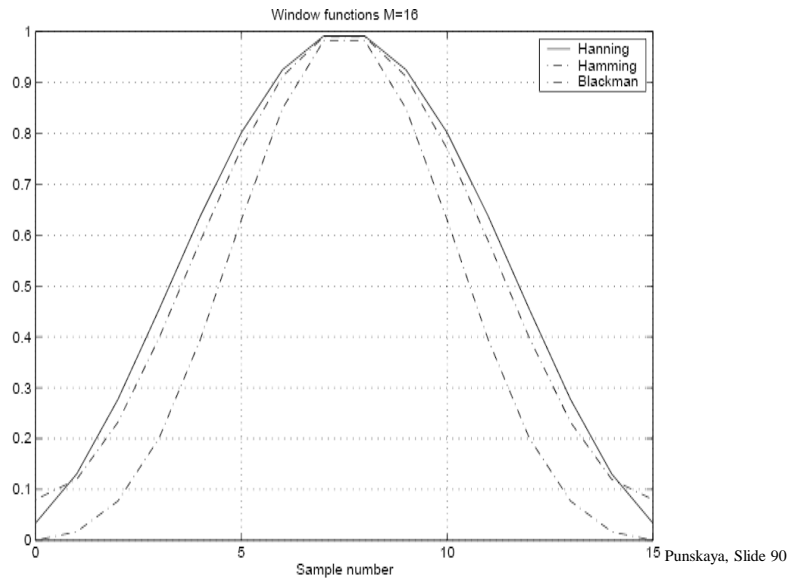
- A DPSS (discrete prolate spheroidal sequence)
- Maximize the energy concentration in the main lobe

$$\rightarrow w(n) = \frac{I_0\left(\pi\alpha\sqrt{1-\left(\frac{2n}{N-1}-1\right)^2}\right)}{I_0(\pi\alpha)}$$

- Where: I_0 is the zero-th order modified Bessel function of the first kind, and usually $\alpha = 3$.

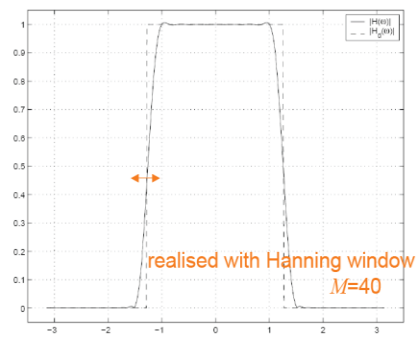
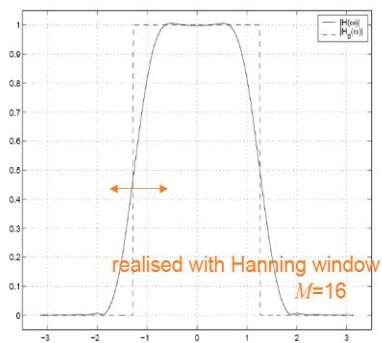


