

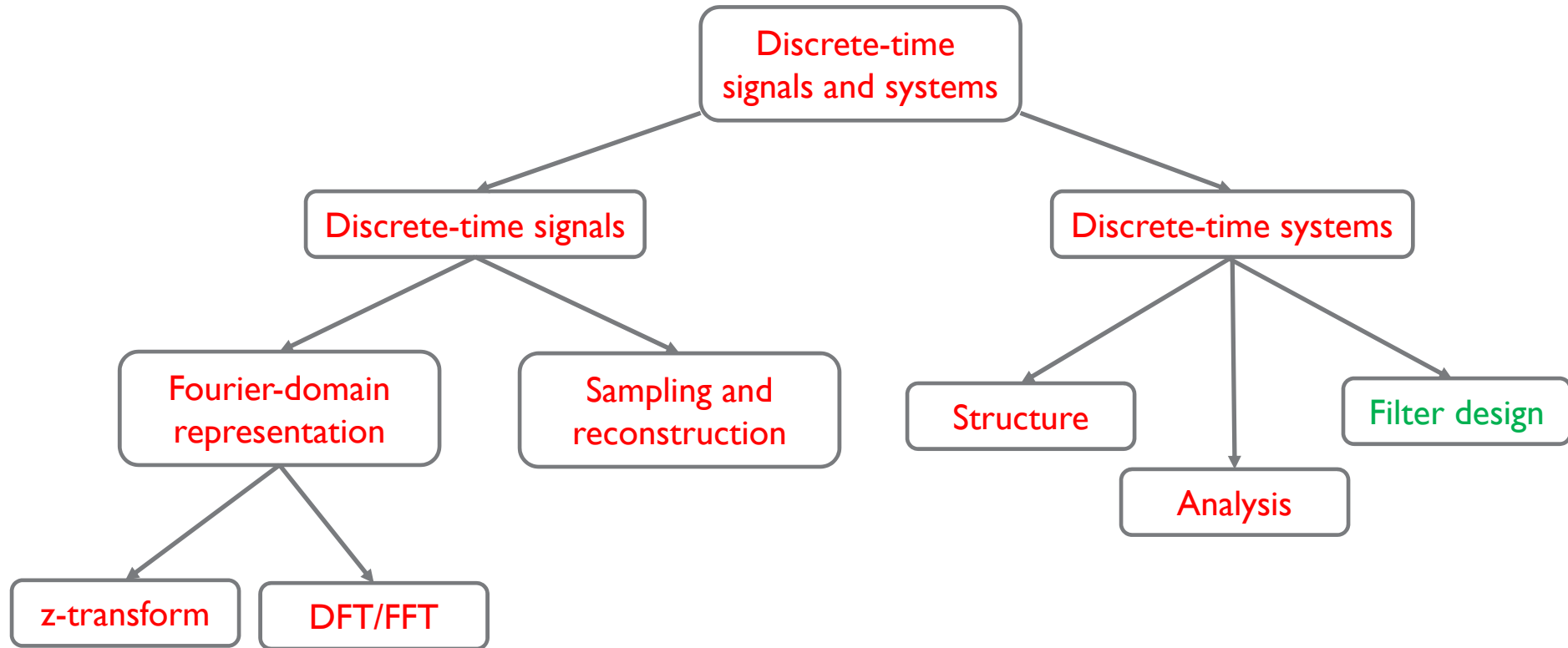
Digital Signal Processing

POSTECH

Department of Electrical Engineering

Junil Choi

Course at glance

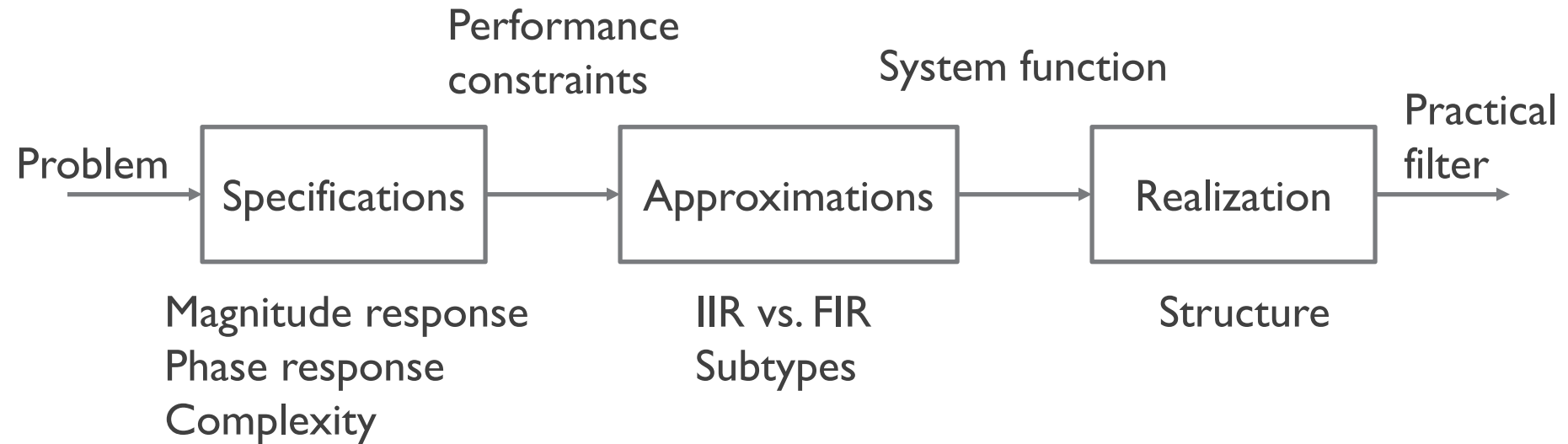


Definition of filter

- ◆ Filter, in broader sense, covers any system
 - ✦ Distortion environments are also filters
- ◆ We denote filters as controllable systems here

Filter design process

◆ Three design steps



◆ Focus on lowpass filters

★ Can be generalized to other frequency-selective filters

Example specifications

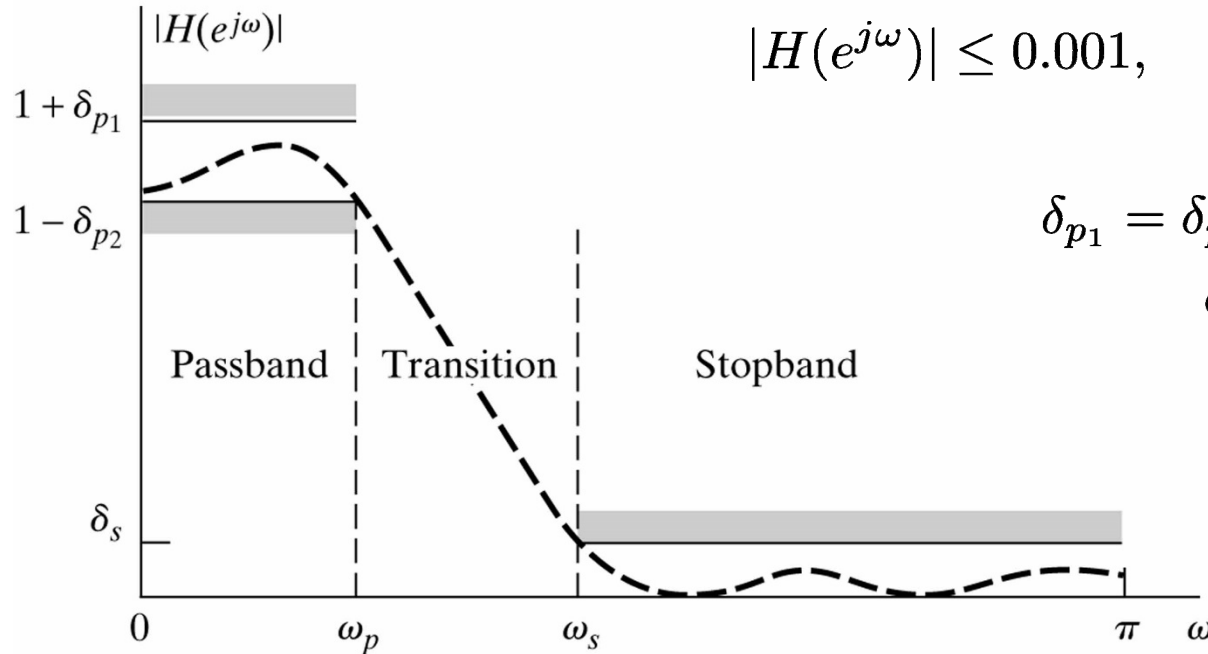
- ◆ Specifications for a discrete-time lowpass filter

$$1 - 0.01 \leq |H(e^{j\omega})| \leq 1 + 0.01, \quad 0 \leq \omega \leq \omega_p$$

$$|H(e^{j\omega})| \leq 0.001, \quad \omega \geq \omega_s$$

$$\delta_{p1} = \delta_{p2} = 0.01$$

$$\delta_s = 0.001$$



Specifications of frequency response

- ◆ Typical lowpass filter specifications in terms of tolerable
 - ✦ Passband distortion → as **smallest** as possible
 - ✦ Stopband attenuation → as **greatest** as possible
 - ✦ Width of transition band → as **narrowest** as possible
- ◆ Improving one often worsens others → tradeoff exists
- ◆ Increasing filter order may improve all → increase complexity

Design a filter

- ◆ Design goal
 - ➔ Find system function to make frequency response meet the specifications (tolerances)

- ◆ Infinite impulse response (IIR) filter
 - ★ Poles inside unit circle due to causality and stability
 - ★ Rational function approximation

- ◆ Finite impulse response (FIR) filter
 - ★ For filters with linear phase requirement
 - ★ Polynomial approximation

Example of IIR filter design

- ◆ For rational (and stable and causal) system function

$$H(z) = \frac{\sum_{k=0}^M b_k z^{-k}}{1 - \sum_{k=1}^N a_k z^{-k}}$$

find the system coefficients such that the corresponding frequency response

$$H(e^{j\omega}) = H(z) \big|_{z=e^{j\omega}}$$

provides a good approximation to a desired response

$$H(e^{j\omega}) \approx H_{\text{desired}}(e^{j\omega})$$

IIR vs. FIR

- ◆ Either FIR or IIR is often dependent on the phase requirements
- ◆ Only FIR filter can be at the same time stable, causal and GLP
- ◆ Design principle
 - ✦ If GLP is required → FIR
 - ✦ If not → IIR preferable because IIR can meet specifications with lower complexity

IIR vs. FIR

◆ IIR

- ✦ Rational system function
- ✦ Poles and zeros
- ✦ Stable/unstable
- ✦ Hard to control phase
- ✦ Low order (4-20)
- ✦ Designed on the basis of analog filter

◆ FIR

- ✦ Polynomial system function
- ✦ Only zeros
- ✦ Always stable
- ✦ Easy to get (generalized) linear phase
- ✦ High order (20-200)
- ✦ Usually unrelated to analog filter designs

IIR Filter Design

Discrete-time IIR filters from continuous-time filters

- ◆ Continuous-time (or analog) IIR filter design is highly advanced
 - ✦ Relatively simple closed-form design possible

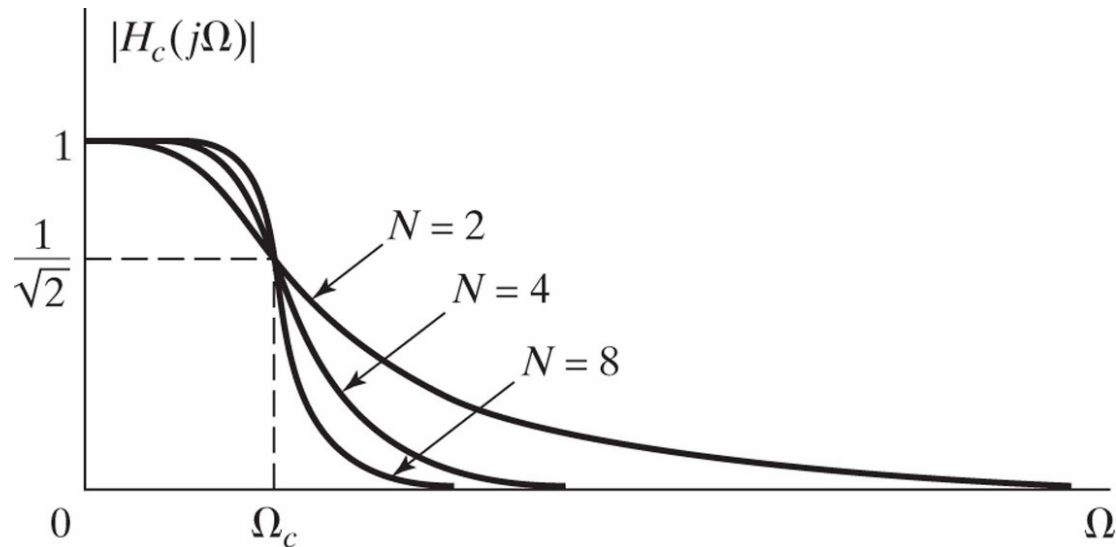
- ◆ Discrete-time IIR filter design
 - ✦ Filter specifications for discrete-time filter
 - ✦ Convert to continuous-time specifications
 - ✦ Design continuous-time filter
 - ✦ Convert to discrete-time filter
 - Impulse invariance method
 - Bilinear transformation method

