

# Overview

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## Part I Sampling

### 1-1 Sampling of Analog Signal

Consider a continuous-time signal  $x_a(t)$  and a discrete-time signal  $x[n]$  is obtained by periodic sampling.

Let  $T$  be the **sampling period**, then we define  $x[n] = x_a(nT)$

If  $x_a(t) = A \cos(2\pi Ft + \phi)$  and the signal is sampled at a rate of  $F_s = \frac{1}{T}$

then  $x[n] = A \cos(2\pi FTn + \phi) = A \cos(2\pi nf + \phi)$

#### Normalized Frequency

the signal with an analog frequency of  $F$  is sampled at a rate of  $F_s$

then the relationship between the **normalized frequency**  $f$  and the **analog frequency**  $F$  is

$$f = \frac{F}{F_s}$$

### 1-2 Sampling Theorem

- If maximum frequency of the signal is known to be  $F_{max}$ , the **minimum** sampling rate would be

$$\text{Nyquist Rate} = 2 \times F_{max}$$

- If the sampling rate is known to be  $F_s$ , the **maximum** frequency in the **signal** must not exceed

$$\text{Nyquist Rate} = \frac{1}{2} F_s$$

## Part II Discrete-Time Signals and Systems

### 2-1 Classification of Discrete-Time Signal and System

#### Energy and Power Signals

$$E = \sum_{-\infty}^{\infty} \|x[n]\|^2$$

$$P = \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^N \|x[n]\|^2$$

#### Periodic Signals

$$x(n+N) = x(n)$$

## Even and Odd Signals

$$\text{Even : } x[-n] = x[n]$$

$$\text{Odd : } x[-n] = -x[n]$$

## Systems

- **FIR: finite** impulse response system
- **IIR: infinite** impulse response system

$$x[n] \rightarrow \delta[n] \quad y[n] \rightarrow h[n]$$

## 2-2 Properties of Discrete-Time Signals

### Linear Property

$$x_1[n] \rightarrow y_1[n] \quad x_2[n] \rightarrow y_2[n]$$

$$ax_1[n] + bx_2[n] \longrightarrow ay_1[n] + by_2[n]$$

### Displacement Invariant Property

$$x[n] \rightarrow y[n]$$

$$x[n - n_0] \rightarrow y[n - n_0]$$

### Causality

The output of the system at some time instant only depends on the time before.

In other words,

$$h(n) = 0 \quad n < 0$$

### Stability

The input is bounded as well as the output is bounded

$$\|x[n]\| \leq M_x < \infty$$

$$\|y[n]\| \leq M_y < \infty$$

## 2-3 Convolution of Discrete-Time Systems

$$y[n] = x[n] * h[n]$$

$$= \sum_{k=-\infty}^{\infty} x[n]h[n - k]$$

## 2-4 Correlation Sequences

the **cross correlation** of two energy  $x(n)$  and  $y(n)$  is defined as

$$r_{xy}(l) = \sum_{n=-\infty}^{\infty} x(n)y(n-l)$$

## Properties

- $r_{yx}(l) = r_{xy}(-l)$
- $r_{xx}(l) = \sum_{n=-\infty}^{\infty} x(n)x(n-l)$
- $r_{xx}(l) = r_{xx}(-l)$
- $r_{xy}(l) = x(l) * y(-l)$

