3 Discrete-Time Fourier Transform

Introduction

- In time-domain, the input-output relation of a linear and time-invariant (LTI) system is characterized by the convolution
- An alternate description of a sequence in terms of complex exponential sequences of the form {e^{-jωn}} where ω is the normalized frequency variable
- The frequency domain representation of the discrete-time sequences and discrete-time LTI systems

Continuous-Time Fourier Transform

• Definition:

The CTFT of a continuous-time signal $x_a(t)$ is given by

$$X_a(j\Omega) = \int_{-\infty}^{+\infty} x_a(t) e^{-j\Omega t} dt$$

 Often referred to as *Fourier spectrum* or simply the *spectrum* of the continuous-time signal

Continuous-Time Fourier Transform

• Definition:

The inverse CTFT of a Fourier transform $X_{a}(j\Omega)$ is given by

$$x_a(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X(j\Omega) e^{j\Omega t} d\Omega$$

• Often referred to as *Fourier integral*

The Continuous-Time Fourier Transform Pair

Analysis equation:

$$X_a(j\Omega) = \int_{-\infty}^{+\infty} x_a(t) e^{-j\Omega t} dt$$

Synthesis equation:

$$x_a(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X_a(j\Omega) e^{j\Omega t} d\Omega$$

A CTFT pair is CTFT also denoted as: $x_a(t) \leftrightarrow X_a(j\Omega)$

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Continuous-Time Fourier Transform

- The Fourier transform or Fourier integral $X_a(j\Omega)$ of $x_a(t)$ is also called the **analysis equation**
- The inverse Fourier transform equation is called the *synthesis equation*
- For aperiodic signals, the complex exponentials occur at a continuum of frequencies
- The transform $X_a(j\Omega)$ of an aperiodic signal $x_a(t)$ is commonly referred to as the **spectrum** of $x_a(t)$

Continuous-Time Fourier Transform

- Variable Ω is real and denotes the continuoustime angular frequency in radians
- In general, the CTFT is a complex function of \varOmega in the range $-\infty < \Omega < \infty$
- It can be expressed in polar form as

$$X_a(j\Omega) = |X_a(j\Omega)| e^{j\theta_a(\Omega)}$$

where

$$\theta_a(\Omega) = \arg\{X_a(j\Omega)\}$$

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Continuous-Time Fourier Transform

$$X_a(j\Omega) = |X_a(j\Omega)| e^{j\theta_a(\Omega)}$$

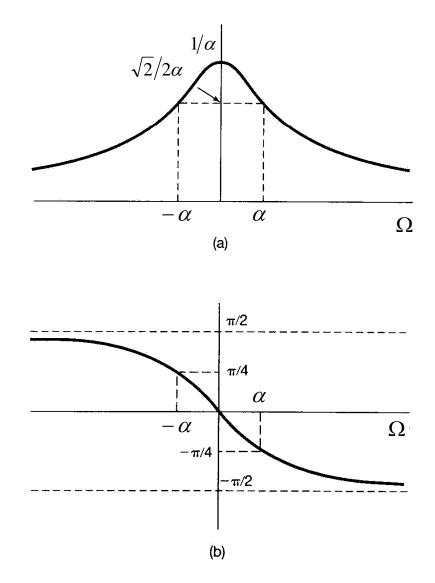
- The quantity $|X_a(j\Omega)|$ is called the *magnitude spectrum*
- The quantity $\theta_{a}(\Omega)$ is called the **phase spectrum**
- Both spectrums are real functions of \varOmega

Example 3.1

The Fourier transform of a causal complex exponential

$$x_a(t) = \begin{cases} e^{-\alpha t}, & t \ge 0\\ 0, & t < 0 \end{cases}$$

$$\begin{aligned} X_a(j\Omega) &= \int_0^\infty e^{-\alpha t} e^{-j\Omega t} dt \\ &= \frac{1}{\alpha + j\Omega}, \quad \alpha > 0 \end{aligned}$$



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The Frequency Response of an LTI Continuous-Time System