Analog filter designs

- ◆ Butterworth filter
- ◆ Type I Chebyshev filter
- ◆ Type II Chebyshev filter
- ◆ Elliptic filter

Butterworth lowpass filter

- ◆ Filter form $|H_c(j\Omega)|^2 = \frac{1}{1 + (\Omega/\Omega_c)^{2N}}$
 - ★ Two parameters
 - Order N
 - Cutoff frequency Ω_c
 - ★ Monotonic in both passband and stopband



Type I Chebyshev lowpass filter

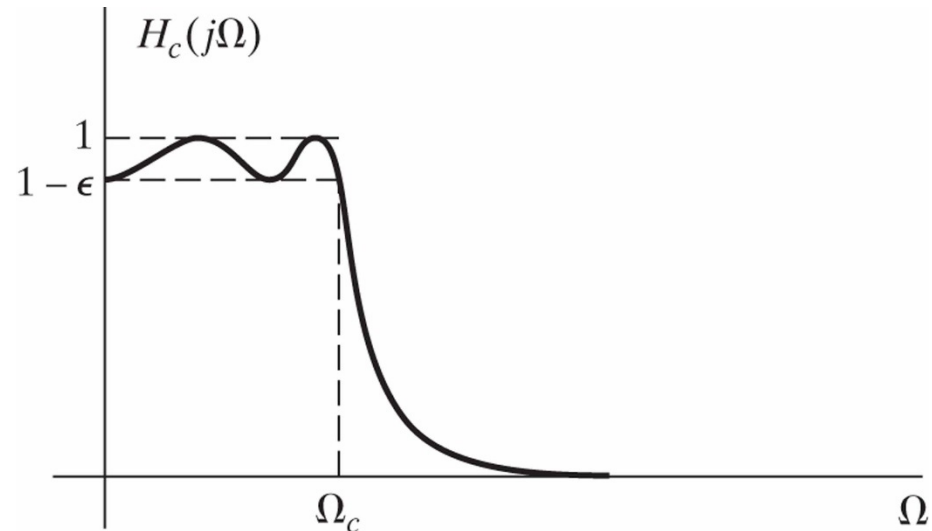
◆ Filter form $|H_c(j\Omega)|^2 = \frac{1}{1 + \epsilon^2 V_N^2(\Omega/\Omega_c)}$

where $V_N(x) = \cos(N \cos^{-1} x)$

★ Three parameters

- Order N
- Cutoff frequency Ω_c
- Allowable passband ripple ϵ

◆ $|H_c(j\Omega)|^2$ has equi-ripple error in passband and monotonic in stopband



Type II Chebyshev lowpass filter

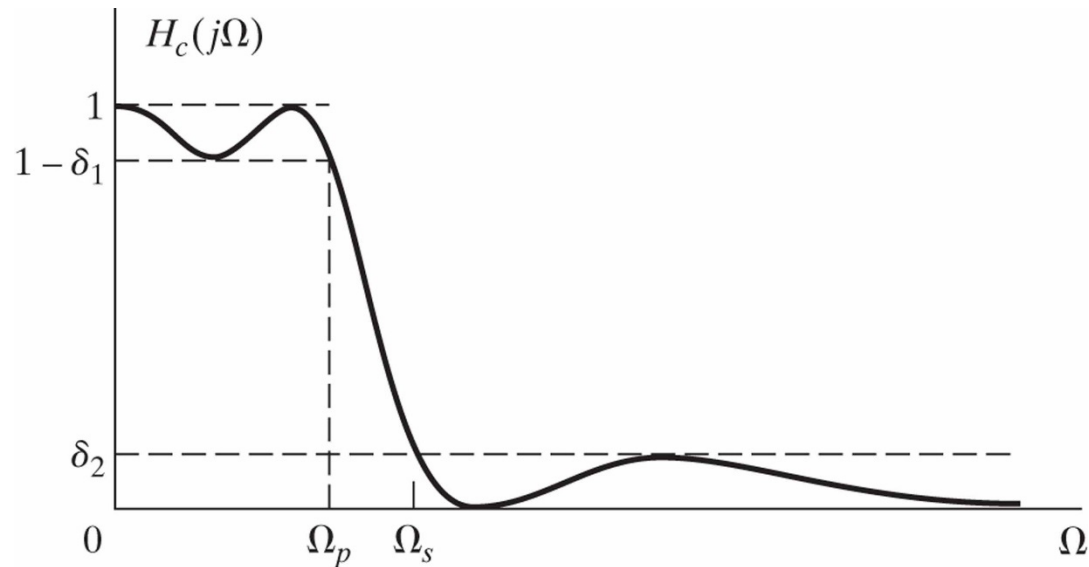
- ◆ Filter form $|H_c(j\Omega)|^2 = \frac{1}{1 + [\epsilon^2 V_N^2(\Omega/\Omega_c)]^{-1}}$
- ◆ Similar to Type I Chebyshev lowpass filter
 - ✦ $|H_c(j\Omega)|^2$ now has equi-ripple error in stopband and flat in passband

Elliptic filter

◆ Filter form $|H_c(j\Omega)|^2 = \frac{1}{1 + \epsilon^2 U_N^2(\Omega)}$

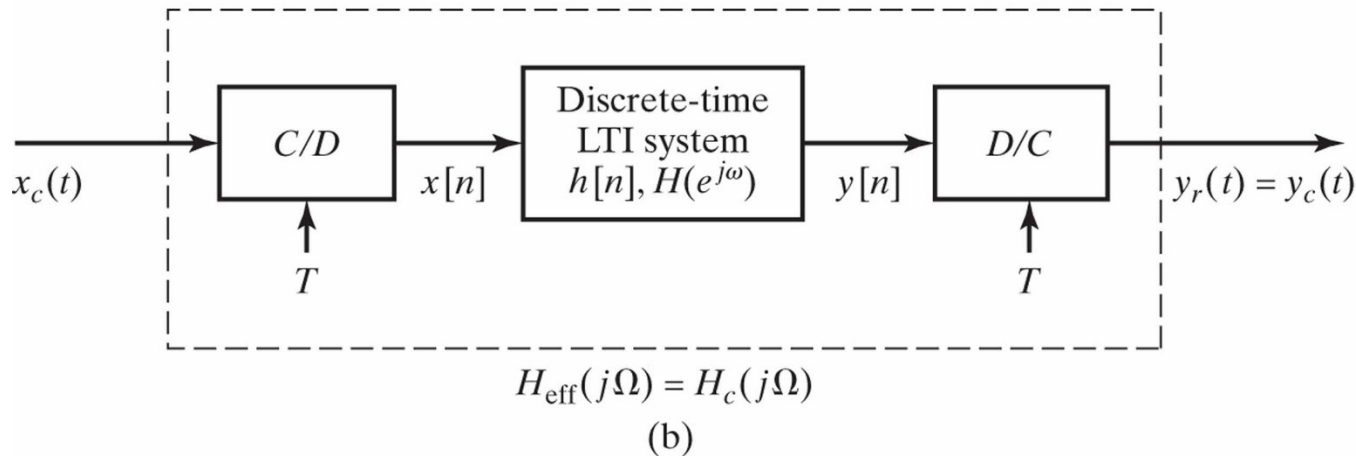
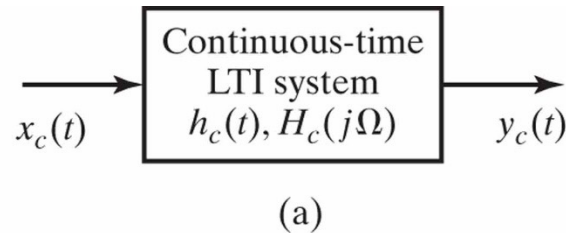
where $U_N(\Omega)$ is a Jacobian elliptic function

◆ $|H_c(j\Omega)|^2$ has equi-ripples in both passband and stopband



Discrete-time IIR filter design – impulse invariance

- ◆ Recall “discrete-time processing of continuous-time signals” in Section 4.4



Output signal

◆ Necessary conditions

- ★ The discrete-time system is LTI
- ★ Continuous-time signal $x_c(t)$ is bandlimited
- ★ Sampling rate Ω_s is at or above the Nyquist rate $2\Omega_N$

◆ If all conditions are satisfied, the output signal becomes

$$Y_r(j\Omega) = H_{\text{eff}}(j\Omega)X_c(j\Omega)$$

where

$$H_{\text{eff}}(j\Omega) = \begin{cases} H(e^{j\Omega T}), & |\Omega| < \pi/T \\ 0, & |\Omega| \geq \pi/T \end{cases}$$

Cutoff frequency of
ideal lowpass filter

$$\omega = \Omega T$$

Impulse invariance

◆ Recall $H_{\text{eff}}(j\Omega) = \begin{cases} H(e^{j\Omega T}), & |\Omega| < \pi/T \\ 0, & |\Omega| \geq \pi/T \end{cases}$

◆ We want to have $H_{\text{eff}}(j\Omega) = H_c(j\Omega)$

➡ $H(e^{j\omega}) = H_c(j\omega/T), \quad |\omega| < \pi$

◆ In time-domain: $h[n] = Th_c(nT)$

$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right)$$

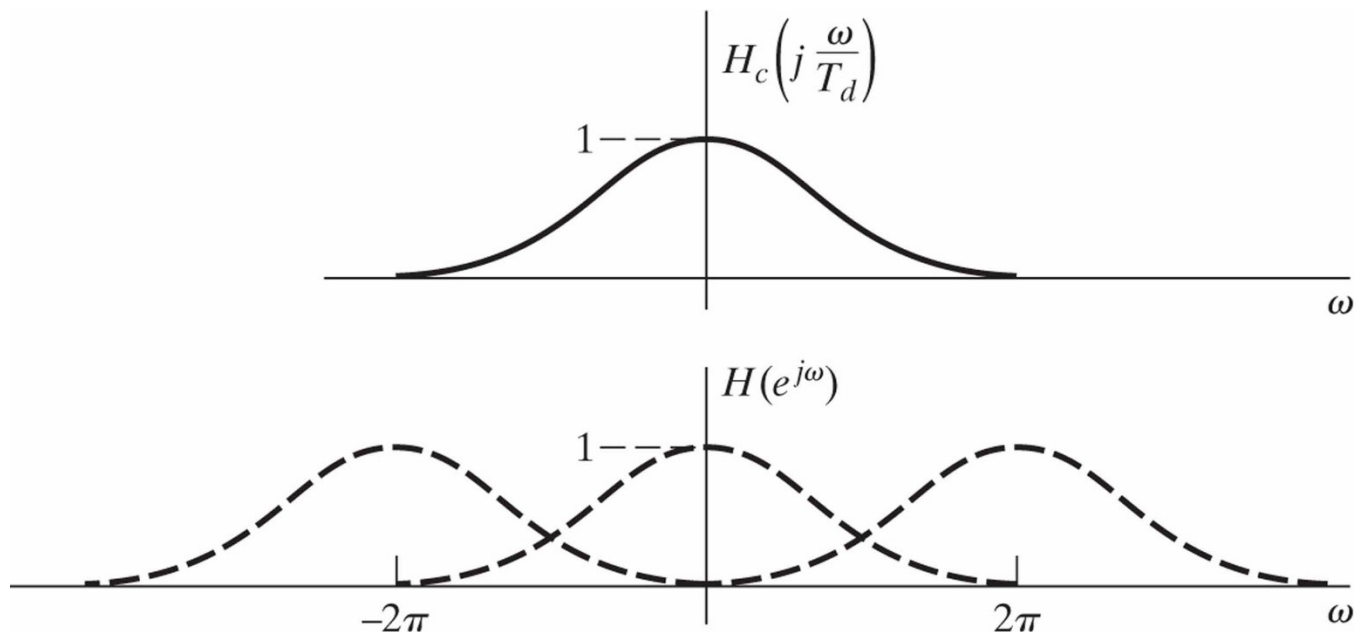
$$\begin{aligned} H(e^{j\omega}) &= T \frac{1}{T} \sum_{k=-\infty}^{\infty} H_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right) \\ &= H_c \left(j \frac{\omega}{T} \right), \quad |\omega| < \pi \end{aligned}$$

Because $H_c(j\Omega) = 0, \quad |\Omega| \geq \pi/T$

Only true when the filter is bandlimited

Impulse invariance - aliasing

- ◆ If the analog filter is not bandlimited (typically the case in practice)
 - ➔ Aliasing occurs in the discrete-time filter
 - ★ Impulse invariance not appropriate for designing highpass filters



How can we avoid the aliasing?