## Normalized Correlation Sequences

the correlation sequences are often normalized to the range  $[-1, 1]$

$$\rho_{xy}(l) = \frac{r_{xy}(l)}{\sqrt{r_{xx}(0)r_{yy}(0)}}$$

## Correlation of Periodic Sequences

$$r_{xy}(l) = \frac{1}{N} \sum_{n=0}^{N-1} x(n)y(n-l)$$

$$r_{xx}(l) = \frac{1}{N} \sum_{n=0}^{N-1} x(n)x(n-l)$$

## Detection of Hidden Periodicity

For a periodic signal  $s(n)$ , suppose that the signal is buried in noise. Let  $\omega(n)$  be the zero-mean white noise. What can we observe is the noisy signal  $x(n)$

$$x(n) = s(n) + \omega(n)$$

Calculate the autocorrelation sequence of  $x(n)$

$$r_{xx}(l) = r_{ss}(l) + r_{s\omega} + r_{\omega s} + r_{\omega\omega}(l)$$

Since  $\omega(n)$  is a zero-mean white noise signal

$$r_{s\omega}(l) = 0 \quad r_{\omega s}(l) = 0 \quad r_{s\omega}(l) = 0 \quad r_{\omega\omega}(l) = \sigma^2 \delta(l)$$

Then we get

$$r_{xx}(l) = r_{ss}(l) + \sigma^2 \delta(l)$$

# Part III The z-Transform

## 3-1 The Direct z-Transform

the z-transform of a general discrete-time signal  $x[n]$  is defined as

$$X(z) = \sum_{n=-\infty}^{\infty} x[n]z^{-n}$$

## 3-2 The Inverse z-Transform

$$X(z) = \sum_{i=1}^m \frac{A_i}{1 - a_i z^{-1}}$$

- ROC outside the pole:  $\frac{A_i}{1 - a_i z^{-1}} \leftrightarrow A_i a_i^n u[n]$
- ROC inside the pole:  $\frac{A_i}{1 - a_i z^{-1}} \leftrightarrow -A_i a_i^n u[-n - 1]$

### Power-series Expansion

$$X(z) = \sum_{n=-\infty}^{\infty} x[n] z^{-n}$$

The coefficients in this power series are the sequence values  $x[n]$  since  $\delta[n + n_0] \leftrightarrow z^{n_0}$

## 3-3 Properties of the z-Transform

### Linearity

$$ax_1[n] + bx_2[n] \longleftrightarrow aX_1(z) + bX_2(z)$$

with the ROC containing  $R_1 \cap R_2$

### Time Shifting

$$x[n - k] \longleftrightarrow z^{-k} X(z)$$

with ROC = R, except for the possible addition or deletion of the origin or infinity

For  $n_0 > 0$ , poles will be introduced at  $z = 0$

For  $n_0 < 0$ , zeros will be introduced at  $z = 0$

### Scaling in the z-Domain

$$a^n x[n] \longleftrightarrow X(a^{-1} z)$$

with ROC =  $|a|R$

### Time Reversal

$$x[-n] \longleftrightarrow X(z^{-1})$$

with ROC =  $\frac{1}{R}$

### Differentiation in the z-Domain

$$nx[n] \longleftrightarrow -z \frac{dX(z)}{dz}$$

with ROC = R

### Convolution Property

$$x_1[n] * x_2[n] \longleftrightarrow X_1(z) X_2(z)$$

### Initial-Value and Final-Value Theorems

if  $x[n]$  is a casual sequence, then

$$x[0] = \lim_{z \rightarrow \infty} X(z)$$

Time-Domain	z-Domain	ROC
$\delta[n]$	1	all z
$u[n]$	$\frac{1}{1-z^{-1}}$	$\ z\  > 1$
$a^n u[n]$	$\frac{1}{1-az^{-1}}$	$\ z\  > \ a\ $
$-a^n u[-n-1]$	$\frac{1}{1-az^{-1}}$	$\ z\  < \ a\ $
$na^n u[n]$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$\ z\  > \ a\ $
$-na^n u[-n-1]$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$\ z\  < \ a\ $
$\cos(\omega_0 n) u[n]$	$\frac{1-z^{-1} \cos(\omega_0)}{1-2z^{-1} \cos \omega_0 + z^{-2}}$	$\ z\  > 1$
$\sin(\omega_0 n) u[n]$	$\frac{z^{-1} \sin(\omega_0)}{1-2z^{-1} \cos \omega_0 + z^{-2}}$	$\ z\  > 1$
$a^n \cos(\omega_0 n) u[n]$	$\frac{1-az^{-1} \cos(\omega_0)}{1-2az^{-1} \cos \omega_0 + a^2 z^{-2}}$	$\ z\  > 1$
$\sin(\omega_0 n) u[n]$	$\frac{az^{-1} \sin(\omega_0)}{1-2az^{-1} \cos \omega_0 + a^2 z^{-2}}$	$\ z\  > 1$

