

Chapter 9 IIR Digital Filter Design

Zhiliang Liu

Zhiliang_Liu@uestc.edu.cn

9.1 Preliminary Considerations

9.1.1 Digital Filter Specifications

Just like analog LPF, it is considered

$$1 - \delta_p \leq |G(e^{j\omega})| \leq 1 + \delta_p, \text{ for } |\omega| \leq \omega_p \cdots (9.1)$$

$$|G(e^{j\omega})| \leq \delta_s, \text{ for } \omega_s \leq |\omega| \leq \pi \cdots (9.2)$$

Passband edge frequency : ω_p

Stopband edge frequency : ω_s

9.1.1 Digital Filter Specifications

ripples : δ_p and δ_s

peak passband ripple :

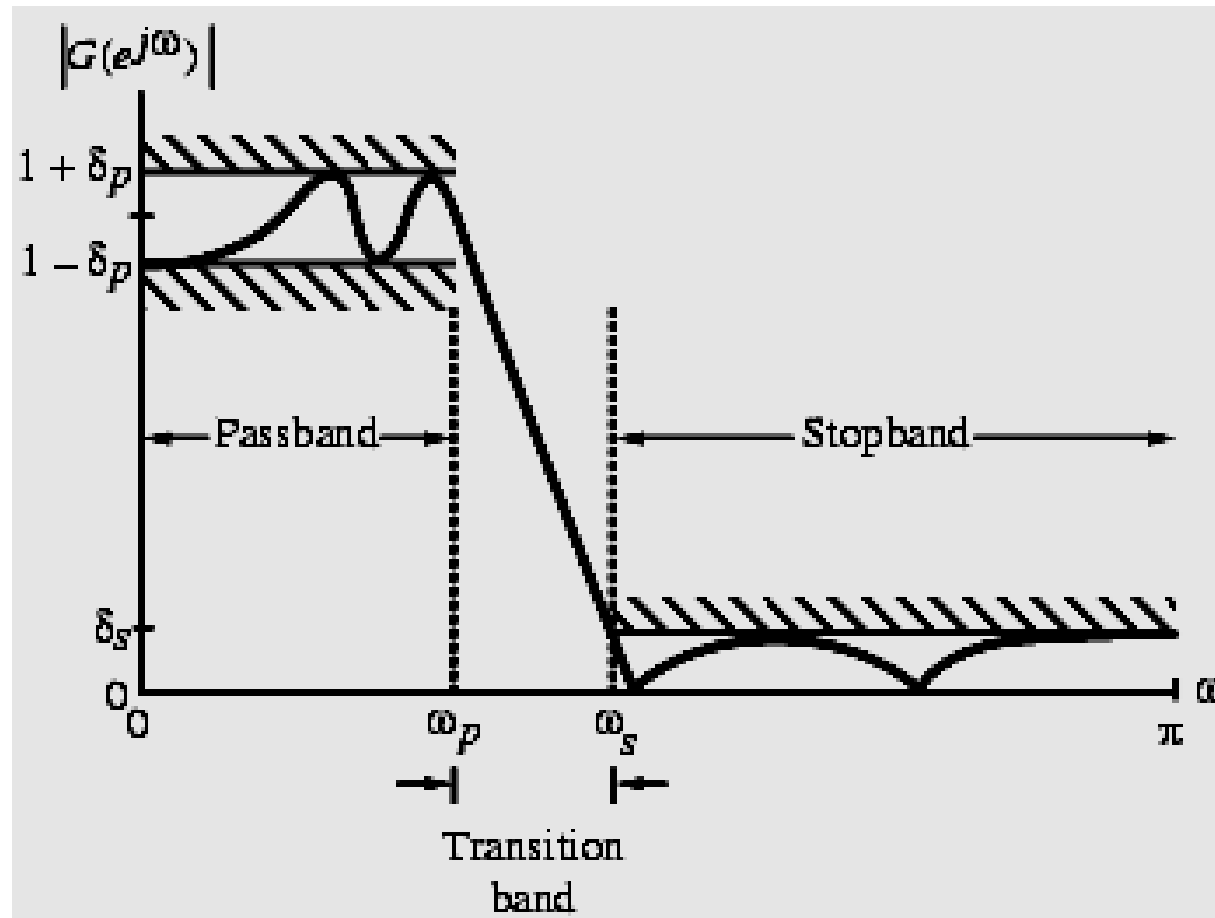
$$\alpha_p = -20 \log_{10}(1 - \delta_p) [dB] \cdots (9.3)$$

minimum stopband attenuation :

$$\alpha_s = -20 \log_{10}(\delta_s) [dB] \cdots (9.4)$$

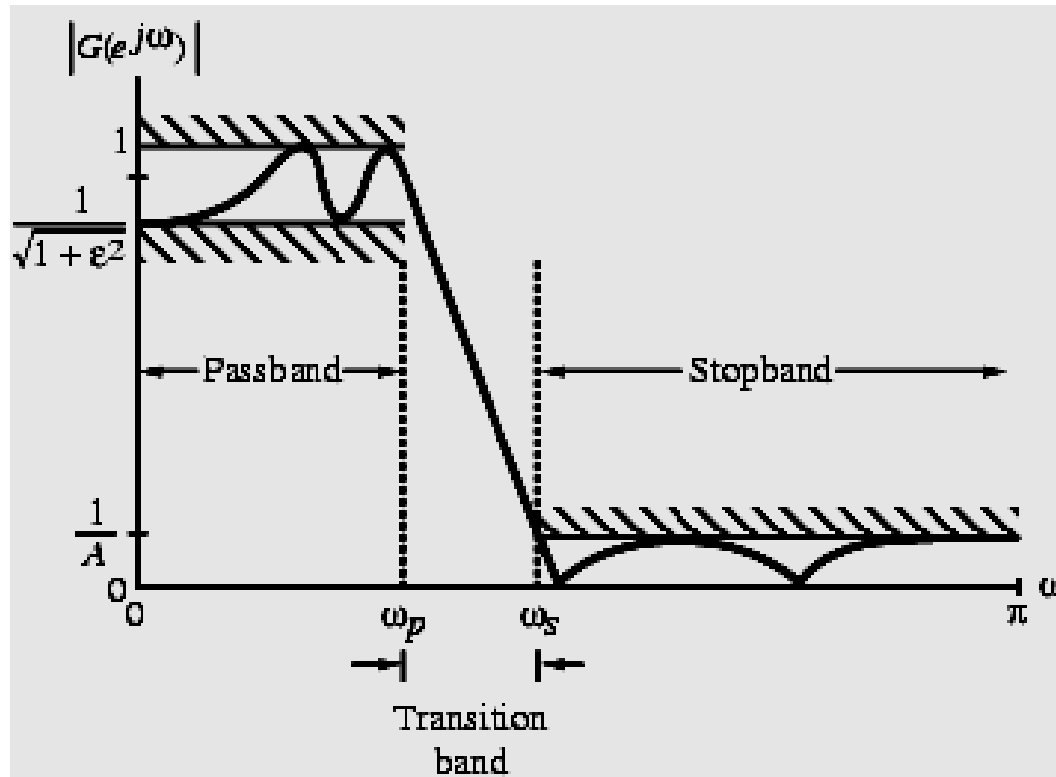
• Specifications are often given in terms of loss function $A(\omega) = -20 \log_{10} |G(e^{j\omega})|$ in dB

9.1.1 Digital Filter Specifications



9.1.1 Digital Filter Specifications

- Magnitude specifications may alternately be given in a normalized form as indicated below



9.1.1 Digital Filter Specifications

- Here, the maximum value of the magnitude in the passband is assumed to be **unity**.
- $1/\sqrt{1+\varepsilon^2}$ - Maximum passband deviation, given by the minimum value of the magnitude in the passband
- $1/A$ - Maximum stopband magnitude

9.1.1 Digital Filter Specifications

- 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

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

9.1.1 Digital Filter Specifications

- In practice, passband edge frequency F_p and stopband edge frequency F_s are specified in Hz
- For digital filter design, normalized band edge frequencies need to be computed from specifications in Hz using

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

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

9.1.1 Digital Filter Specifications

- **Example:** Let $F_p = 7$ kHz, $F_s = 3$ kHz, and $F_T = 25$ kHz
- Then

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

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

9.1.2 Selection of the Filter Type

The *objective* of digital filter design is to develop a casual transfer function $H(z)$ meeting the frequency response specifications.

Whether IIR or FIR should be selected?

IIR:

low order to reduce
computational complexity,
but must be stable.

$$H(z) = \frac{\sum_{k=0}^M p_k z^{-k}}{\sum_{k=0}^N d_k z^{-k}}$$

FIR:

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

with linear phase ,but high order.

9.1.3 Basic Approaches to IIR Digital Filter Design

IIR design

The most common practice is to transform $H_a(s)$ into the desired digital transfer function $G(z)$.

**convert the digital filter specifications to
into analog lowpass prototype filter
specifications**



**determine the analog lowpass filter
transfer function meeting these
specifications**



**transform the analog filter into the
desired digital transfer function**

9.1.3 Basic Approaches to IIR Digital Filter Design

This approach has been widely used for many reasons:

- (a) Analog approximation techniques are highly advanced.**
- (b) They usually yield closed-form solutions.**
- (c) Extensive tables are available for analog filter design.**
- (d) Many applications require the digital simulation of analog filters.**

9.1.3 Basic Approaches to IIR Digital Filter Design

The basic idea is to apply a mapping from s -domain to the z -domain so that the essential properties of the analog frequency response are preserved.

This implies that the mapping function should be such that

- (1) The **imaginary axis** in the s -plane be mapped onto the **unit circle** of the z -plane.
- (2) A **stable** analog transfer function be transferred into a **stable** digital transfer function.

9.1.4 Estimation of the Filter Order

For the design of IIR LPF, the order of $H_a(s)$ is estimated from its specifications using the appropriate formula given in Eq. (4.35), (4.43), or (4.54), depending on which approximation desired. the order of $G(z)$ is determined automatically from the transformation being used to convert $H_a(s)$ into $G(z)$.