Comparison of Alternative Windows –Time Domain

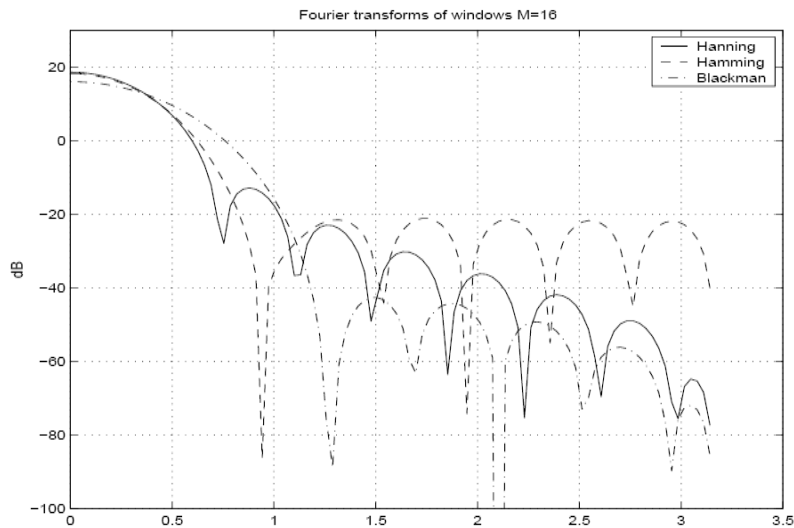


Adding Order

- + Transition and Smoothness
- Increased Size



Comparison of Alternative Windows Frequency Domain



Punskaya, Slide 91



Summary Characteristics of Common Window Functions

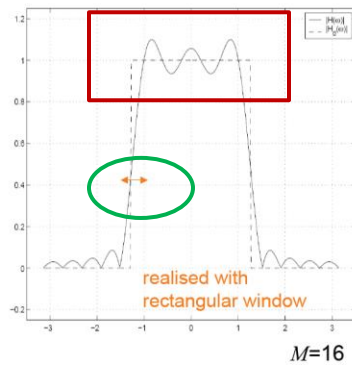
No.	Window $w(t)$	Mainlobe Width	Rolloff Rate (dB/oct)	Peak Sidelobe level (dB)	Peak $20\log_{10}\delta$
1	Rectangular: $\text{rect}\left(\frac{t}{T}\right)$	$\frac{4\pi}{T}$	-6	-13.3	-21dB
2	Bartlett: $\Delta\left(\frac{t}{2T}\right)$	$\frac{8\pi}{T}$	-12	-26.5	
3	Hanning: $0.5\left[1 + \cos\left(\frac{2\pi t}{T}\right)\right]$	$\frac{8\pi}{T}$	-18	-31.5	-44dB
4	Hamming: $0.54 + 0.46\cos\left(\frac{2\pi t}{T}\right)$	$\frac{8\pi}{T}$	-6	-42.7	-53dB
5	Blackman: $0.42 + 0.5\cos\left(\frac{2\pi t}{T}\right) + 0.08\cos\left(\frac{4\pi t}{T}\right)$	$\frac{12\pi}{T}$	-18	-58.1	-74dB
6	Kaiser: $\frac{I_0\left[\alpha\sqrt{1-4\left(\frac{t}{T}\right)^2}\right]}{I_0(\alpha)}$ $0 \leq \alpha \leq 10$	$\frac{11.2\pi}{T}$	-6	-59.9 ($\alpha = 8.168$)	

