

# **CHAPTER 4:**

# **DISCRETE-TIME FOURIER TRANSFORM**

## **(DTFT)**

**Lecture 6:** Discrete-Time Fourier Transform (DTFT)

**Lecture 7:** Digital signal spectra

**Lecture 8:** Frequency response and filter

# Lecture 6

## Discrete-Time Fourier Transform (DTFT)

- **Outline:**

1. DTFT basics
2. Inversion of DTFT
3. Properties of DTFT

# DTFT formula

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

- Continuous function in frequency and periodic with period of  $2\pi$
- Gives the complex frequency spectrum of DT signal
- Not all DTFT is converge

# Convergence of the DTFT

We always have:

$$\left| \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n} \right| \leq \sum_{n=-\infty}^{\infty} |x[n] e^{-j\Omega n}|$$
$$\leq \sum_{n=-\infty}^{\infty} |x[n]| |e^{-j\Omega n}|$$
$$= \sum_{n=-\infty}^{\infty} |x[n]|$$

→ DTFT exists when:

$$\boxed{\sum_{n=-\infty}^{\infty} |x[n]| < \infty}$$

# From ZT to DTFT

Recall ZT of  $x[n]$ :

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

Evaluating  $X(z)$  on the unit circle (if the unit circle is in the ROC of  $X(z)$ ):

$$X(z) \Big|_{z=e^{j\Omega}} = \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega n} = X(\Omega)$$

# From ZT to DTFT

DTFT is the Z-transform of  $x[n]$  evaluated on the **unit circle**

$$X(\Omega) = X(z) \Big|_{z=e^{j\Omega}}$$

If the ROC of the ZT contains the unit circle, we can get the DTFT from the ZT by substitution  **$z = e^{j\Omega}$**

# Example of calculating DTFT

Find DTFT of  $x(n)$  where  $x[n] = a^n u[n]$

If  $|a| \geq 1$ , DTFT does not exist

If  $|a| < 1$ :

$$X(\Omega) = \sum_{n=0}^{\infty} a^n e^{-j\Omega n} = \sum_{n=0}^{\infty} (ae^{-j\Omega})^n = \frac{1}{1 - ae^{-j\Omega}} = \frac{e^{j\Omega}}{e^{j\Omega} - a}$$

$$X(z) = \frac{z}{z-a}, ROC : |z| > |a| \Rightarrow X(\Omega) = \frac{e^{j\Omega}}{e^{j\Omega} - a}, |a| < 1$$

# Lecture 6

## DTFT

- **Outline:**

1. DTFT basics
2. Inversion of DTFT
3. Properties of DTFT

# Inversion of DTFT

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

$$\begin{aligned} \rightarrow \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\Omega) e^{j\Omega l} d\Omega &= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left[ \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n} \right] e^{j\Omega l} d\Omega \\ &= \sum_{n=-\infty}^{\infty} x[n] \left[ \frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\Omega(l-n)} d\Omega \right] : \end{aligned}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\Omega) e^{j\Omega n} d\Omega$$

# Examples of calculating the inverse DTFT

1. Find  $x[n]$  from its DTFT  $X(\Omega)$ : 
$$X(\Omega) = \begin{cases} 1, & |\Omega| \leq \Omega_c \\ 0, & \Omega_c < |\Omega| < \pi \end{cases}$$

$$x[n] = \frac{1}{2\pi} \int_{-\Omega_c}^{\Omega_c} 1 \cdot e^{j\Omega n} d\Omega$$

## Examples of calculating the inverse DTFT

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2. Find  $x(n)$  from its DTFT  $X(\Omega)$ :  $X(\Omega) = \cos^2 \Omega$

# Examples of calculating of the inverse of DTFT

3. Find  $x(n)$  from its DTFT  $X(\Omega)$ :
- $$X(\Omega) = \frac{e^{j\Omega}}{e^{j\Omega} - 2}$$

# Lecture 6

## DTFT

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2. Inversion of DTFT
3. Properties of DTFT

# Linearity

$$ax[n] + by[n] \xleftrightarrow{\text{DTFT}} aX(\Omega) + bY(\Omega)$$

The DTFT of a linear combination of two or more signals is equal to the same linear combination of the DTFT of the individual signals.

# Time shifting

$$x[n - n_0] \xleftrightarrow{\text{DTFT}} e^{-j\Omega n_0} X(\Omega)$$

**Proof:** infer from the shifting property of ZT

$$x[n - n_0] \xleftrightarrow{\text{ZT}} z^{-n_0} X(z)$$

→ A shift in time causes **a linear phase shift** in frequency – **no change in DTFT magnitude**

# Frequency shifting and modulation

$$e^{j\Omega_0 n} x[n] \xleftrightarrow{\text{DTFT}} X(\Omega - \Omega_0)$$

$$\cos(\Omega_0 n) x[n] \xleftrightarrow{\text{DTFT}} \frac{1}{2} X(\Omega - \Omega_0) + \frac{1}{2} X(\Omega + \Omega_0)$$

$$e^{j\Omega_0 n} x[n] \xleftrightarrow{\text{DTFT}} \sum_{n=-\infty}^{\infty} (e^{j\Omega_0 n} x[n]) e^{-j\Omega n} = \sum_{n=-\infty}^{\infty} x[n] e^{-j(\Omega - \Omega_0)n} = X(\Omega - \Omega_0)$$

→ Modulation causes **a shift** in frequency

# Convolution in time domain

$$x_1[n] * x_2[n] \xleftrightarrow{\text{DTFT}} X_1(\Omega) \cdot X_2(\Omega)$$

Convolution in time  $\leftrightarrow$  Multiplication in frequency

# Convolution in frequency domain

$$x_1[n] \cdot x_2[n] \xleftrightarrow{\text{DTFT}} \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(\lambda) X_2(\Omega - \lambda) d\lambda = \frac{1}{2\pi} X_1(\Omega) * X_2(\Omega)$$

$x_1[n] \cdot x_2[n]$   $\xleftrightarrow{\text{DTFT}}$   $\sum_{n=-\infty}^{\infty} (x_1[n] \cdot x_2[n]) e^{-j\Omega n}$

Multiplication  
in time

$$= \sum_{n=-\infty}^{\infty} \left( \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(\lambda) e^{j\lambda n} d\lambda \right) x_2[n] e^{-j\Omega n}$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(\lambda) \left( \sum_{n=-\infty}^{\infty} x_2[n] e^{-j(\Omega - \lambda)n} \right) d\lambda$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(\lambda) X_2(\Omega - \lambda) d\lambda$$

Convolution in  
frequency

# HW

**Prob.1** Compute the DTFT of the following signals

a)  $x[n] = \left\{ -2, -1, \underset{\uparrow}{0}, 1, 2 \right\}$

b)  $y[n] = \begin{cases} 2 - (\frac{1}{2})n & |n| \leq 4 \\ 0 & |n| > 4 \end{cases}$

