

Lecture I213E – Class 6

Discrete Signal Processing

Sakriani Sakti



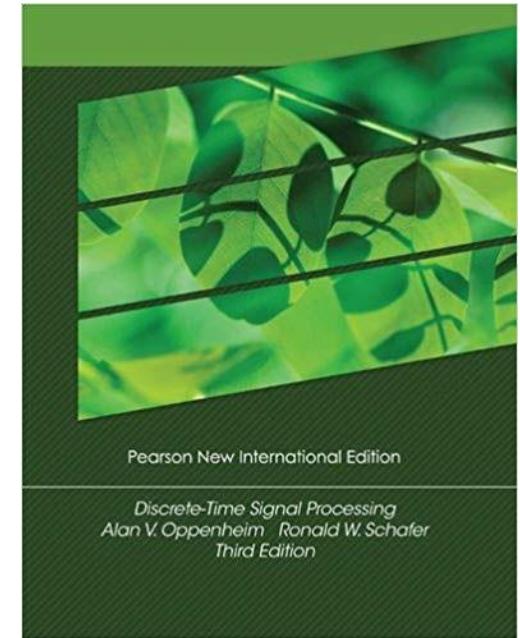
Course Materials

■ Materials

- Lecture notes will be uploaded before each lecture
<https://jstorage-2018.jaist.ac.jp/s/PGXRrC7iFmN2FWo>
Pass: dsp-i213e-2022
(Slide Courtesy of Prof. Nak Young Chong)

■ References

- Chi-Tsong Chen:
Linear System Theory and Design, 4th Ed.,
Oxford University Press, 2013.
- Alan V. Oppenheim and Ronald W. Schafer:
Discrete-Time Signal Processing, 3rd Ed.,
Pearson New International Ed., 2013.



Related Courses & Prerequisite

■ Related Courses

- I212 Analysis for Information Science
- I114 Fundamental Mathematics for Information Science

■ Prerequisite

- None

Evaluation

■ Viewpoint of evaluation

→ Students are able to understand:

- Basic principles in modeling and analysis of linear time-invariant systems
- Applications of mathematical methods and tools to different signal processing problems.

■ Evaluation method

→ Homework, term project, midterm exam, and final exam

■ Evaluation criteria

→ Homework/labs (30%), term project (30%)
midterm exam (15%), and final exam (25%)

Contact

- **Lecturer**

- Sakriani Sakti

- **TA**

- Tutorial hours & Term project**

- WANG Lijun (s2010026)

- TANG Bowen (s2110411)

- Homework**

- PUTRI Fanda Yuliana (s2110425)

- **Contact Email**

- dsp-i213e-2022@ml.jaist.ac.jp

Schedule

- December 8th, 2022 – February 9th, 2023

- Lecture Course Term 2-2

- Tuesday 9:00 – 10:40
- Thursday 10:50 – 12:30

- Tutorial Hours

- Tuesday 13:30-15:10

Schedule

Sun	Mon	Tue	Wed	Thu	Fri	Sat
				1	2	3
4	5	6	7	8	9	10
11	12	13	14	15	16	17
18	19	20	21	22	23	24
25	26	27	28	29	30	31

Dec

Sun	Mon	Tue	Wed	Thu	Fri	Sat
1	2	3	4	✗	6	7
8	9	10	11	12	13	14
15	16	17	18	19	20	21
22	23	✗	25	26	27	28
29	30	31				

Jan

Sun	Mon	Tue	Wed	Thu	Fri	Sat
			1	2	3	4
5	6	7	8	9	10	11
12	13	14	15	16	17	18
19	20	21	22	23	24	25
26	27	28				

Feb

Lecture:
 Tuesday 9:00 — 10:40
 Thursday 10:50 — 12:30

Tutorial:
 Tuesday 13:30 — 15:10

Midterm & final exam
 Thursday 10:50 — 12:30

Course review &
 term project evaluation
 (on tutorial hours)