### 3-4 Analysis of LTI System Using z-Transform

$$Y(z) = X(z)H(z)$$

where  $H(z)$  is the system function/transfer function

#### Causality

**Definition:** if the ROC of its system function is the exterior of a circle, including infinity

A system with rational system with rational system function  $H(z)$  is casual if and only if:

- the ROC is the exterior of a circle outside the outermost pole
- for  $H(z)$ , the order of numerator cannot be greater than the order of the denominator

#### Stability

**Definition:** the ROC of its system function  $H(z)$  includes the unit circle,  $|z| = 1$

A system with rational system function  $H(z)$  is stable if and only if all of the poles of  $H(z)$  lie inside the i=unit circle

### 3-5 The Inverse z-Transform Methods

- Inspection Method
  - using the properties and transform paris to get the inverse transform directly
- Power Series Method
  - has only few terms
  - can be found easily using Taylor series expansion/long division
- **Partial Fraction Expansion Method**
  - Simple poles

- Repeated poles

## Part IV The Discrete Fourier Transform

### 4-1 N-Point DFT

The formula for the DFT is given by

$$X(k) = \sum_{n=0}^{N-1} x(n)W_N^{nk} \quad W_N = e^{-\frac{2j\pi}{N}}$$

and also could be expressed as the matrix form

$$\mathbf{x}_N = \begin{bmatrix} x[0] \\ x[1] \\ \dots \\ x[N-1] \end{bmatrix} \quad \mathbf{W}_N = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W_N^1 & W_N^2 & \dots & W_N^{N-1} \\ 1 & W_N^2 & W_N^4 & \dots & W_N^{2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & W_N^{N-1} & W_N^{2(N-1)} & \dots & W_N^{(N-1)(N-1)} \end{bmatrix}$$

where the N-point DFT could be expressed in the matrix form as

$$\mathbf{X}_N = \mathbf{W}_N \mathbf{x}_N$$

### 4-2 Circular or Linear Convolution Using DFT

For example, do the circular convolution of the following two sequences

$$x_1[n] = \{2, 1, 2, 1\} \quad x_2[n] = \{1, 2, 3, 4\}$$

Then we can get  $x_3$  by

$$x_3[n] = \begin{bmatrix} 1 & 4 & 3 & 2 \\ 2 & 1 & 4 & 3 \\ 3 & 2 & 1 & 4 \\ 4 & 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 2 \\ 1 \end{bmatrix} = \{14, 16, 14, 16\}$$

in other way

$$X_1(k) = 2 + W_4^k + 2W_4^{2k} + W_4^{3k} \quad X_2(k) = 1 + 2W_4^k + 3W_4^{2k} + 4W_4^{3k}$$

then according to  $Y(k) = X_1(k)X_2(k)$

$$Y(k) = 2 + 5W_4^k + 10W_4^{2k} + 16W_4^{3k}$$

$$12W_4^{4k} + 11W_4^{5k} + 4W_4^{6k} = 14 + 16W_4^k + 14W_4^{2k} + 16W_4^{3k}$$

## Part V The Design of Filter

## 5-1 FIR Filter Design

Filter Type	Impulse Response
Lowpass	$\frac{\sin(\omega_c n)}{\pi n}$
Highpass	$\delta(n) - \frac{\sin(\omega_c n)}{\pi n}$
Bandpass	$\frac{\sin(\omega_2 n) - \sin(\omega_1 n)}{\pi n}$
Bandstop	$\frac{\sin(\omega_1 n) - \sin(\omega_2 n)}{\pi n}$

