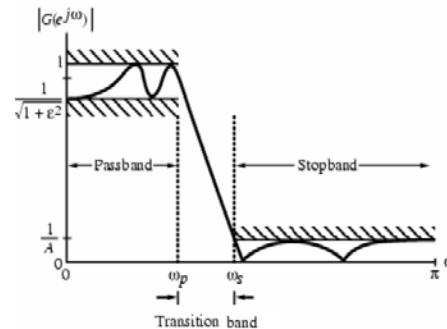
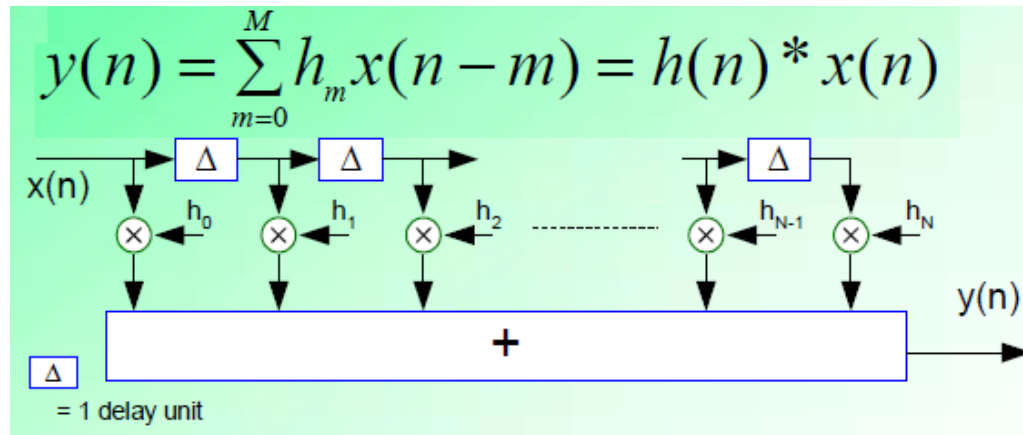


Specifications of Filters





Digital Filter Design

Objective: Determining a **realizable** transfer function $G(z)$ **approximating** a given frequency response specification.

- Generally classified into categories of finite impulse response (FIR) and infinite impulse response (IIR).
- $G(z)$ of IIR filter should be a *stable* and *real rational* function of two polynomials of z .
- $G(z)$ of FIR filter is a polynomial and is always stable, why?
- Digital filter design is the process of deriving the transfer function $G(z)$ to **approximate** the intended specifications.
- Either the *magnitude* and/or the *phase (delay)* response of the digital filter are specified for most applications
- The problem of interest is the development of a ***realizable approximation*** to a given magnitude response specification
- We mainly discuss the **magnitude approximation** of **FIR** filter

Digital Filter Design – Filter Specifications

- The magnitude response specifications of a digital filter in the passband and the stopband are given with some acceptable tolerances
- In addition, a *transitional* band is specified between the passband and stopband
- For example, the magnitude response $G(e^{j\omega})$ of a digital *lowpass* filter may be given as indicated in the Figure
- Similarly, other types of filters also have the transitional band.

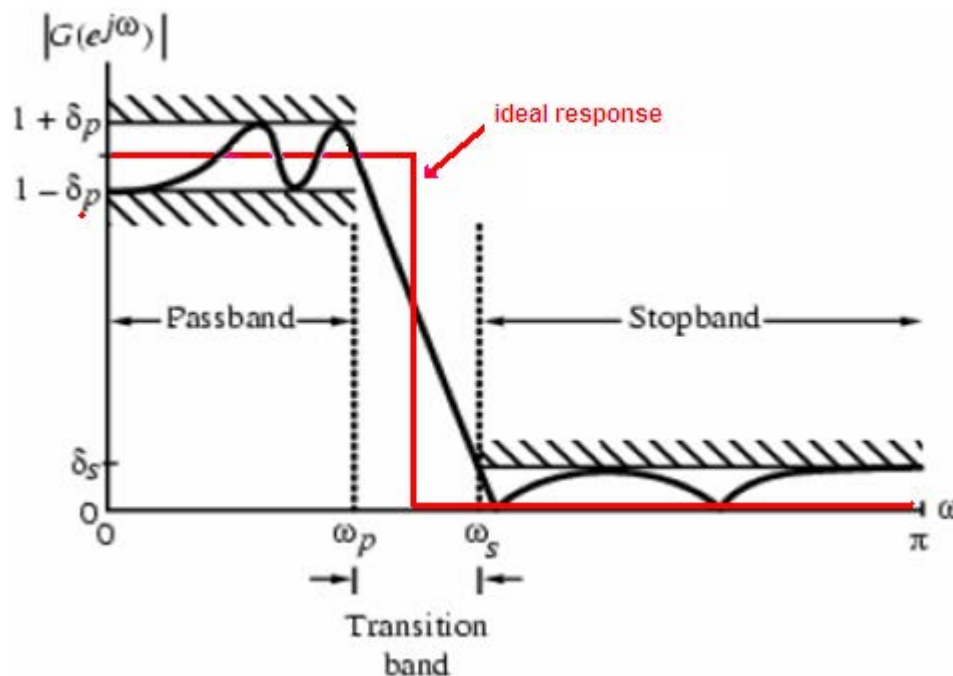


Figure 31 Filter specification parameters

Digital Filter Design – Filter Specifications

- The passband is defined by $0 \leq \omega \leq \omega_p$,
 $|G(e^{j\omega})| \approx 1$ with an error $\pm \delta_p$, i.e.,

$$1 - \delta_p \leq |G(e^{j\omega})| \leq 1 + \delta_p, \quad |\omega| \leq \omega_p$$

