

#### Module IA

# Review of Discrete-Time Signal Processing



#### Overview

- Linearity
- Time-Invariance
- Impulse Response/Convolution
- Stability
- Fourier transform
- z-transform
- Sampling Theorem
- Upsampling/Downsampling
- Difference Equations



## Linearity

 A discrete-time system is *linear* if superposition holds. Define a system T such that

$$y(n) = \mathcal{T}[x(n)]$$

T is linear if

$$\mathcal{T}[a_1x_1(n) + a_2x_2(n)] = a_1\mathcal{T}[x_1(n)] + a_2\mathcal{T}[x_2(n)]$$



#### Time-Invariance

 A discrete-time system is time-invariant if the system response does not depend on the time at which the input is applied. If

$$y(n) = T[x(n)]$$

then

$$y(n-n_0) = T[x(n-n_0)]$$





## Example

The discrete-time system

$$y(n) = [x(n)]^2$$

is time-invariant but non-linear since

$$T[x_1(n) + x_2(n)] = [x_1(n)]^2 + [x_2(n)]^2 + 2x_1(n)x_2(n)$$
  
and

$$\mathcal{T}[x_1(n)] + \mathcal{T}[x_2(n)] = [x_1(n)]^2 + [x_2(n)]^2$$



# Impulse Response/Convolution

 A discrete-time system that is both *linear* and *time-invariant* (LTI) can be completely characterized by its response to an impulse input. Let

$$h(n) = \mathcal{T}[\delta(n)]$$
 impulse response

$$y(n) = \sum_{k=-\infty}^{\infty} x(k)h(n-k) = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$$

convolution sum



## Computing the Convolution Sum

- Reflect h(k) about the origin to obtain h(-k)
- Translate h(-k) by n samples to form h(n-k)
- Multiply sample by sample with x(k) for the selected value of n
- Sum the result over all k
- Repeat for all other values of n



## **Stability**

 Every bounded input must produce a bounded output

$$|x(n)| \le B_X \implies |y(n)| \le B_Y$$

 A discrete-time LTI system is stable in the bounded-input bound-output (BIBO) sense if

$$\sum_{n=-\infty}^{\infty} |h(n)| < \infty$$



## Example

The discrete-time accumulator with impulse response

$$h(n) = \begin{cases} 1, & \text{if } n \ge 0 \\ 0, & \text{if } n < 0 \end{cases}$$

is not BIBO stable since

$$\sum_{n=-\infty}^{\infty} |h(n)|$$
 is unbounded



#### **Fourier Transform**

 The Fourier transform of the discrete sequence x(n) is defined by

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

with 
$$0 \le \omega \le 2\pi$$



## Fourier Transform Properties (1 of 2)

$$x(n-n_0) \stackrel{\mathcal{F}}{\longleftrightarrow} e^{-j\omega n_0} X\left(e^{j\omega}\right)$$
 Time shift

$$e^{j\omega_0 n} x(n) \stackrel{\mathcal{F}}{\longleftrightarrow} X\left(e^{j(\omega-\omega_0)}\right)$$
 Frequency modulation

$$x(n) * y(n) \stackrel{\mathcal{F}}{\longleftrightarrow} X\left(e^{j\omega}\right) Y\left(e^{j\omega}\right)$$
 Convolution



## Fourier Transform Properties (2 of 2)

$$x(n)w(n) \stackrel{\mathcal{F}}{\longleftrightarrow} \frac{1}{2\pi} \int_{-\pi}^{\pi} X\left(e^{j\theta}\right) W\left(e^{j(\omega-\theta)}\right) d\theta$$
 Windowing

$$\sum_{n=-\infty}^{\infty} |x(n)|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega$$

Parseval's Theorem



#### Some Fourier Transform Pairs (1 of 2)

$$\delta(n) \stackrel{\mathcal{F}}{\longleftrightarrow} 1$$

$$\delta(n-n_0) \stackrel{\mathcal{F}}{\longleftrightarrow} e^{-j\omega n_0}$$

$$u(n) \stackrel{\mathcal{F}}{\longleftrightarrow} \frac{1}{1 - e^{-j\omega}} + \sum_{k = -\infty}^{\infty} \pi \delta(\omega + 2\pi k)$$
unit step

$$\frac{\sin \omega_c n}{\pi n} \stackrel{\mathcal{F}}{\longleftrightarrow} X\left(e^{j\omega}\right) = \begin{cases} 1, & \text{for } |\omega| \leq \omega_c \\ 0, & \text{for } \omega_c < |\omega| \leq \pi \end{cases}$$



#### Some Fourier Transform Pairs (2 of 2)

$$x(n) = \begin{cases} 1, & 0 \le n \le M \\ 0, & otherwise \end{cases} \xrightarrow{\mathcal{F}} \frac{\sin(\omega(M+1)/2)}{\sin(\omega/2)} e^{-j\omega M/2}$$

$$e^{j\omega_0 n} \stackrel{\mathcal{F}}{\longleftrightarrow} \sum_{k=-\infty}^{\infty} \pi \delta(\omega - \omega_0 + 2\pi k)$$