• The output response of $y_a(t)$ of an initially relaxed linear, time-invariant continuous-time system characterized by an impulse response $h_a(t)$ for an input signal $x_a(t)$ is given by the convolution integral

$$y_a(t) = \int_{-\infty}^{+\infty} h_a(t-\tau) x_a(\tau) d\tau$$

Applying CTFT to both sides

$$Y_a(j\Omega) = H_a(j\Omega)X_a(j\Omega)$$

• $H_a(j\Omega)$ is the **frequency response** of the system

The Discrete-Time Fourier Transform

- The discrete-time Fourier transform (DTFT) of a discrete-time sequence x[n] is a representation of the sequence in terms of the complex exponential sequence {e^{-jωn}} where ω is the real frequency variable
- The DTFT representation of a sequence, if it exists, is unique and the original sequence can be computed from its DTFT by an inverse transform operation

The Discrete-Time Fourier Transform

The *discrete-time Fourier transform* (DTFT)
 X(e^{jw}) of a sequence x[n] is defined by:

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

- The Fourier transforms of most practical discrete-time sequences can be expressed in terms of a sum of a convergent geometric series
- They can be summed in a simple closed form

Example:

Consider a causal sequence: $x[n] = \alpha^n \mu[n], |\alpha| < 1$

The Fourier transform $X(e^{j\omega})$ is obtained as:

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} \alpha^n \mu[n] \ e^{-j\omega n} = \sum_{n=0}^{\infty} \alpha^n e^{-j\omega n}$$
$$= \sum_{n=0}^{\infty} (\alpha e^{-j\omega})^n = \frac{1}{1 - \alpha e^{-j\omega}}$$

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Discrete-Time Fourier Transform (DTFT)

- As can be seen from definition, DTFT X(e^{jω}) of a sequence x[n] is a continuous function of ω
- Unlike the continuous-time Fourier transform, DTFT is a periodic function in ω with a period 2π

$$X(e^{j(\omega+2\pi k)}) = \sum_{n=-\infty}^{+\infty} x[n]e^{-j(\omega+2\pi k)n} = \sum_{n=-\infty}^{+\infty} x[n]e^{-j\omega n}e^{-j2\pi k n}$$
$$= \sum_{n=-\infty}^{+\infty} x[n]e^{-j\omega n} = X(e^{j\omega}), \text{ for all values of } k$$

where $e^{-j2\pi kn} = 1$

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Inverse Discrete-Time Fourier Transform

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

• The inverse discrete-time Fourier transform can be interpreted as a linear combination of infinitesimally small complex exponential signals of the form $\frac{1}{2\pi}e^{j\omega n}d\omega$, weighted by the complex constant $X(e^{j\omega})$ over the angular frequency range from $-\pi$ to π

The Discrete-Time Fourier Transform (DTFT) Pair

Analysis equation, denoted by operator $\mathcal{F}{x[n]}$:

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

Synthesis equation, denoted by operator $\mathcal{F}^{-1}{x[n]}$:

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

Basic Properties of the DTFT

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

• $X(e^{j\omega})$ is a complex function the real variable ω :

$$X(e^{j\omega}) = X_{re}(e^{j\omega}) + jX_{im}(e^{j\omega})$$

 $X(e^{j\omega}) = |X(e^{j\omega})| e^{j\theta(\omega)}, \text{ where } \theta(\omega) = \arg\{X(e^{j\omega})\}$

- $|X(e^{j\omega})|$ is the magnitude function
- $\theta(e^{j\omega})$ is called the phase function