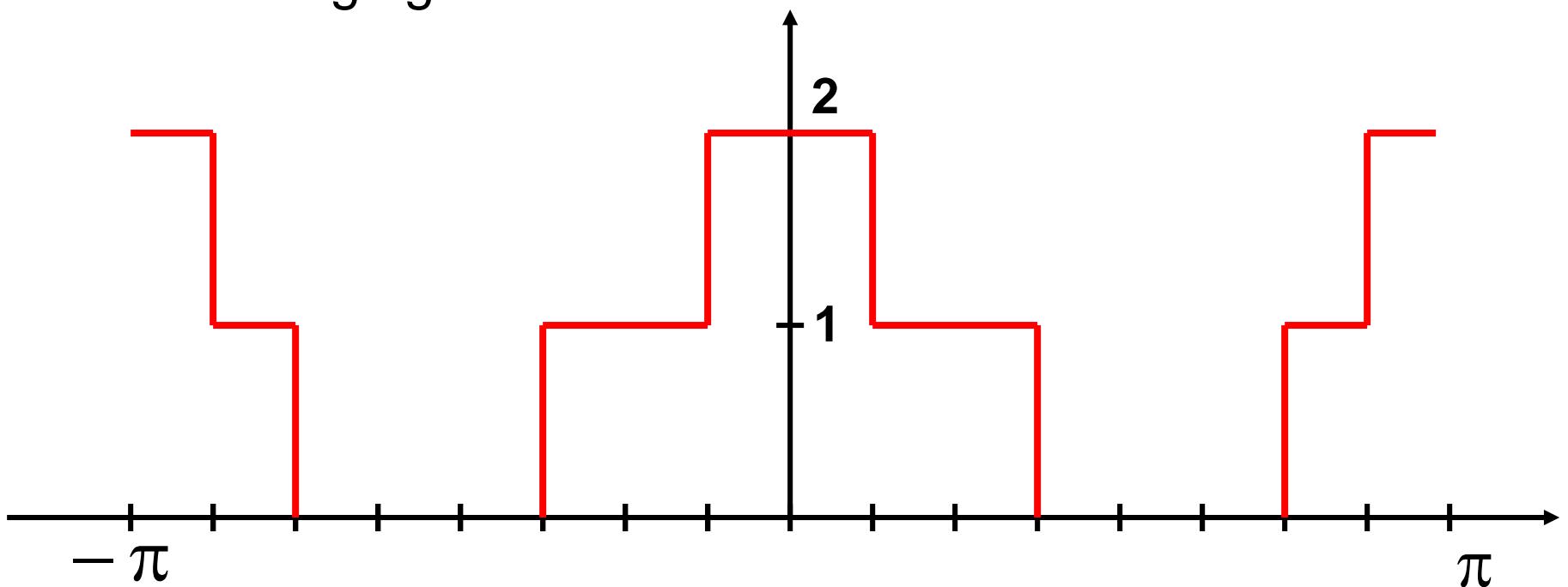
c)  $v[n] = 2^n u[-n]$

d)  $h[n] = a^n \sin(\Omega_0 n) u[n]$

e)  $w[n] = u[n] - u[n - 6]$

# HW

**Prob.2** Determine the signal having the DTFT shown in following figure



# HW

**Prob.3** A signal  $x(n)$  has the following DTFT:

$$X(\Omega) = \frac{1}{1 - ae^{-j\Omega}}$$

Determine the DTFT of the following signals:

$$(a) x[n] * x[n - 1]$$

$$(b) x[n]\cos(0.3\pi n)$$

$$(c) e^{j\pi n/2}x[n + 2]$$

# Lecture 7

# Digital signal spectra

- **Outline:**

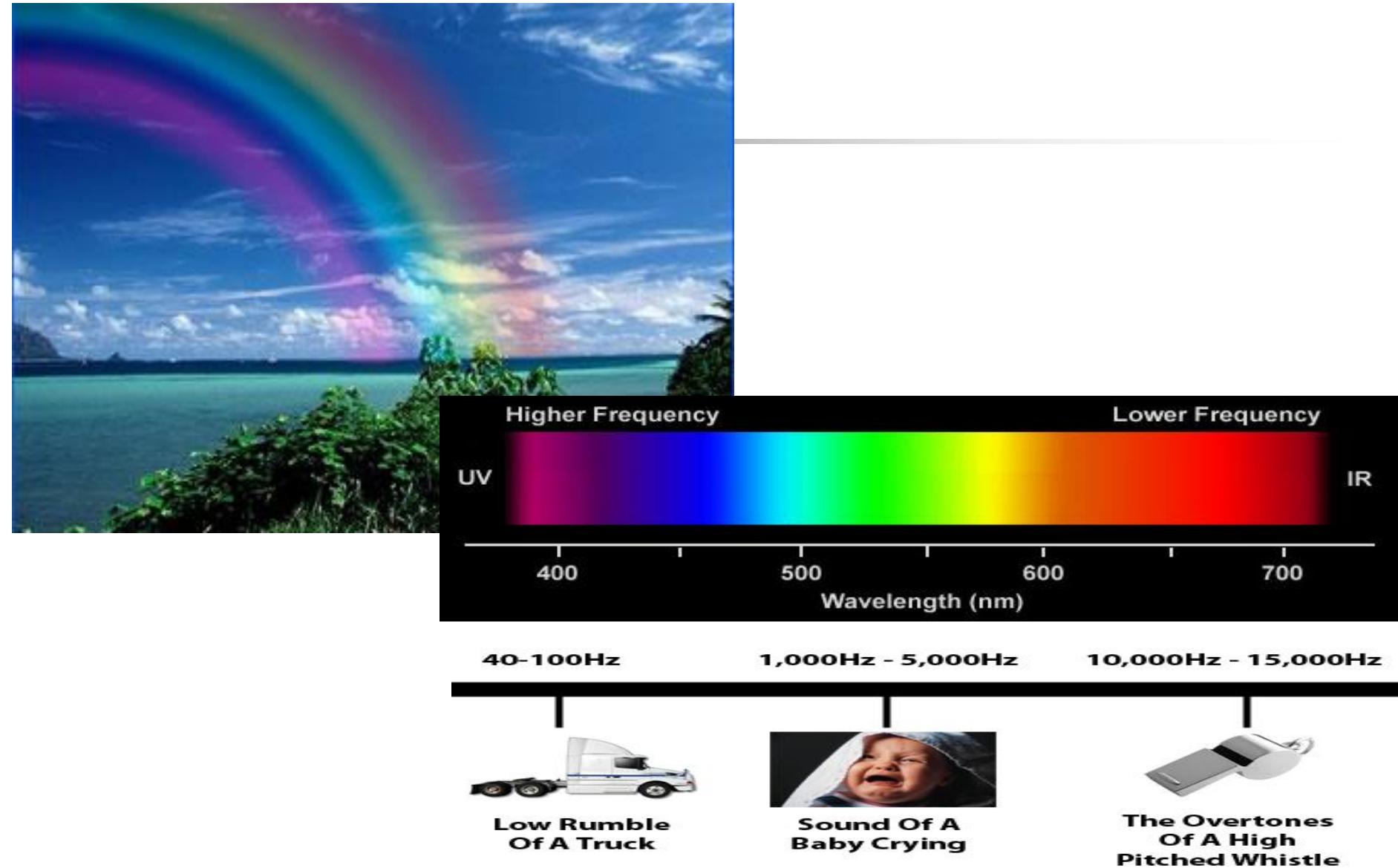
1. Frequency spectrum
2. Nonperiodic signals
3. Periodic signals

# Frequency spectrum

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- Detailed description of the frequency components the signal contains.
- Two parts: magnitude spectrum & phase spectrum.
- Tools for calculating the spectrum: DTFT for nonperiodic signals and DFS for periodic signals.