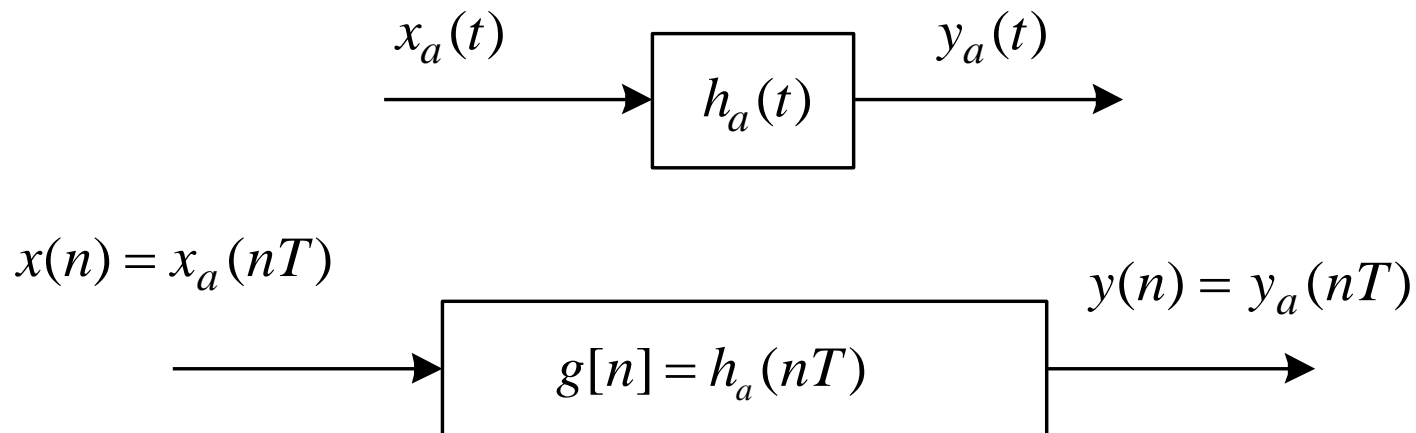
The scaled transfer function $G_t(z) = KG(z)$, $K = 1/G_{\max}(z_0)$ is BR function.

9.2 The Design Methods of IIR Filter

9.2.1 The Impulse Invariance Method

1. The Main Idea (Problem 9.6)

It is desired $g[n] = h_a(nT)$. If $h_a(t) \leftrightarrow H_a(s)$, $g[n] \leftrightarrow G(z)$. From the relation between LT and ZT, we know that



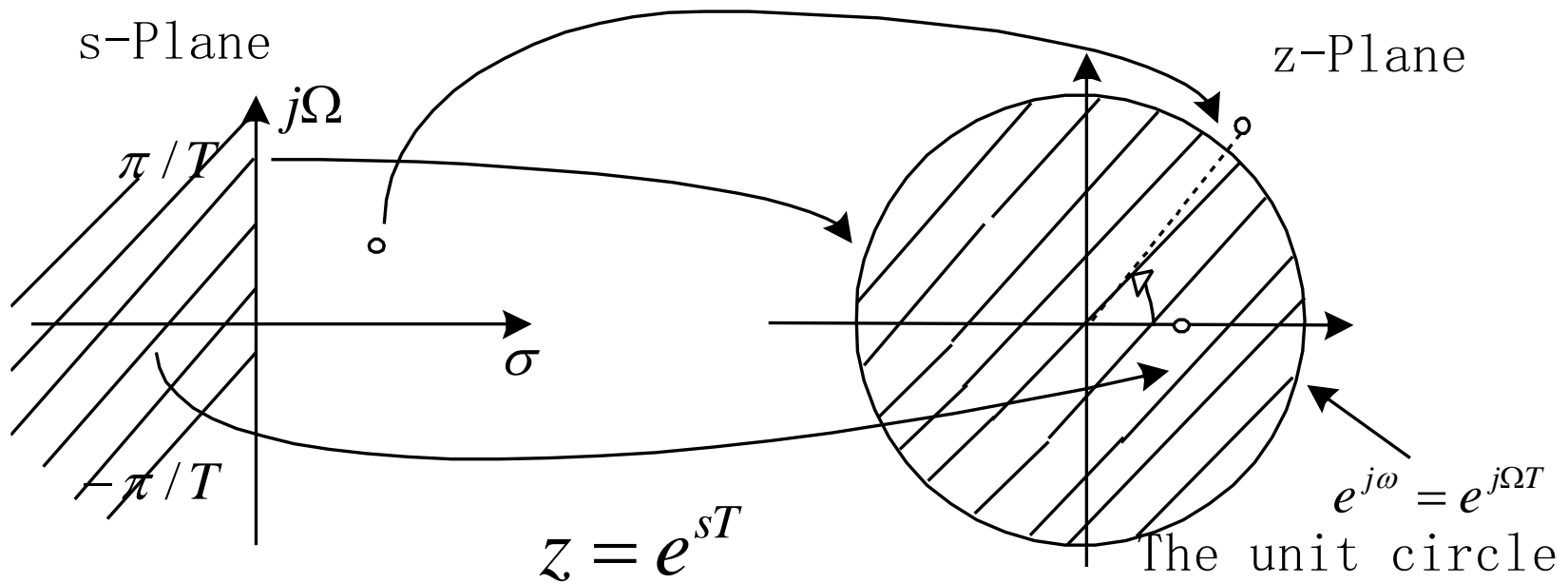
9.2.1 The Impulse Invariance Method

The frequency response

$$G(z) \big|_{z=e^{sT}} = \frac{1}{T} \sum_{m=-\infty}^{\infty} H_a\left(s - jm\frac{2\pi}{T}\right)$$

**The mapping function is $z = e^{sT}$
or $s = \frac{1}{T} \ln z$.**

9.2.1 The Impulse Invariance Method



Discussion: (1) The mapping satisfies the essential properties ($j\Omega$ -axis mapped onto the unit circle, stable region in s -plane mapped into the stable region in z -plane).

9.2.1 The Impulse Invariance Method

(2) The exact relation between $j\Omega$ -axis in the s -plane and the unit circle in z -plane is $\omega = \Omega T$.

(3) When Nyquist theorem is satisfied,

$$G(e^{j\omega}) = \frac{1}{T} H_a(j\Omega), (|\omega| < \pi)$$

If $H_a(j\Omega)$ is not bandlimited, $G(e^{j\omega})$ obtained by this method is overlapped in frequency-domain.

9.2.1 The Impulse Invariance Method

2. The Design Steps

**(1) $H_a(s)$ is partial-fractional expressed as:
($h_a(t)$ is derived)**

$$H_a(s) = \sum_{i=1}^N \frac{A_i}{s - s_i}, (\text{Re}(s_i)_{\max} < 0)$$

(2) From $g[n] = h_a(nT)$, we can get:

$$G_1(z) = \sum_{i=1}^N \frac{A_i}{1 - e^{s_i T} z^{-1}}$$

9.2.1 The Impulse Invariance Method

(3) Usually, we scale T to $G(z)$ and get:

$$G(z) = \sum_{i=1}^N \frac{TA_i}{1 - e^{s_i T} z^{-1}}$$

Proof: Because

$$h_a(t) = L^{-1}\{H_a(s)\} = \sum_{i=1}^N A_i e^{s_i t} \mu(t)$$

$$g[n] = h_a(nT) = \sum_{i=1}^N A_i e^{s_i nT} \mu[n]$$

$$G_1(z) = \sum_{i=1}^N \frac{A_i}{1 - e^{s_i T} z^{-1}}$$

9.2.1 The Impulse Invariance Method

Example If $H_a(s) = \frac{2}{s^2 + 3s + 2}$, the
sampling period is T . Then

$$H_a(s) = \frac{2}{s+1} - \frac{2}{s+2}$$

The poles: $s_1 = -1$, $s_2 = -2$.

Using the impulse invariance method,
we get:

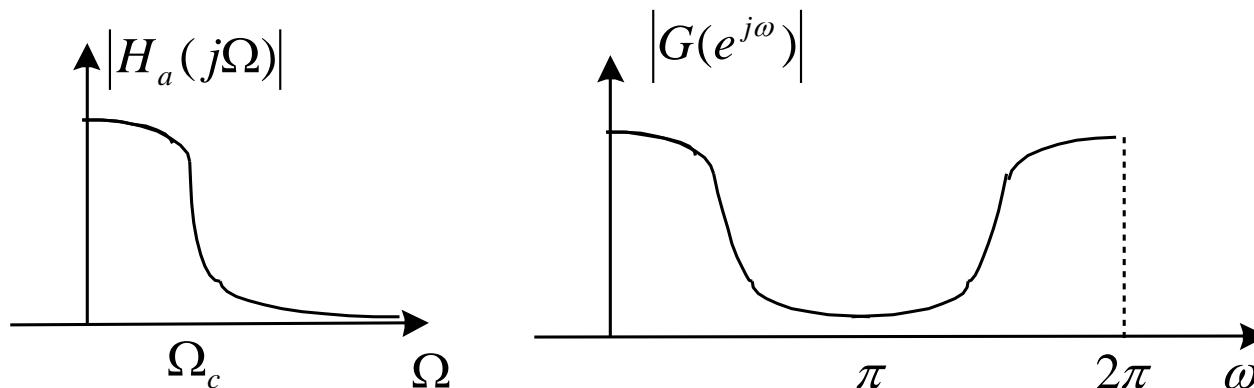
$$\begin{aligned} G(z) &= T \frac{2}{1 - e^{-T} z^{-1}} - T \frac{2}{1 - e^{-2T} z^{-1}} \\ &= \frac{2T(e^{-T} - e^{-2T})z^{-1}}{1 - (e^{-2T} + e^{-T})z^{-1} + e^{-3T}z^{-2}} \end{aligned}$$

9.2.1 The Impulse Invariance Method

When $T = 1$,

$$G(z) = \frac{0.4651 z^{-1}}{1 - 0.5032 z^{-1} + 0.04979 z^{-2}}$$

Note: $G(e^{j\omega})$ is overlapped in the frequency-domain.



9.2.2 The Bilinear Translation Method

The Bilinear Transformation

The mapping from s-plane to z-plane:

$$s = \frac{2}{T} \left(\frac{1 - z^{-1}}{1 + z^{-1}} \right) \cdots (9.14)$$

The relation between the digital transfer function and the parent analog transfer function:

$$G(z) = H_a(s) \Big|_{s = \frac{2}{T} \left(\frac{1 - z^{-1}}{1 + z^{-1}} \right)} \cdots (9.15)$$

9.2.2 The Bilinear Translation Method

- Digital filter design consists of 3 steps:
 - (1) Develop the specifications of $H_a(s)$ by applying the inverse bilinear transformation to specifications of $G(z)$
 - (2) Design $H_a(s)$
 - (3) Determine $G(z)$ by applying bilinear transformation to $H_a(s)$
- As a result, the parameter T has no effect on $G(z)$ and $T = 2$ is chosen for convenience

9.2.2 The Bilinear Translation Method

- Inverse bilinear transformation for $T=2$ is

$$z = \frac{1+s}{1-s}$$

For $s = \sigma_0 + j\Omega_0$

$$z = \frac{(1 + \sigma_0) + j\Omega_0}{(1 - \sigma_0) - j\Omega_0} \Rightarrow |z|^2 = \frac{(1 + \sigma_0)^2 + \Omega_0^2}{(1 - \sigma_0)^2 + \Omega_0^2}$$