Basic Properties of the DTFT

$$X(e^{j\omega}) = \left| X(e^{j\omega}) \right| e^{j\theta(\omega)}$$

- In many applications, the Fourier transform X(e^{jω}) is called the *Fourier spectrum*
- $|X(e^{j\omega})|$ is called the *magnitude spectrum* and
- $\theta(\omega)$ is the **phase spectrum**
- It is usually assumed that the phase function $\theta(\omega)$ is restricted to the *principal value*

$$-\pi \leq \theta(\omega) < \pi$$

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Commonly Used DTFT Pairs

Sequence		DTFT
$\delta[n]$	\leftrightarrow	1
1	\leftrightarrow	$\sum_{k=-\infty}^{\infty} 2\pi \delta(\omega + 2\pi k)$
$\mu[n]$	\leftrightarrow	$\frac{1}{1-e^{-j\omega}}$
$e^{j\omega_0 n}$	\leftrightarrow	$\sum_{k=0}^{\infty} 2\pi \delta(\omega - \omega_0 + 2\pi k)$
$\alpha^n \mu[n]$, $(\alpha < 1)$	\leftrightarrow	$\frac{1}{1 - \alpha e^{-j\omega}}$

DTFT Properties

- There are a number of important properties of the DTFT that are useful in signal processing applications
- These are listed here without proof
- Their proofs are straightforward
- The applications of some of the properties are illustrated

Table 3.1: DTFT Properties: Symmetry Relations

Sequence	Discrete-Time Fourier Transform		
<i>x</i> [<i>n</i>]	$X(e^{j\omega})$		
x[-n]	$X(e^{-j\omega})$		
$x^{*}[-n]$	$X^*(e^{j\omega})$		
$\operatorname{Re}\{x[n]\}$	$X_{\rm cs}(e^{j\omega}) = \tfrac{1}{2} \{ X(e^{j\omega}) + X^*(e^{-j\omega}) \}$		
$j \operatorname{Im} \{x[n]\}$	$X_{\rm ca}(e^{j\omega}) = \tfrac{1}{2} \{ X(e^{j\omega}) - X^*(e^{-j\omega}) \}$		
$x_{cs}[n]$	$X_{\rm re}(e^{j\omega})$		
$x_{ca}[n]$	$jX_{\rm im}(e^{j\omega})$		

Note: $X_{cs}(e^{j\omega})$ and $X_{ca}(e^{j\omega})$ are the conjugate-symmetric and conjugate-antisymmetric parts of $X(e^{j\omega})$, respectively. Likewise, $x_{cs}[n]$ and $x_{ca}[n]$ are the conjugate-symmetric and conjugate-antisymmetric parts of x[n], respectively.

x[n]: A complex sequence

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Table 3.2: DTFT Properties: Symmetry Relations

Sequence	Discrete-Time Fourier Transform
x[n]	$X(e^{j\omega}) = X_{\rm re}(e^{j\omega}) + jX_{\rm im}(e^{j\omega})$
$x_{ev}[n]$	$X_{\rm re}(e^{j\omega})$
$x_{\mathrm{od}}[n]$	$jX_{\rm im}(e^{j\omega})$
	$X(e^{j\omega})=X^*(e^{-j\omega})$
	$X_{\rm re}(e^{j\omega}) = X_{\rm re}(e^{-j\omega})$
Symmetry relations	$X_{\rm im}(e^{j\omega})=-X_{\rm im}(e^{-j\omega})$
	$ X(e^{j\omega}) = X(e^{-j\omega}) $
	$\arg\{X(e^{j\omega})\} = -\arg\{X(e^{-j\omega})\}$

Note: $x_{ev}[n]$ and $x_{od}[n]$ denote the even and odd parts of x[n], respectively.

x[n]: A real sequence

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Important DTFT Theorems

- There are a number of important theorems of the DTFT that are useful in analysis and synthesis of discrete-time LTI systems
- Many algorithms in signal processing applications are based on these theorems
- Their proofs are straightforward based on the definitions
- Assume that:

$$g[n] \stackrel{F}{\leftrightarrow} G(e^{j\omega}) \text{ and } h[n] \stackrel{F}{\leftrightarrow} H(e^{j\omega})$$

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Table 3.4:General Properties of DTFT