- The stopband is defined by $\omega_s \leq \omega \leq \pi$,
 $|G(e^{j\omega})| \approx 0$ with an error δ_s , that is

$$|G(e^{j\omega})| \leq \delta_s, \quad \omega_s \leq |\omega| \leq \pi$$

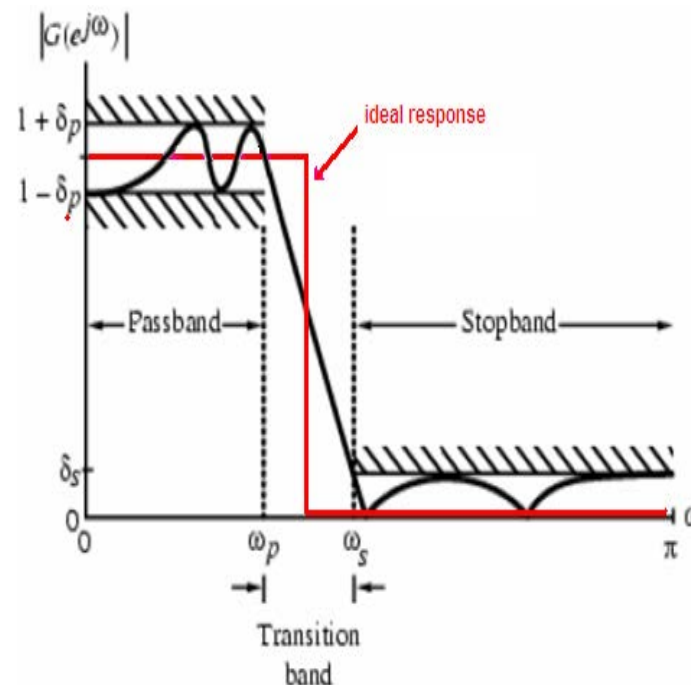
where

ω_p - passband edge frequency;

ω_s - stopband edge frequency,

δ_p - peak ripple in the passband;

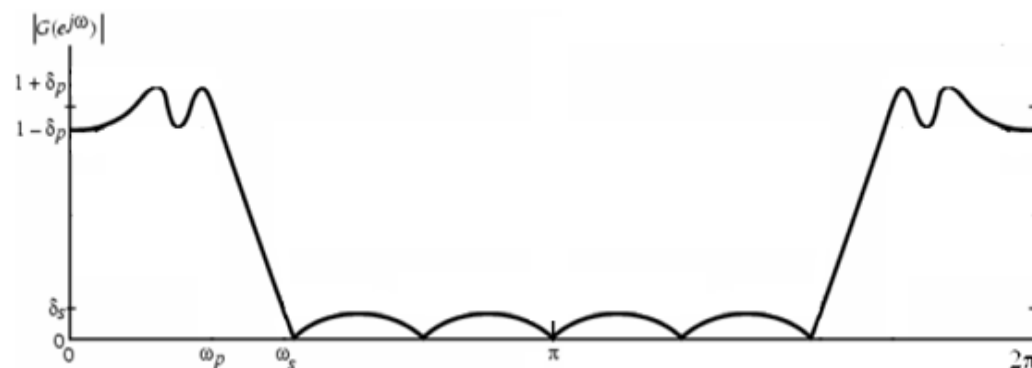
δ_s - peak ripple in the stopband



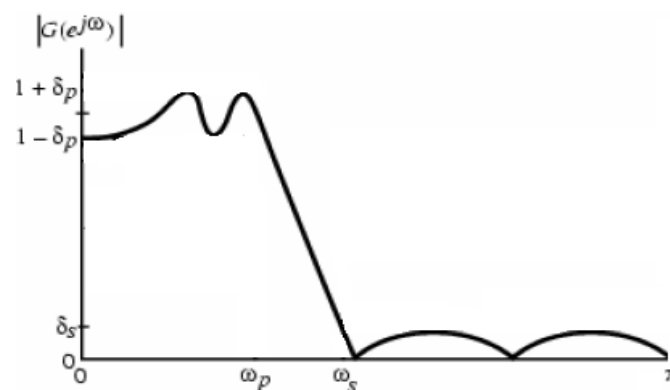
Digital Filter Design – Filter Specifications

■ Example

- $G(e^{j\omega})$ is a *periodic* function of ω , and $|G(e^{j\omega})|$ is the spectrum magnitude of a *real-coefficient* filter and an *even function* of ω .
- As a result, filter specifications are only for the frequency range $0 \leq \omega \leq \pi$



(a) A complete view of filter magnitude response



(b) Simplified view of filter magnitude response if coefficients are real

Figure 32 Representation of filter magnitude response

Digital Filter Design – Filter Specifications

- Specifications are often given in terms of *loss function* $A(\omega) = -20 \log |G(e^{j\omega})|$ in dB
- Peak passband ripple $\alpha_p(\omega) = -20 \log_{10} |1 - 2\delta_p|$ dB
- The minimum stopband attenuation is $\alpha_s(\omega) = -20 \log_{10} \delta_s$ dB
- Magnitude specifications may be alternately given in a *normalized* form as indicated