### Given

- Passband ripple:  $A_p = 20 \log(1 + \delta_p)$
- Passband edge:  $f_p = \omega_p / 2\pi$
- Stopband attenuation:  $A_s = -20 \log \delta_s$
- Stopband edge:  $f_s = \omega_s / 2\pi$
- Sample rate:  $f$

### Window Method Steps

1. Get  $\delta_p$  and  $\delta_s$  using  $A_p$  and  $A_s$
2. Calculate  $A = -20 \log(\min(\delta_s, \delta_p))$
3. Find the appropriate window type which  $A_w \geq A$
4. The transition bandwidth  $\Delta\omega = 2\pi(f_s - f_p)/f \geq \alpha\pi/N$  to get integer N
5. Get cut-off frequency  $\omega_c = 2\pi(f_s + f_p)/f$  then the ideal lowpass filter  $h_s(n) = \sin(\omega_c n)/(\pi n)$
6. For window function  $N_1 = N/2$  or  $(N - 1)/2$
7. The windowed filter equals the multiply of the window and the filter
8. Delay the impulse response by  $N_1$  to obtain the final filter

### Kaiser Window Steps

1. Get  $\delta_p$  and  $\delta_s$  using  $A_p$  and  $A_s$
2. Calculate  $A = -20 \log(\min(\delta_s, \delta_p))$
3. Determine  $\beta$  for the Kaiser window using  $A$
4. Compute the filter length  $N + 1 \geq (A - 7.95)/(2.285\Delta\omega)$
5. Get cut-off frequency  $\omega_c = 2\pi(f_s + f_p)/f$  then the ideal lowpass filter  $h_s(n) = \sin(\omega_c n)/(\pi n)$
6. For window function  $N_1 = N/2$  or  $(N - 1)/2$
7. The windowed filter equals the multiply of the window and the filter
8. Delay the impulse response by  $N_1$  to obtain the final filter

### Band Filter Design

1. Get  $\delta_p$  and  $\delta_s$  using  $A_p$  and  $A_s$
2. Calculate  $A = -20 \log(\min(\delta_s, \delta_p))$
3. Find the appropriate window type which  $A_w \geq A$
4. Calculate  $\Delta f = \min(f'_{p_1} - f_s, f'_s - f_{p_2})$
5. The transition bandwidth  $\Delta\omega = 2\pi\Delta f/f \geq \alpha\pi/N$  to get integer N
6. Get cut-off frequency  $\omega_{c_1} = 2\pi(f'_{p_1} - \Delta f/2)/f$  and  $\omega_{c_2} = 2\pi(f_{p_2} + \Delta f/2)/f$
7. Then the ideal bandpass filter has the impulse response  $h_d(n)$
8. For window function  $N_1 = N/2$  or  $(N - 1)/2$



9. The windowed filter equals the multiply of the window and the filter
10. Delay the impulse response by  $N_1$  to obtain the final filter

## 5-2 IIR Filter Design

$N$	Normalized Butterworth LPF $H(s) = 1/D(s) \quad \Omega_C = 1$
1	$D(s) = s + 1$
2	$D(s) = s^2 + \sqrt{2}s + 1$
3	$D(s) = s^3 + 2s^2 + 2s + 1$
4	$D(s) = (s^2 + 0.7653s + 1)(s^2 + 1.6180s + 1)$
5	$D(s) = (s + 1)(s^2 + 0.618s + 1)(s^2 + 1.618s + 1)$

### Given

- Passband ripple:  $A_p = 20 \log(1 + \delta_p)$
- Passband edge:  $f_p = \omega_p/2\pi$
- Stopband attenuation:  $A_s = -20 \log \delta_s$
- Stopband edge:  $f_s = \omega_s/2\pi$
- Sample rate:  $f$