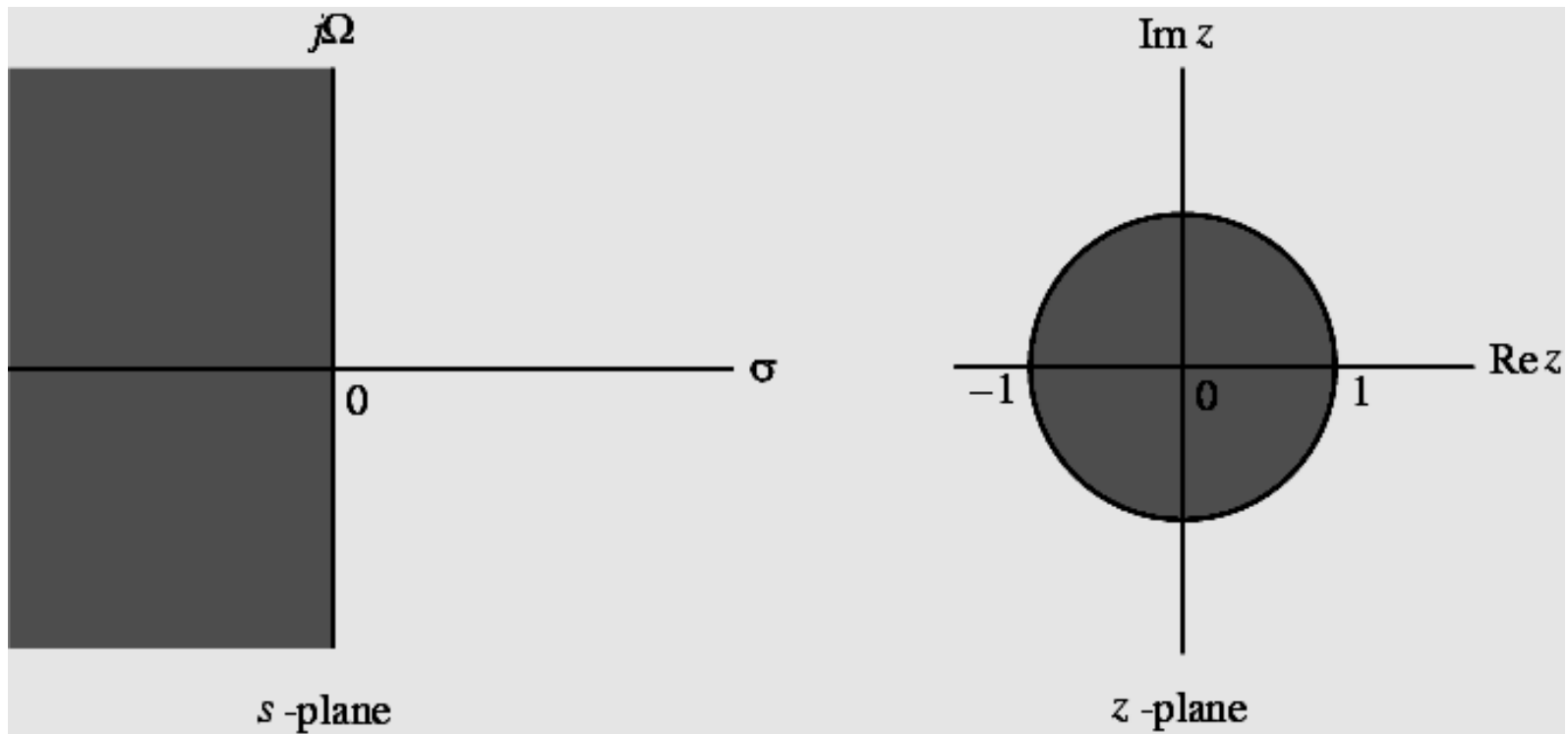
Thus, $\sigma_0 = 0 \rightarrow |z| = 1$

$\sigma_0 < 0 \rightarrow |z| < 1$

$\sigma_0 > 0 \rightarrow |z| > 1$

9.2.2 The Bilinear Translation Method

- Mapping of s -plane into the z -plane

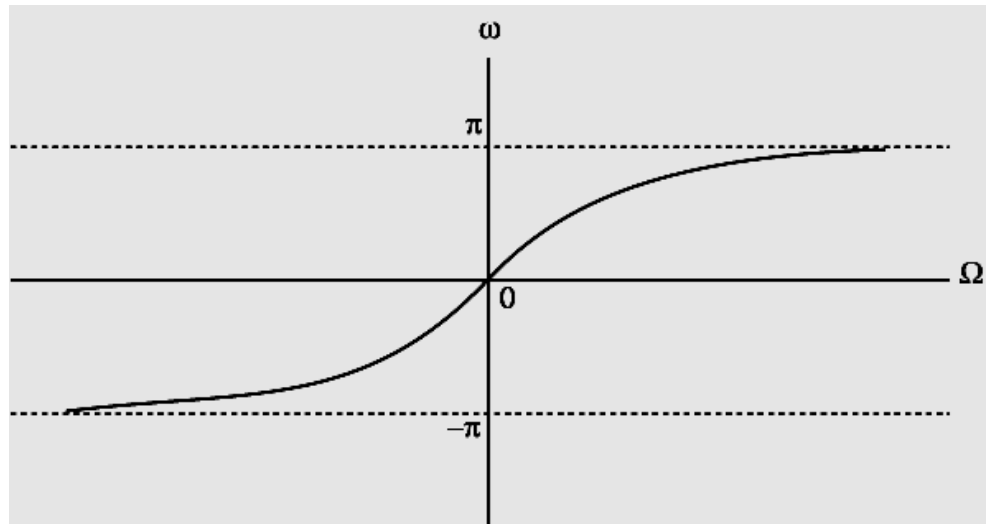


9.2.2 The Bilinear Translation Method

- For $z = e^{j\omega}$ with $T = 2$ we have

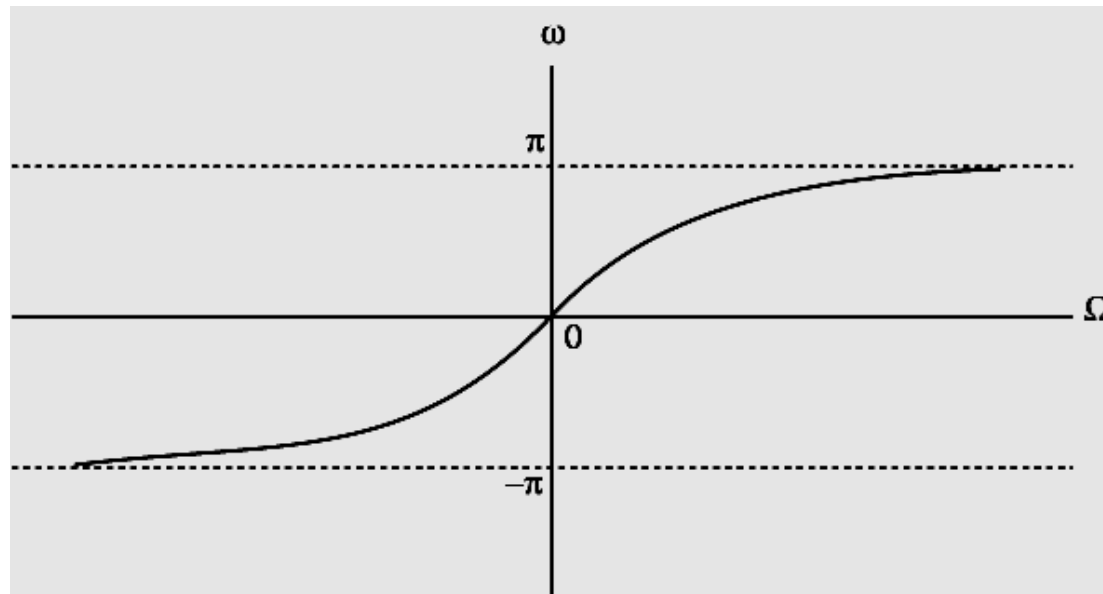
$$\begin{aligned} j\Omega &= \frac{1 - e^{-j\omega}}{1 + e^{-j\omega}} = \frac{e^{-j\omega/2}(e^{j\omega/2} - e^{-j\omega/2})}{e^{-j\omega/2}(e^{j\omega/2} + e^{-j\omega/2})} \\ &= \frac{j2\sin(\omega/2)}{2\cos(\omega/2)} = j\tan(\omega/2) \end{aligned}$$

or $\Omega = \tan(\omega/2)$



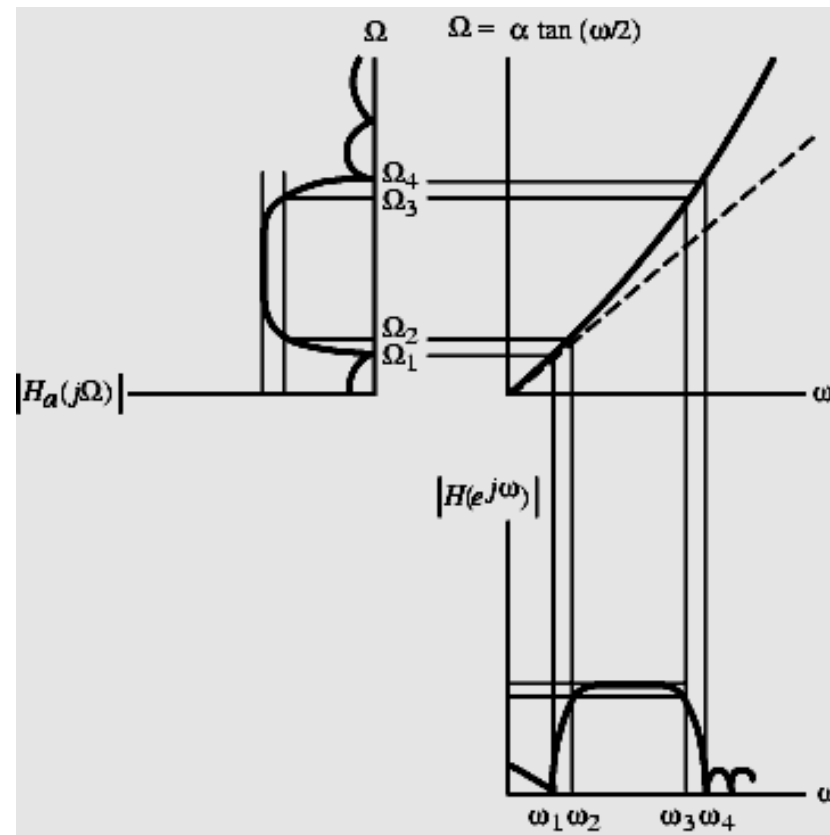
The exact relation between $j\Omega$ -axis in the s -plane and the unit circle in z -plane (from Eq. (9.18) with $T = 2$):

$$\Omega = \tan\left(\frac{\omega}{2}\right) \cdots (9.18)$$



This introduces a distortion the frequency axis called *frequency warping*.

Thus, to develop a digital filter meeting a specified magnitude, we must ***first prewarp*** the critical bandedge frequencies (ω_p and ω_s) to find their analog Equivalents (Ω_s and Ω_p) using Eq.(9.18).