Syllabus

Class	Date	Lecture Course Tue 9:00 — 10:40 / Thr 10:50 — 12:30	Tutorial Hours Tue 13:30 — 15:10
1	12/08	Introduction to Linear Systems with Applications to Signal Processing	
2	12/13	State Space Description	○
3	12/15	Linear Algebra	
4	12/20	Quantitative Analysis (State Space Solutions) and Qualitative Analysis (Stability)	○
5	12/22	Discrete-time Signals and Systems	
X	01/05		
6	01/10	Discrete-time Fourier Analysis	
7	01/10*	Review of Discrete-time Linear Time-Invariant Signals and Systems (on Tutorial Hours)	
	01/12	Midterm Exam	
8	01/17	Sampling and Reconstruction of Analog Signals	○
9	01/19	z-Transform	
X	01/24		○
10	01/26	Discrete Fourier Transform	
11	01/31	FFT Algorithms	○
12	02/02	Implementation of Digital Filters	
13	02/07	Digital Signal Processors and Design of Digital Filters	
14	02/07*	Review of the Course and Term Project Evaluation (on Tutorial Hours)	
	02/09	Final exam	

Class 6

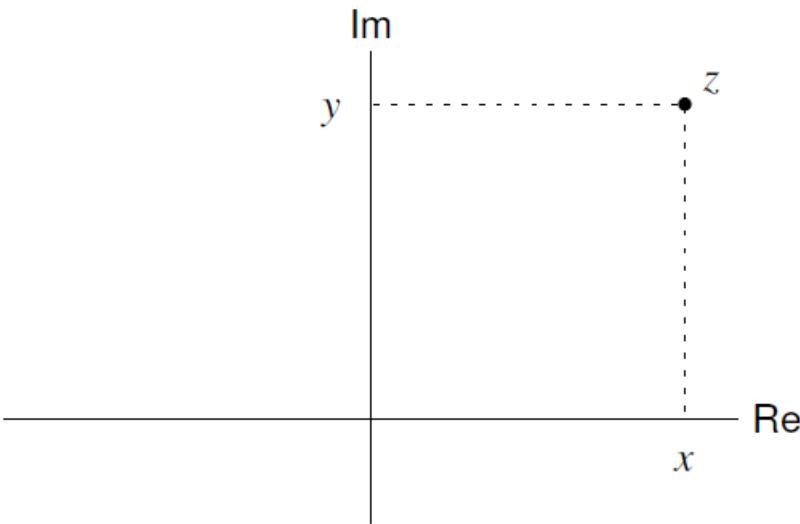
Discrete-time

Fourier Analysis

Review of Complex Analysis

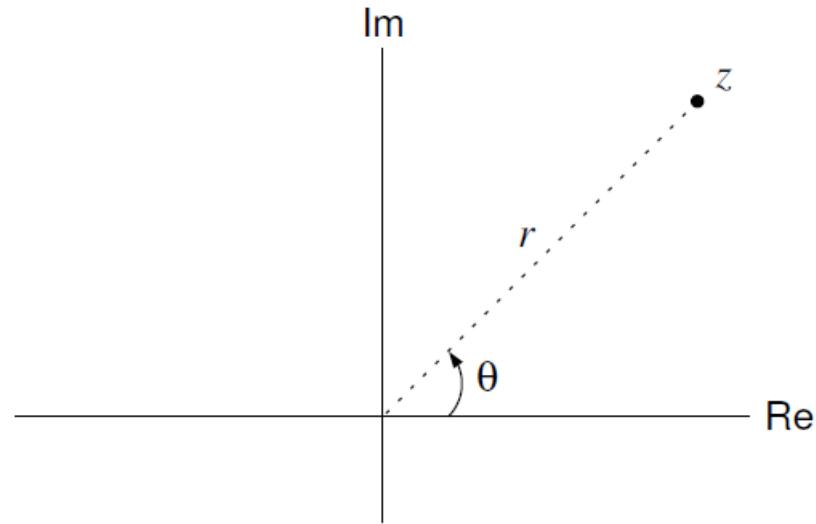
Complex Numbers

- Cartesian (Rectangular) Form
- Polar Form



$$z = x + jy$$

where $x = \text{Re } z$ and $y = \text{Im } z$



$$z = r(\cos \theta + j \sin \theta) = re^{j\theta}$$

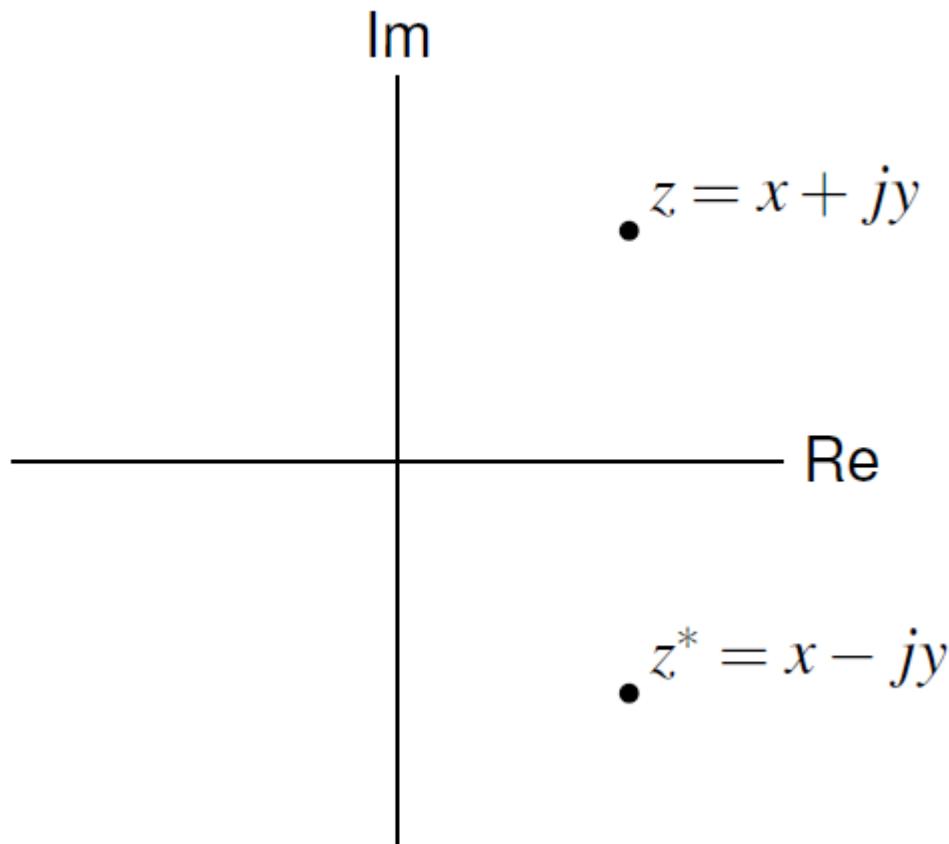
where $r = |z|$ and $\theta = \arg z$

Conjugate

■ Conjugate Complex Numbers

→ The conjugation operation reflects a point in the complex plane about the real axis