### Butterworth Approximation Steps

1. Specify the filter requirements and get  $\Omega_p$  and  $\Omega_s$  by equation  $\omega/T$

using  $T = 1$  for impulse invariant method

2. From  $\|H(j\omega)\|^2 = (1 + (\Omega/\Omega_c)^{2N})^{-1} = (1 + \varepsilon^2(\Omega/\Omega_p)^{2N})^{-1}$
3. we get  $\|H(j\Omega_p)\| = (1 - \delta_p)^2 = 10^{-\frac{A_p}{10}}$
4. and  $\|H(j\Omega_s)\| = \delta_s^2 = 10^{-\frac{A_s}{10}}$  to solve for  $N$  and  $\Omega_c$
5. Find the transfer function of the analog filter from the table and get roots by partial fraction expression
6. Find the impulse response  $h(t)$  and transfer function  $H(z)$

### Bilinear Transform Steps

1. Get  $\omega_{p,s} = 2\pi(f_{p,s}/f)$  then use bilinear transform to get  $\Omega_{p,s} = \frac{2}{T} \tan(\omega_{p,s}/2)$

using  $T = 1$  for impulse invariant method

2. Using  $\|H(j\omega)\|^2 = (1 + (\Omega/\Omega_c)^{2N})^{-1} = (1 + \varepsilon^2(\Omega/\Omega_p)^{2N})^{-1}$
3. We get  $\varepsilon^2 \leq 10^{-\frac{A_p}{10}} - 1$
4. and  $(\Omega_s/\Omega_p)^{2N} \geq \frac{10^{\frac{A_s}{10}} - 1}{\varepsilon^2}$  solve for  $\varepsilon^2$  and  $N$
5. The analog filter is mapped into the desired digital filter  $H(z) = H_a(s)|_{s=\frac{2}{T} \frac{1-z^{-1}}{1+z^{-1}}}$ , where  $H_a(s) = \frac{1}{s+1}$

Type of Transform	Transformation	Parameters (Prototype low pass filter has band edge frequency $\omega_p$ )
Lowpass	$z^{-1} \rightarrow \frac{z^{-1} - a}{1 - az^{-1}}$	$\omega'_p$ = band edge frequency of newfilter $a = \frac{\sin((\omega_p - \omega'_p)/2)}{\sin((\omega_p + \omega'_p)/2)}$
Highpass	$z^{-1} \rightarrow \frac{z^{-1} + a}{1 + az^{-1}}$	$\omega'_p$ = band edge frequency of newfilter $a = \frac{\cos((\omega_p + \omega'_p)/2)}{\cos((\omega_p - \omega'_p)/2)}$
Bandpass	$z^{-1} \rightarrow -\frac{z^{-2} - a_1 z^{-1} + a_2}{a_2 z^{-2} - a_1 z^{-1} + 1}$	$\omega_l$ = lower band edge frequency $\omega_u$ = upper band edge frequency $a_1 = 2\alpha K / (K + 1)$ $a_2 = (K - 1) / (K + 1)$ $\alpha = \frac{\cos((\omega_u + \omega_l)/2)}{\cos((\omega_u - \omega_l)/2)}$ $K = \cot[(\omega_u - \omega_l)/2] \tan(\omega_p/2)$
Bandstop	$z^{-1} \rightarrow -\frac{z^{-2} - a_1 z^{-1} + a_2}{a_2 z^{-2} - a_1 z^{-1} + 1}$	$\omega_l$ = lower band edge frequency $\omega_u$ = upper band edge frequency $a_1 = 2\alpha / (K + 1)$ $a_2 = (1 - K) / (1 + K)$ $\alpha = \frac{\cos((\omega_u + \omega_l)/2)}{\cos((\omega_u - \omega_l)/2)}$ $K = \tan[(\omega_u - \omega_l)/2] \tan(\omega_p/2)$