# Frequency spectrum



# Lecture 7

# Digital signal spectra

- **Outline:**

1. Frequency spectrum
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# DTFT to calculate spectrum of a nonperiodic signal

$$X(\Omega) = |X(\Omega)| e^{j\theta(\Omega)}$$

Amplitude spectrum

Phase spectrum

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

$$\Rightarrow X(\Omega) = X^*(-\Omega)$$

$$\Rightarrow |X(\Omega)| = |X(-\Omega)| \quad \text{and} \quad \angle X(\Omega) = -\angle X(-\Omega)$$

# Example of finding amplitude spectrum and phase spectrum

Find and plot amplitude spectrum and phase spectrum:

$$x[n] = u[n] - u[n - 4]$$

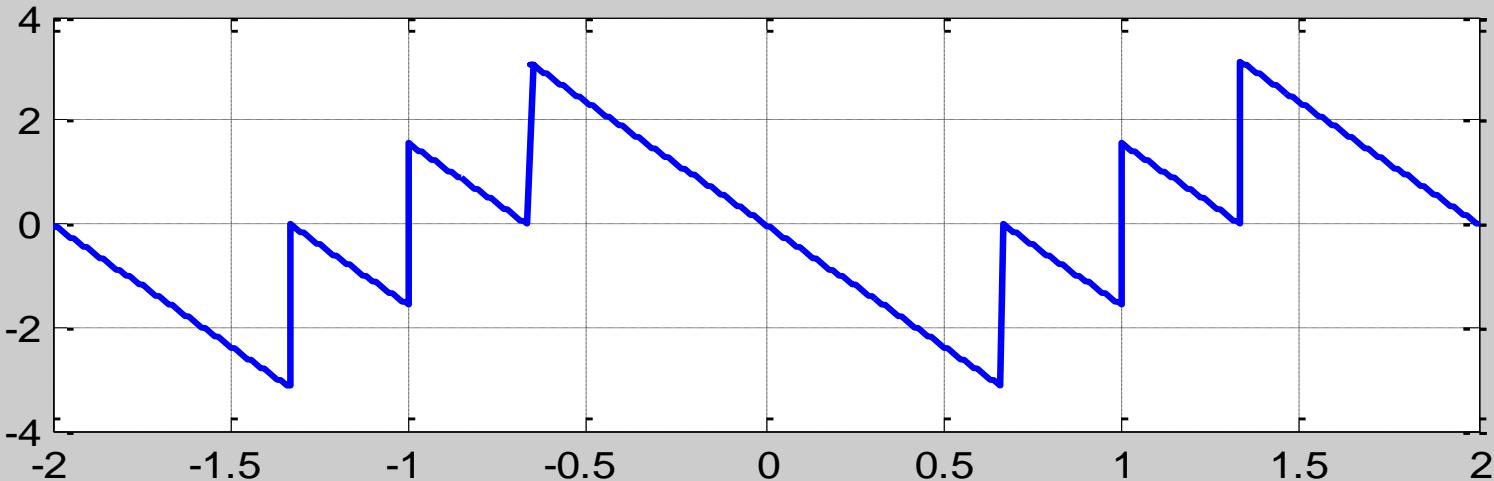
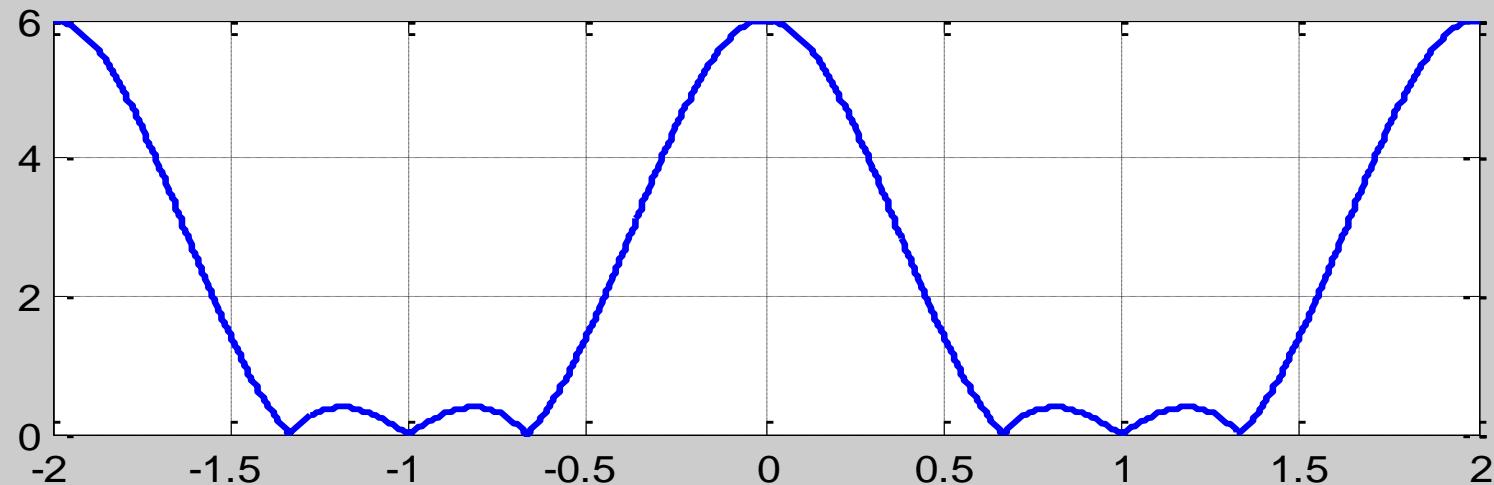
$$\begin{aligned} X(\Omega) &= \sum_{n=0}^3 e^{-j\Omega n} = \frac{1 - e^{-j4\Omega}}{1 - e^{-j\Omega}} \\ &= \frac{\sin(2\Omega)}{\sin(\Omega/2)} e^{-j3\Omega/2} \end{aligned}$$

# Using Matlab to plot amplitude spectrum and phase spectrum

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```
w = -2*pi:pi/255:2*pi; % freq. -2π → 2π, resolution of π/255  
X =4*sinc(2*w/pi)./sinc(w/(2*pi)).*exp(-j*1.5*w);  
  