- ◆ Consider higher sampling frequency for analog filter $\Omega_s = 1/T$
- ◆ Will this work? No!
 - ★ Filter specifications given from discrete-time filter requirements
 - ★ The specifications transformed to continuous-time by $\Omega = \omega/T$
 - ★ Continuous-time filter designed by continuous-time specifications
 - ★ Final discrete-time filter obtained by impulse invariance method (sampling)

$$H(e^{j\omega}) = \sum_{k=-\infty}^{\infty} H_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right)$$

➔ Effect of $\Omega_s = 1/T$ cancels out

- ◆ Aliasing can be avoided by overdensing analog filter

Interpretation using system functions

- ◆ Transformation from continuous-time system to discrete-time system is easy to carry out using system functions
- ◆ After partial fraction expansion

$$H_c(s) = \sum_{k=1}^N \frac{A_k}{s - s_k}$$

$$h_c(t) = \begin{cases} \sum_{k=1}^N A_k e^{s_k t}, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

$$h[n] = T_d h_c(nT_d)$$

$$= \sum_{k=1}^N T_d A_k e^{s_k n T_d} u[n]$$

$$= \sum_{k=1}^N T_d A_k (e^{s_k T_d})^n u[n]$$

$$\xleftrightarrow{Z} H(z) = \sum_{k=1}^N \frac{T_d A_k}{1 - e^{s_k T_d} z^{-1}}$$

Interpretation using system functions

- ◆ Mapping from $H_c(s)$ to $H(z)$
 - ★ Pole of $H_c(s)$ at $s = s_k$ maps to pole of $H(z)$ at $z = e^{s_k T_d}$
 - ➔ Stability and causality preserved
 - ★ Continuous-time: $\text{Re}\{s_k\} < 0$
 - ★ Discrete-time: $z = e^{s_k T_d}$ inside the unit circle

Impulse invariance with Butterworth filter

- ◆ Specifications: $0.89125 \leq |H(e^{j\omega})| \leq 1, \quad 0 \leq |\omega| \leq 0.2\pi$
 $|H(e^{j\omega})| \leq 0.17783, \quad 0.3\pi \leq |\omega| \leq \pi$

- ◆ Since the effect $\Omega_s = 1/T$ cancels out, set $T=1$ and $\omega = \Omega$

- ◆ Transformed analog specifications

$$0.89125 \leq |H_c(j\Omega)| \leq 1, \quad 0 \leq |\Omega| \leq 0.2\pi$$
$$|H_c(j\Omega)| \leq 0.17783, \quad 0.3\pi \leq |\Omega| \leq \pi$$

- ◆ Due to monotonicity of Butterworth filter

$$|H_c(j0.2\pi)| \geq 0.89125$$

$$|H_c(j0.3\pi)| \leq 0.17783$$

Impulse invariance with Butterworth filter

- ◆ The magnitude-squared function of Butterworth filter

$$|H_c(j\Omega)|^2 = \frac{1}{1 + (\Omega/\Omega_c)^{2N}}$$

- ◆ From the specifications $|H_c(j0.2\pi)| \geq 0.89125$, $|H_c(j0.3\pi)| \leq 0.17783$

$$1 + \left(\frac{0.2\pi}{\Omega_c}\right)^{2N} = \left(\frac{1}{0.89125}\right)^2, \quad 1 + \left(\frac{0.3\pi}{\Omega_c}\right)^{2N} = \left(\frac{1}{0.17783}\right)^2$$

- ★ Simultaneous solutions are $N = 5.8858$, $\Omega_c = 0.70474$

Should be integer

- ◆ Let $N = 6$ and $\Omega_c = 0.7032$ to exactly meet the passband specifications

- ★ Stopband specification exceeded → margin for aliasing

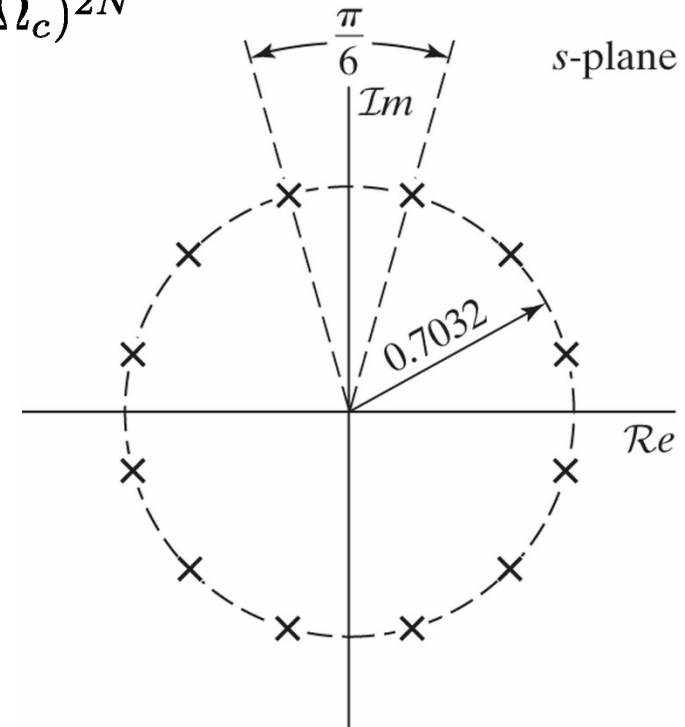
Impulse invariance with Butterworth filter

- ◆ Rewrite the magnitude-squared function

$$H_c(s)H_c(-s) = \frac{1}{1 + (s/j\Omega_c)^{2N}}$$

- ★ The system function has 12 poles

- ◆ To have a stable filter, $H_c(s)$ should have three pole pairs in the left half of s-plane



Impulse invariance with Butterworth filter

- ◆ With three pole pairs

$$H_c(s) = \frac{0.12093}{(s^2 + 0.3640s + 0.4945)(s^2 + 0.9945s + 0.4945)(s^2 + 1.3585s + 0.4945)}$$

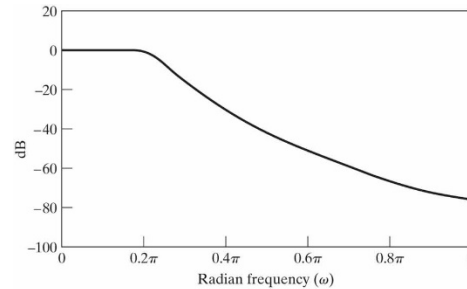
- ◆ After partial fraction, use the transformation

$$H_c(s) = \sum_{k=1}^N \frac{A_k}{s - s_k} \quad \Rightarrow \quad H(z) = \sum_{k=1}^N \frac{T_d A_k}{1 - e^{s_k T_d} z^{-1}}$$

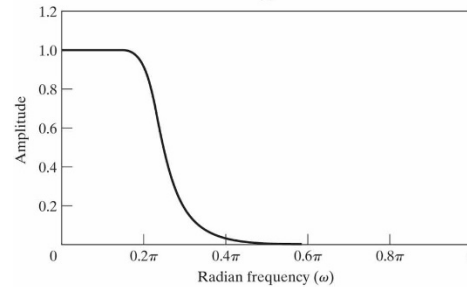
- ◆ Final discrete-time filter

$$H(z) = \frac{0.2871 - 0.4466z^{-1}}{1 - 1.2971z^{-1} + 0.6949z^{-2}} + \frac{-2.1428 + 1.1455z^{-1}}{1 - 1.0691z^{-1} + 0.3699z^{-2}} + \frac{1.8557 - 0.6303z^{-1}}{1 - 0.9972z^{-1} + 0.2570z^{-2}}$$

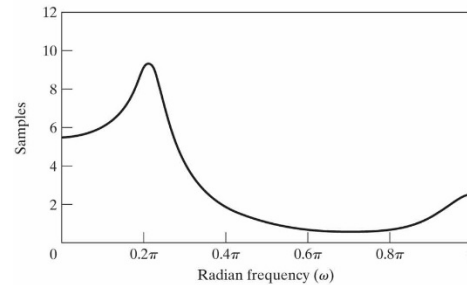
Impulse invariance with Butterworth filter



(a)



(b)



(c)

Discrete-time IIR filter design – bilinear transformation

- ◆ Continuous-time (analog) filter designed using s-plane (Laplace transform)

$$s = \sigma + j\Omega$$

$$H_c(s) = \int_{-\infty}^{\infty} h(t)e^{-st}dt$$

$$H_c(j\Omega) = \int_{-\infty}^{\infty} h(t)e^{-j\Omega t}dt$$

$$z = re^{-j\omega}$$

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

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

- ◆ Mapping between s-plane and z-plane

$$s = \frac{2}{T_d} \left(\frac{1 - z^{-1}}{1 + z^{-1}} \right) \quad \Rightarrow \quad H(z) = H_c \left(\frac{2}{T_d} \left(\frac{1 - z^{-1}}{1 + z^{-1}} \right) \right)$$

Rational behind bilinear transformation

◆ Recall $H_c(s) = \int_{-\infty}^{\infty} h(t)e^{-st}dt$ and $H(z) = \sum_{n=-\infty}^{\infty} h[n]z^{-n}$

$$z = e^{sT}$$

T : numerical integration step size of the trapezoidal rule

$$s = \frac{1}{T} \ln(z)$$

Series based on area hyperbolic tangent function

$$= \frac{2}{T} \left[\frac{z-1}{z+1} + \frac{1}{3} \left(\frac{z-1}{z+1} \right)^3 + \frac{1}{5} \left(\frac{z-1}{z+1} \right)^5 + \dots \right]$$

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

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

Bilinear transformation - concept

◆ Given $s = \sigma + j\Omega$

$$z = \frac{1 + (T_d/2)s}{1 - (T_d/2)s} = \frac{1 + \sigma T_d/2 + j\Omega T_d/2}{1 - \sigma T_d/2 - j\Omega T_d/2}$$

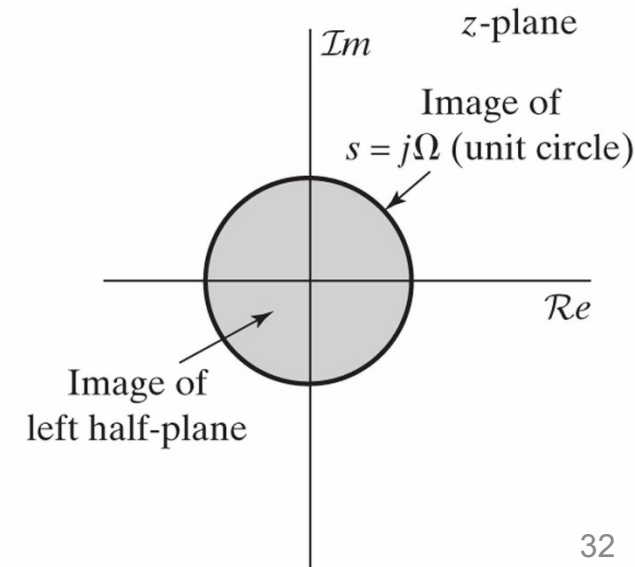
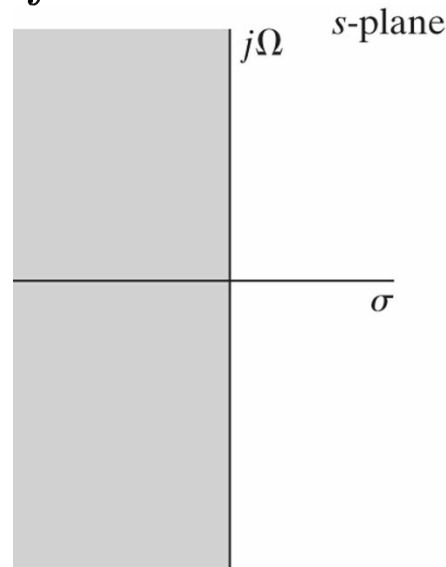
★ If $\sigma < 0$, $|z| < 1$ for any Ω

★ If $\sigma > 0$, $|z| > 1$ for any Ω

◆ Given $s = j\Omega$

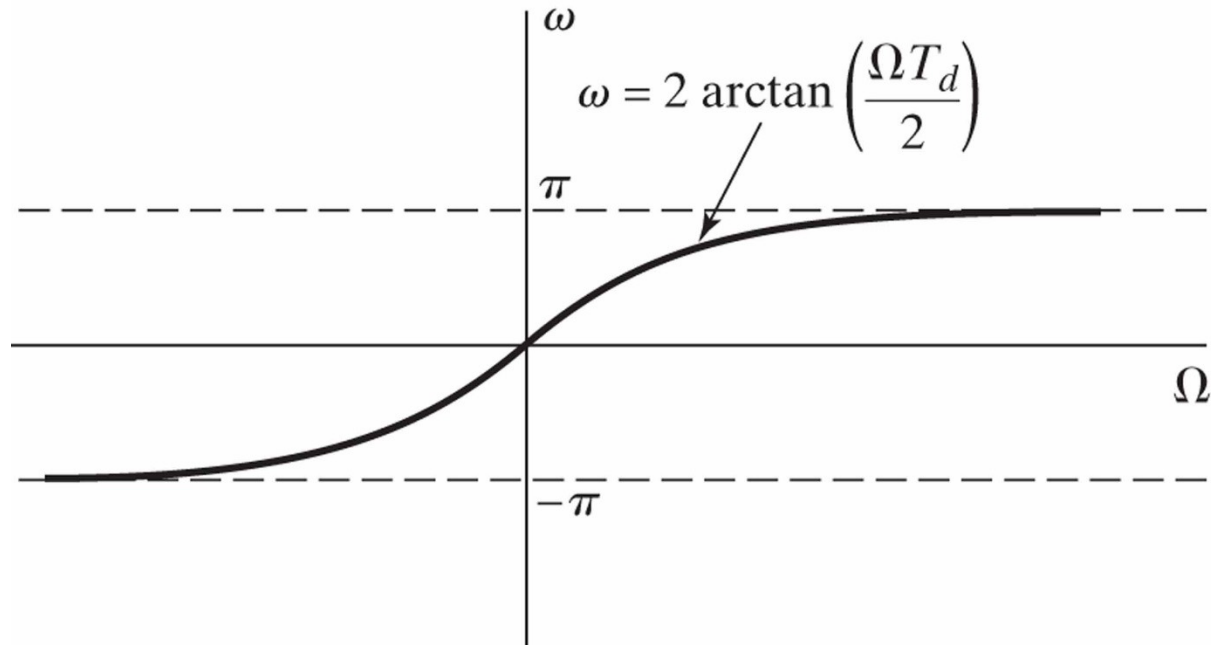
$$z = \frac{1 + j\Omega T_d/2}{1 - j\Omega T_d/2}$$