Type of Property	Sequence	Discrete-Time Fourier Transform
	g[n] h[n]	$G(e^{j\omega})$ $H(e^{j\omega})$
Linearity	$\alpha g[n] + \beta h[n]$	$\alpha G(e^{j\omega}) + \beta H(e^{j\omega})$
Time-shifting	$g[n - n_0]$	$e^{-j\omega n_o}G(e^{j\omega})$
Frequency-shifting	$e^{j\omega_0 n}g[n]$	$G\left(e^{j\left(\omega-\omega_{o}\right)}\right)$
Differentiation in frequency	ng[n]	$j \frac{dG(e^{j\omega})}{d\omega}$
Convolution	$g[n] \circledast h[n]$	$G(e^{j\omega})H(e^{j\omega})$
Modulation	g[n]h[n]	$\frac{1}{2\pi}\int_{-\pi}^{\pi}G(e^{j\theta})H(e^{j(\omega-\theta)})d\theta$
Parseval's relation	$\sum_{n=-\infty}^{\infty} g[n]h^*[$	$n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} G(e^{j\omega}) H^*(e^{j\omega}) d\omega$

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The Frequency Response of an LTI Discrete-Time System

• Time-Domain:

An LTI discrete-time system is completely characterized by its impulse response sequence {*h*[*n*]}

 Transform-Domain: Alternative representations of an LTI discretetime system using the DTFT (and the *z*transform)

The Frequency Response - Definition

- An important property of an LTI system is that for certain types of input signals, called eigenfunctions, the output signal is the input signal multiplied by a complex constant
- We consider one such eigenfunction, the complex exponential sequence
- In general, for CT and DT systems:
 - Continuous-time: $e^{sT} \rightarrow H(s) e^{sT}$
 - Discrete-time: $z^n \rightarrow H(z) z^n$

The Frequency Response

Superposition property:

The response of an LTI system to a linear combination of complex exponential signals can be determined by knowing its response to a single complex exponential signal

The response of the LTI system to a complex exponential input is considered

Frequency Response is a transform-domain representation of the LTI discrete-time system

Complex Exponential Input

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

Input:
$$x[n] = e^{j\omega n}, -\infty < n < \infty$$

Output:
$$y[n] = \sum_{k=-\infty}^{\infty} h[k] e^{j\omega(n-k)}$$

$$=e^{j\omega n}\sum_{k=-\infty}^{\infty}h[k]e^{-j\omega k}=x[n]\left(\sum_{k=-\infty}^{\infty}h[k]e^{-j\omega k}\right)$$

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The Frequency Response

Define:

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

- *H*(*e^{j w}*) is called the frequency response of the LTI discrete-time system
- $H(e^{j\omega})$ is the DTFT of h[n]
- For a complex exponential input:

$$y[n] = H(e^{j\omega})e^{j\omega n}$$

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The Response to a Complex Exponential

- For a fixed frequency $\omega = \omega_0$: $y[n] = H(e^{j\omega_0})e^{j\omega_0 n}$
- For a complex exponential input *x*[*n*] of angular frequency ω₀, the output *y*[*n*] is a complex exponential sequence of the same angular frequency ω₀ weighted by a complex constant *H*(*e*^{*j*ω₀})
- In general, the frequency response H(e^{jω}) is a function of the angular frequency and can be evaluated at all input frequencies ω
- *H*(*e^{jω}*) completely characterizes the behavior of an LTI discrete-time system in frequency domain

The Frequency Response

• $H(e^{j\omega})$ is a complex function of ω with a period 2π

$$H(e^{j\omega}) = H_{re}(e^{j\omega}) + jH_{im}(e^{j\omega})$$
$$= \left| H(e^{j\omega}) \right| e^{j\theta(\omega)}$$

where $\theta(\omega) = \arg\{H(e^{j\omega})\}$

- $|H(e^{j\omega})|$ is called the *magnitude response*
- $\theta(\omega)$ is called the **phase response**

The Frequency Response

In some cases, the magnitude function is defined in decibels

$$G(\omega) = 20\log_{10} \left| H(e^{j\omega}) \right| \, \mathrm{dB}$$

- $G(\omega)$ is called the **gain** function
- The negative of the gain function,
 A(ω) -G(ω) is called the *attenuation* or *loss function*

