

Lecture 10

**z-Transform  
&  
Sampling and Reconstruction  
from Samples**

School of ECE  
Center for Signal and Information Processing  
Georgia Institute of Technology

Overview of Lecture

- Repeated Roots in PFE
- z-transform Theorems
- System Function of a DE
- Causality, stability and z-transform
- Chapter 4--Sampling
  - Bandlimited signals
  - Sampling a sine wave
  - Ideal Continuous-to-Discrete conversion (sampling)
  - Derivation of frequency-domain formula
  - Interpretation of frequency-domain formula
    - Oversampling
    - Aliasing distortion

Repeated Roots in PFE

Suppose  $H(z)$  has a pole of order  $s$  at  $z=d_i$  and all the other poles are first order:

$$H(z) = \underbrace{\left[ \sum_{r=0}^{(M-N)} B_r z^{-r} \right]}_{\text{only if } M \geq N} + \sum_{k=1}^N \frac{A_k}{1 - d_k z^{-1}} + \sum_{m=1}^s \frac{C_m}{(1 - d_i z^{-1})^m}$$

• The coefficients  $A_k$  and  $B_r$  are found as before.  $C_m$  is found by:

$$C_m = \frac{1}{(s-m)!(-d_i)^{s-m}} \left\{ \frac{d^{s-m}}{dw^{s-m}} [(1 - d_i w)^s H(w^{-1})] \right\}_{w=d_i^{-1}}$$

for  $m = 1, \dots, s$  →  $(s-m)^{\text{th}}$  order derivative

z-Transform Theorems

• The delay or shift property:

$$x[n - n_0] \Leftrightarrow z^{-n_0} X(z)$$

• The convolution property:

$$y[n] = x[n] * h[n] \Leftrightarrow Y(z) = H(z)X(z)$$

• Example:

$$y[n] = x[n] * \delta[n - n_0] \Leftrightarrow Y(z) = z^{-n_0} X(z)$$

## An IIR System

- Difference equation:

$$y[n] = ay[n-1] + x[n] \Leftrightarrow Y(z) = az^{-1}Y(z) + X(z)$$

- System function:

$$H(z) = \frac{Y(z)}{X(z)} = \frac{1}{1 - az^{-1}} = \frac{z}{z - a} \quad |z| > |a| \quad \leftarrow \text{ROC if causal}$$

- Frequency response:

$$H(e^{j\omega}) = H(z)|_{z=e^{j\omega}} = \frac{1}{1 - ae^{-j\omega}} \quad \leftarrow \text{Assumes } |a| < 1$$

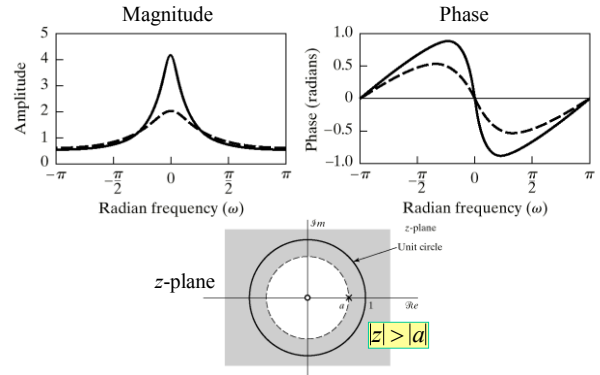
- Impulse response:

$$h[n] = a^n u[n]$$

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## IIR Frequency Response



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## System Function of a DE

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

$$\sum_{k=0}^N a_k Y(z) z^{-k} = \sum_{k=0}^M b_k X(z) z^{-k}$$

$$\left( \sum_{k=0}^N a_k z^{-k} \right) Y(z) = \left( \sum_{k=0}^M b_k z^{-k} \right) X(z)$$

$$H(z) = \frac{Y(z)}{X(z)} = \frac{\left( \sum_{k=0}^M b_k z^{-k} \right)}{\left( \sum_{k=0}^N a_k z^{-k} \right)}$$

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## Frequency Response of a DE

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

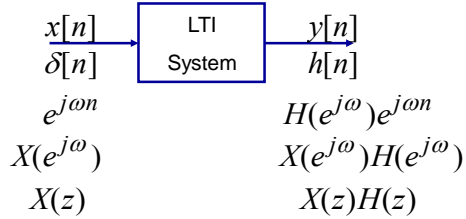
$$H(e^{j\omega}) = H(z)|_{z=e^{j\omega}} = \frac{\left( \sum_{k=0}^M b_k e^{-j\omega k} \right)}{\left( \sum_{k=0}^N a_k e^{-j\omega k} \right)}$$