→ $|z|=1$ for any s



Bilinear transformation – frequency relationship

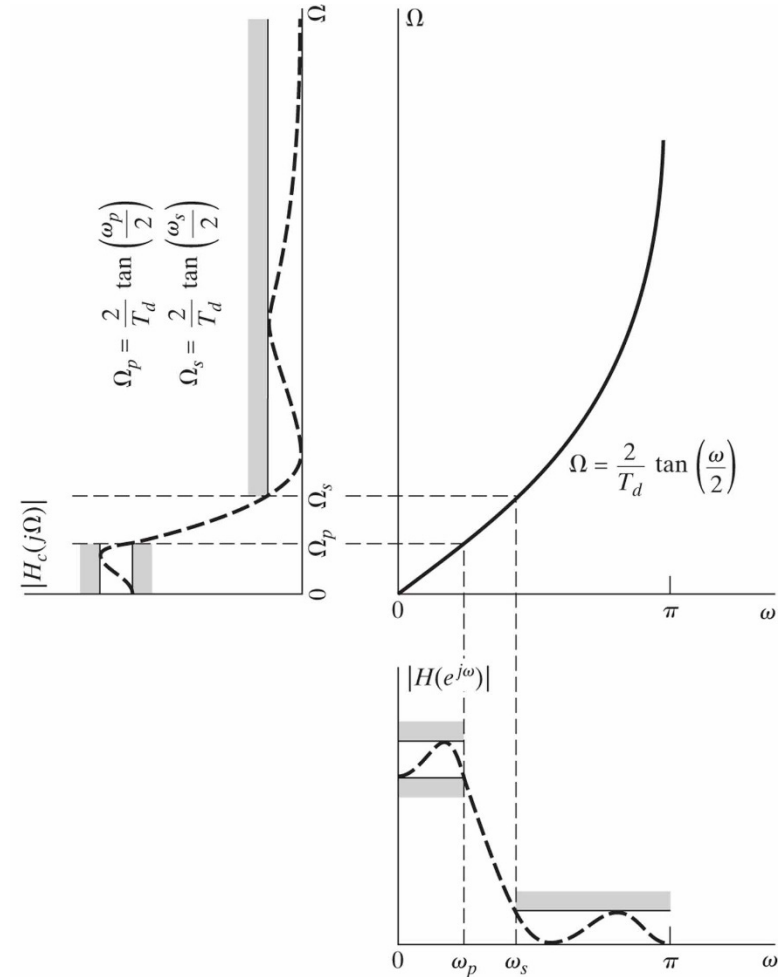
◆ $\Omega = \frac{2}{T_d} \tan(\omega/2), \quad \omega = 2 \arctan(\Omega T_d/2)$



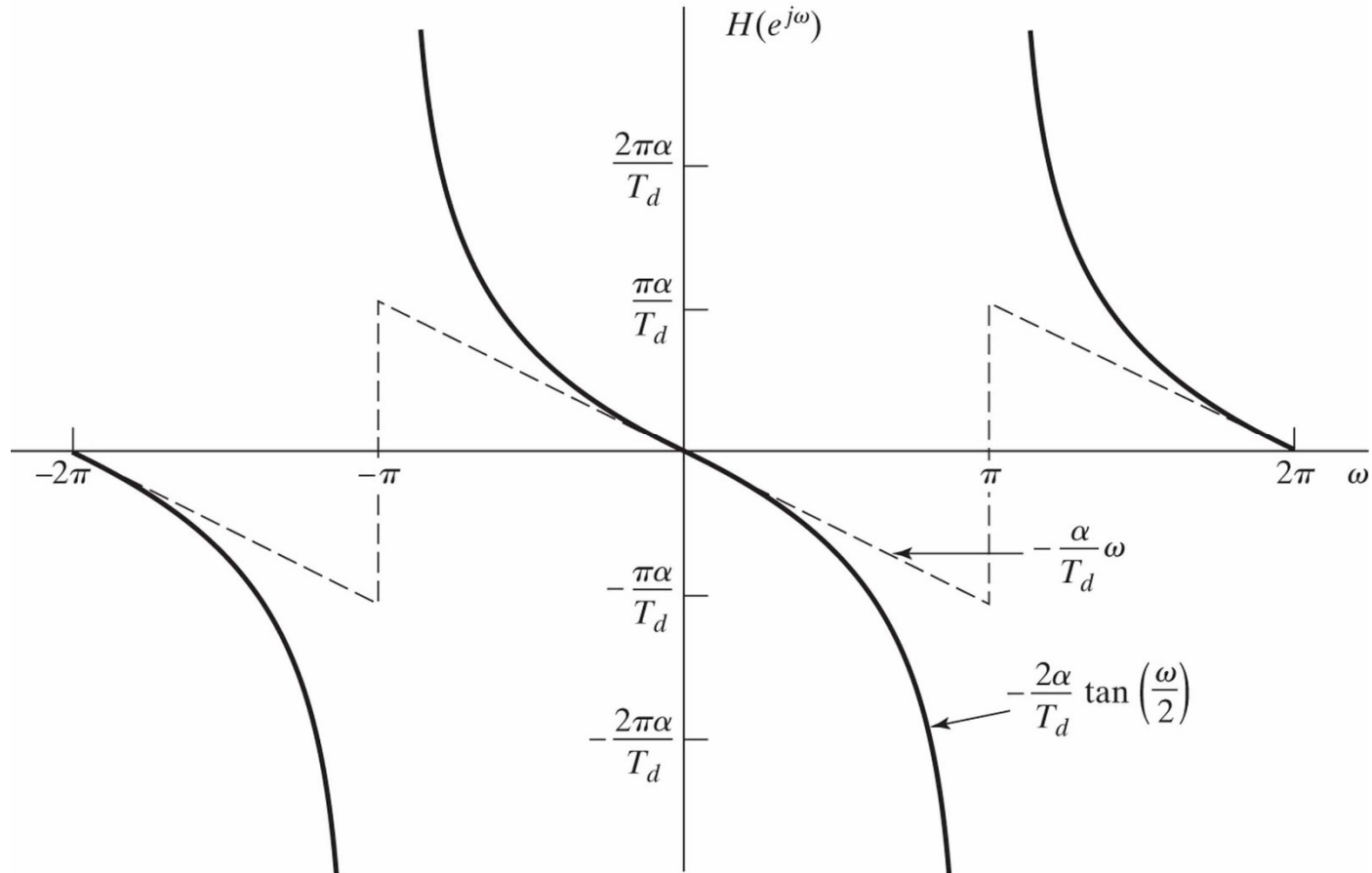
Frequency warping

Bilinear transformation

- ◆ No problem of aliasing compared to impulse invariance method
 - ★ Good for highpass filter design
- ◆ There exists the nonlinear compression of the frequency axis
 - ★ Suitable for piecewise-constant magnitude response filters
 - ★ Linear phase analog filters may lose linear phase property after transformation



Effect on phase response



Impulse invariance vs. bilinear transformation

◆ Bilinear transformation

- ✦ No aliasing effect
- ✦ Not good for preserving phase response

◆ Impulse invariance

- ✦ Aliasing happens due to sampling
- ✦ Possible to preserve linear phase of analog filter
 - Suitable to differentiator that requires linear phase

Bilinear transformation with Butterworth filter

- ◆ Specifications: $0.89125 \leq |H(e^{j\omega})| \leq 1, \quad 0 \leq |\omega| \leq 0.2\pi$
 $|H(e^{j\omega})| \leq 0.17783, \quad 0.3\pi \leq |\omega| \leq \pi$

- ◆ Transformed analog specifications

$$0.89125 \leq |H_c(j\Omega)| \leq 1, \quad 0 \leq |\Omega| \leq \frac{2}{T_d} \tan\left(\frac{0.2\pi}{2}\right)$$
$$|H_c(j\Omega)| \leq 0.17783, \quad \frac{2}{T_d} \tan\left(\frac{0.3\pi}{2}\right) \leq |\Omega| \leq \infty$$

- ◆ Due to monotonicity of Butterworth filter

$$|H_c(j2 \tan(0.1\pi))| \geq 0.89125, \quad |H_c(j2 \tan(0.15\pi))| \leq 0.17783$$

Bilinear transformation with Butterworth filter

- ◆ Using similar approach as in impulse invariance method, we get $N=5.305$

- ◆ Let $N = 6$, $\Omega_c = 0.766$, which now satisfies the stopband specification

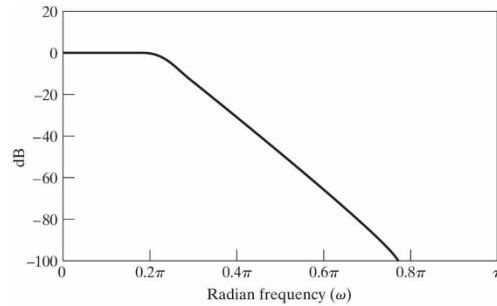
$$|H_c(j2 \tan(0.15\pi))| \leq 0.17783$$

- ◆ This is reasonable for bilinear transformation due to lack of aliasing
 - ★ Possible to have the desired stopband edge

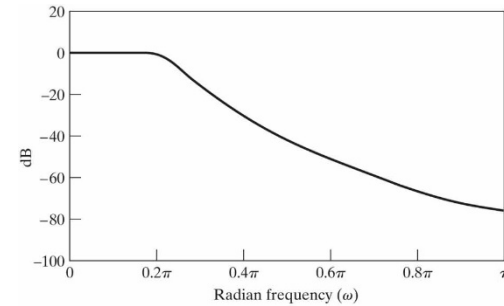
- ◆ Derive stable system function $H_c(s)$ and apply bilinear transformation

$$s = \frac{2}{T_d} \left(\frac{1 - z^{-1}}{1 + z^{-1}} \right)$$

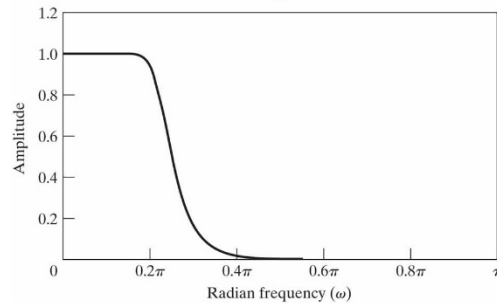
Impulse invariance vs. bilinear transformation



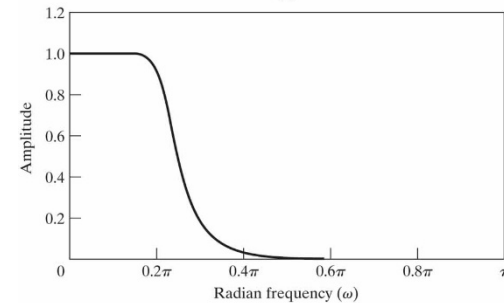
(a)



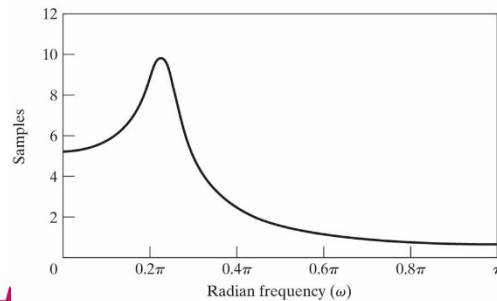
(a)



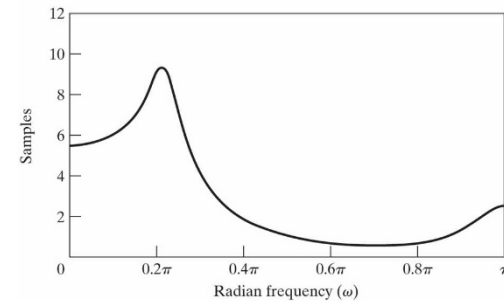
(b)



(b)