Frequency-Domain Characterization of LTI Systems

• Input-output relation in frequency-domain

$$Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega}) = \left(\sum_{k=-\infty}^{\infty} h[k]e^{-j\omega k}\right)X(e^{j\omega})$$

 Convolution in the time-domain transforms into product in the frequency-domain

$$H(e^{j\omega}) = \frac{Y(e^{j\omega})}{X(e^{j\omega})}$$

 The frequency response of an LTI discrete-time system is the ratio of Y(e^{jw}) and X(e^{jw})

Frequency Responses of LTI FIR Discrete-Time Systems

• Input-output relation of the LTI FIR discrete-time system N_2

$$y[n] = \sum_{k=N_1}^{2} h[k] x[n-k], \quad N_1 < N_2$$

• Applying the discrete-time Fourier transform (DTFT) results in the transform-domain input-output relation

$$Y(e^{j\omega}) = \left(\sum_{k=N_1}^{N_2} h[k] e^{-j\omega k}\right) X(e^{j\omega}) = H(e^{j\omega}) X(e^{j\omega})$$

where $Y(e^{j\omega})$ and $X(e^{j\omega})$ are the DTFTs of the output and input sequences Frequency Responses of LTI FIR Discrete-Time Systems

• The frequency response of the LTI FIR discrete-time system is thus

$$H(e^{j\omega}) = \sum_{k=N_1}^{N_2} h[k] e^{-j\omega k}$$

• The frequency response of the LTI FIR discretetime system is a polynomial in $e^{-j\omega}$

Frequency Responses of LTI IIR Discrete-Time Systems

Input-output relation of the LTI IIR discrete-time system

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

 Applying the discrete-time Fourier transform (DTFT) results in the transform-domain inputoutput relation

$$\sum_{k=0}^{N} d_k e^{-j\omega k} Y(e^{j\omega}) = \sum_{k=0}^{M} p_k e^{-j\omega k} X(e^{j\omega})$$

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Frequency Responses of LTI IIR Discrete-Time Systems

• The frequency-domain relation can be written in the form

$$\left(\sum_{k=0}^{N} d_{k}e^{-j\omega k}\right) Y(e^{j\omega}) = \left(\sum_{k=0}^{M} p_{k}e^{-j\omega k}\right) X(e^{j\omega})$$

- Solving the ratio $H(e^{j\omega}) = \frac{Y(e^{j\omega})}{X(e^{j\omega})} = \frac{\sum_{k=0}^{M} p_k e^{-j\omega k}}{\sum_{k=0}^{N} d_k e^{-j\omega k}}$
- The frequency response of the LTI IIR discretetime system is a polynomial in $e^{-j\omega}$

Example: Simple IIR Discrete-Time System

• Consider the first order recursive or infinite impulse response (IIR) filter

 $y[n] - \alpha y[n-1] = x[n]$, with $|\alpha| < 1$

• The frequency response of this system is obtained by the Fourier transform

$$Y(e^{j\omega}) + \alpha Y(e^{j\omega})e^{-j\omega} = X(e^{j\omega})$$

- Solving the ratio: $H(e^{j\omega}) = \frac{Y(e^{j\omega})}{X(e^{j\omega})} = \frac{1}{1 \alpha e^{-j\omega}}$
- The impulse response is: $h[n] = \alpha^n \mu[n]$

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- In practice, the excitation to an LTI discrete-time system is usually a causal sequence applied at some finite sample index $n = n_0$
- The output for such an input when observed at sample instants beginning at n = n₀ will consist of a transient part along with a steady-state component
- Assume that the input is a causal exponential sequence applied at n = 0, i.e., $x[n] = e^{j\omega n}\mu[n]$