Lathi, Table 7.3
Punskaya, Slide 92

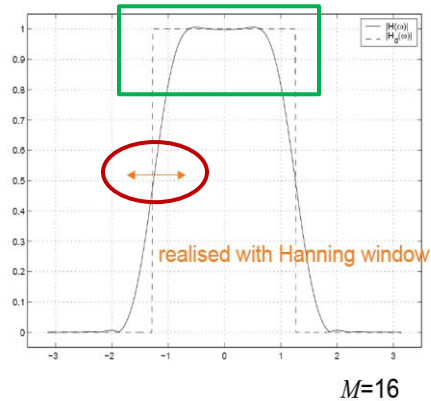


FIR: Rectangular & Hanning Windows

- Rectangular



- Hanning

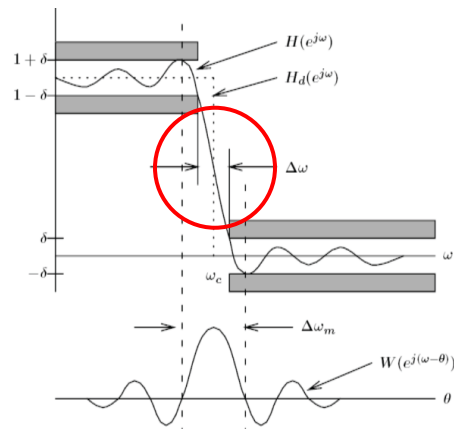


→ Hanning: Less ripples, but wider transition band

Punskaya, Slide 93



Windowed FIR Property 1: Equal transition bandwidth

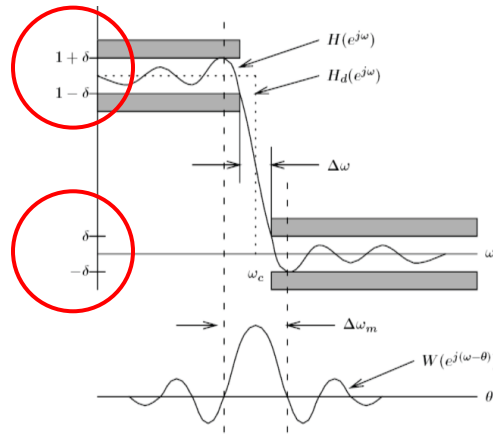


Punskaya, Slide 96

- Equal transition bandwidth on both sides of the ideal cutoff frequency



Windowed FIR Property 2: Peak Errors same in Passband & Stopband

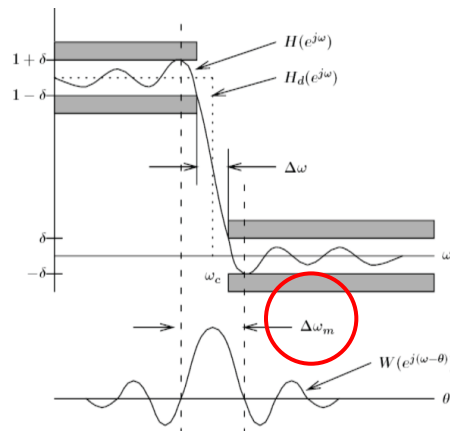


Punskaya, Slide 96

- Peak approximation error in the passband ($1+\delta \rightarrow 1-\delta$) is equal to that in the stopband ($\delta \rightarrow -\delta$)



Windowed FIR Property 3: Mainlobe Width

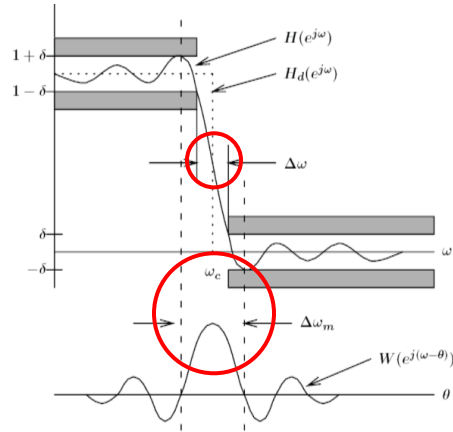


Punskaya, Slide 99

- The distance between approximation error peaks is approximately equal to the width of the mainlobe $\Delta\omega_m$



Windowed FIR Property 4: Mainlobe Width [2]

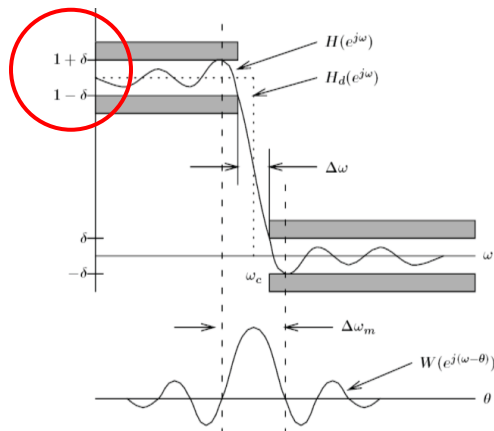


Punskaya, Slide 96

- The width of the mainlobe is wider than the transition bandwidth



Windowed FIR Property 5: Peak $\Delta\delta$ is determined by the window shape

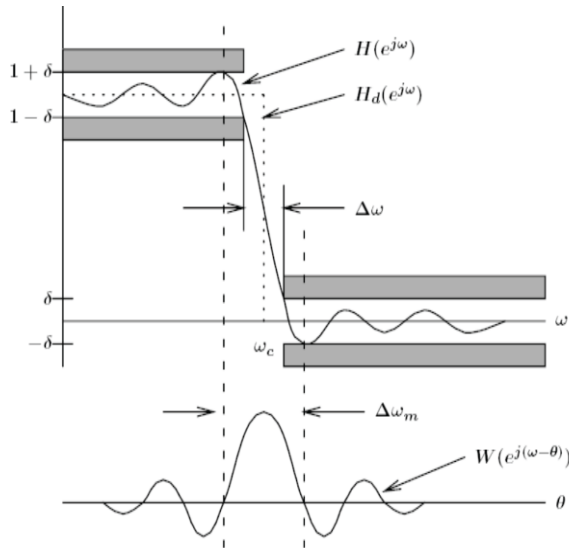


Punskaya, Slide 96

- peak approximation error is determined by the window shape, independent of the filter order