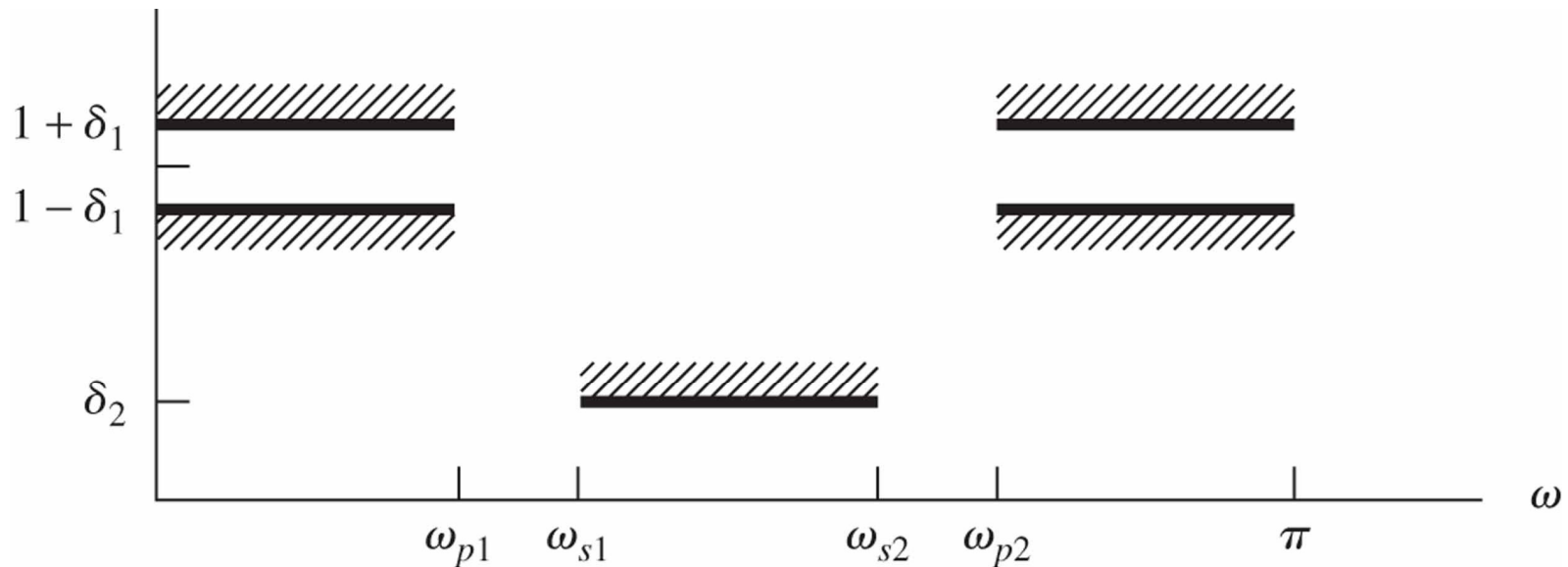
(c)



(c)

Frequency transformation of lowpass IIR filter

- ◆ So far, we have focused on lowpass IIR filter
- ◆ How can we implement general bandpass (multiband) filters?



Possible approaches

- ◆ Transform from analog multiband filter
 - ✦ Acceptable only with bilinear transformation
 - ✦ Impulse invariance suffers from aliasing
 - ➔ Hard to implement highpass (or multiband) filters

- ◆ Transform from discrete-time lowpass filter
 - ✦ Works for both impulse invariance and bilinear transformation

Transformation table

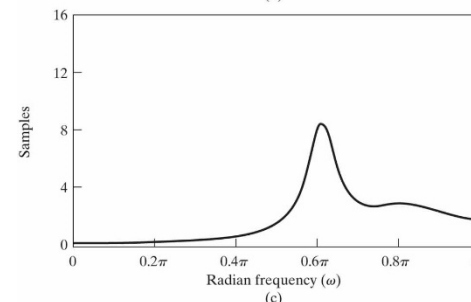
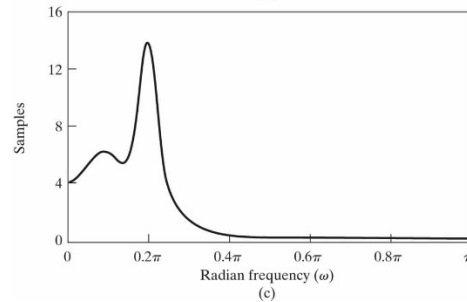
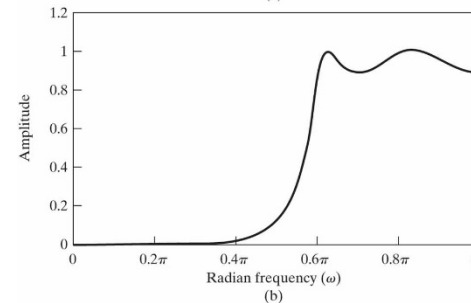
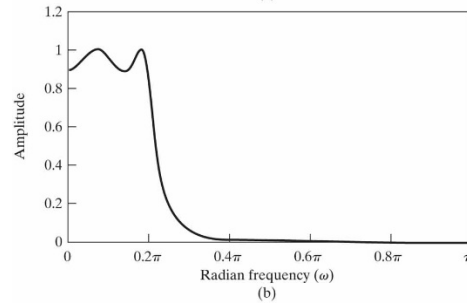
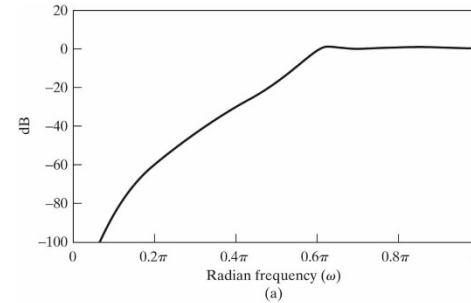
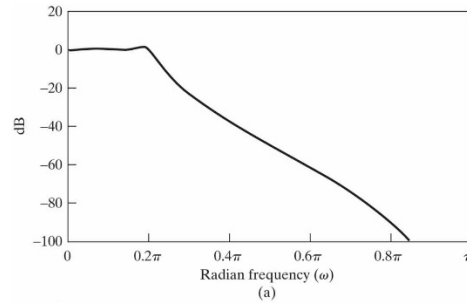
TABLE 7.1 TRANSFORMATIONS FROM A LOWPASS DIGITAL FILTER PROTOTYPE OF CUTOFF FREQUENCY θ_p TO HIGHPASS, BANDPASS, AND BANDSTOP FILTERS

Z-plane of prototype
lowpass filter

z-plane of
desired filter

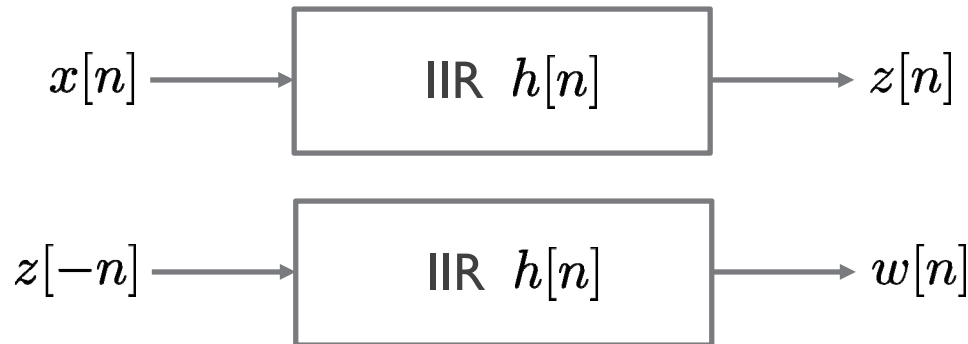
Filter Type	Transformations	Associated Design Formulas
Lowpass	$Z^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}$	$\alpha = \frac{\sin\left(\frac{\theta_p - \omega_p}{2}\right)}{\sin\left(\frac{\theta_p + \omega_p}{2}\right)}$ $\omega_p = \text{desired cutoff frequency}$
Highpass	$Z^{-1} = -\frac{z^{-1} + \alpha}{1 + \alpha z^{-1}}$	$\alpha = -\frac{\cos\left(\frac{\theta_p + \omega_p}{2}\right)}{\cos\left(\frac{\theta_p - \omega_p}{2}\right)}$ $\omega_p = \text{desired cutoff frequency}$
Bandpass	$Z^{-1} = -\frac{z^{-2} - \frac{2\alpha k}{k+1}z^{-1} + \frac{k-1}{k+1}}{\frac{k-1}{k+1}z^{-2} - \frac{2\alpha k}{k+1}z^{-1} + 1}$	$\alpha = \frac{\cos\left(\frac{\omega_{p2} + \omega_{p1}}{2}\right)}{\cos\left(\frac{\omega_{p2} - \omega_{p1}}{2}\right)}$ $k = \cot\left(\frac{\omega_{p2} - \omega_{p1}}{2}\right) \tan\left(\frac{\theta_p}{2}\right)$ $\omega_{p1} = \text{desired lower cutoff frequency}$ $\omega_{p2} = \text{desired upper cutoff frequency}$
Bandstop	$Z^{-1} = \frac{z^{-2} - \frac{2\alpha}{1+k}z^{-1} + \frac{1-k}{1+k}}{\frac{1-k}{1+k}z^{-2} - \frac{2\alpha}{1+k}z^{-1} + 1}$	$\alpha = \frac{\cos\left(\frac{\omega_{p2} + \omega_{p1}}{2}\right)}{\cos\left(\frac{\omega_{p2} - \omega_{p1}}{2}\right)}$ $k = \tan\left(\frac{\omega_{p2} - \omega_{p1}}{2}\right) \tan\left(\frac{\theta_p}{2}\right)$ $\omega_{p1} = \text{desired lower cutoff frequency}$ $\omega_{p2} = \text{desired upper cutoff frequency}$

Lowpass to highpass filter transformation



IIR filter with linear phase

- ◆ IIR filters generally have nonlinear phases
- ◆ Possible to have linear phase IIR filters for non real-time applications



$$y[n] = w[-n]$$

Frequency-domain analysis

◆ $Z(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega})$

$$W(e^{j\omega}) = H(e^{j\omega})Z^*(e^{j\omega})$$

$$= H(e^{j\omega})H^*(e^{j\omega})X^*(e^{j\omega})$$

$$= |H(e^{j\omega})|^2 X^*(e^{j\omega})$$

$$z[-n] \xleftrightarrow{\mathcal{DTFT}} Z^*(e^{j\omega})$$

◆ Since $y[n] = w[-n]$

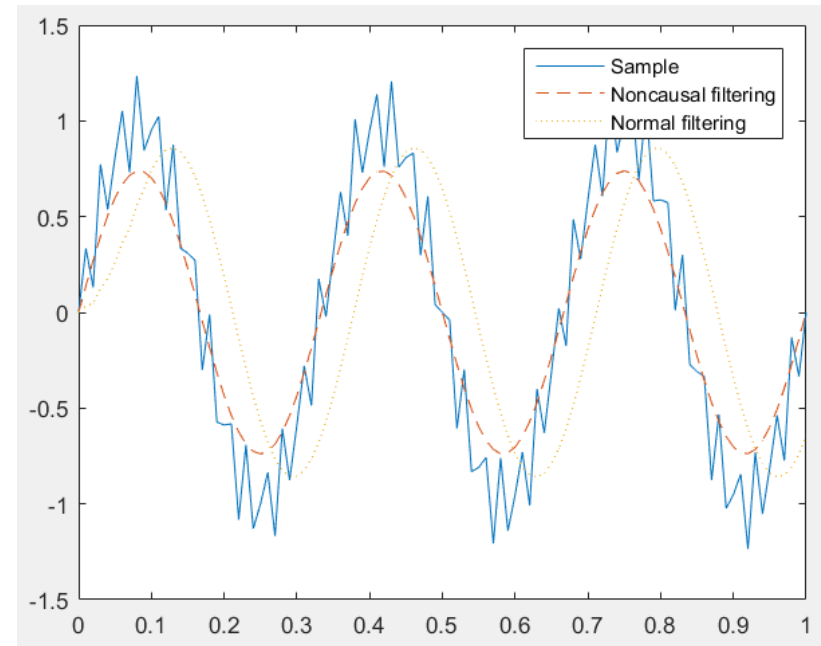
$$Y(e^{j\omega}) = W^*(e^{j\omega}) = |H(e^{j\omega})|^2 X(e^{j\omega})$$

Real number → no phase distortion at all!

Matlab example

```
% Linear phase IIR filter example from Mathworks.com
fs = 100;
t = 0:1/fs:1;
x = sin(2*pi*t*3)+.25*sin(2*pi*t*40);

b = ones(1,10)/10;           % 10 point averaging filter
y = filtfilt(b,1,x);         % Noncausal filtering
yy = filter(b,1,x);          % Normal filtering
plot(t,x,t,y,'--',t,yy,':')
```



FIR Filter Design

FIR filter design

- ◆ Design problem: FIR system function

$$H(z) = \sum_{k=0}^M b_k z^{-k}$$
$$h[n] = \begin{cases} b_n, & 0 \leq n \leq M \\ 0, & \text{otherwise} \end{cases}$$

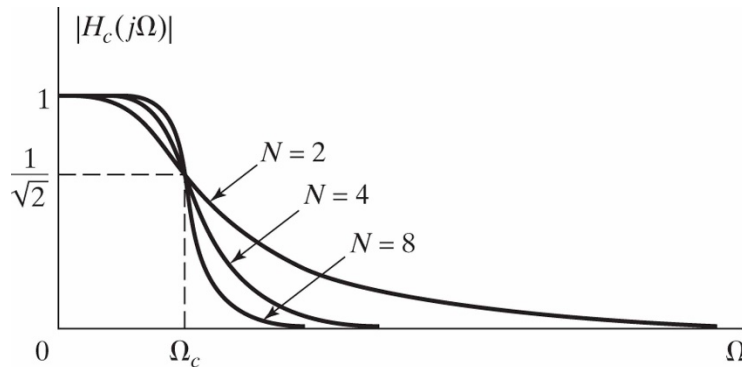
- ◆ Need to find
 - ★ Degree M
 - ★ Filter coefficients b_n (or $h[n]$) for $0 \leq n \leq M$
to approximate the desired frequency response

Lowpass filter as an example

- ◆ Ideal lowpass filter $H_{lp}(e^{j\omega}) = \begin{cases} 1, & |\omega| < \omega_c \\ 0, & \omega_c < \omega \leq \pi \end{cases}$

$$h[n] = \frac{\sin \omega_c n}{\pi n}, \quad -\infty < n < \infty$$

- ◆ Discrete-time IIR filter: transform from continuous-time IIR lowpass filter



$$|H_c(j\Omega)|^2 = \frac{1}{1 + (\Omega/\Omega_c)^{2N}}$$

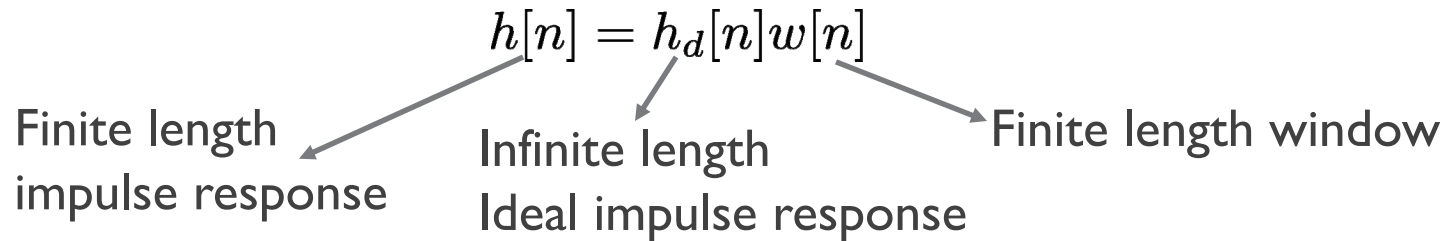
Get discrete-time filter using either impulse invariance or bilinear transformation

- ◆ How to get FIR filter? $h[n]$ is non-causal and infinite!