ROC must Contain the Unit circle

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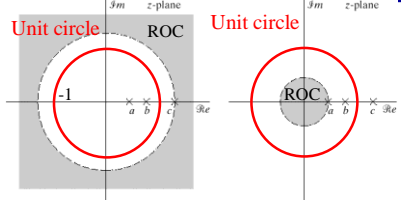
## LTI Systems



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

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

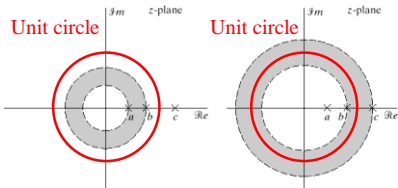
## Stability, Causality, ROC



- |   |  |
|---|--|
| <ul style="list-style-type: none"> <li>• May be Causal</li> <li>• Not stable</li> </ul> | <ul style="list-style-type: none"> <li>• Not Causal</li> <li>• Not stable</li> </ul> |
|---|--|

## Stability, Causality, ROC

- |  |  |
|--|--|
| <ul style="list-style-type: none"> <li>• Not Causal</li> <li>• Not stable</li> </ul> | <ul style="list-style-type: none"> <li>• Not Causal</li> <li>• Stable</li> </ul> |
|--|--|



## ROC of Z-Transform and Causality

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## ROC of Z-Transform and Causality

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## Digital Processing of Analog Signals



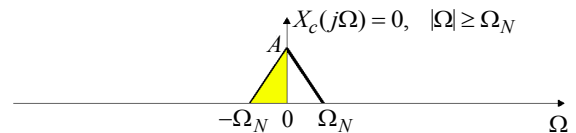
- **A-to-D conversion:** sampling and quantization
- **Numerical algorithm:** convolution, difference equations, DFT, LPC
  - Implemented on DSP chips, computers or ASICs with finite-precision arithmetic
- **D-to-A conversion:** quantization and filtering

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## Bandlimited Signals

- A bandlimited continuous-time (analog) signal has a Fourier transform that is zero for all frequencies above some highest frequency (called the *Nyquist* frequency  $\Omega_N$ ).



- Simple example is a sinusoidal signal

$$x_c(t) = A \cos(\Omega_0 t + \phi) \quad \Omega_N = \Omega_0 + \varepsilon$$

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## Sampling a Sinusoidal Signal



$$x_1(t) = A \cos(\Omega_0 t + \phi)$$

$$x_1[n] = A \cos(\Omega_0 nT + \phi) = A \cos(\omega_0 n + \phi) \quad \omega_0 = \Omega_0 T$$

$$x_2(t) = A \cos[(\Omega_0 + 2\pi/T)t + \phi]$$

$$\begin{aligned} x_2[n] &= A \cos[(\Omega_0 + 2\pi/T)nT + \phi] \\ &= A \cos(\Omega_0 nT + 2\pi n + \phi) = A \cos(\Omega_0 nT + \phi) \end{aligned}$$

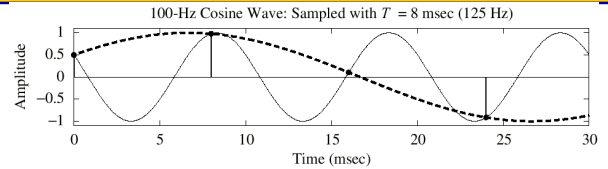
$$x_3(t) = A \cos[(2\pi/T - \Omega_0)t - \phi]$$

$$\begin{aligned} x_3[n] &= A \cos[(2\pi/T - \Omega_0)nT - \phi] \\ &= A \cos(2\pi n - \Omega_0 nT - \phi) = A \cos(\Omega_0 nT + \phi) \end{aligned}$$

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## Aliasing and Reconstruction



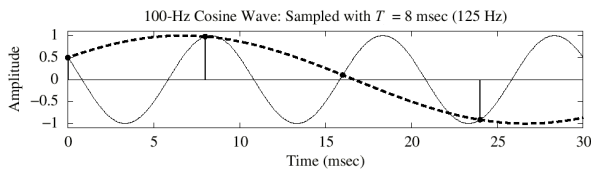
- Each of the sinusoidal components of a signal has an **infinite** number of aliases.
- The ideal D-to-C converter reconstructs the **lowest** frequency alias of each sinusoidal component. This is why we need to constrain our sampling rate so that

$$\frac{2\pi}{T} \geq 2\Omega_N$$

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## The Sampling Theorem



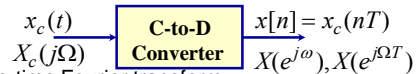
A bandlimited signal with Nyquist (highest) frequency  $\Omega_N$  can be reconstructed exactly from samples taken with sampling frequency

$$\frac{2\pi}{T} \geq 2\Omega_N$$

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## Sampling (C-to-D Conversion)