Window Design Method Design Terminology



Where:

- ω_c : cutoff frequency
- δ : maximum passband ripple
- $\Delta\omega$: transition bandwidth
- $\Delta\omega_m$: width of the window mainlobe

Punskaya, Slide 96



Passband / stopband ripples

ω_s and ω_p : Corner Frequencies

Passband / stopband ripples are often expressed in dB:

- passband ripple = $20 \log_{10} (1 + \delta_p)$ dB
- peak-to-peak passband ripple $\cong 20 \log_{10} (1 + 2\delta_p)$ dB
- minimum stopband attenuation = $-20 \log_{10} (\delta_s)$ dB



Passband / stopband ripples

ω_s and ω_p : Corner Frequencies

Passband / stopband ripples are often expressed in dB:

- passband ripple = ~~$20 \log_{10} (1 + \delta_p)$ dB~~ = $20 \log_{10} (\delta_p)$ dB
- peak-to-peak passband ripple \cong ~~$20 \log_{10} (1 + 2\delta_p)$ dB~~
 $\cong 20 \log_{10} (2\delta_p)$ dB
- minimum stopband attenuation = ~~$20 \log_{10} (\delta_s)$ dB~~
 $= 20 \log_{10} (\delta_s)$ dB



Summary of Design Procedure

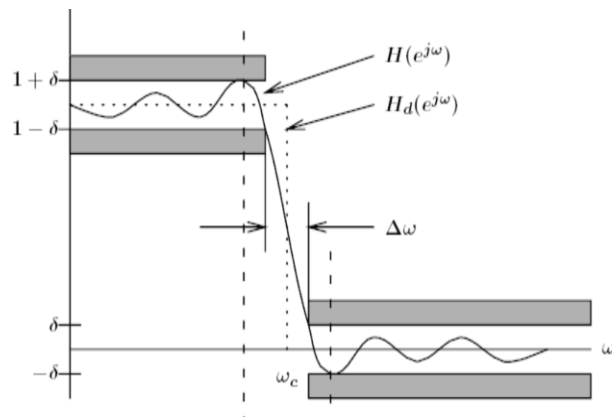
1. Select a suitable window function
2. Specify an ideal response $H_d(\omega)$
3. Compute the coefficients of the ideal filter $h_d(n)$
4. Multiply the ideal coefficients by the window function to give the filter coefficients
5. Evaluate the frequency response of the resulting filter and iterate if necessary (e.g. by increasing M if the specified constraints have not been satisfied).



Windowed Filter Design Example

- Design a type I low-pass filter with:

- $\omega_p = 0.2\pi$
- $\omega_s = 0.3\pi$
- $\delta = 0.01$



Windowed Filter Design Example: Step I: Select a suitable Window Function

No.	Window $w(t)$	Mainlobe Width	Rolloff Rate (dB/oct)	Peak Sidelobe level (dB)	Peak $20\log_{10}\delta$
1	Rectangular: $\text{rect}\left(\frac{t}{T}\right)$	$\frac{4\pi}{T}$	-6	-13.3	-21dB
2	Bartlett: $\Delta\left(\frac{t}{2T}\right)$	$\frac{8\pi}{T}$	-12	-26.5	
3	Hanning: $0.5\left[1 + \cos\left(\frac{2\pi t}{T}\right)\right]$	$\frac{8\pi}{T}$	-18	-31.5	-44dB
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6	Kaiser: $\frac{I_0\left[\alpha\sqrt{1 - 4\left(\frac{t}{T}\right)^2}\right]}{I_0(\alpha)}$ $0 \leq \alpha \leq 10$	$\frac{11.2\pi}{T}$	-6	-59.9 ($\alpha = 8.168$)	