subplot(2,1,1);  
plot(w/pi,abs(X)); % plot amplitude spectrum  
  
subplot(2,1,2);  
plot(w/pi,phase(X)); % plot phase spectrum
```

## Amplitude spectrum



## Phase spectrum

# Lecture 7

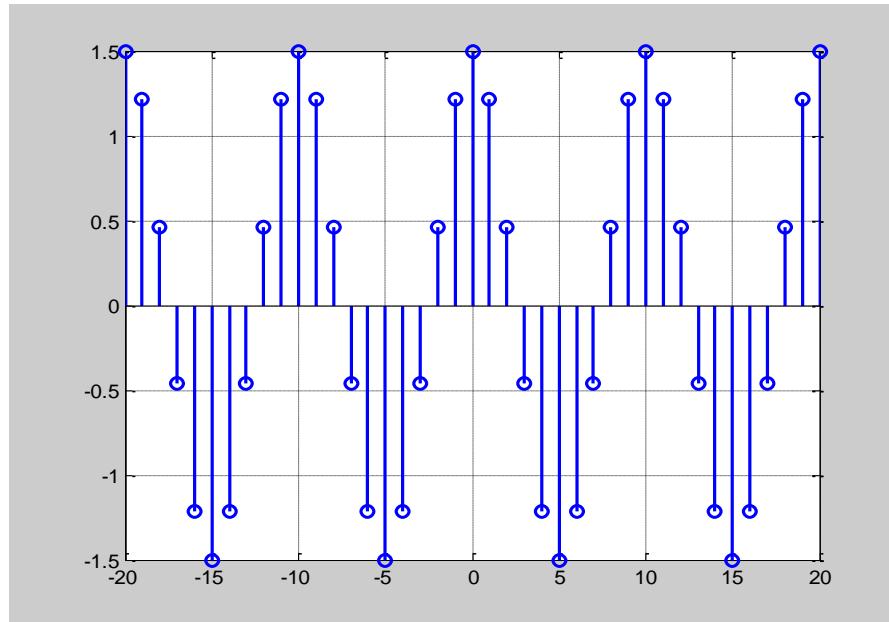
# Digital signal spectra

- **Outline:**

1. Frequency spectrum
2. Nonperiodic signals
3. Periodic signals

# Periodic sinusoidal signals

- Digital sinusoidal signal is periodic only if its frequency  $F$  is rational number  $x(n) = A \cos(\Omega n + \theta), -\infty < n < +\infty$   
 $= A \cos(2\pi F n + \theta), -\infty < n < +\infty$



# Fourier expansion

- CT periodic signals with period T:

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}; \quad a_k = \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt$$

$$\omega_0 = \frac{2\pi}{T}$$

- DT periodic signals with period N:

$$x[n] = \sum_{k \in \langle N \rangle} a_k e^{jk\Omega_0 n}; \quad a_k = \frac{1}{N} \sum_{n \in \langle N \rangle} x[n] e^{-jk\Omega_0 n}$$

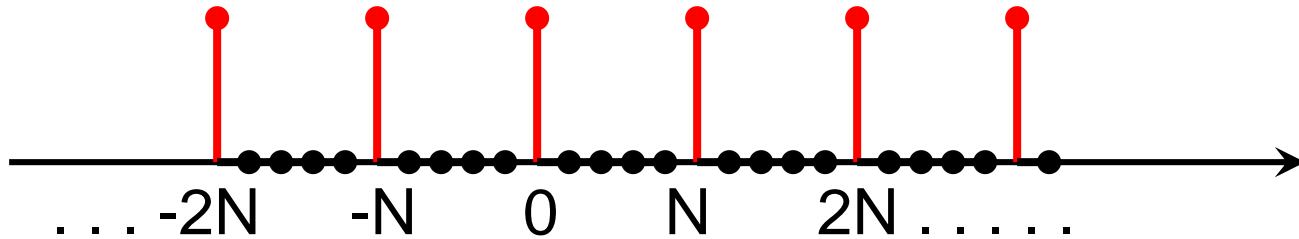
$$\Omega_0 = \frac{2\pi}{N}$$

**Note:** finite sums over an interval length of one periodic N

$$e^{jk\Omega_0 n} = e^{jk\frac{2\pi}{N} n} = e^{j(k+N)\frac{2\pi}{N} n} = e^{j(k+N)\Omega_0 n}$$

# Example of Fourier Series expansion

Given a DT periodic signals with period N:  $p[n] = \sum_{k=-\infty}^{\infty} \delta[n - kN]$ .



$$p[n] = \sum_{k=0}^{N-1} a_k e^{jk \frac{2\pi}{N} n}$$

$$a_k = \frac{1}{N} \sum_{n=0}^{N-1} p[n] e^{-jk 2\pi n / N} = \frac{1}{N}$$

# Spectrum of periodic signals

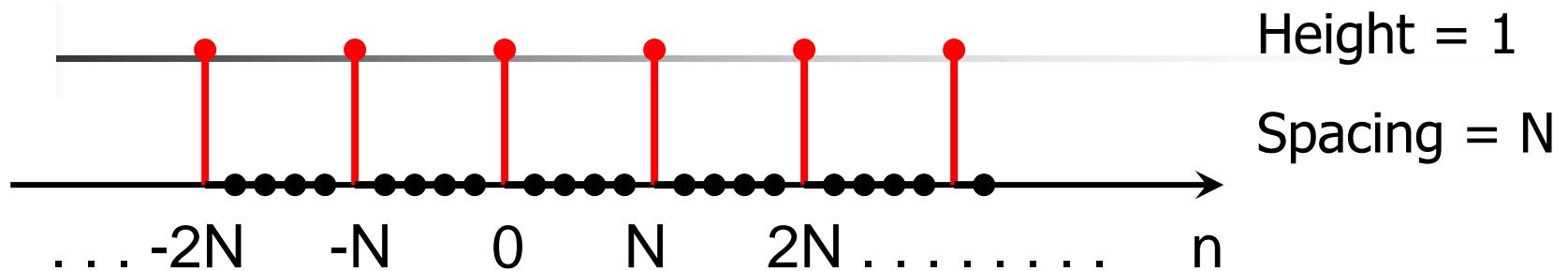
- CT periodic signals with period T:

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} \xrightarrow{F} X(\omega) = 2\pi \sum_{k=-\infty}^{\infty} a_k \delta(\omega - k\omega_0)$$

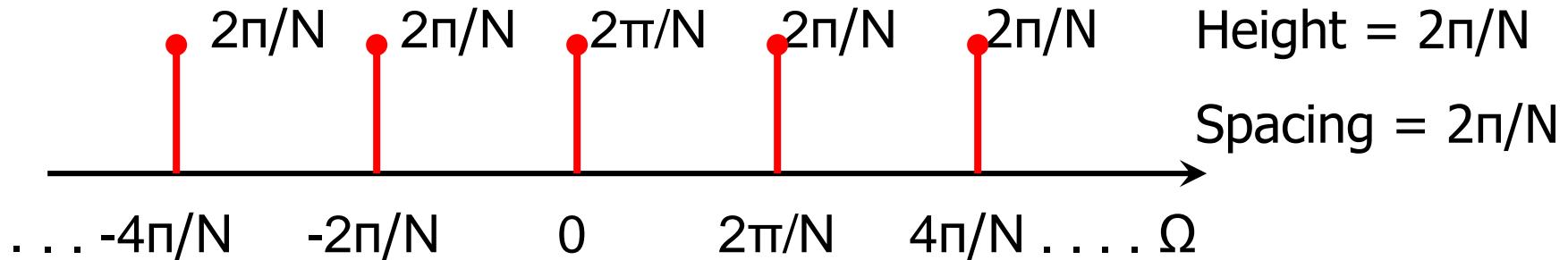
- DT periodic signals with period N:

$$x[n] \xrightarrow{F} X(\Omega) = 2\pi \sum_{k=-\infty}^{\infty} a_k \delta(\Omega - k\Omega_0)$$

# Example of calculating DTFT of periodic signals (DFS)

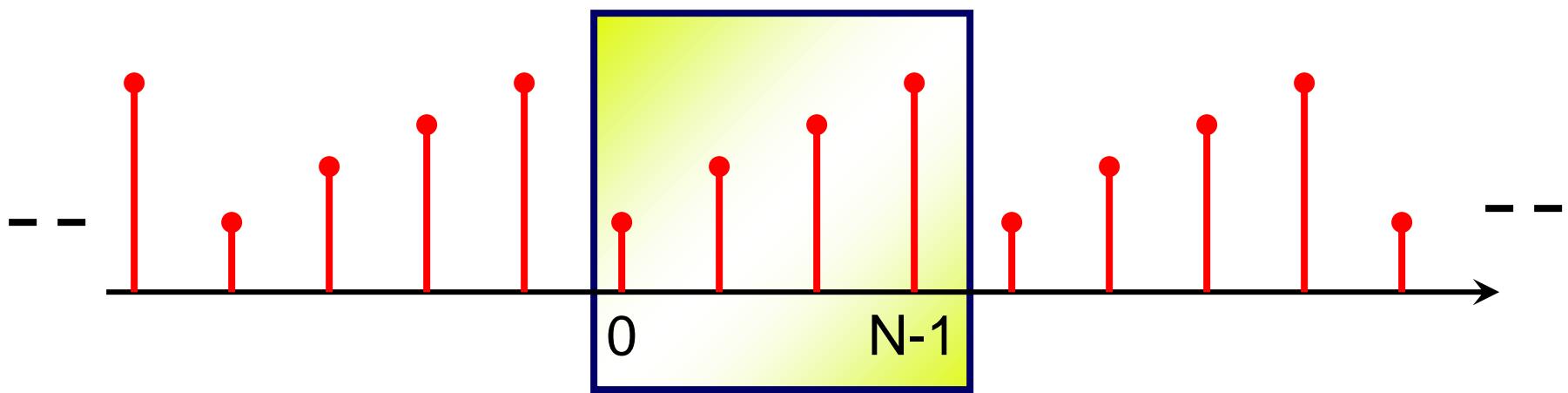


$P(\Omega)$



# Another approach to get DTFT of periodic signals (DFS)

$x(n)$  is periodic signal;  $x_0(n)$  is a part of  $x(n)$  that is repeated



$$x_0[n] = \begin{cases} x[n], & 0 \leq n \leq N-1 \\ 0, & \text{otherwise.} \end{cases}$$

# Another approach to get DTFT of periodic signals (DFS)

$$x_0[n] = \begin{cases} x[n], & 0 \leq n \leq N-1 \\ 0, & \text{otherwise.} \end{cases}$$

$$x[n] = \sum_{k=-\infty}^{\infty} x_0[n-kN] = \sum_{k=-\infty}^{\infty} x_0[n] * \delta[n-kN] = x_0[n] * \sum_{k=-\infty}^{\infty} \delta[n-kN]$$

p(n) in previous example

$$x[n] = x_0[n] * p[n] \xrightarrow{F} X_0(\Omega)P(\Omega) = X(\Omega)$$