For n > 0, the output is obtained using the convolution sum

$$y[n] = \sum_{k=0}^{\infty} h[k] e^{j\omega(n-k)} \mu[n-k] = \left(\sum_{k=0}^{n} h[k] e^{-j\omega k}\right) e^{j\omega n}$$

as $\mu[n-k] = 0$ for k > n

• Rewriting the last expression of the equation

$$y[n] = \left(\sum_{k=0}^{\infty} h[k] \ e^{-j\omega k}\right) e^{j\omega n} - \left(\sum_{k=n+1}^{\infty} h[k] \ e^{-j\omega k}\right) e^{j\omega n}$$
$$= H(e^{j\omega}) \ e^{j\omega n} - \left(\sum_{k=n+1}^{\infty} h[k] \ e^{-j\omega k}\right) e^{j\omega n}, \quad n \ge 0$$

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$$y[n] = H(e^{j\omega}) e^{j\omega n} - \left(\sum_{k=n+1}^{\infty} h[k] e^{-j\omega k}\right) e^{j\omega n}, \quad n \ge 0$$

Steady-state response

$$y_{sr}[n] = H(e^{j\omega}) e^{j\omega n}$$

$$y_{tr}[n] = -\left(\sum_{k=n+1}^{\infty} h[k] e^{-j\omega k}\right) e^{j\omega n}$$

The effect of the transient response on the output is

$$y_{tr}[n] = \left| \sum_{k=n+1}^{\infty} h[k] e^{-j\omega(k-n)} \right| \le \sum_{k=n+1}^{\infty} |h[k]| \le \sum_{k=0}^{\infty} |h[k]|$$

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Response to a Causal Exponential Sequence $|y_{tr}[n]| = \left|\sum_{k=n+1}^{\infty} h[k] e^{-j\omega(k-n)}\right| \le \sum_{k=n+1}^{\infty} |h[k]| \le \sum_{k=0}^{\infty} |h[k]|$

- For a causal and stable IIR LTI discrete-time system, the impulse response is absolutely summable
- As a result the transient response y_{tr}[n] is a bounded sequence
- Moreover, as $n \to \infty$, $\sum_{k=n+1}^{\infty} |h[k]| \to 0$ the transient response decays to zero as *n* gets very large

- In most practical cases, the transient response becomes negligibly small after some finite amount of time, and the system can be assumed to be in a steady-state
- For a causal FIR LTI discrete-time system with an impulse response of length *N*+1, *h*[*n*]=0 for *n* > *N* and, thus, *y*_{tr}[*n*]=0 for *n* > *N*-1
- It should be noted that transients will occur whenever an input is applied or changed

The Concept of Filtering

- A *digital filter* is a discrete-time system that passes certain frequency components in an input sequence without any distortion and blocks other frequency components
- The key to the filtering process is the inverse discrete-time Fourier transform which expresses an arbitrary sequence as a linear weighted sum of an infinite number of exponential (sinusoidal) sequences
- By appropriately choosing the frequency response (or its magnitude) of the LTI digital filter the individual sinusoidal components can be attenuated or amplified independent of each other

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The Concept of Filtering

• Consider a real coefficient LTI discrete-time system characterized by a magnitude function

$$|H(e^{j\omega})| \cong \begin{cases} 1, & 0 \le \omega \le \omega_c \\ 0, & \omega_c < \omega \le \pi \end{cases}$$

• An input sequence

$$x[n] = A\cos(\omega_1 n) + B\cos(\omega_2 n),$$

with $0 < \omega_1 < \omega_c < \omega_2 < \pi$

is applied to the system

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The Concept of Filtering

• The output sequence is given by

$$y[n] = A |H(e^{j\omega_1})| \cos(\omega_1 n + \theta(\omega_1))$$

+ $B |H(e^{j\omega_2})| \cos(\omega_2 n + \theta(\omega_2))$

• Making use of $|H(e^{j\omega})|$ the output is

$$y[n] \cong A |H(e^{j\omega_1})| \cos(\omega_1 n + \theta(\omega_1))$$

• The LTI system is a lowpass filter

Response to a Sinusoidal Sequence

• Consider the sinusoidal input to an LTI discretetime system with the frequency response $H(e^{j\omega})=|H(e^{j\omega})|e^{j\theta(\omega)}$

$$x[n] = A\cos(\omega_0 n + \phi)$$

$$y[n] = A |H(e^{j\omega_0})| \cos(\omega_0 n + \theta(\omega_0) + \phi)$$

- The output signal y[n] has the same sinusoidal waveform as the input x[n] with two differences
 - The amplitude is multiplied by the constant value $|H(e^{j\omega_0})|$
 - The output has a phase **lag** by amount $\theta(\omega_0)$