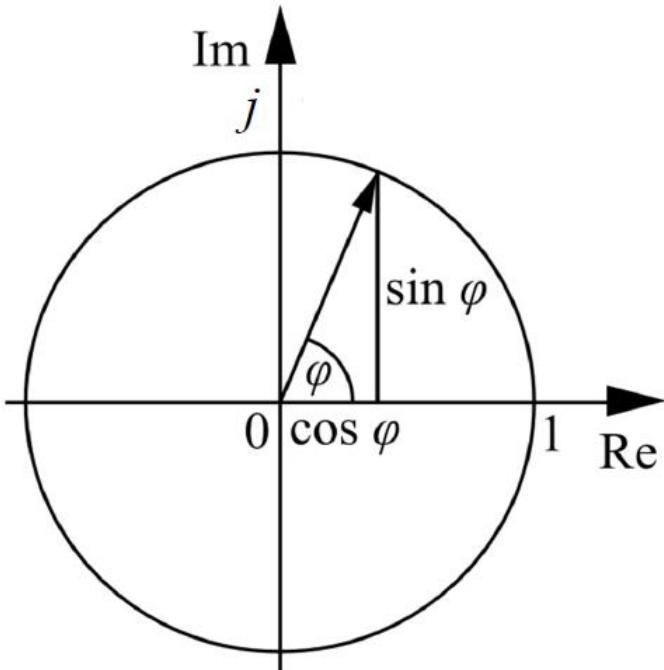
$$z^* = x - jy$$



Complex Exponential

Euler's Relation

$$e^{j\varphi} = \cos \varphi + j \sin \varphi$$



$$\begin{aligned} e^{j\varphi} &= 1 + j\varphi + \frac{(j\varphi)^2}{2!} + \frac{(j\varphi)^3}{3!} \dots = \sum_{k=0}^{\infty} \frac{(j\varphi)^k}{k!} \\ &= \boxed{1 - \frac{\varphi^2}{2!} + \frac{\varphi^4}{4!} - \dots} + j \left(\boxed{\varphi - \frac{\varphi^3}{3!} + \frac{\varphi^5}{5!} - \dots} \right) \\ &= \cos \varphi + j \sin \varphi \end{aligned}$$

Even function Odd function

$$\sin \varphi = \varphi - \frac{\varphi^3}{3!} + \frac{\varphi^5}{5!} - \dots = \sum_{k=0}^{\infty} \frac{(-1)^k}{(2k+1)!} \varphi^{2k+1}$$

$$\cos \varphi = 1 - \frac{\varphi^2}{2!} + \frac{\varphi^4}{4!} - \dots = \sum_{k=0}^{\infty} \frac{(-1)^k}{(2k)!} \varphi^{2k}$$

Complex Exponential

■ Euler's Relation

$$e^{j\omega t} = \cos \omega t + j \sin \omega t$$

1. Convenient representations of sinusoidal signals
2. A mixture of two orthogonal signals
 - the **in-phase** component $\cos \omega t$
 - the **phase-quadrature** component $\sin \omega t$