Design of FIR filter by windowing

- ◆ Most straightforward approach

- ★ Truncate the ideal impulse response by windowing (to have finite length) and do time shifting (to make it causal)



- ◆ Use rectangular window for simple truncation

$$w[n] = \begin{cases} 1, & 0 \leq n \leq M \\ 0, & \text{otherwise} \end{cases}$$

Lowpass filter example

- ◆ Ideal lowpass filter with zero delay

$$H_{lp}(e^{j\omega}) = \begin{cases} 1, & |\omega| < \omega_c \\ 0, & \omega_c < |\omega| \leq \pi \end{cases}$$

$$h_{lp}[n] = \frac{\sin \omega_c n}{\pi n}, \quad -\infty < n < \infty$$

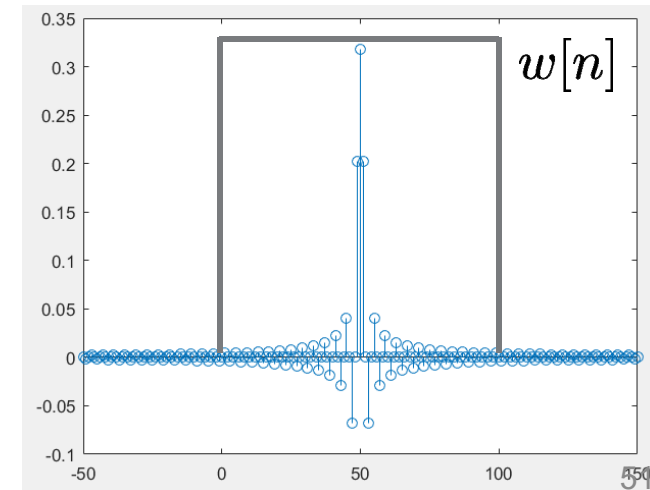
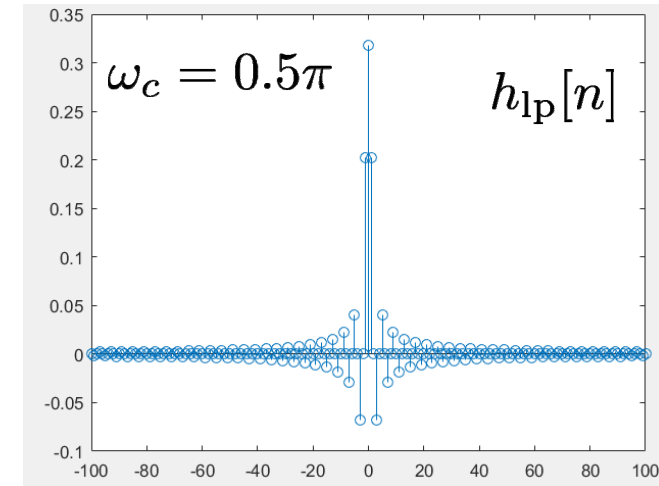
- ◆ Ideal lowpass filter with delay

$$H_{lp}(e^{j\omega}) = \begin{cases} e^{-j\omega n_d}, & |\omega| < \omega_c \\ 0, & \omega_c < |\omega| \leq \pi \end{cases}$$

$$h_{lp}[n] = \frac{\sin \omega_c (n - n_d)}{\pi (n - n_d)}, \quad -\infty < n < \infty$$

- ◆ Ideal lowpass filter with delay and truncation

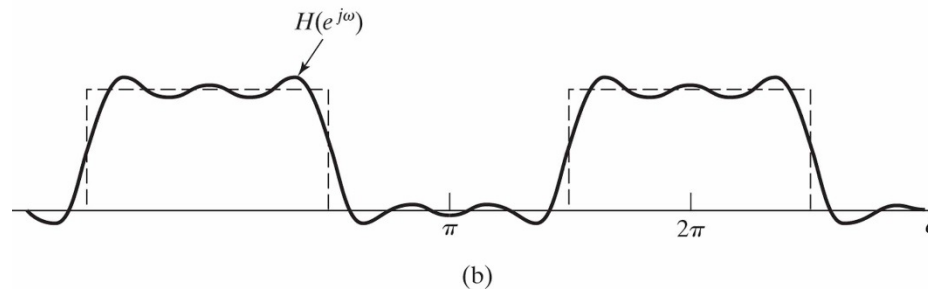
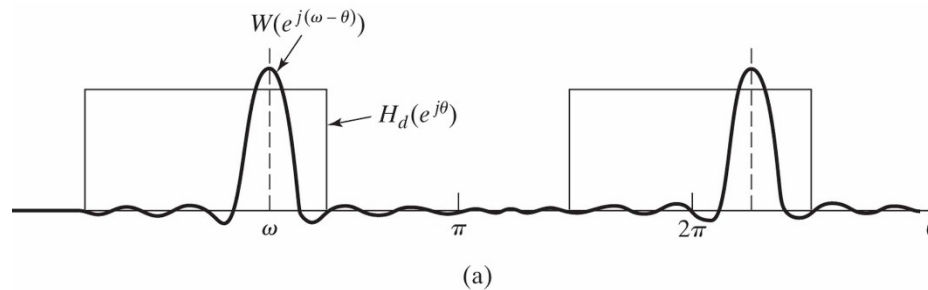
$$h_{lp}[n] = \frac{\sin \omega_c (n - n_d)}{\pi (n - n_d)}, \quad -n_1 < n < n_2$$



Frequency-domain representation

- ◆ From modulation (or windowing) theorem (2.9.7)

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

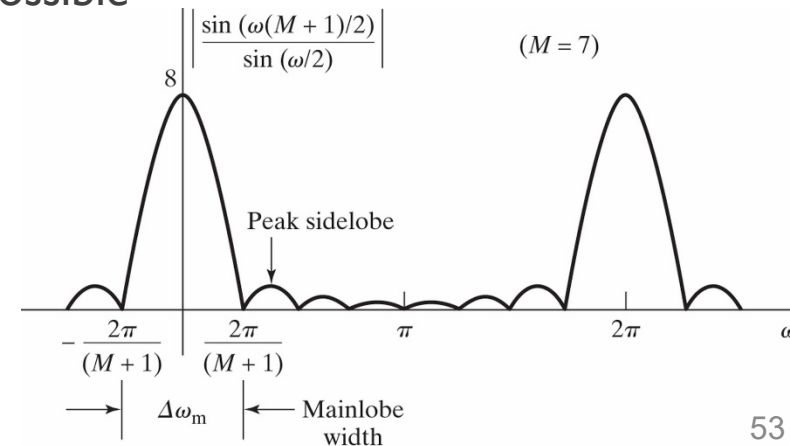


Tradeoff relationship

- ◆ If $w[n] = 1$ for all n , no truncation
 - ➔ $W(e^{j\omega})$ becomes a periodic impulse train with period 2π
 - ➔ $H(e^{j\omega}) = H_d(e^{j\omega})$

- ◆ How to choose the window size?
 - ★ As short as possible to minimize implementation computation
 - ★ Have $W(e^{j\omega})$ like an impulse as much as possible
 - ➔ Requires long window size

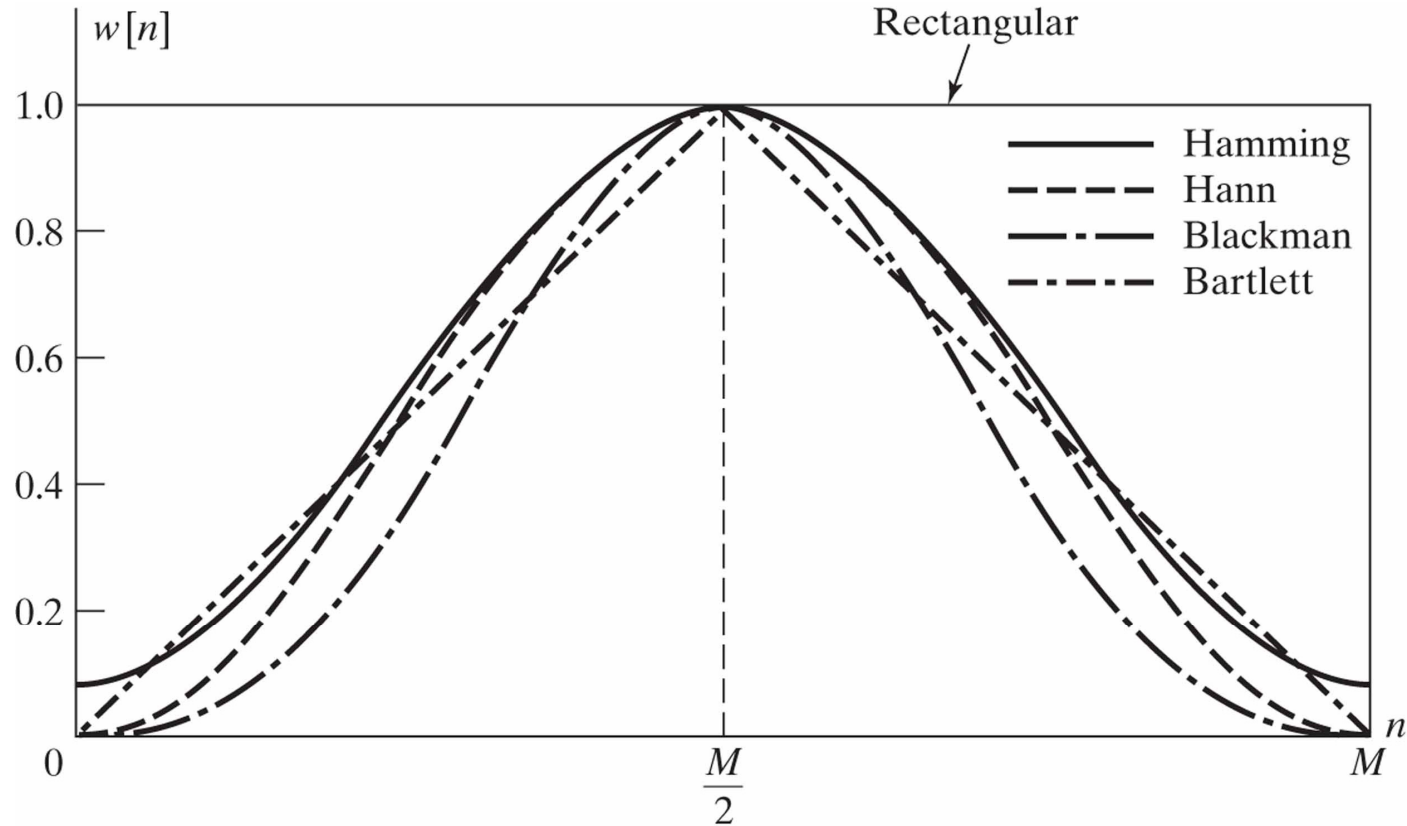
- ◆ Abrupt change in window
 - ➔ More ripples in frequency-domain