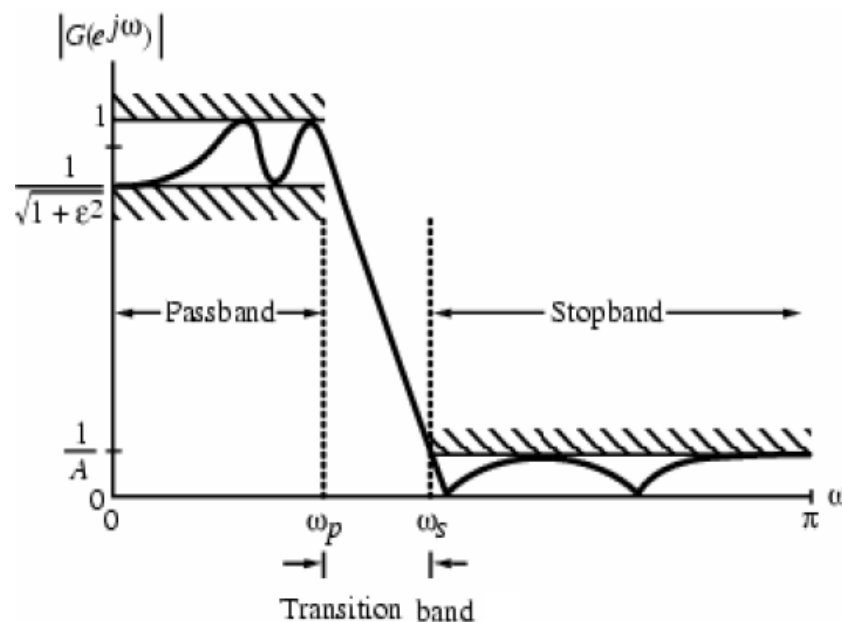


Figure 33 Normalized filter specification

Digital Filter Design – Filter Specifications

- Here, the maximum value of the magnitude in the passband is assigned to be unity (or 0 dB)
- $1/\sqrt{1+\varepsilon^2}$ is the *maximum passband deviation* given by the *minimum* value of the magnitude in the passband. ε^2 is a value larger than 0.
- $1/A$ is the *maximum stopband magnitude*
- For the normalized specification, maximum value of the *gain function* or the minimum value of the *loss function* is 0 dB

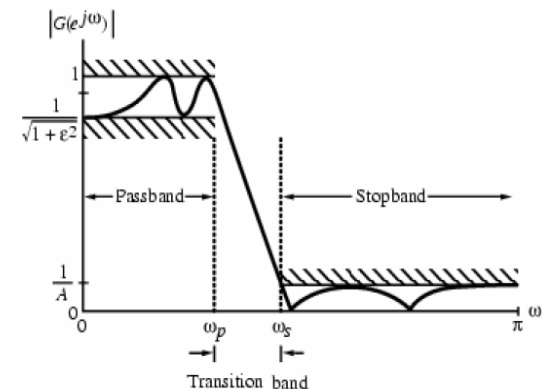
- Maximum passband attenuation*

$$\alpha_{\max} = 20\log_{10}\left(\sqrt{1+\varepsilon^2}\right) \text{ dB}$$

- For $\delta_p \ll 1$, it can be shown that

$$\alpha_{\max} \cong -20\log_{10}(1-2\delta_p) \text{ dB}$$

- In practice, passband edge frequency F_p and stopband edge frequency F_s are specified in Hz





Digital Filter Design – Filter Specifications

- For digital filter design, *normalized* band edge frequencies need to be computed from specifications in *Hz* using

$$\omega_p = \frac{2\pi F_p}{F_T} = 2\pi F_p T$$

$$\omega_s = \frac{2\pi F_s}{F_T} = 2\pi F_s T$$

where T is the sample interval.

- Example :** Let $F_s = 7 \text{ kHz}$, $F_p = 3 \text{ kHz}$ and $F_T = 25 \text{ kHz}$. Then

$$\omega_s = \frac{2\pi(7 \times 10^3)}{25 \times 10^3} = 0.56\pi$$

$$\omega_p = \frac{2\pi(3 \times 10^3)}{25 \times 10^3} = 0.24\pi$$



Digital Filter Design – FIR Filters

- For FIR digital filter design, the transfer function is a polynomial of z^{-1} with real coefficients:

$$H(z) = \sum_{n=0}^N h[n]z^{-n}$$

- To minimize computational complexity, order N of $H(z)$ must be as small as possible
- If a *linear phase* is desired, the filter coefficients must satisfy the constraint:

$$h[n] = \pm h[N - n], n = 0, 1, \dots, N/2-1 \text{ or } (N+1)/2-1.$$

- Advantages of FIR filter :
 - Can be designed with exact linear phase,
 - Filter structure is always stable with quantized coefficients
- Disadvantages of FIR filter :
 - Order of an FIR filter is considerably higher than the order of an equivalent IIR filter meeting the same *amplitude specifications*, and FIR filter has thus *higher* computational complexity (why?)



How Are the Filter Coefficients Calculated to Meet the Specifications?

Windowed Fourier Series Approach

Digital Filter Design – Basic Approach

- FIR filter design is based on a direct approximation of the specified *magnitude response*, with the added requirement of *linear phase*.
- The design of an FIR filter of order N may be accomplished by finding either the length- $(N+1)$ impulse response samples $\{h[n]\}$ or the $(N+1)$ samples of its frequency response $H(e^{j\omega})$
- Three commonly used approaches to FIR filter design
 - Windowed Fourier series approach
 - Frequency sampling approach
 - Computer-based optimization methods
- We next describe the design approach based on windowed Fourier series approach.