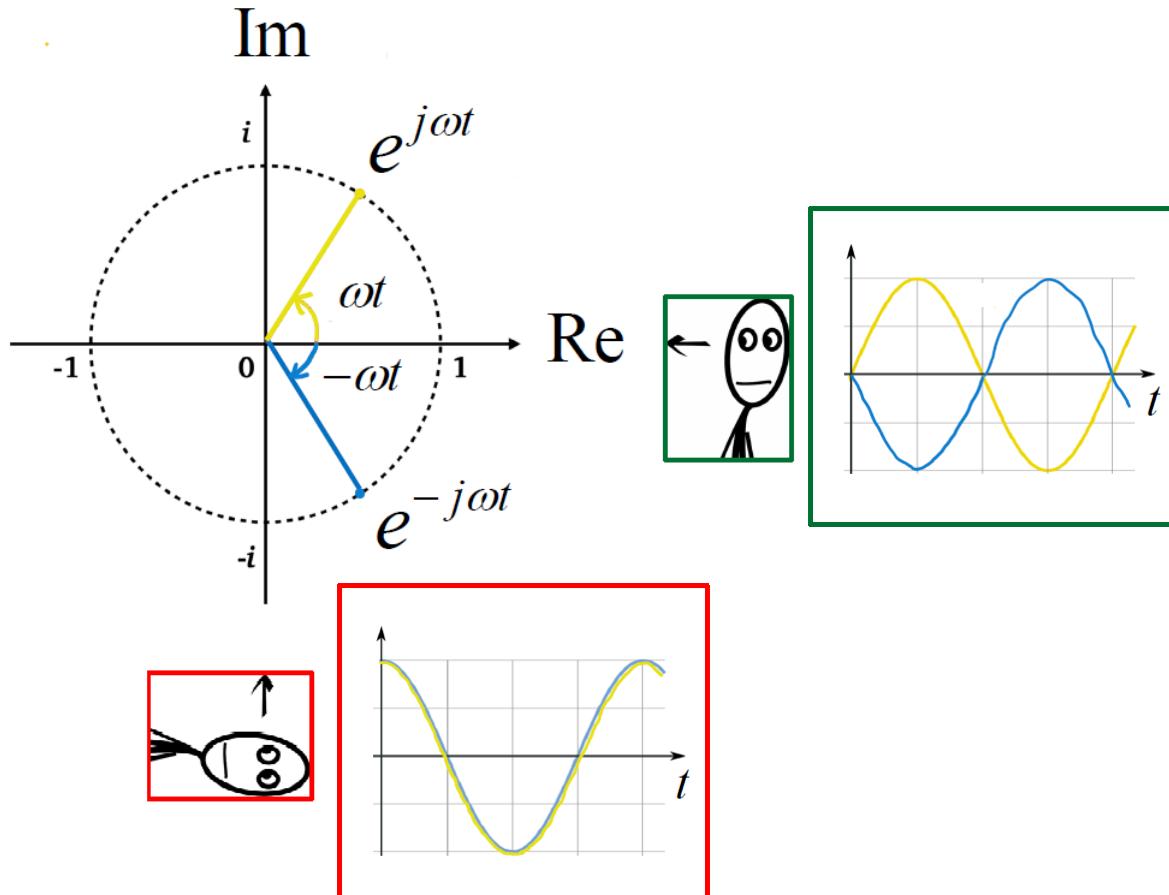


90 degrees out of phase,
A relative phase shift of $\pm \pi / 2$

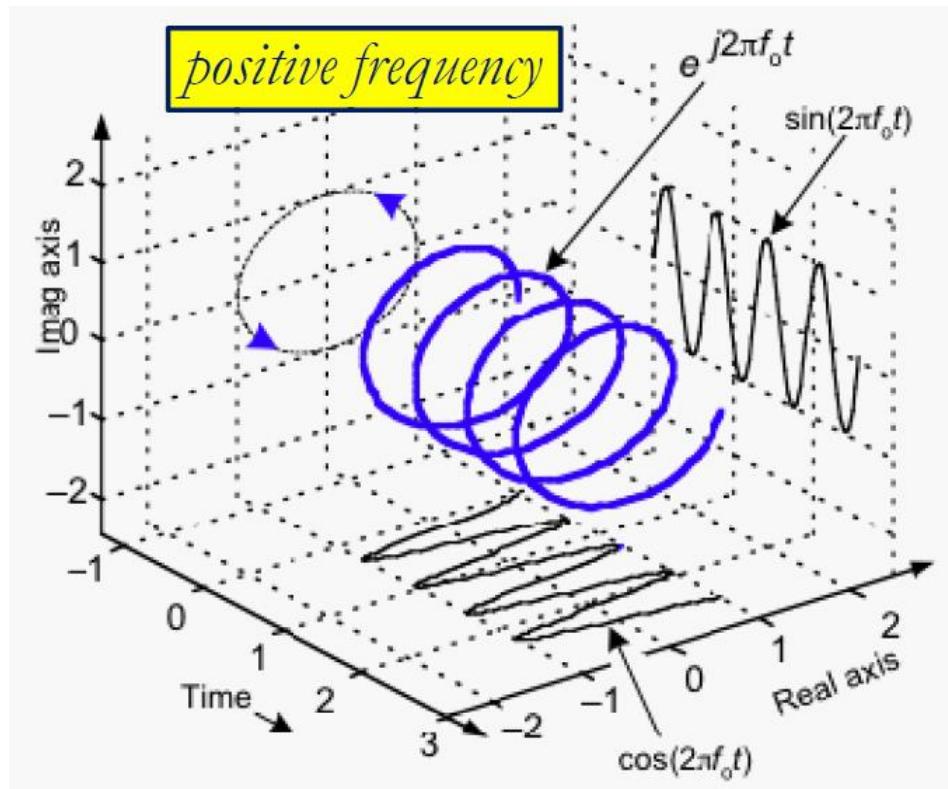
Complex Exponential

■ Euler's Relation

$$e^{j\omega t} = \cos \omega t + j \sin \omega t$$



Complex Exponential



Fourier Analysis

Fourier Analysis

■ Fourier