- LP with: $\omega_p = 0.2\pi$, $\omega_s = 0.3\pi$, $\delta = 0.01$

- $\delta = 0.01$: The required peak error spec: $-20\log_{10}(\delta) = -40$ dB

} Hanning Window

- Main-lobe width:

$$\omega_s - \omega_p = 0.3\pi - 0.2\pi = 0.1\pi \rightarrow 0.1\pi = 8\pi / M$$

\rightarrow Filter length $M \geq 80$ & Filter order $N \geq 79$

- BUT, Type-I filters have even order so $N = 80$



Windowed Filter Design Example: Step 2: Specify the Ideal Response

- From Property 1 (Midpoint rule)

$$\rightarrow \omega_c = (\omega_s + \omega_p)/2 = (0.2\pi + 0.3\pi)/2 = 0.25\pi$$

∴ An ideal response will be:

$$H_d(\omega) = \begin{cases} 1 & \text{if } |\omega| \leq 0.25\pi \\ 0 & \text{if } 0.25\pi < |\omega| < \pi \end{cases}$$



Windowed Filter Design Example: Step 3: Compute the coefficients of the ideal filter

- The ideal filter coefficients h_d are given by the Inverse **Discrete time** Fourier transform of $H_d(\omega)$

$$\begin{aligned} x(n) &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\omega) e^{j\omega n} d\omega = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} e^{j\omega n} d\omega \\ &= \frac{\omega_c \sin \omega_c n}{\pi \omega_c n}. \end{aligned}$$

+ Delayed impulse response (to make it causal)

$$\tilde{h}(n) = \hat{h}\left(n - \frac{N-1}{2}\right)$$

→ Coefficients of the ideal filter (via equation or IFFT):

$$h(n) = \frac{\sin(0.25\pi(n-40))}{\pi(n-40)}$$



Windowed Filter Design Example:
Step 4: Multiply to obtain the filter coefficients

$$\rightarrow h(n) = \frac{\sin(0.25\pi(n-40))}{\pi(n-40)}$$

- Multiply by a Hamming window function for the passband:

$$w(n) = 0.54 - 0.46 \cos\left(\frac{2\pi n}{M}\right)$$



Windowed Filter Design Example:
Step 5: Evaluate the Frequency Response and Iterate

- The frequency response is computed as the DFT of the filter coefficient vector
- **If** the resulting filter does not meet the specifications, **then**:
 - Adjust the ideal filter frequency response (for example, move the band edge) and repeat (step 2)
 - Adjust the filter length and repeat (step 4)
 - change the window (& filter length) (step 4)
- And/Or consult with Matlab:
 - **FIR1** and **FIR2**
 - **B=FIR2(N,F,M)** : Designs a Nth order FIR digital filter with



Windowed Filter Design Example: Consulting Matlab:

- **FIR1** and **FIR2**
 - **B=FIR2(N,F,M)** : Designs a Nth order FIR digital filter
 - **F** and **M** specify frequency and magnitude breakpoints for the filter such that **plot(N,F,M)** shows a plot of desired frequency
 - Frequencies **F** must be in increasing order between 0 and $F_s/2$, with F_s corresponding to the sample rate.
 - **B** is the vector of length $N+1$, it is real, has linear phase and symmetric coefficients
 - Default window is Hamming – others can be specified

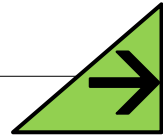


In Conclusion

- FIR Filters are digital (can not be implemented in analog) and exploit the difference and delay operators
- A window based design builds on the notion of a truncation of the “ideal” box-car or rectangular low-pass filter in the Frequency domain (which is a sinc function in the time domain)
- Other Design Methods exist:
 - Least-Square Design
 - Equiripple Design
 - Remez method
 - The Parks-McClellan Remez algorithm
 - Optimisation routines ...



Next Time...



- **Digital Filters**
- **Review:**
 - Chapter 12 of Lathi
- **Ponder?**

$$y[k] = f[k] * h[k] \qquad Y(\Omega) = F(\Omega)H(\Omega)$$

where $F(\Omega)$, $Y(\Omega)$, and $H(\Omega)$ are DTFTs of $f[k]$, $y[k]$, and $h[k]$, respectively; that is,

$$f[k] \Leftrightarrow F(\Omega), \quad y[k] \Leftrightarrow Y(\Omega), \quad \text{and} \quad h[k] \Leftrightarrow H(\Omega)$$