Design of FIR Digital Filters – Basic Approach

- FIR *filter order* can be estimated by the Kaiser's formula as follows:

$$N \cong \frac{-10\log_{10}(\delta_p \delta_s) - 13}{14.6(\omega_s - \omega_p) / 2\pi} \quad \begin{aligned} \delta_p &= 0.5(1 - 10^{-\alpha_p/20}) \\ \delta_s &= 10^{-(\alpha_s/20)} \end{aligned}$$

- **Example:** Estimate the order of a linear-phase lowpass FIR filter with the specifications: passband edge $F_p = 1.8$ kHz, stopband edge $F_s = 2$ kHz, peak passband ripple $\alpha_p = 0.1$ dB, minimum stopband attenuation $\alpha_s = 35$ dB, and sampling rate $F_T = 12$ kHz.
- We get $\delta_p = 0.005723$ and $\delta_s = 0.01778279$. Substituting these values and band edge frequencies in the above equation, we get

$$N \cong \frac{-10\log_{10}(0.00010177) - 13}{14.6(2000 - 1800) / 12,000} = 110.78 \cong 111$$

- Other estimation method for the filter order such as Hermann's formula also exists.

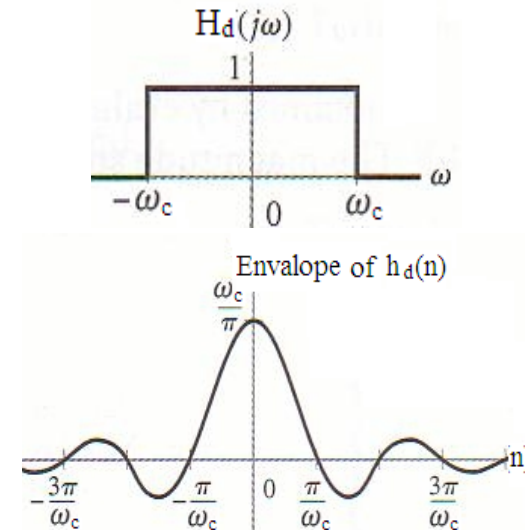
Design of FIR Digital Filters – LSE Approach

- Let $H_d(e^{j\omega})$ be the desired frequency response, and is a periodic function of ω with a period 2π , expressed as a Fourier series

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

where

$$h_d(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} H_d(e^{j\omega}) e^{j\omega n} d\omega, \\ -\infty < n < \infty$$



- In this case, $\{h_d[n]\}$ is of *infinite* length and *non-causal*
- Objective:** Find a *finite-duration* $\{h_t[n]\}$ of length $N+1$ whose DTFT $H_t(e^{j\omega})$ approximates the desired DTFT $H_d(e^{j\omega})$ in *some sense*.

Design of FIR Digital Filters – LSE Approach

- One commonly used approximation criterion is the *integral-squared error*

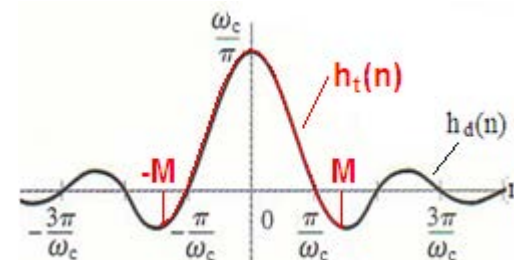
$$\Phi = \frac{1}{2\pi} \int_{-\pi}^{\pi} |H_t(e^{j\omega}) - H_d(e^{j\omega})|^2 d\omega$$

where

$$H_t(e^{j\omega}) = \sum_{n=-M}^M h_t[n] e^{-j\omega n}$$

- Using Parseval's relation we can write

$$\begin{aligned} \Phi &= \sum_{n=-\infty}^{\infty} |h_t[n] - h_d[n]|^2 \\ &= \sum_{n=-M}^M |h_t[n] - h_d[n]|^2 + \sum_{n=-\infty}^{-(M+1)} h_d^2[n] + \sum_{n=M+1}^{\infty} h_d^2[n] \end{aligned}$$



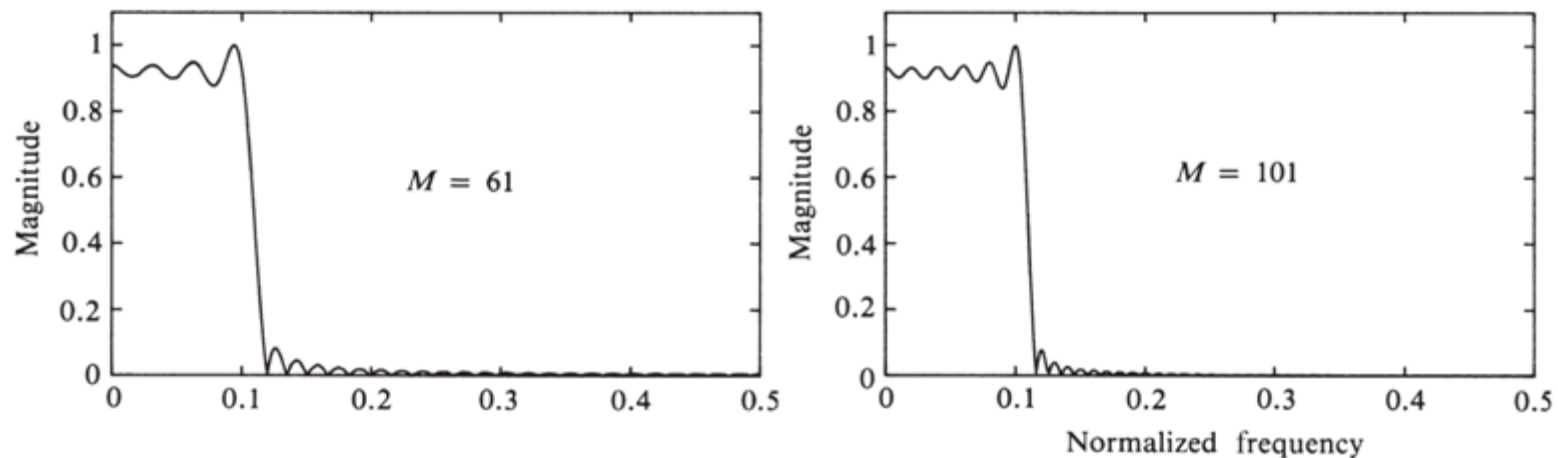
- It follows from above that Φ is minimum when