9.2.2 The Bilinear Translation Method

- **Steps in the design of a digital filter**
 - (1) Prewarp (ω_p, ω_s) to find their analog equivalents (Ω_p, Ω_s)**
 - (2) Design the analog filter $H_a(s)$**
 - (3) Design the digital filter $G(z)$ by applying bilinear transformation to $H_a(s)$**
- **Transformation can be used only to design digital filters with prescribed magnitude response with piecewise constant values**
- **Transformation does not preserve phase response of analog filter**

IIR Digital Filter Design Using Bilinear Transformation

•Example - Consider

$$H_a(s) = \frac{\Omega_c}{s + \Omega_c}$$

Applying bilinear transformation to the above we get the transfer function of a first-order digital lowpass Butterworth filter

$$G(z) = H_a(s) \Big|_{s=\frac{1-z^{-1}}{1+z^{-1}}} = \frac{\Omega_c(1+z^{-1})}{(1-z^{-1}) + \Omega_c(1+z^{-1})}$$

IIR Digital Filter Design Using Bilinear Transformation

- Rearranging terms we get

$$G(z) = \frac{1-\alpha}{2} \cdot \frac{1+z^{-1}}{1-\alpha z^{-1}}$$

where

$$\alpha = \frac{1-\Omega_c}{1+\Omega_c} = \frac{1-\tan(\omega_c/2)}{1+\tan(\omega_c/2)}$$

IIR Digital Filter Design Using Bilinear Transformation

• Example - Consider the second-order analog notch transfer function

$$H_a(s) = \frac{s^2 + \Omega_o^2}{s^2 + B s + \Omega_o^2}$$

for which $|H_a(j\Omega_0)| = 0$

$$|H_a(j0)| = |H_a(j\infty)| = 1$$

- Ω_0 is called the **notch frequency**
- If $|H_a(j\Omega_2)| = |H_a(j\Omega_1)| = 1/\sqrt{2}$ then $B = \Omega_2 - \Omega_1$ is the 3-dB notch bandwidth

IIR Digital Filter Design Using Bilinear Transformation

- **Then**

$$\begin{aligned} G(z) &= H_a(s) \Big|_{s=\frac{1-z^{-1}}{1+z^{-1}}} \\ &= \frac{(1+\Omega_o^2) - 2(1-\Omega_o^2)z^{-1} + (1+\Omega_o^2)z^{-2}}{(1+\Omega_o^2+B) - 2(1-\Omega_o^2)z^{-1} + (1+\Omega_o^2-B)z^{-2}} \\ &= \frac{1+\alpha}{2} \cdot \frac{1-2\beta z^{-1} + z^{-2}}{1-\beta(1+\alpha)z^{-1} + \alpha z^{-2}} \end{aligned}$$

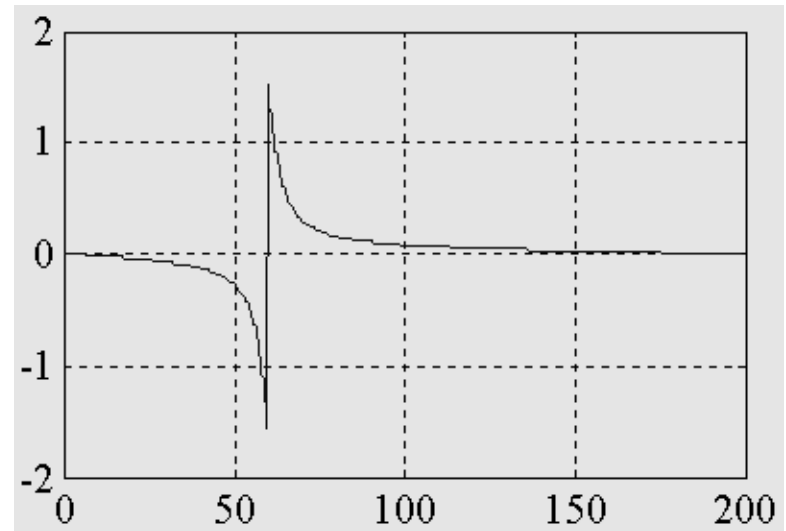
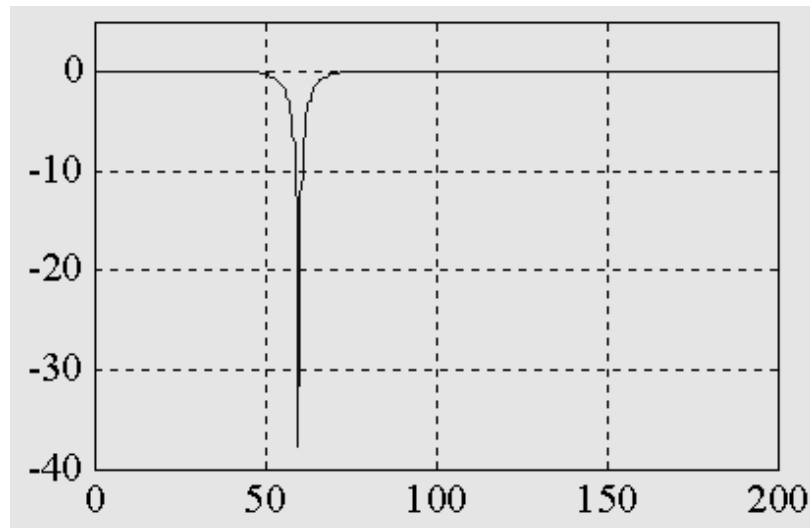
where

$$\alpha = \frac{1+\Omega_o^2-B}{1+\Omega_o^2+B} = \frac{1-\tan(B_w/2)}{1+\tan(B_w/2)}, \quad \beta = \frac{1-\Omega_o^2}{1+\Omega_o^2} = \cos \omega_o$$

Example

$$G(z) = \frac{0.954965 - 1.1226287z^{-1} + 0.954965z^{-2}}{1 - 1.1226287z^{-1} + 0.90993z^{-2}}$$

The gain and phase responses are shown below



9.3 Design of Lowpass IIR Digital Filters

Consider $G(z)$ with a maximally flat magnitude, and a pass ripple not exceeding 0.5dB,

$$\omega_P = 0.25\pi, \omega_S = 0.55\pi$$

and the minimum stopband attenuation 15dB.

$$20 \log_{10} |G(e^{j0.25\pi})| \geq -0.5[dB] \quad \dots(9.34a)$$

$$20 \log_{10} |G(e^{j0.55\pi})| \leq -15[dB] \quad \dots(9.34b)$$

1. The Bilinear Translation Method

(1) Prewarpping

$$\Omega_P = \tan\left(\frac{\omega_P}{2}\right) = \tan\left(\frac{0.25\pi}{2}\right) = 0.4142136$$

$$\Omega_S = \tan\left(\frac{\omega_S}{2}\right) = \tan\left(\frac{0.55\pi}{2}\right) = 1.1708496$$

(2) Design the parent analog filter $H_a(s)$

From the specifications we obtain

pass ripple 0.5dB  $\varepsilon^2 = 0.1220185$

stopband attenuation 15dB  $A^2 = 31.622777$

Using Eqs. (4.31), (4.32),

$$k = \frac{\Omega_s}{\Omega_p} = \frac{1.1708496}{0.4142136} = 2.8266809$$

$$k_1 = \frac{\varepsilon}{\sqrt{A^2 - 1}} = 15.841979$$

Using Eq. (4.35).

$$N = \frac{\log_{10}(1/k_1)}{\log_{10}(1/k)} = \frac{\log_{10}(15.841979)}{\log_{10}(2.8266814)} = 2.6586997$$

The least order of Butterworth LPF is $N = 3$.

Using Eq. (4.34a) we get the 3-dB frequency

$$\Omega_c = (\varepsilon)^{-1/N} \Omega_p = 1.419915(\Omega_p) = 0.588148$$

third-order normalized lowpass Butterworth transfer function

$$H_{an}(s) = \frac{1}{(s+1)(s^2+s+1)} = \frac{1}{1+2s+2s^2+s^3}$$

which has 3-dB frequency at $\Omega_c = 1$.

Therefore, the **parent** transfer function

$$H_a(s) = H_{an}\left(\frac{s}{\Omega_c}\right) = \frac{0.203451}{(s + 0.588148)(s^2 + 0.588148s + 0.345918)}$$

(3) Design the digital filter $G(z)$ by applying bilinear transformation to $H_A(s)$

$$G(z) = H_A(s) \Big|_{s = \frac{1-z^{-1}}{1+z^{-1}}}$$

$$G(z) =$$

$$\frac{0.0662272 (1 + z^{-1})^3}{(1 - 0.2593284 z^{-1})(1 - 0.6762858 z^{-1} + 0.3917468 z^{-2})}$$

.....(9.35)

Note: If this digital filter is used to process an analog signal, it has the equivalent passband-edge frequency at

$$f_p = \frac{\omega_P}{2\pi} \times f_T = 10 \text{ kHz}$$

when the sampling frequency is $f_T = 80 \text{ kHz}$.

2. The Impulse Invariance Method

The third-order lowpass Butterworth transfer function which has 3-dB frequency at Ω_c .

$$H_a(s) = H_{an}\left(\frac{s}{\Omega_c}\right) = \frac{1}{1 + 2\frac{s}{\Omega_c} + 2\left(\frac{s}{\Omega_c}\right)^2 + \left(\frac{s}{\Omega_c}\right)^3}$$

We can get the poles and represent it as

$$H_a(s) = \frac{\Omega_c^3}{(s + \Omega_c)(s - \Omega_c e^{j\frac{2\pi}{3}})(s - \Omega_c e^{-j\frac{2\pi}{3}})}$$

It is partial-fractional expressed as:

$$H_a(s) = \frac{\Omega_c}{s + \Omega_c} + \frac{-(\Omega_c / \sqrt{3})e^{j\pi/6}}{s + \Omega_c(1 - j\sqrt{3})/2} + \frac{-(\Omega_c / \sqrt{3})e^{-j\pi/6}}{s + \Omega_c(1 + j\sqrt{3})/2}$$