- Discrete-time Fourier transform:

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

- Frequency-domain relation:

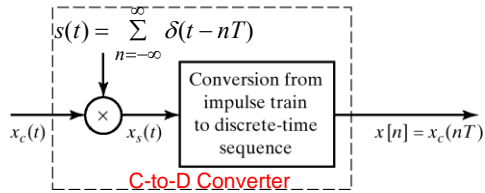
$$X(e^{j\Omega T}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega T n} = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

- Sampling frequency:  $\Omega_s = 2\pi/T$
- Normalized frequency:  $\omega = \Omega T$

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### Derivation of Basic FT Formula - I



$$x_s(t) = x_c(t) \cdot \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} x_c(t) \delta(t - nT)$$

$$x_s(t) = \sum_{n=-\infty}^{\infty} x_c(nT) \delta(t - nT) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT)$$

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### Derivation of Basic FT Formula - II

$$x_s(t) = x_c(t) \cdot s(t) = x_c(t) \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT) \Leftrightarrow S(j\Omega) = \sum_{k=-\infty}^{\infty} \frac{2\pi}{T} \delta(\Omega - k \frac{2\pi}{T})$$

$$X_s(j\Omega) = \frac{1}{2\pi} X_c(j\Omega) * S(j\Omega)$$

$$X_s(j\Omega) = \frac{1}{2\pi} X_c(j\Omega) * \sum_{k=-\infty}^{\infty} \frac{2\pi}{T} \delta(\Omega - k \frac{2\pi}{T})$$

$$X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(\Omega - k \frac{2\pi}{T})$$

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### Derivation of Basic FT Formula - III

$$x_s(t) = x_c(t) \sum_{n=-\infty}^{\infty} \delta(t - nT) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT)$$

$$X_s(j\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n T}$$

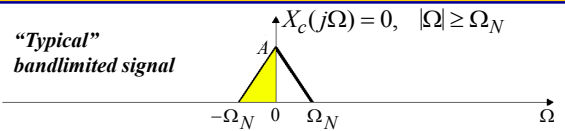
$$X_s(j\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j(\Omega T)n} = X(e^{j\Omega T})$$

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

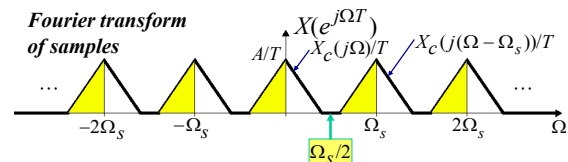
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### Oversampling



$$X(e^{j\Omega T}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s)) \quad \Omega_s = 2\pi/T$$

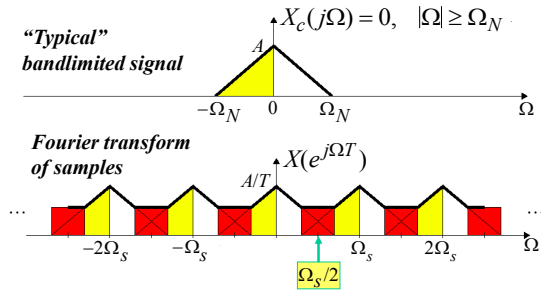


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## Undersampling (Aliasing Distortion)

- If  $\Omega_s < 2\Omega_N$ , the copies of  $X_c(j\Omega)$  overlap, and we have **aliasing distortion**.



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## Sampling Theorem

$$\begin{array}{ccc} x_c(t) & \xrightarrow{\text{C-to-D}} & x[n] = x_c(nT) \\ X_c(j\Omega) & \xrightarrow{\text{Converter}} & X(e^{j\omega}), X(e^{j\Omega T}) \end{array}$$

- Frequency-domain representation

$$X(e^{j\Omega T}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega T n} = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

- Sampling theorem justification:

If  $X_c(j\Omega) = 0, |\Omega| \geq \Omega_N$  and  $\Omega_s/2 = \pi/T \geq \Omega_N$ ,

- Therefore we should be able to recover  $x_c(t)$ !

$$\text{then } X(e^{j\Omega T}) = \frac{1}{T} X_c(j\Omega), \quad |\Omega| \leq \pi/T = \Omega_s/2$$

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