$$h_t[n] = h_d[n] \quad \text{for } -M \leq n \leq M$$

- Best finite-length approximation to ideal infinite-length impulse response in the *mean-square sense* is obtained by *truncation*

Design of FIR Digital Filters – Gibbs Phenomenon

- Gibbs phenomenon - Oscillatory behavior in the magnitude responses of causal FIR filters obtained by truncating the impulse response coefficients of ideal filters



Lowpass filter designed with a rectangular window: (a) $M = 61$ and (b) $M = 101$.

- The number of ripples in both passband and stopband increases as the length of the lowpass filter is increased, with a corresponding decrease in the ripple widths



Design of FIR Digital Filters – Gibbs Phenomenon

- Height of the largest ripples remain the same, i.e., being independent of filter length
- Similar oscillatory behavior observed in the magnitude responses of the truncated versions of other types (such as high pass, bandpass and bandstop) of ideal filters
- Gibbs phenomenon can be explained by treating the truncation operation as a windowing operation with $w[n]$:

$$h_t[n] = h_d[n] \cdot w[n]$$

- In the frequency domain, we have the convolution

$$H_t(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} H_d(e^{j\varphi}) W(e^{j(\omega-\varphi)}) d\varphi$$

where $H_t(e^{j\omega})$ and $W(e^{j\omega})$ are the DTFTs of $h_t[n]$ and $w[n]$, respectively

Design of FIR Digital Filters – Gibbs Phenomenon

- Thus, $H_t(e^{j\omega})$ is obtained from a continuous convolution of $H_d(e^{j\omega})$ and $W(e^{j\omega})$