Then we get digital transfer function as:

$$G(z) = \frac{\omega_c}{1 - e^{-\omega_c} z^{-1}} + \frac{-(\omega_c / \sqrt{3})e^{j\pi/6}}{1 - e^{-\omega_c(1 - j\sqrt{3})/2} z^{-1}} + \frac{-(\omega_c / \sqrt{3})e^{-j\pi/6}}{1 - e^{-\omega_c(1 + j\sqrt{3})/2} z^{-1}}$$

$$\omega_c = \pi / 4$$

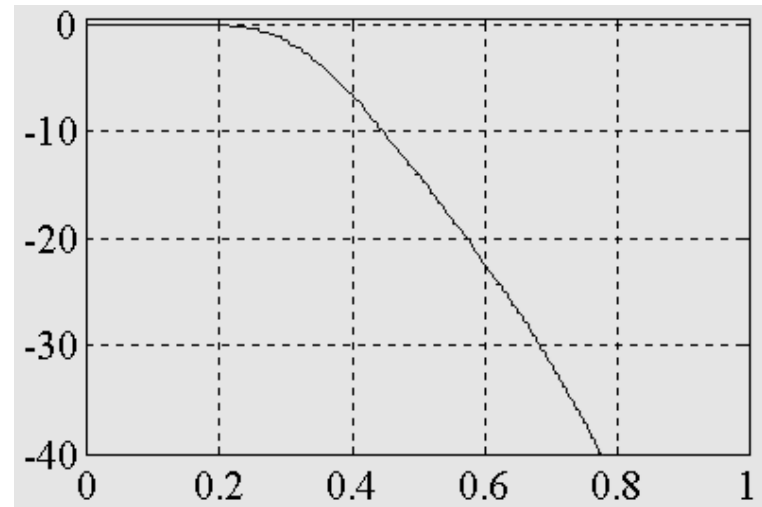
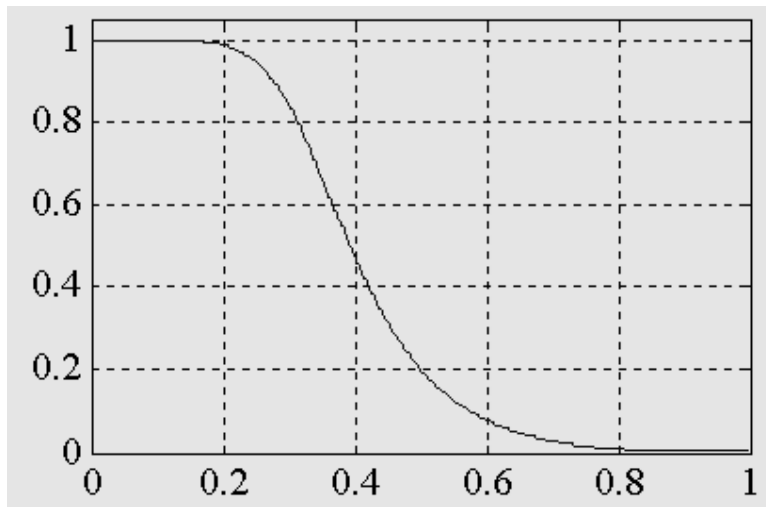
$$G(z) = \frac{0.785398}{1 - 0.455938z^{-1}} + \frac{-0.785398 + 0.604884z^{-1}}{1 - 1.0500756z^{-1} + 0.455938z^{-2}}$$

Finally,

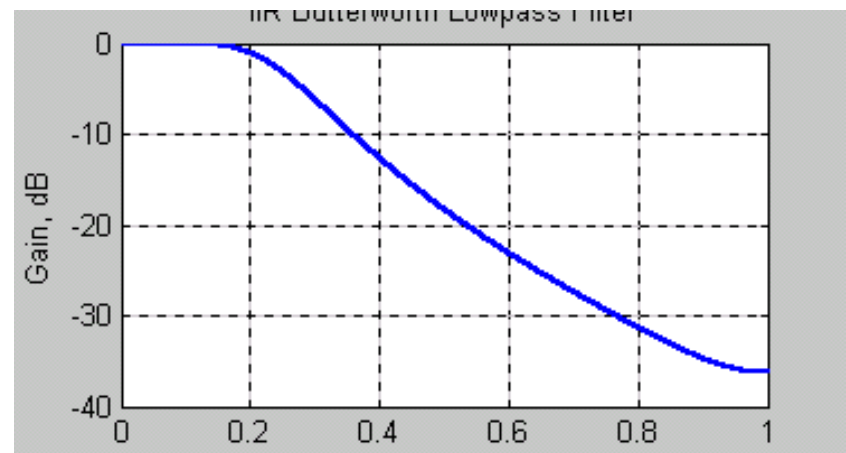
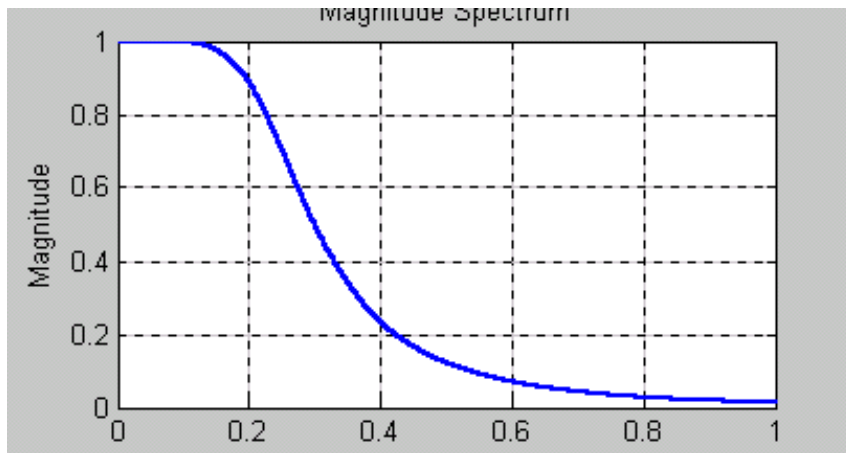
$$G(z) = \frac{0.13825z^{-1} + 0.08230z^{-2}}{1 - 1.50601z^{-1} + 0.93471z^{-2} - 0.20788z^{-3}}$$

Discussion:

Magnitude and gain responses of $G(z)$ derived by The Impulse Invariance Method and The Bilinear Translation Method are shown below:



The Bilinear Translation Method



The Impulse Invariance Method

9.4 Design of Highpass, Bandpass, and Bandstop IIR Digital Filters

Read and exercise by yourself !

9.5 Spectral Transformations of IIR Filters

Read Table 9.1 and exercise by yourself !

$$z = F(\hat{z}) \quad \dots(9.36)$$

$$G_D(\hat{z}) = G_L(F(\hat{z})) \quad \dots(9.37)$$

where $z^{-1} = F^{-1}(\hat{z}) = \frac{1}{F(\hat{z})}$ **Must be an allpass function.**

9.6 IIR Digital Filter Design Using MATLAB

Read and exercise by yourself!

Note: LPF design **directly** in digital form

1. Order Estimation

```
[N,Wn] = buttord(Wp, Ws, Rp, Rs);
```

2. Filter Design

```
[b,a] = butt(N,Wn);
```

```
% Getting:  $G(z) = B(z) / A(z)$ 
```

3. Other type filter

```
[b,a] = cheby1(N,Rp,Wn,'high');
```

```
%Wn=[W1, W2]
```

Note: Digital LPF design in **analog** form by **Bilinear Transform**

1. Order Estimation

```
[N,Wn] = ellipord(Wp,Ws,Rp,Rs,'s');
```

2. Filter Design

```
[bt,at] = ellip(N,Wn);
```

```
%Getting:  $H_a(z) = B_t(z) / A_t(z)$ 
```

3. **Bilinear** Transform

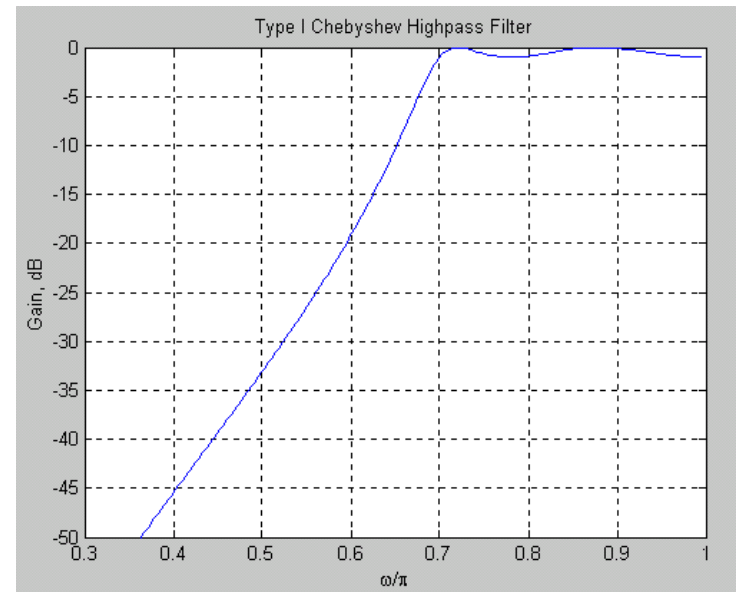
```
[num,den]= bilinear(b,a,0.5);
```

```
% 0.5 means  $T=2, F_s=0.5$ 
```

```
% Getting:  $G(z) = B(z) / A(z)$ 
```

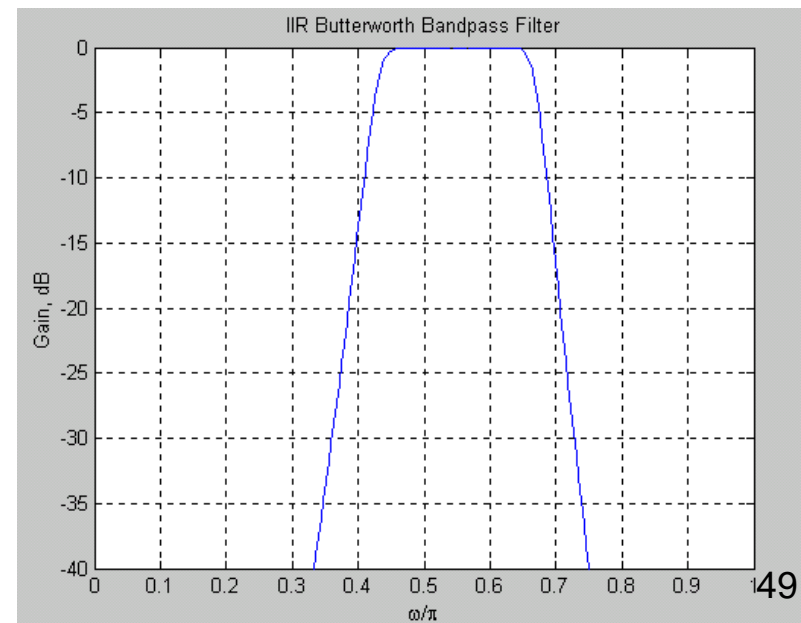

Example (HPF)

(Chebyshev 1)



Example (BPF)

($N = 6$, 12th-order)



§ 9 Analysis of Finite Word-length Effects

There are three types of finite word-length effect in DSP.