Phase and Group Delays

• Let us rewrite the output to a sinusoidal input as

$$y[n] = A |H(e^{j\omega_0})| \cos\left(\omega_0 \left(n + \frac{\theta(\omega_0)}{\omega_0}\right) + \phi\right)$$
$$= A |H(e^{j\omega_0})| \cos\left(\omega_0 \left(n - \tau_p(\omega_0)\right) + \phi\right)$$

where $\tau(\omega_0) = -\frac{\theta(\omega_0)}{\omega_0}$ is called the **phase delay**

• The output *y*[*n*] is a time-delayed version of the input *x*[*n*]

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Example: Linear combination of sinusoidal signals

Consider the signal: $x(t) = 1 + \frac{1}{2}\cos 2\pi t + \cos 4\pi t + \frac{2}{3}\cos 6\pi t$

The same sinusoidal components with phase shifts:

$$x(t) = 1 + \frac{1}{2}\cos(2\pi t + \phi_1) + \cos(4\pi t + \phi_2) + \frac{2}{3}\cos(6\pi t + \phi_3)$$

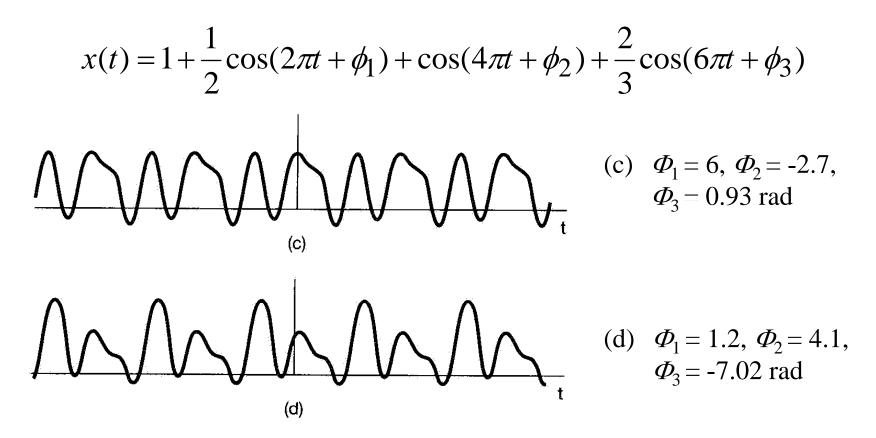
$$\underbrace{\sqrt{(a)}}_{t} \qquad (a) \quad \Phi_1 = \Phi_2 = \Phi_3 = 0$$

 $\underbrace{\bigwedge_{\text{(b)}}}_{\text{t}} (b) \underbrace{\varphi_1 = 4, \ \varphi_2 = 8,}_{\text{t}} \text{ and } \varphi_3 = 12 \text{ rad}$

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Example: Linear combination of sinusoidal signals



The resulting signals differ significantly for different relative phases

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The Group Delay

- When the input signal contains many sinusoidal components with different frequencies that are not harmonically related, each component will go through different phase delays when processed by a frequency-selective LTI discrete-time system
- The delay is determined using a different parameter called the *group delay* defined as

$$\tau_g(\omega) = -\frac{d\theta(\omega)}{d\omega}$$

 Group delay has a physical interpretation in calculating the responses of discrete-time systems

The Group Delay

• Group delay function provides a measure of the linearity of the phase response

$$\tau_g(\omega) = -\frac{d\theta(\omega)}{d\omega}$$

• For a moving average filter of length *M*, the phase response is *linear*

$$\theta(\omega) = -\frac{M-1}{2}\omega$$

and the group delay is *constant*

$$\tau_g(\omega) = \frac{M-1}{2}$$

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