# Another approach to get DTFT of periodic signals (DFS)

$$x[n] = x_0[n] * p[n] \xrightarrow{F} X_0(\Omega)P(\Omega) = X(\Omega)$$

$$X(\Omega) = X_0(\Omega) \left( \frac{2\pi}{N} \sum_k \delta(\Omega - k \frac{2\pi}{N}) \right)$$

$$= \frac{2\pi}{N} \sum_k X_0(k \frac{2\pi}{N}) \delta(\Omega - k \frac{2\pi}{N})$$

N samples—you are sampling the DTFT at N equal intervals around the unit circle

→ It has **N** distinct values at  $k = 0, 1, \dots, N-1$

# Procedure to calculate spectrum (DFS) of periodic signals

## Step 1:

Start with  $x_0(n)$  – one period of  $x(n)$ , with zero everywhere else

## Step 2:

Find the DTFT  $X_0(\Omega)$  of the signal  $x_0[n]$  above

## Step 3:

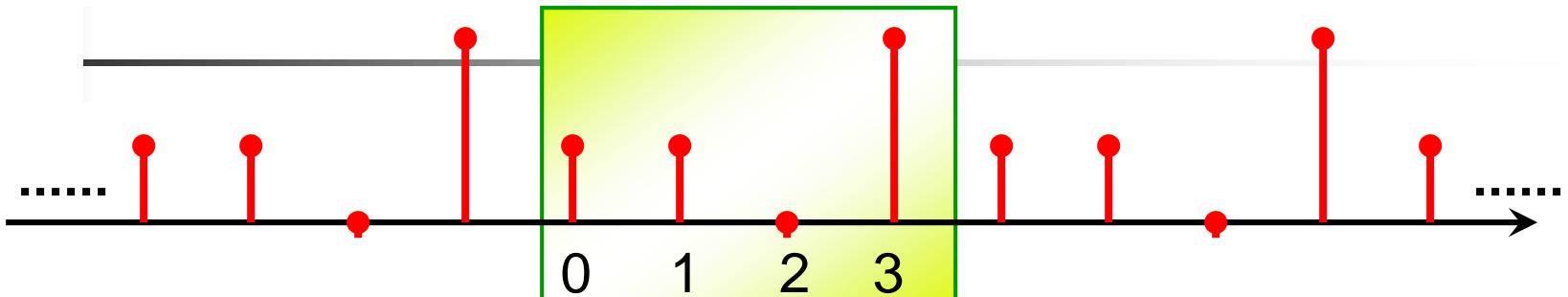
Find  $X_0(\Omega)$  at  $N$  equally spacing frequency points  $X_0(k2\pi/N)$

## Step 4:

Obtain the DTFT(DFS):

$$X(\Omega) = \frac{2\pi}{N} \sum_k X_0\left(k \frac{2\pi}{N}\right) \delta\left(\Omega - k \frac{2\pi}{N}\right)$$

# Example of calculating DFS of periodic signals



$$X_0(\Omega) = \sum_{n=0}^3 x_0(n)e^{-j\Omega n} = 1 + e^{-j\Omega} + 2e^{-j3\Omega}$$

$$X_o\left(\frac{2\pi k}{4}\right) = 1 + e^{-j\frac{2\pi k}{4}} + 2e^{-j\frac{3\pi k}{4}} \quad k = 0, 1, 2, 3$$

$$k = 0 \rightarrow X_0(0) = 4; \quad k = 1 \rightarrow X_0(1) = 1+j$$

$$k = 2 \rightarrow X_0(2) = -2; \quad k = 3 \rightarrow X_0(3) = 1-j$$

# Example of calculating DFS of periodic signals

$$X(\Omega) = \frac{2\pi}{N} \sum_{k=-\infty}^{\infty} X_o\left(\frac{2\pi k}{N}\right) \delta\left(\Omega - \frac{2\pi k}{N}\right)$$

$$= \frac{2\pi}{4} \sum_{k=-\infty}^{\infty} X_o\left(\frac{2\pi k}{4}\right) \delta\left(\Omega - \frac{2\pi k}{4}\right)$$

For one period     $0 \leq \Omega < 2\pi$

$$\frac{\pi}{2} \left\{ 4\delta(\Omega) + (1+j)\delta\left(\Omega - \frac{\pi}{2}\right) - 2\delta(\Omega - \pi) + (1-j)\delta\left(\Omega - \frac{3\pi}{2}\right) \right\}$$

# HW

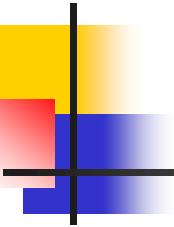
**Prob.4** Determine and sketch the magnitude and phase response of the following signals

$$a) \quad x[n] = \frac{1}{2}(\delta[n] - \delta[n-2])$$

$$b) \quad x[n] = \frac{1}{2}(\delta[n] - 2\delta[n-1] + \delta[n-2])$$

$$c) \quad x[n] = \frac{1}{8}(\delta[n] + 3\delta[n-1] + 3\delta[n-2] + \delta[n-3])$$

# HW



## Prob.5

Given a periodic signal  $y(n)$  with  $N = 3$  with associated

$$y_0[n] = \delta[n] + 2\delta[n-1] + 3\delta[n-2]$$

Find  $Y_0(\Omega)$  and  $Y(\Omega)$ . Do the test with the inverse DTFT

# Lecture 8

## Frequency response and filter

- **Outline:**

1. Frequency response
2. Response to complex exponential signals
3. Filters