- (1) A/D Conversion noise**
- (2) Coefficient Quantization of Filter**
- (3) The Quantization of Arithmetic Operations**

Read 8.4~8.6 by yourself to understand the representation of **Fixed-Point number and arithmetic operations of binary data.**

§ 9.1 The Quantization Process and Errors

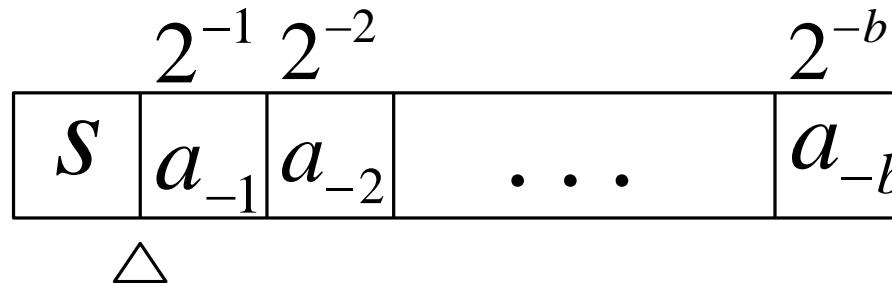


Fig. 9.2 A general $(b+1)$ -bit fixed-point fraction

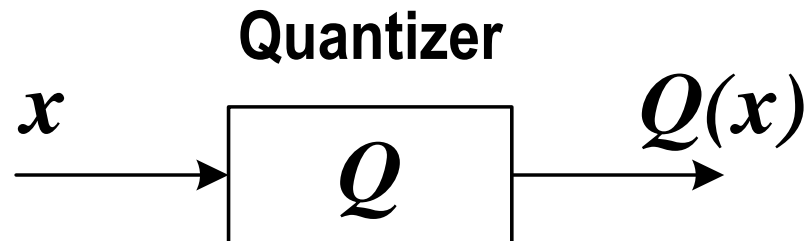


Fig. 9.3: The quantization process model

2^{-b} is called as ***quantization step***.

x : the original data

$Q(x)$: $(b+1)$ -bit (employed *truncation* or *rounding*)

§ 9.2 Quantization of Fixed-Point Numbers

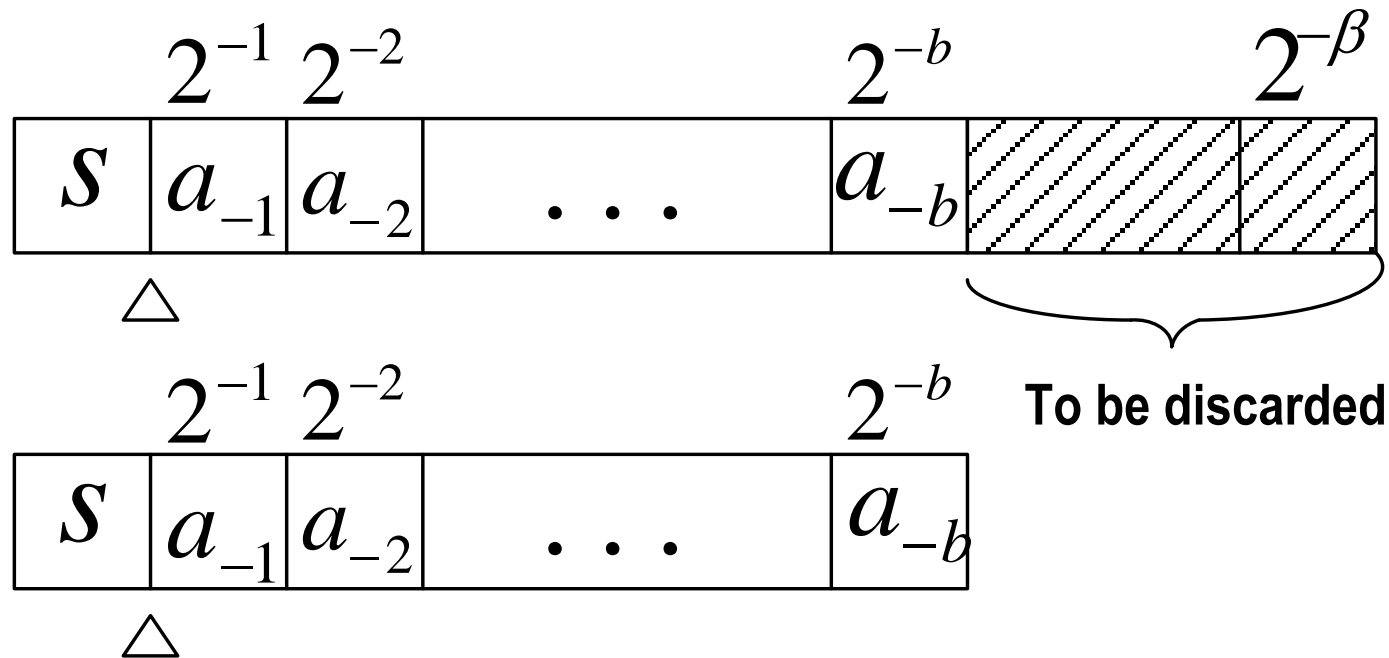
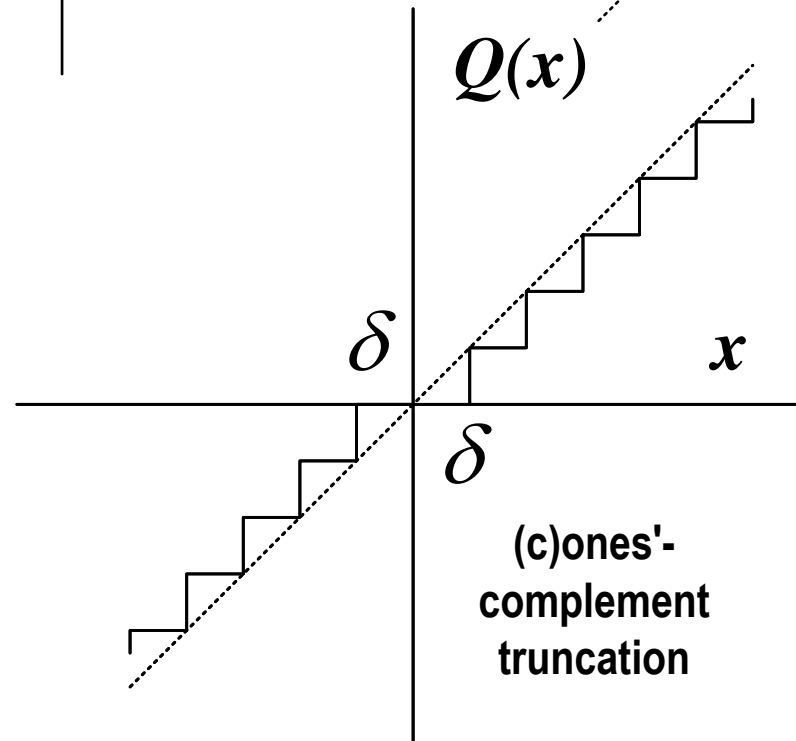
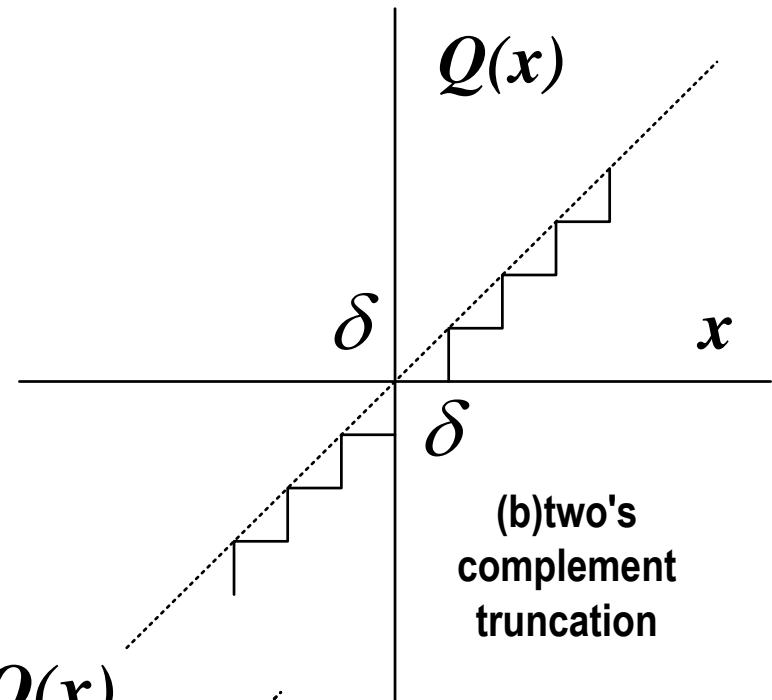
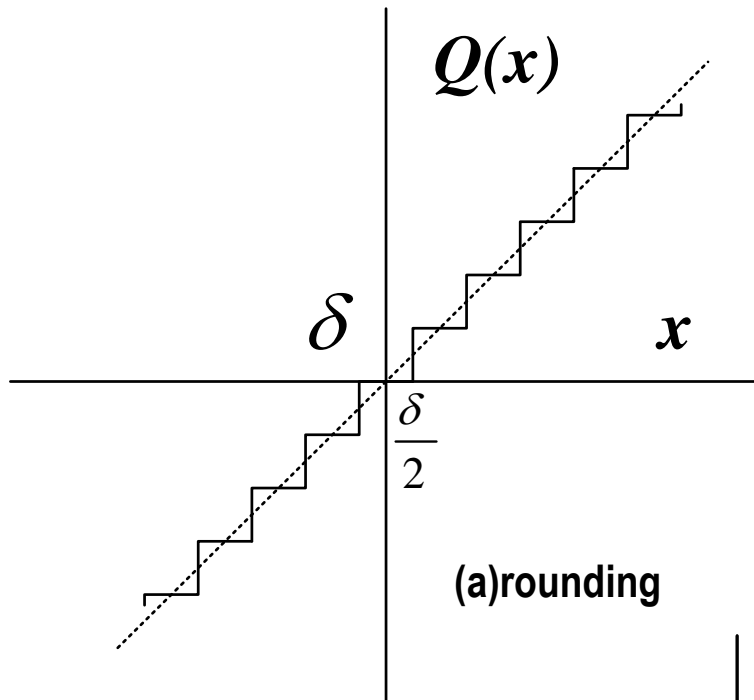


Fig 9.4 and 9.5 illustrate quantization process.



$$\delta = 2^{-b}$$

Table 9.1 Range of quantization error.

Rounding: $-\frac{1}{2}\delta \leq \varepsilon_r \leq \frac{1}{2}\delta$

Truncation $0 \leq \varepsilon_t \leq \delta$ **or** $-\delta \leq \varepsilon_t \leq 0$