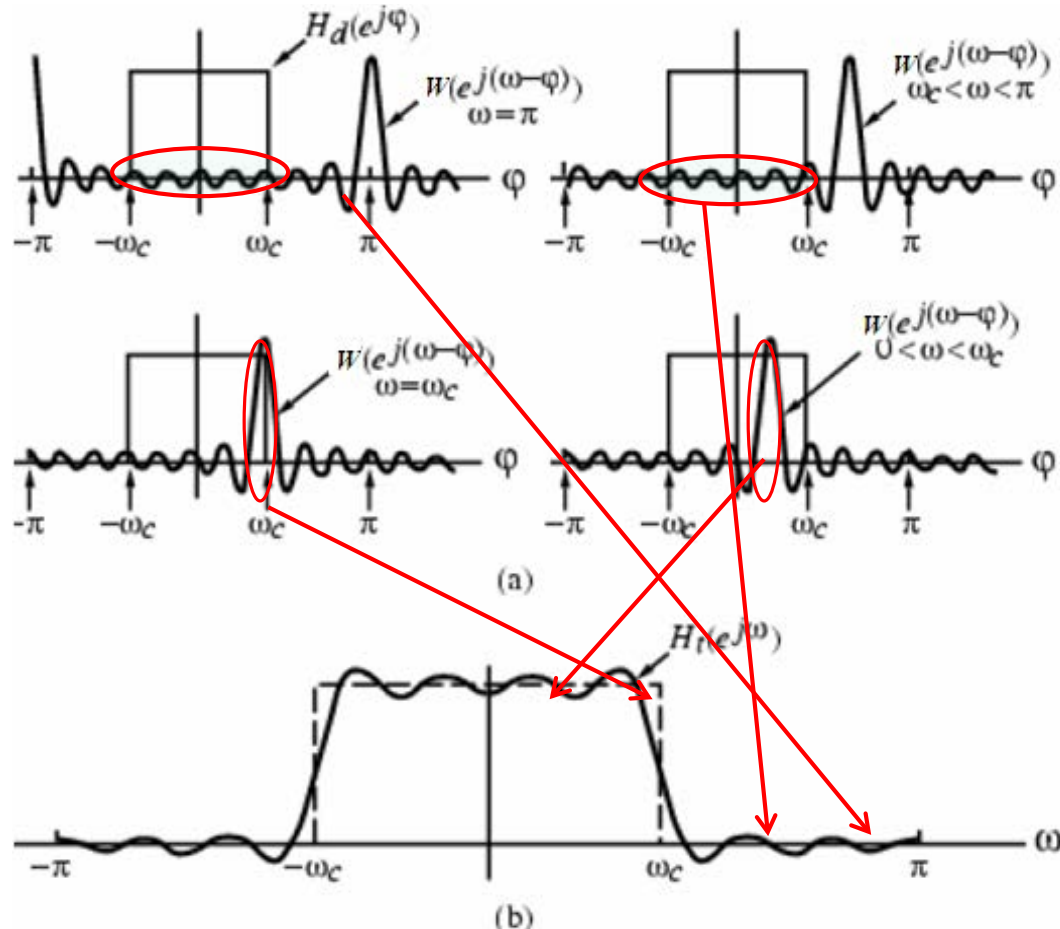


Figure 35 Window based method for generating filter coefficients



Design of FIR Digital Filters – Gibbs Phenomenon

- If $W(e^{j\omega})$ is a very narrow pulse centered $\omega = 0$ (ideally a delta function) compared to variations in $H_d(e^{j\omega})$, then $H_t(e^{j\omega})$ will approximate $H_d(e^{j\omega})$ very closely, which means length $N=2M+1$ of $w[n]$ should be very large.
- On the other hand, length $2M+1$ of $h_t[n]$ should be *as small as possible* to reduce computational complexity
- Gibbs phenomenon can be reduced either:
 - using a window that tapers smoothly to zero at each end, or
 - providing a smooth transition from passband to stopband in the magnitude specifications



Design of FIR Digital Filters – Fixed Window Functions

- Using a tapered window causes the height of the sidelobes to diminish, with a corresponding increase in the main lobe width resulting in a *wider transition* at the discontinuity



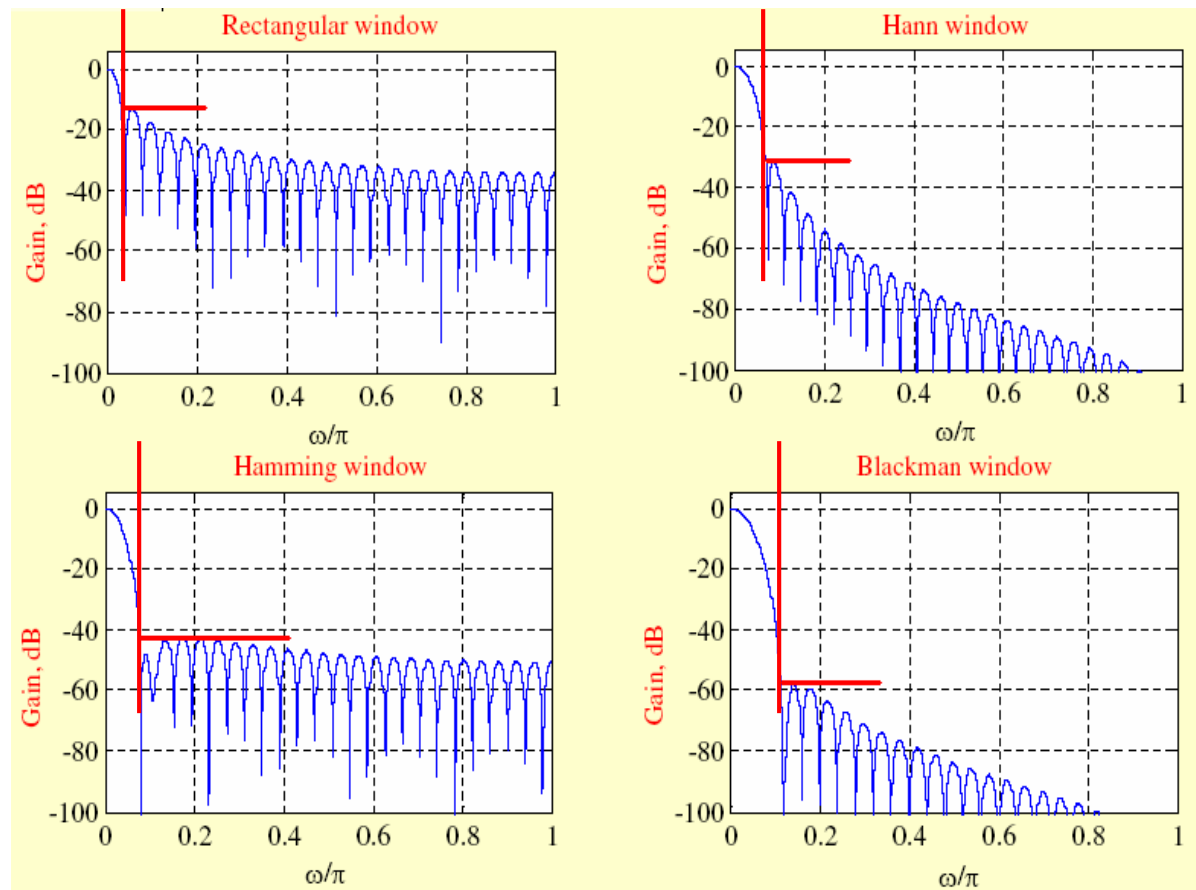
- Illustration of windows

Assuming the window length $N = 2M+1$, we have

- Hann $w[n] = 0.5 + 0.5 \cos\left(\frac{2\pi n}{2M+1}\right), \quad -M \leq n \leq M$
- Hamming: $w[n] = 0.54 + 0.46 \cos\left(\frac{2\pi n}{2M+1}\right), \quad -M \leq n \leq M$
- Blackman $w[n] = 0.42 + 0.5 \cos\left(\frac{2\pi n}{2M+1}\right) + 0.08 \cos\left(\frac{4\pi n}{2M+1}\right),$

Design of FIR Digital Filters – Fixed Window Functions

- Plots of magnitudes of the DTFTs of these windows for $M = 25$ are shown below:

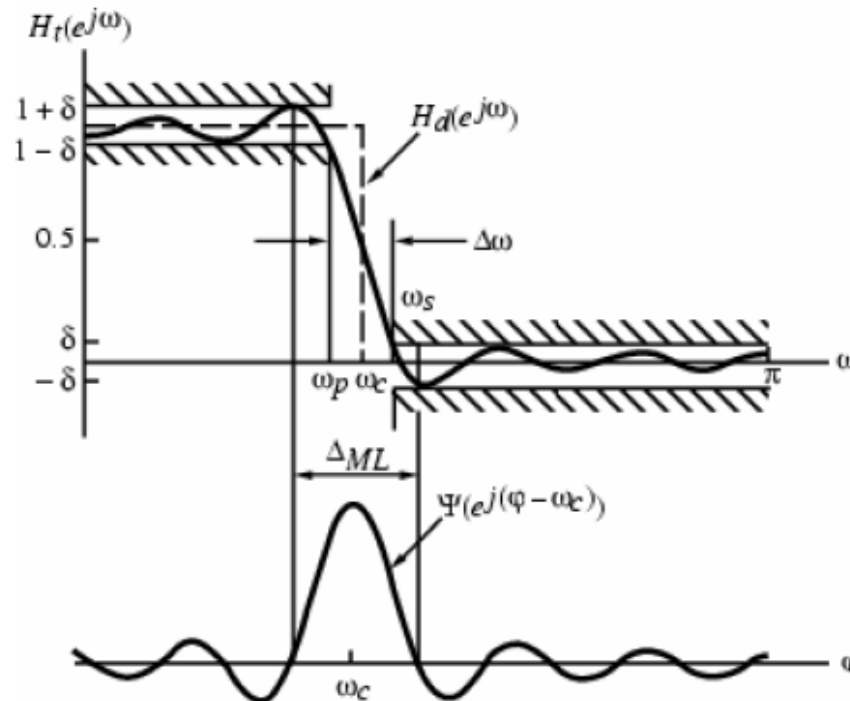


- As the side lobes reduced, the width of the main lobe increases

Design of FIR Digital Filters – Fixed Window Functions

- Magnitude spectrum of each window characterized by a main lobe centered at $\omega = 0$ followed by a series of sidelobes with decreasing amplitudes
- Parameters predicting the performance of a window in filter design are
 - Main lobe width
 - Relative sidelobe level

Design of FIR Digital Filters – Fixed Window Functions

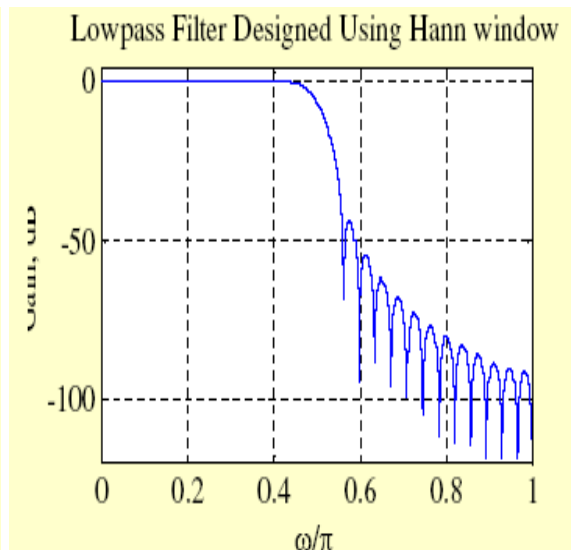
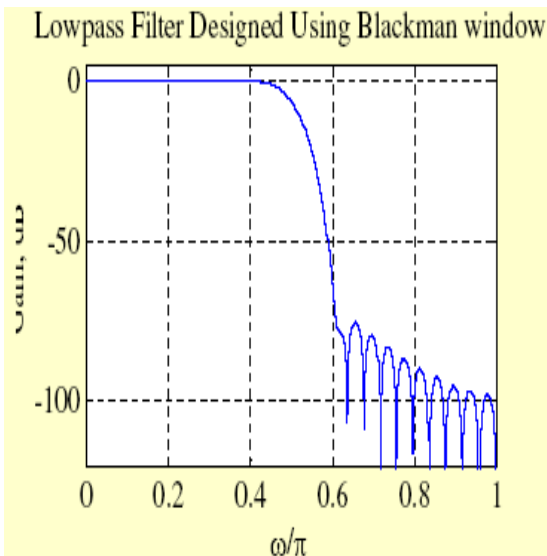
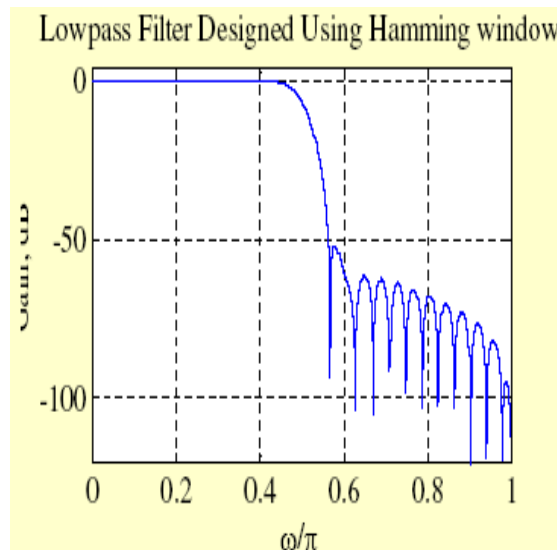


- Observe
$$H_t(e^{j(\omega_c + \Delta\omega)}) + H_t(e^{j(\omega_c - \Delta\omega)}) \cong 1$$
- Thus
$$H_t(e^{j\omega_c}) \cong 0.5$$
- Ripples in passband and stopband are about the same.