# Frequency response

- For impulse response,  $h(n)$ , its DTFT is often called **frequency response  $H(\Omega)$**
- **$H(\Omega)$**  completely characterizes a LTI system in the frequency domain
- **$H(\Omega)$**  allows us to determine the steady-state response of the system to any arbitrary weighted linear combination of sinusoids or complex exponential

# Example of frequency response

- A LTI causal system is described by the following equation:

$$y[n] + 0.1y[n-1] + 0.85y[n-2] = x[n] - 0.3x[n-1]$$

- First, checking the stability of the system (by using Matlab):

$$b = [1 -0.3];$$

$$a = [1 0.1 0.85];$$

`zplane(b,a) % plot zeros and poles to check if all poles are inside the unit circle`

- Second, take DTFT for two sides:

$$H(\Omega) = \frac{1 - 0.3e^{-j\Omega}}{1 + 0.1e^{-j\Omega} + 0.85e^{-j2\Omega}}$$

# Example of amplitude & phase responses

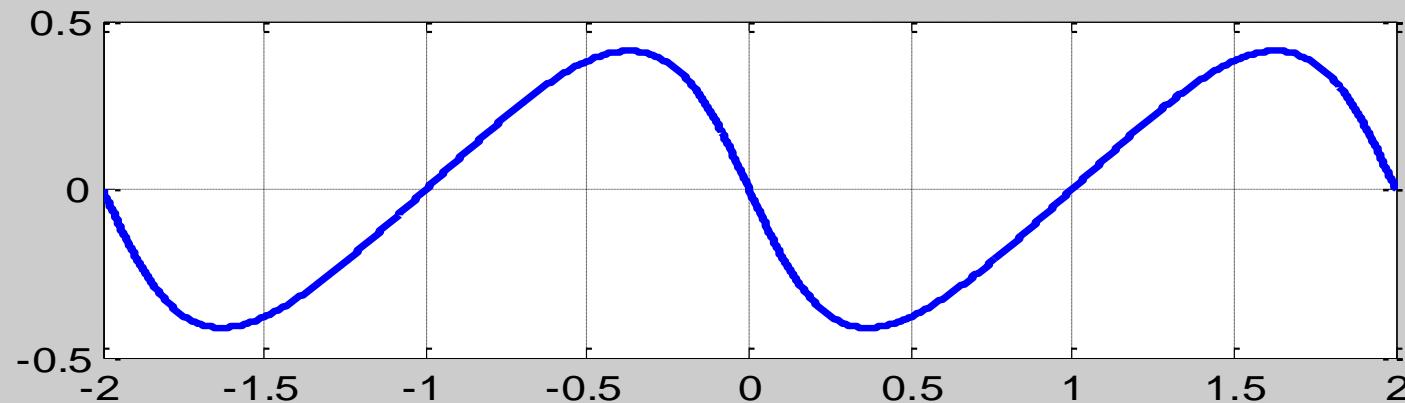
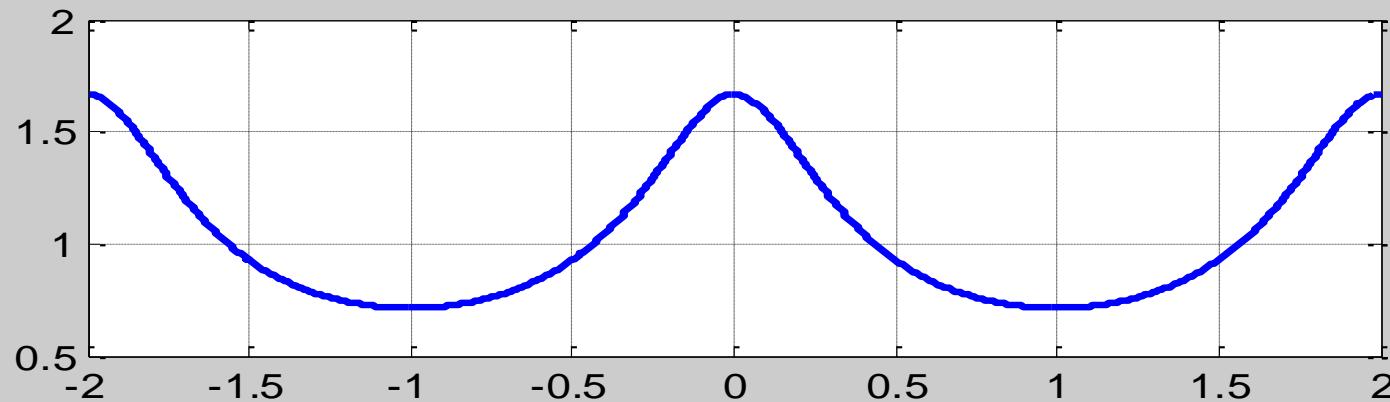
$$H(\Omega) = \frac{1}{1 - 0.4e^{-j\Omega}}$$

$$\begin{aligned}|H(\Omega)| &= \frac{1}{|1 - 0.4e^{-j\Omega}|} = \frac{1}{|1 - 0.4\cos\Omega + j0.4\sin\Omega|} \\&= \frac{1}{\sqrt{(1 - 0.4\cos\Omega)^2 + (0.4\sin\Omega)^2}} \\&= \frac{1}{\sqrt{1.16 - 0.8\cos\Omega}}\end{aligned}$$

$$\angle H(\Omega) = 0 - \angle(1 - 0.4e^{-j\Omega}) = -\arctg\left(\frac{0.4\sin\Omega}{1 - 0.4\cos\Omega}\right)$$

# Example of amplitude & phase responses

$$H(\Omega) = \frac{1}{1 - 0.4e^{-j\Omega}}$$



# Lecture 8

## Frequency response and filter

- **Outline:**

1. Frequency response
2. Response to complex exponential signals
3. Filters

# Response to complex exponential signals

$$x[n] = A e^{j\Omega_0 n}, \quad -\infty < n < \infty$$


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

$$\begin{aligned} y[n] &= \sum_{k=-\infty}^{\infty} h[k] \left( A e^{j\Omega_0(n-k)} \right) \\ &= A \left[ \sum_{k=-\infty}^{\infty} h[k] \left( e^{-j\Omega_0 k} \right) \right] e^{j\Omega_0 n} \\ &= (A e^{j\Omega_0 n}) H(\Omega_0) = x[n] H(\Omega_0) \end{aligned}$$

# Example of determining response to complex exponential signals

Determine the output signal of system  $h[n] = (1/2)^n u[n]$  to the input signal below

$$x[n] = Ae^{\frac{j\pi}{2}n}, \quad -\infty < n < \infty$$

$$H\left(\frac{\pi}{2}\right) = \frac{1}{1 + j\frac{1}{2}} = \frac{2}{\sqrt{5}} e^{-j26.6^0}$$

$$y[n] =$$

# Response to sinusoidal signals

$$x[n] = A \cos(\Omega_0 n) = \frac{A}{2} e^{j\Omega_0 n} + \frac{A}{2} e^{-j\Omega_0 n}, \quad -\infty < n < \infty$$

$$y[n] = \frac{A}{2} e^{j\Omega_0 n} H(\Omega_0) + \frac{A}{2} e^{-j\Omega_0 n} H(-\Omega_0)$$