§ 9.3 (9.5) A/D Conversion Noise

9.3.1 Quantization Noise Model

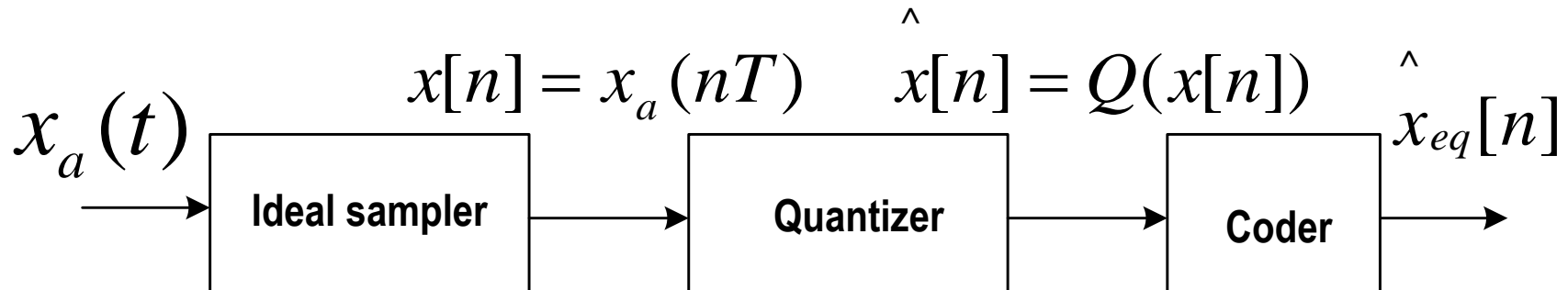
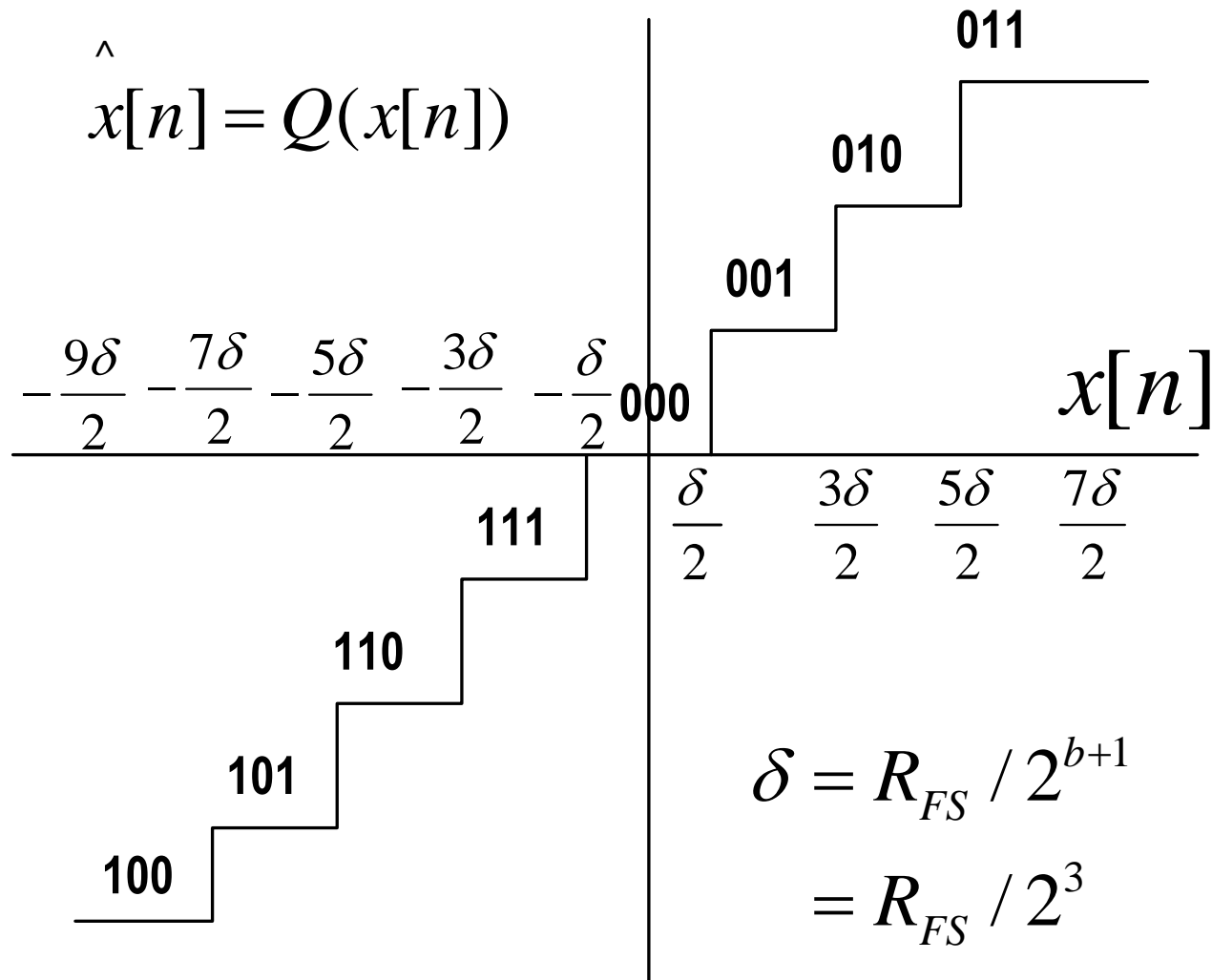


Fig 9.13: Model of a practical A/D conversion system

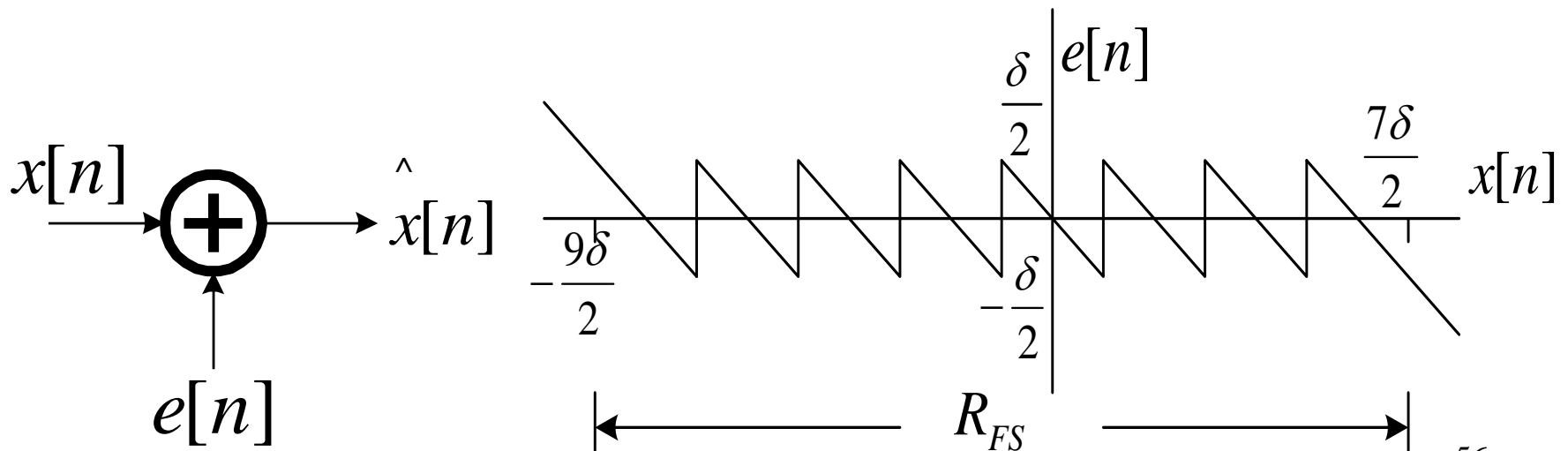
For example 3-bit bipolar A/D conversion with two's-complement representation (Fig 9.14).



$$-1 \leq \hat{x}_{eq}[n] < 1 \quad \hat{x}_{eq}[n] = \frac{2 \hat{x}[n]}{R_{FS}}$$

R_{FS} : *full-scale range*.

$$e[n] = Q(x[n]) - x[n] \quad \text{and} \quad -\frac{\delta}{2} < e[n] \leq \frac{\delta}{2}$$



The analysis of error is based on assumptions:

- (a) $e[n]$ is a wide-sense stationary random process, white noise, with uniformly distribution(as Fig 9.17)
- (b) $e[n]$ is uncorrelated with $x[n]$
- (c) $x[n]$ is a sample sequence of a stationary random process.

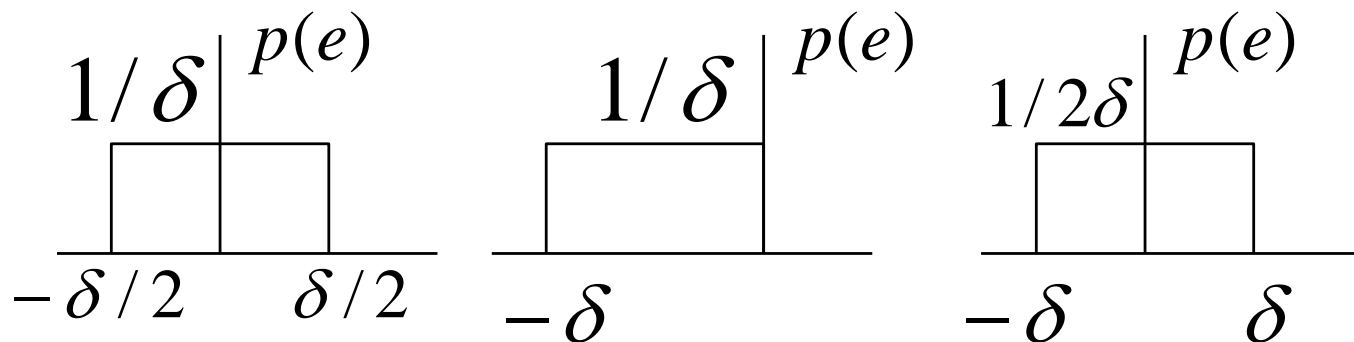


Fig 9.17

In the case of rounding error, the mean and variance are given by

$$m_e = E[e[n]] = \int_{-\infty}^{\infty} ep(e)de$$

$$m_e = \int_{-\delta/2}^{\delta/2} e \frac{1}{\delta} de = 0$$

$$\sigma_e^2 = E[(e[n] - m_e)^2] = \int_{-\infty}^{\infty} (e - m_e)^2 p(e)de$$

$$\sigma_e^2 = \delta^2 / 12 \quad \delta = 2^{-b}$$

The corresponding parameters for the two's-complement truncation are as follows:

$$m_e = \int_{-\delta}^0 e \frac{1}{\delta} de = -\delta / 2 \quad \sigma_e^2 = \delta^2 / 12$$

9.3.2 Signal-to-Quantization Noise Ratio

$$SNR_{A/D} = 10 \log_{10}(\sigma_x^2 / \sigma_e^2) (dB) \cdots (9.72)$$

where σ_x^2 is the input signal variance representing the signal power.