Commonly used windows

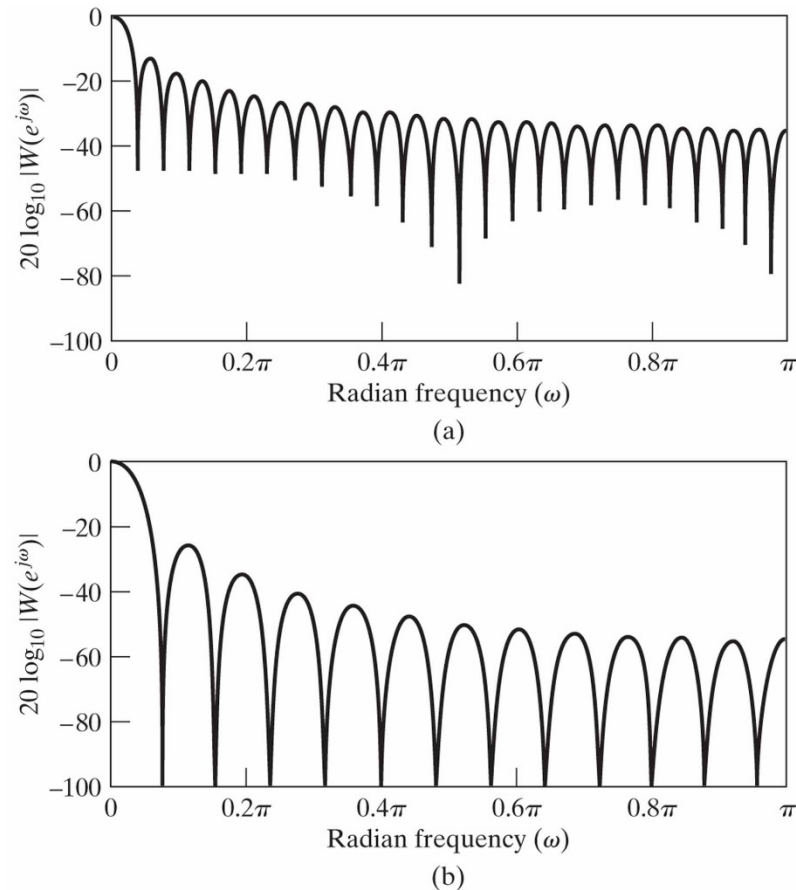
- ◆ Check Section 7.5.1
 - ✦ Rectangular
 - ✦ Bartlett (triangular)
 - ✦ Hann
 - ✦ Hamming
 - ✦ Blackman

Commonly used windows



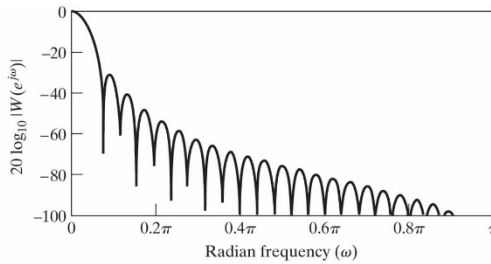
Commonly used windows

Figure 7.30 Fourier transforms (log magnitude) of windows of Figure 7.29 with $M = 50$. (a) Rectangular. (b) Bartlett.

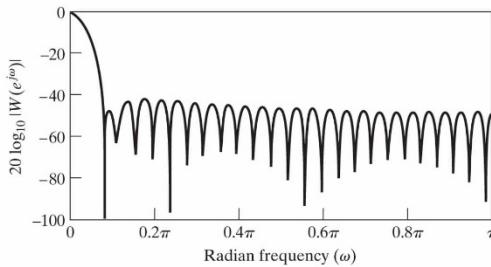


Commonly used windows

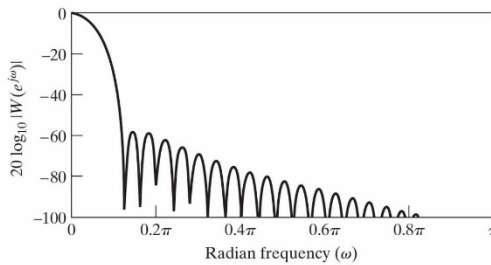
Figure 7.30 (continued) (c) Hann. (d) Hamming. (e) Blackman.



(c)



(d)



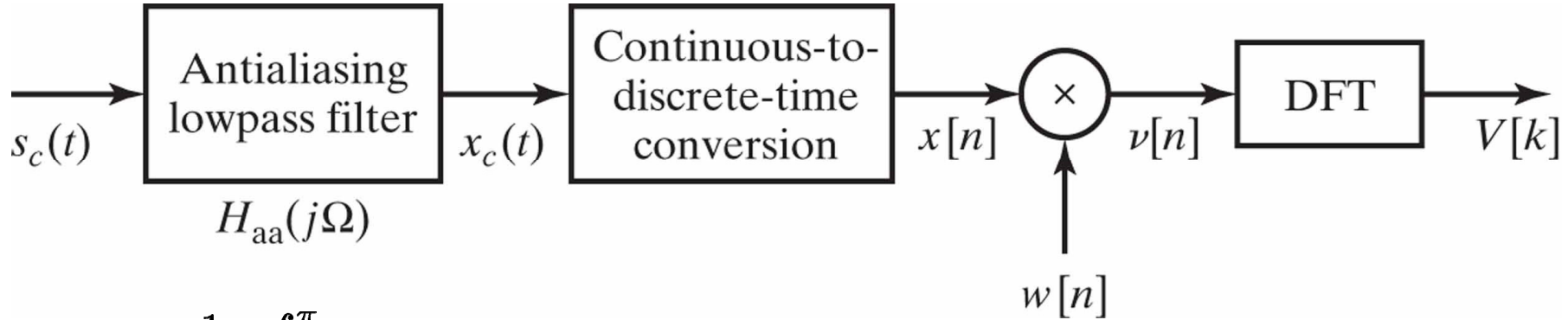
(e)

Comparison of standard windows

TABLE 7.2 COMPARISON OF COMMONLY USED WINDOWS

Type of Window	Peak Side-Lobe Amplitude (Relative)	Approximate Width of Main Lobe	Peak Approximation Error, $20 \log_{10} \delta$ (dB)	Equivalent Kaiser Window, β	Transition Width of Equivalent Kaiser Window
Rectangular	-13	$4\pi / (M + 1)$	-21	0	$1.81\pi / M$
Bartlett	-25	$8\pi / M$	-25	1.33	$2.37\pi / M$
Hann	-31	$8\pi / M$	-44	3.86	$5.01\pi / M$
Hamming	-41	$8\pi / M$	-53	4.86	$6.27\pi / M$
Blackman	-57	$12\pi / M$	-74	7.04	$9.19\pi / M$

Processing steps of Fourier analysis



$$V(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\theta}) W(e^{j(\omega-\theta)}) d\theta$$

$$V[k] = \sum_{n=0}^{N-1} v[n] e^{-j(2\pi/N)kn}, \quad k = 0, 1, \dots, N-1$$

$$= V(e^{j\omega}) \big|_{\omega=2\pi k/N}$$

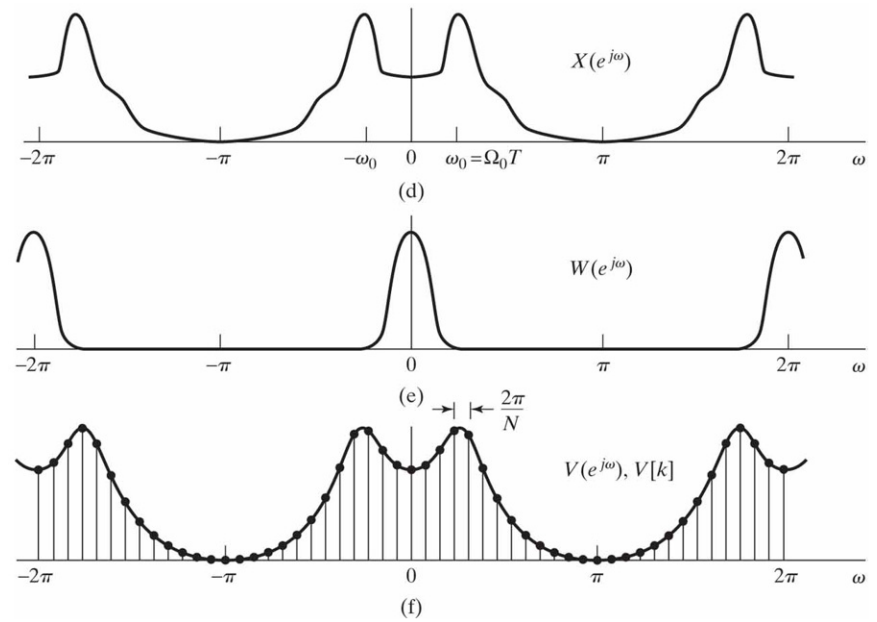
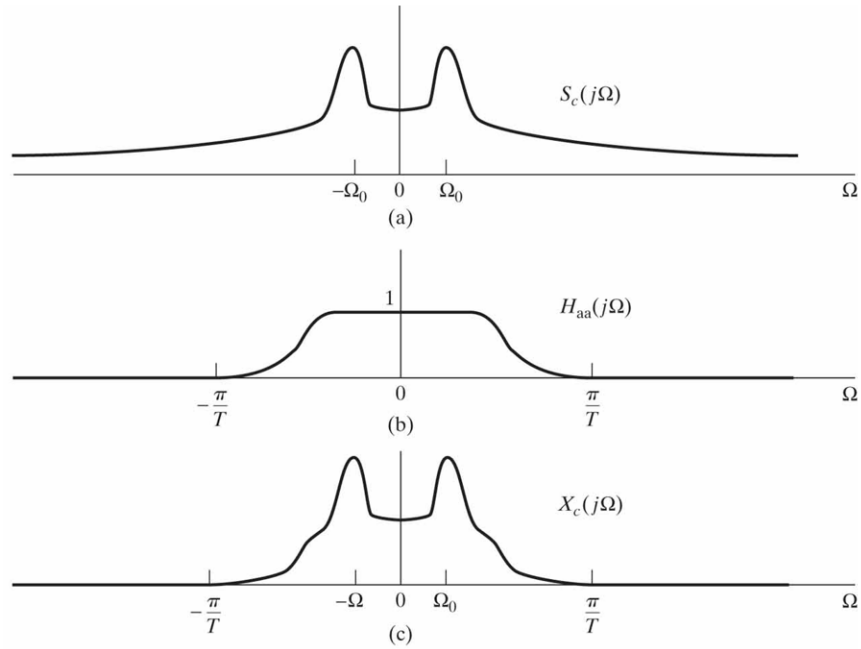
The k-th DFT frequency

$$\omega_k = 2\pi k/N$$

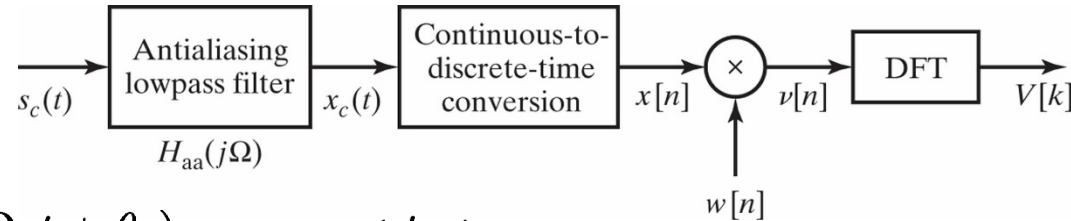
Corresponding continuous-time frequency

$$\Omega_k = \frac{2\pi k}{NT}$$

Illustration



DFT analysis of sinusoidal signals



$$s_c(t) = A_0 \cos(\Omega_0 t + \theta_0) + A_1 \cos(\Omega_1 t + \theta_1), \quad -\infty < t < \infty$$

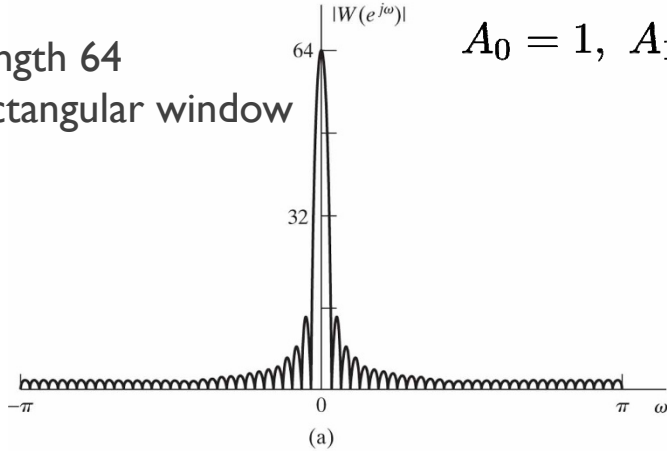
$$x[n] = A_0 \cos(\omega_0 n + \theta_0) + A_1 \cos(\omega_1 n + \theta_1), \quad -\infty < n < \infty$$

$$\begin{aligned} v[n] &= A_0 w[n] \cos(\omega_0 n + \theta_0) + A_1 w[n] \cos(\omega_1 n + \theta_1) \\ &= \frac{A_0}{2} w[n] e^{j\theta_0} e^{j\omega_0 n} + \frac{A_0}{2} w[n] e^{-j\theta_0} e^{-j\omega_0 n} + \frac{A_1}{2} w[n] e^{j\theta_1} e^{j\omega_1 n} + \frac{A_1}{2} w[n] e^{-j\theta_1} e^{-j\omega_1 n} \end{aligned}$$

$$V(e^{j\omega}) = \frac{A_0}{2} e^{j\theta_0} W(e^{j(\omega-\omega_0)}) + \frac{A_0}{2} e^{-j\theta_0} W(e^{j(\omega+\omega_0)}) + \frac{A_1}{2} e^{j\theta_1} W(e^{j(\omega-\omega_1)}) + \frac{A_1}{2} e^{-j\theta_1} W(e^{j(\omega+\omega_1)})$$