$$\begin{aligned} &= \frac{A}{2} e^{j\Omega_0 n} |H(\Omega_0)| e^{j\angle H(\Omega_0)} + \frac{A}{2} e^{-j\Omega_0 n} |H(\Omega_0)| e^{-j\angle H(\Omega_0)} \\ &= A |H(\Omega_0)| \cos(\Omega_0 n + \angle H(\Omega_0)) \end{aligned}$$

# Example of determining response to sinusoidal signals

$$h[n] = (1/2)^n u[n]$$

$$x[n] = 10 - 5 \sin \frac{\pi}{2} n + 20 \cos \pi n, \quad -\infty < n < \infty$$

$$H(0) = \frac{1}{1 - 0.5} = 2; H\left(\frac{\pi}{2}\right) = \frac{2}{\sqrt{5}} e^{-j26.6^0}; H(\pi) = \frac{1}{1 + 0.5} = \frac{2}{3}$$

$$y[n] = 10 |H(0)| - 5 \left| H\left(\frac{\pi}{2}\right) \right| \sin \left( \frac{\pi}{2} n + \angle H\left(\frac{\pi}{2}\right) \right)$$

$$+ 20 |H(\pi)| \cos(\pi n)$$

$$= 20 - 2\sqrt{5} \sin\left(\frac{\pi}{2}n - 26.6^0\right) + \frac{40}{3} \cos(\pi n)$$

# Lecture 8

## Frequency response and filter

- **Outline:**

1. Frequency response
2. Response to complex exponential signals
3. Filters

# What is digital filters?

- Systems that perform mathematical operations on a DT signal to reduce or enhance certain aspects of that signal.
- Provide a convenient means to change the nature of a signal.
- **Change the frequency characteristics** of a signal in a specific way, letting some frequencies in the signal pass while blocking other

B  
E  
F  
O  
R  
E



A  
F  
T  
E  
R

# Characterizations of digital filter

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1. Impulse response  **$h(n)$** : FIR and IIR filters
2. Transfer function  **$H(z)$**
3. Frequency response  **$H(\Omega)$** : amplitude response and phase response
4. Difference equation:

$$y[n] + a_1 y[n-1] + \dots + a_N y[n-N] = b_0 x[n] + b_1 x[n-1] + \dots + b_M x[n-M]$$

# Typical applications of digital filters

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- **Noise Suppression:** radio signal, biomedical signal, analog media signal...
- **Enhancement of Selected Frequency Range:** treble/bass control, equalizers in audio, image edge enhancement...
- **Bandwidth Limiting:** aliasing prevention, interference avoidance...
- **Removal of Specific Frequencies:** DC removal, 60 Hz signal removal, notch filter...
- **Special Operations:** differentiation, integration, Hilbert transform...

# **Basic digital filter types**

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- **Low pass filter (LPF):** lets low frequencies through while blocking high frequencies;
- **High pass filter (HPF):** lets high frequencies while blocking low frequencies;
- **Band pass filter (BPF):** allows a band of frequencies to pass;
- **Band stop filter (BSF):** allows all frequencies outside a band to pass