In the case of a bipolar $(b+1)$ -bit A/D converter, for rounding

Hence $\delta = 2^{-(b+1)} R_{FS} \quad \sigma_e^2 = 2^{-2b} (R_{FS})^2 / 48$

So,

$$\begin{aligned} SNR_{A/D} &= 10 \log_{10} (48 \sigma_x^2 / (2^{-2b} R_{FS}^2)) (dB) \\ &= 6.02b + 16.81 - 20 \log_{10} (R_{FS} / \sigma_x) (dB) \\ &\quad \dots(9.74) \end{aligned}$$

σ_x : the RMS value of input signal

This expression can be used to determine the minimum word-length of an A/D converter needed to meet a specified $SNR_{A/D}$

- **Note: $SNR_{A/D}$ increases by 6 dB for each added bit to the word-length**

- **Computed values of the SNR for various values of K are as given below:**

Table 9.3

	$b = 7$	$b = 9$	$b = 11$	$b = 13$	$b = 15$
$K = 4$	46.91	58.95	70.99	83.04	95.08
$K = 6$	43.39	55.43	67.47	79.51	91.56
$K = 8$	40.89	52.93	64.97	77.01	89.05

9.3.3 Effect of Input Scaling on SNR

$$SNR_{A/D} = 6.02b + 16.81 - 20 \log_{10}(K) + 20 \log_{10}(A)(dB)$$

...(9.77)

- For a given wordlength, the actual SNR depends on σ_x , the RMS value of the input signal amplitude and the full-scale range R_{FS} of the A/D converter
- Example - Determine the SNR in the digital equivalent of an analog sample $x[n]$ with a zero-mean Gaussian distribution using a $(b+1)$ -bit A/D converter having $R_{FS} = K\sigma_x$

$$SNR_{A/D} = 6.02b + 16.81 - 20 \log_{10}(K) \text{ (dB)}$$

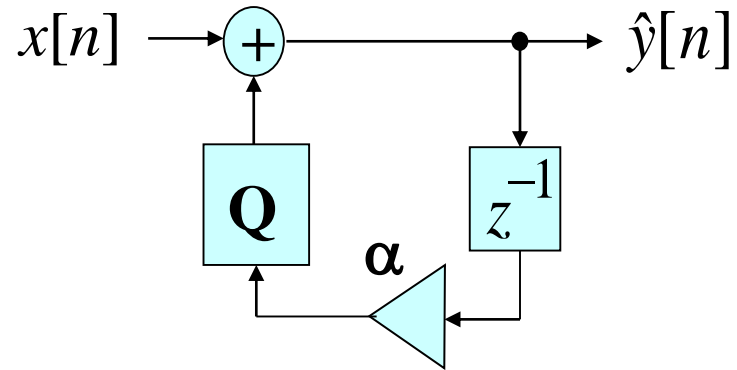
$$K = R_{FS} / \sigma_x \cdots (9.75)$$

Limit Cycles in IIR Digital Filters

- This type of instability usually results in an oscillatory periodic output called a **limit cycle**
- The system **remains** in this condition **until** an input of sufficiently large amplitude is applied to move the system into a more conventional operation
- Limit cycles occur in **IIR filters** due to the presence of **feedback**
- Such oscillations are absent in **FIR filters** as they do **not have any feedback** path

Limit Cycles in IIR Digital Filters

- Consider the first-order IIR filter as shown right



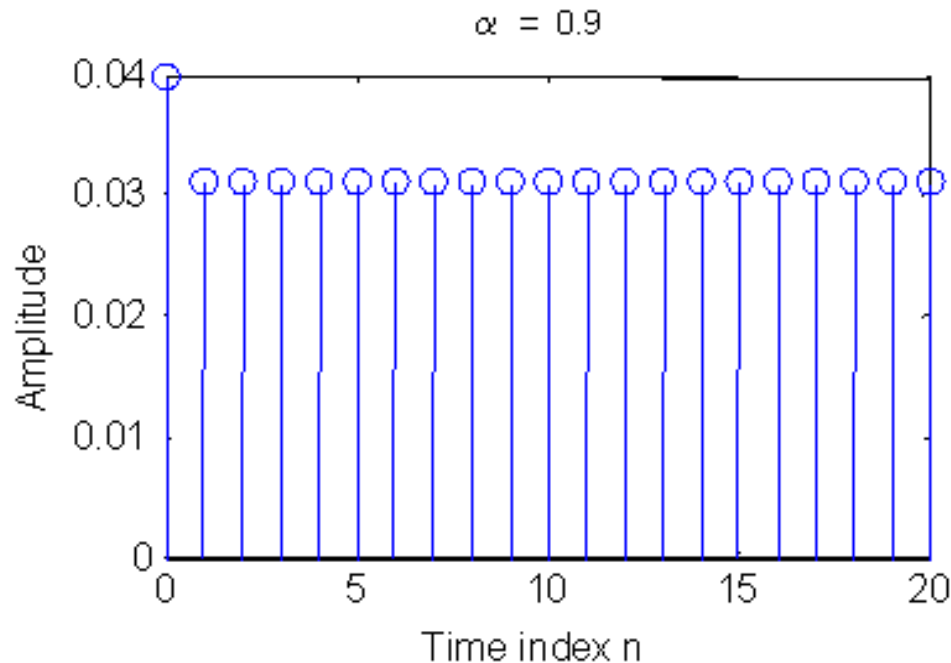
Assume the quantization operation to be rounding and the filter to be implemented with a signed 6-bit fractional arithmetic

- The nonlinear difference equation characterizing the filter is given by

$$\hat{y}[n] = Q(\alpha \cdot \hat{y}[n-1]) + x[n]$$

Limit Cycles in IIR Digital Filters

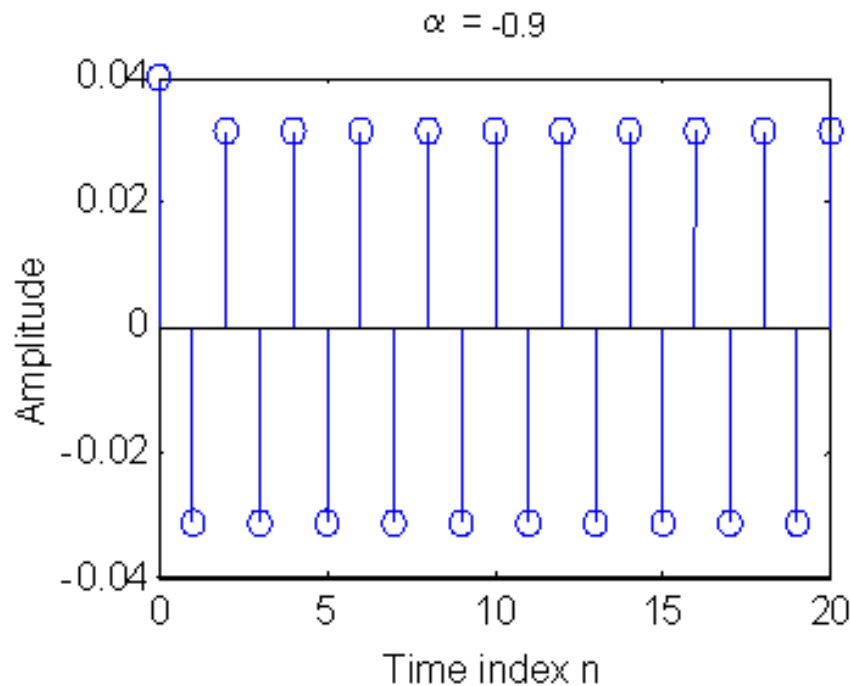
- For $x[n] = 0.03\delta[n]$, $\hat{y}[-1] = 0$, and $\alpha = 0.9$, the output of the filter is as shown below



- The limit cycle generated has a period of 1

Limit Cycles in IIR Digital Filters

- For $x[n] = 0.03\delta[n]$, $\hat{y}[-1] = 0$, and the output of the filter is as shown below



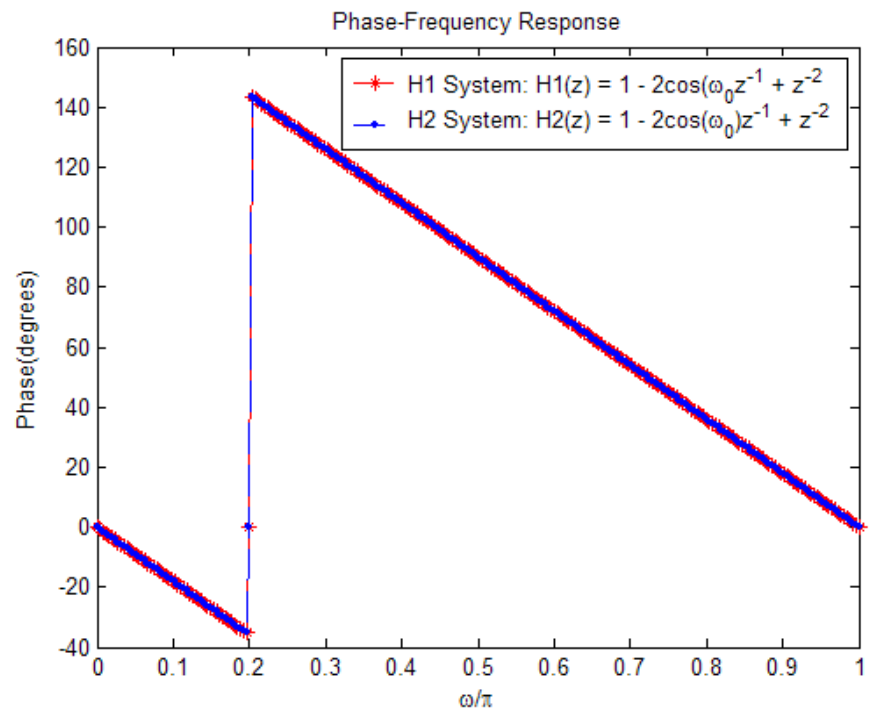
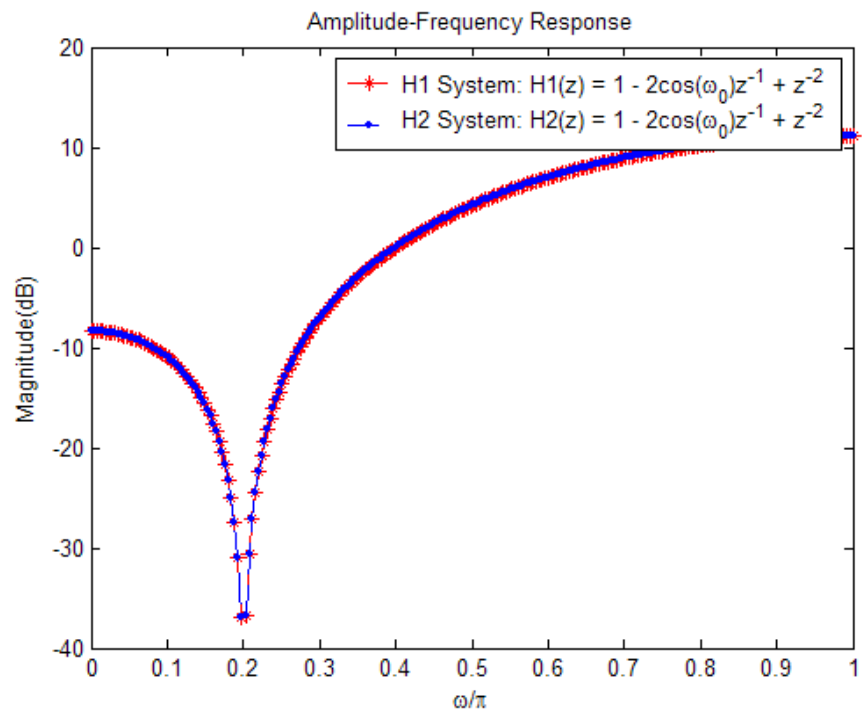
- The limit cycle generated has a period of 2

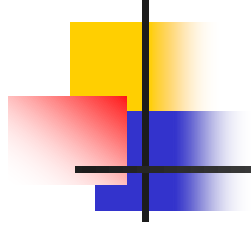
Coefficient Quantization of Filter

- **Notch filter design**
 - **No coefficient quantization error**
 - **No pole**

$$H_1(z) = 1 - 2 \cos(0.2\pi) z^{-1} + z^{-2}$$

$$H_2(z) = 1 - 2 \cos(0.2\pi) z^{-1} + z^{-2}$$





Thanks!

Any questions?