Effect of windowing

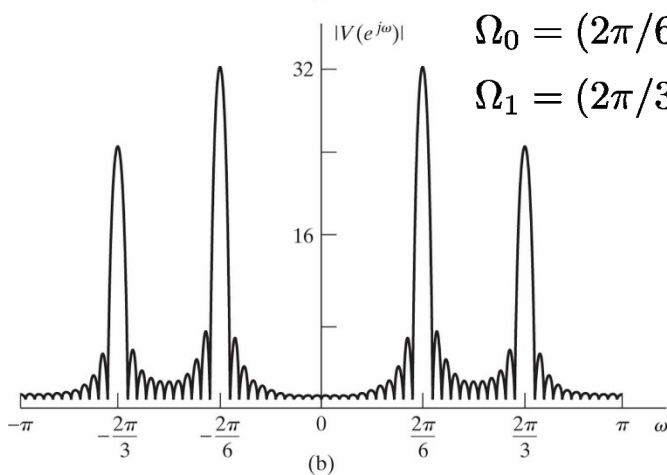
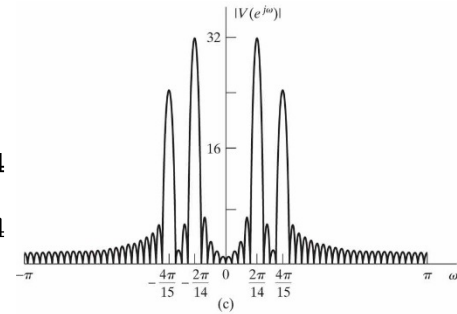
Length 64
rectangular window



$$A_0 = 1, A_1 = 0.75, \theta_0 = \theta_1 = 0, 1/T = 10 \text{ kHz}$$

$$\Omega_0 = (2\pi/14) \times 10^4$$

$$\Omega_1 = (4\pi/15) \times 10^4$$

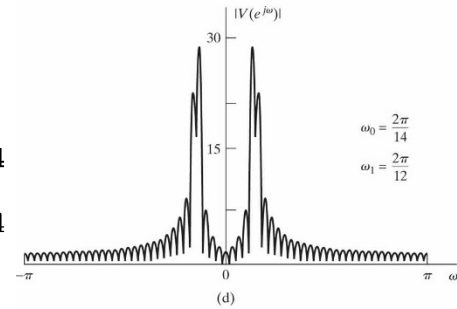


$$\Omega_0 = (2\pi/6) \times 10^4$$

$$\Omega_1 = (2\pi/3) \times 10^4$$

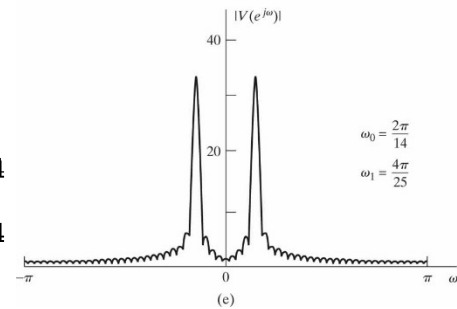
$$\Omega_0 = (2\pi/14) \times 10^4$$

$$\Omega_1 = (2\pi/12) \times 10^4$$



$$\Omega_0 = (2\pi/14) \times 10^4$$

$$\Omega_1 = (4\pi/25) \times 10^4$$



Effect of windowing

- ◆ Two primary effects
 - ✦ Reduced resolution for close frequencies
 - ➔ Related to width of main lobe
 - ✦ Leakage from one frequency component to another
 - ➔ Related to relative amplitude ratio of main lobe to side lobes

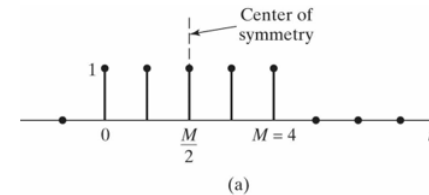
- ◆ Need to carefully select window depending on applications

Generalized linear phase FIR filter by windowing

◆ Recall four classes of FIR systems with generalized linear-phase

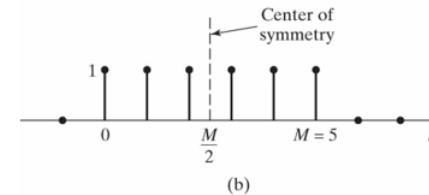
★ Type I

- Symmetric: $h[n] = h[M - n]$
- M even



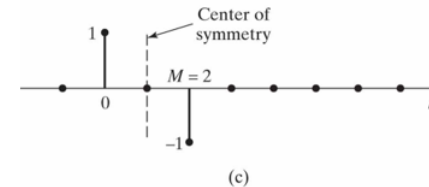
★ Type II

- Symmetric: $h[n] = h[M - n]$
- M odd



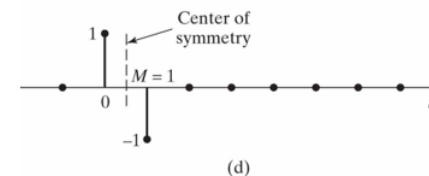
★ Type III

- Antisymmetric: $h[n] = -h[M - n]$
- M even



★ Type IV

- Antisymmetric: $h[n] = -h[M - n]$
- M odd



Generalized linear phase FIR filter by windowing

- ◆ Often aim at designing a causal system with a generalized linear phase
 - ✦ Stability is not a problem with FIR systems
- ◆ With (anti-)symmetric (possibly infinite length) impulse response,

$$h_d[M - n] = h_d[n]$$

← Symmetric at $M/2$

choose windows being symmetric at $M/2$

$$w[n] = \begin{cases} w[M - n], & 0 \leq n \leq M \\ 0, & \text{otherwise} \end{cases}$$

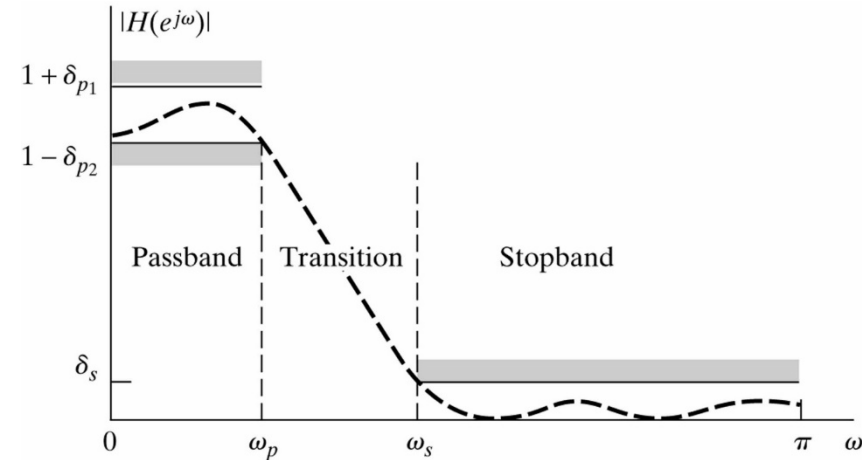
- ◆ Truncated filter $h[n] = h_d[n]w[n]$ still (anti-)symmetric
 - ➔ Generalized linear phase!!!

FIR filter design procedure

- ◆ Specifications to meet
 - ★ Transition bandwidth: $\Delta\omega = \omega_s - \omega_p$
 - ★ Ripple levels: δ_s , δ_p

- ◆ Window design method
 - ★ Choose window shape
 - ★ Adjust window length M
 - ➔ One parameter to adjust

- ◆ Need to perform trial and error
 - ➔ Not a good method



Kaiser window

◆ Formalization of window design method

- ★ No need for trial and error
- ★ Easy way to find the trade-off between the main-lobe width and side-lobe area
- ★ Design with two parameters: M and β
- ★ Can control the tradeoff between side-lobe amplitude and main-lobe width
 - This was not possible for previous windows

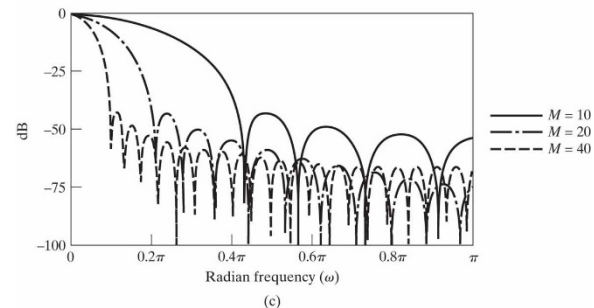
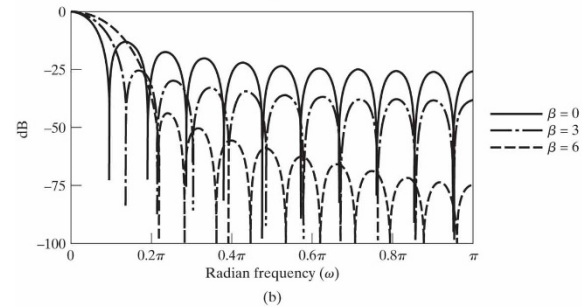
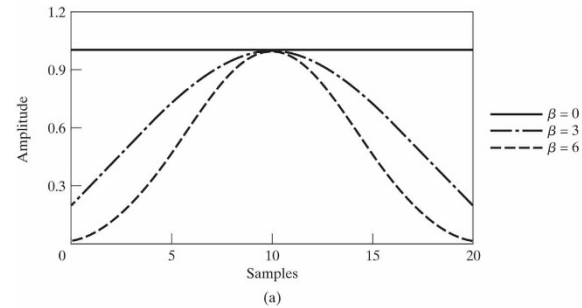
◆ Kaiser window expression Zeroth-order Bessel function of the first kind

$$w[n] = \begin{cases} I_0[\beta(1 - [(n - \alpha)/\alpha]^2)^{1/2}] / I_0(\beta), & 0 \leq n \leq M \\ 0, & \text{otherwise} \end{cases}$$

with $\alpha = M/2$ and $\beta \geq 0$

Kaiser window

- ◆ Becomes rectangular window when $\beta = 0$



$M = 20$

$\beta = 6$

Design FIR filter by Kaiser window

- ◆ Calculate M and β to meet the filter specifications

★ The peak approximation error δ is determined by β

- ◆ Define $A = -20 \log_{10} \delta$

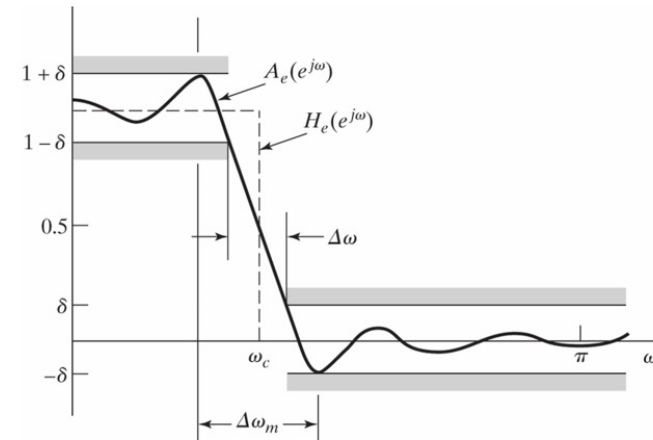
$$\beta = \begin{cases} 0.1102(A - 8.7), & A > 50 \\ 0.5842(A - 21)^{0.4} + 0.07886(A - 21), & 21 \leq A \leq 50 \\ 0, & A < 21 \end{cases}$$

- ◆ Passband/stopband frequencies also defined by δ

$$|H(e^{j\omega_p})| \geq 1 - \delta, \quad |H(e^{j\omega_s})| \leq \delta$$

★ Transition width becomes $\Delta\omega = \omega_s - \omega_c$

- ◆ Possible to show $M = \frac{A - 8}{2.285\Delta\omega}$ to satisfy the specifications



Comparison of standard windows

TABLE 7.2 COMPARISON OF COMMONLY USED WINDOWS

Type of Window	Peak Side-Lobe Amplitude (Relative)	Approximate Width of Main Lobe	Peak Approximation Error, $20 \log_{10} \delta$ (dB)	Equivalent Kaiser Window, β	Transition Width of Equivalent Kaiser Window
Rectangular	-13	$4\pi / (M + 1)$	-21	0	$1.81\pi / M$
Bartlett	-25	$8\pi / M$	-25	1.33	$2.37\pi / M$
Hann	-31	$8\pi / M$	-44	3.86	$5.01\pi / M$
Hamming	-41	$8\pi / M$	-53	4.86	$6.27\pi / M$
Blackman	-57	$12\pi / M$	-74	7.04	$9.19\pi / M$

Matlab examples

```
% Butterworth filter example
fc = 300;      % cutoff frequency (in Hz)
fs = 1000;    % sampling frequency (in Hz)
[b,a] = butter(6,fc/(fs/2));
freqz(b,a)

dataIn = randn(1000,1);
dataOut = filter(b,a,dataIn);

figure(2)
plot(1:1000,dataIn,1:1000,dataOut,'r')
legend('Random samples','Filtered samples')
```

Matlab examples

```
% designfilt example

lpFilt = designfilt('lowpassfir','PassbandFrequency',0.25, ...
'StopbandFrequency',0.35,'PassbandRipple',0.5, ...
'StopbandAttenuation',65,'DesignMethod','kaiserwin');
fvtool(lpFilt)
dataIn = rand(1000,1);
dataOut = filter(lpFilt,dataIn);
plot(1:1000,dataIn,1:1000,dataOut,'r')
```