## Ideal digital filters

$$y[n] = \begin{cases} Cx[n - n_0], & \omega_1 < \omega < \omega_2 \\ 0, & \omega \neq \end{cases}$$

→  $Y(e^{j\omega}) = CX(e^{j\omega})e^{-j\omega n_0} = X(e^{j\omega})H(e^{j\omega}), \quad \omega_1 < \omega < \omega_2$

→  $H(e^{j\omega}) = \begin{cases} Ce^{-j\omega n_0}, & \omega_1 < \omega < \omega_2 \\ 0, & \omega \neq \end{cases}$

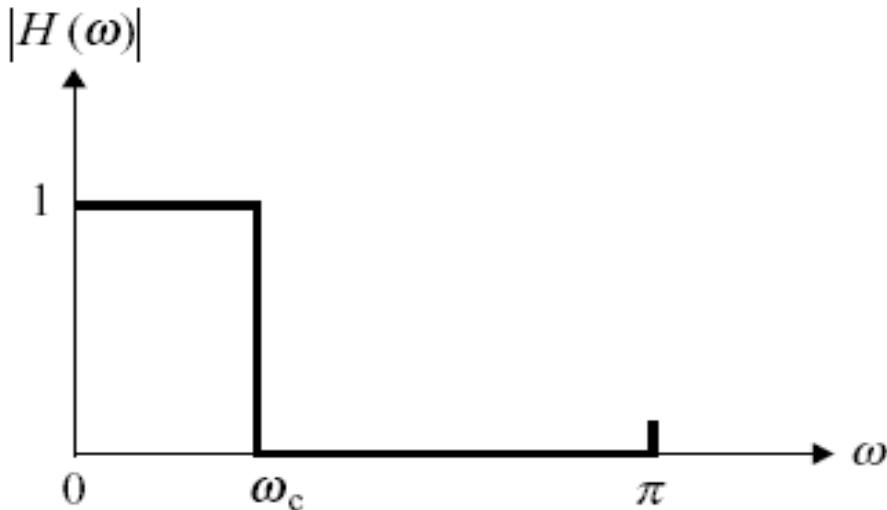
→  $|H(e^{j\omega})| = C, \quad \omega_1 < \omega < \omega_2$

$\theta(\omega) = -\omega n_0, \quad \omega_1 < \omega < \omega_2$

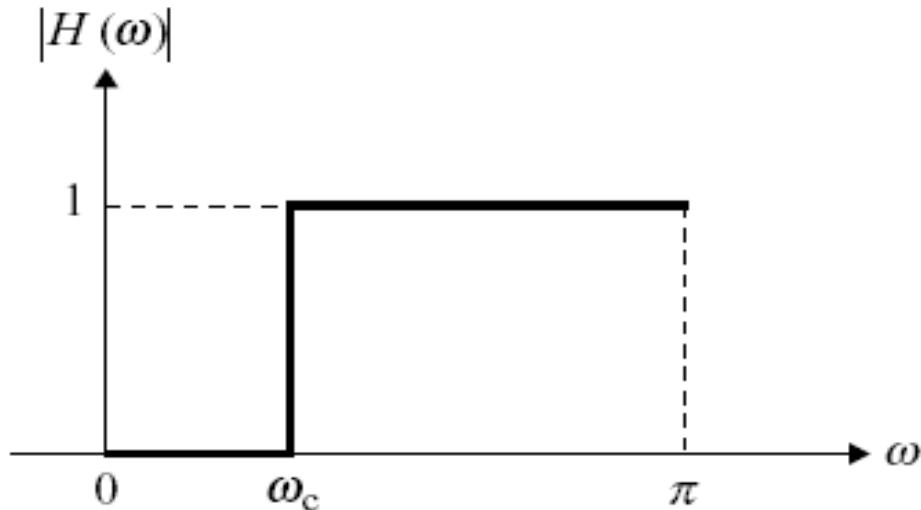
Linear phase response

Constant amplitude response

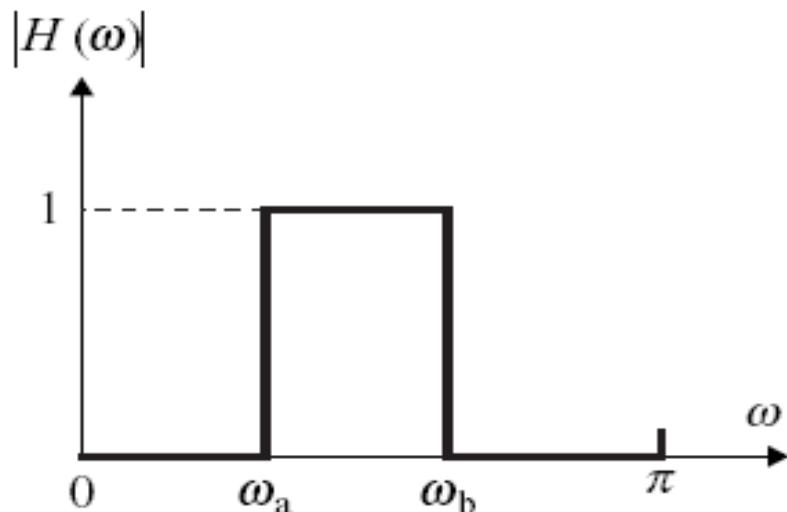
# Ideal digital filters



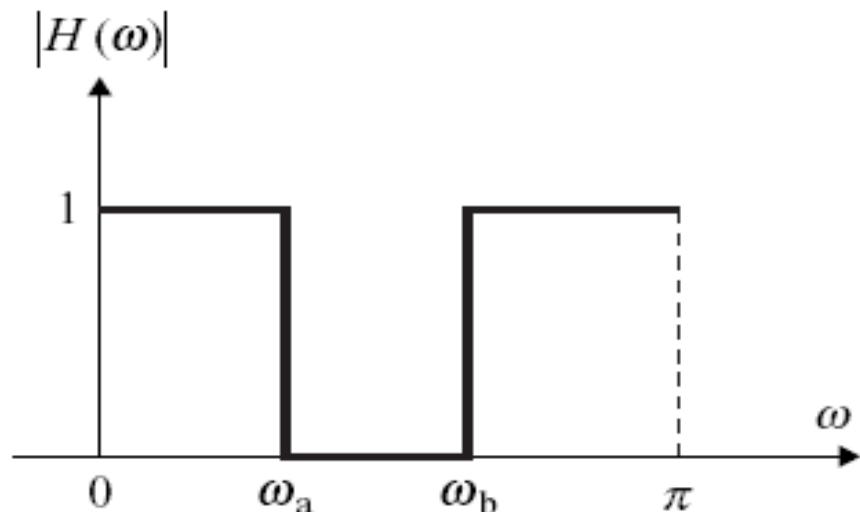
(a) Lowpass filter



(b) Highpass filter



(c) Bandpass filter



(d) Bandstop filter

# Block diagram representation

- Being built based on input-output relation or transform function by connecting the basic components in the specific way:

$$y[n] + a_1 y[n-1] + \dots + a_N y[n-N] = b_0 x[n] + b_1 x[n-1] + \dots + b_M x[n-M]$$

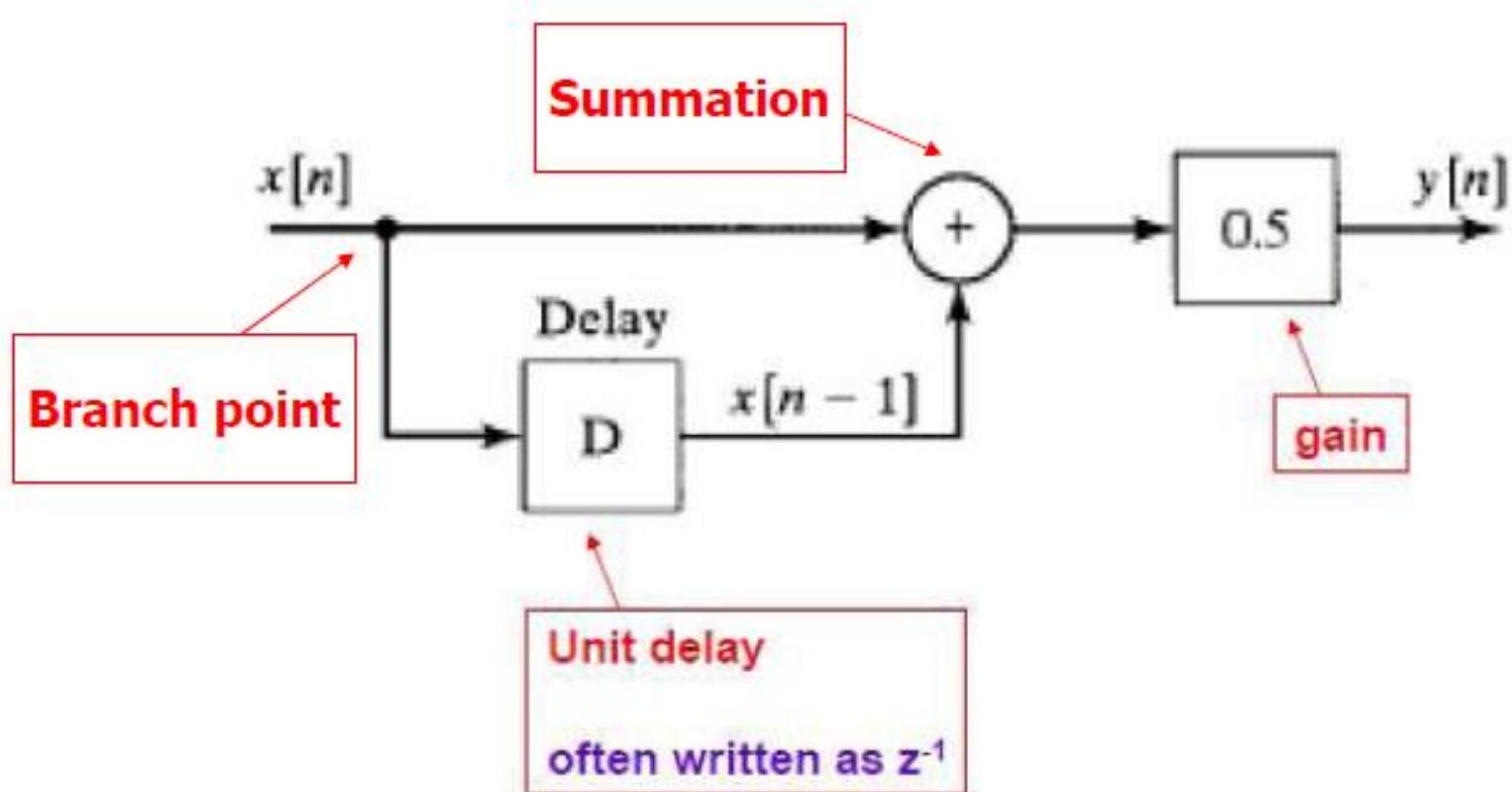
$$\leftrightarrow \sum_{k=0}^N a_k y[n-k] = \sum_{r=0}^M b_r x[n-r]$$

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

- Some advantages: easy to determine the relation between input and output, easy to manipulate the diagram to the other equivalent one, easy to determine the hardware...

# Components for block diagram

**EX:** an averaging system  $y[n] = 0.5(x[n] + x[n-1])$



# HW

**Prob.6** An FIR filter is described by the difference equation

$$y[n] = x[n] - x[n-10]$$

- a) Compute and sketch its magnitude and phase response
- b) Determine its response to the inputs

$$(1) \quad x[n] = \cos \frac{\pi}{10} n + 3 \sin \left( \frac{\pi}{3} n + \frac{\pi}{10} \right) \quad -\infty < n < \infty$$

$$(2) \quad x[n] = 10 + 5 \cos \left( \frac{2\pi}{5} n + \frac{\pi}{2} \right) \quad -\infty < n < \infty$$

# HW

**Prob.7 (a)** Determine the coefficients of a linear-phase FIR filter

$$y[n] = b_0x[n] + b_1x[n - 1] + b_2x[n - 2]$$

such that:

- It rejects completely a frequency component at  $\Omega_0 = 2\pi/3$
- Its frequency response is normalized so that  $H(0) = 1$

(b) Compute and sketch the magnitude and phase response of the filter to check if it satisfies the requirements

- $w = -\pi:\pi/255:\pi;$
- $b_0 = 1/3 ; b_1 = 1/3 ; b_2 = 1/3 ;$
- $H = b_0 + b_1 * \exp(-j * w) + b_2 * \exp(-j * 2 * w);$
- $Habs = \text{abs}(H);$
- $Hphase = \text{phase}(H);$
- $\text{subplot}(2,1,1);$
- $\text{plot}(w/\pi, Habs, \text{'linewidth'}, 2); \text{grid};$
- $\text{subplot}(2,1,2);$
- $\text{plot}(w/\pi, Hphase, \text{'linewidth'}, 2); \text{grid};$