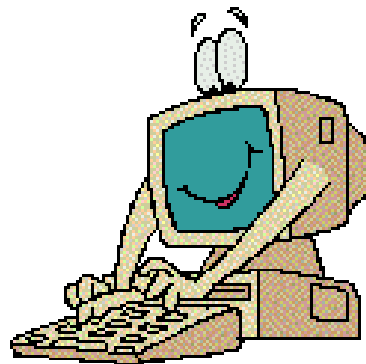
Design of FIR Digital Filters – Design Examples

Design steps:

1. Set $\omega_c = (\omega_p + \omega_s)/2$
2. Choose window based on specified α_s
3. Estimating M using $\Delta\omega \approx \frac{c}{M}$ where c is a constant value for a given window
4. Lowpass filter of length 51 and $\omega_c = \pi/2$ (See the Figures).



Optimization of Filter Design

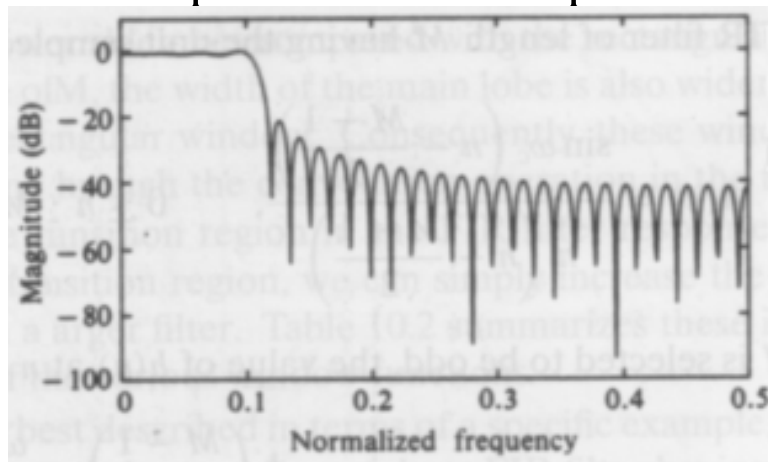


Computer-Aided Design of Digital Filters

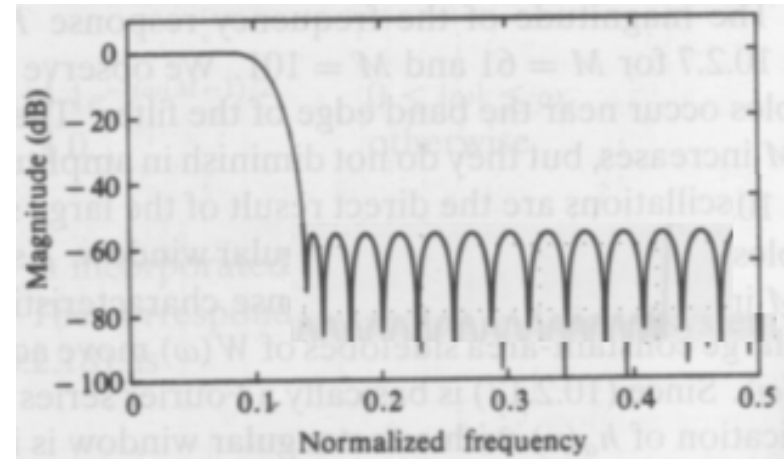
- The FIR filter design techniques discussed so far can be easily implemented with computer programs
- In addition, there are a number of FIR filter design algorithms that rely on some optimization techniques to minimize the errors between the desired frequency response and that of the computer-generated filter
- Let $H(e^{j\omega})$ denote the frequency response of the digital filter $H(z)$ approximating the desired frequency response $D(e^{j\omega})$, given as a piecewise linear function of ω , in some sense
- **Objective** - Determine iteratively the coefficients of $H(z)$ so that the difference between $H(e^{j\omega})$ and $D(e^{j\omega})$ over closed subintervals of $0 < \omega < \pi$ is minimized

Computer-Aided Design of Digital Filters

- One obvious way of optimizing the filter design is to achieve equal ripples so that the side lobe magnitude can be significantly minimized.
- The figures below shows the comparison between the widow-based and the optimized filter response.



Rectangular window



Equal ripple

- Obviously, the optimized one has much lower sidelobe and wider transitional band.

Computer-Aided Design of Digital Filters

- This algorithm is based on iteratively adjusting the coefficients of $\hat{H}(\omega)$ until the peak absolute value of $E(\omega)$ is minimized
- The Parks-McClellan Algorithm or (equi-ripples) widely used for various types of FIR filters.



Example: The LPF specification are

passband edge $f_p = 800$ Hz, stopband edge $f_s = 1000$ Hz,

passband ripple $R_p = 0.5$ dB, stop band attenuation $A_s = 40$ dB,

sampling frequency $F_s = 4000$ Hz,

Transitional bandwidth $\Delta f = 200$ Hz.

Find the coefficients with the Parks-McClellan algorithm.

Summary of Review Materials



- Understanding the expressions of filter specifications in different ways and being able to use these expressions
- Understanding the effects of different windows on the characteristics of the designed filters
- Being able to estimate the filter order, or the computational complexity, from a set of given filter specifications.
- Generating filter coefficients is not required.
- Main steps of filter design
- For specifying other type of FIR filters such as highpass and bandpass, the only difference is the passband and stopband frequencies. Other parameters are the same.