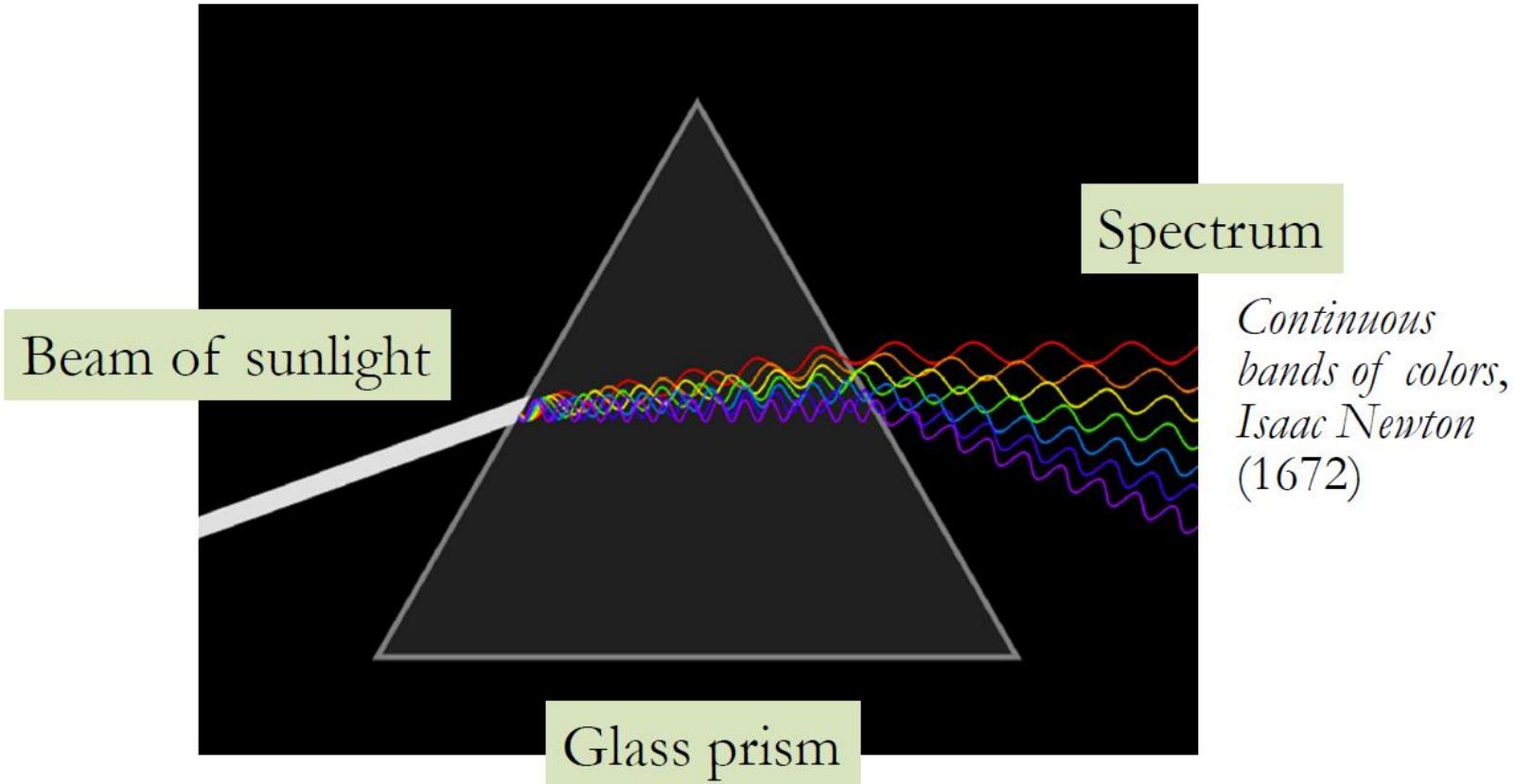
- Named after **Jean Baptiste Joseph Fourier**
(1768-1830)
- French mathematician and physicist

■ Fourier Analysis

- The analysis of a complex waveform expressed as a series of sinusoidal functions, the frequencies of which form a harmonic series

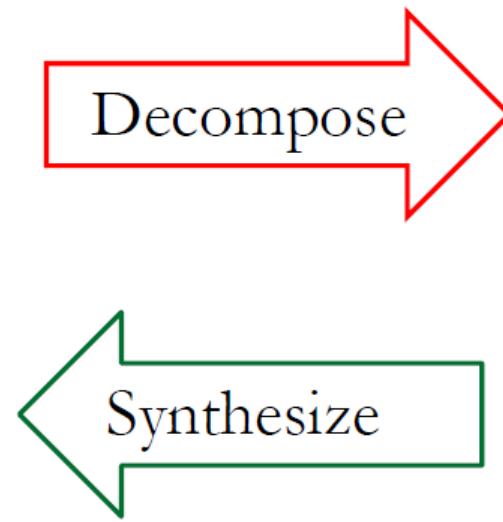
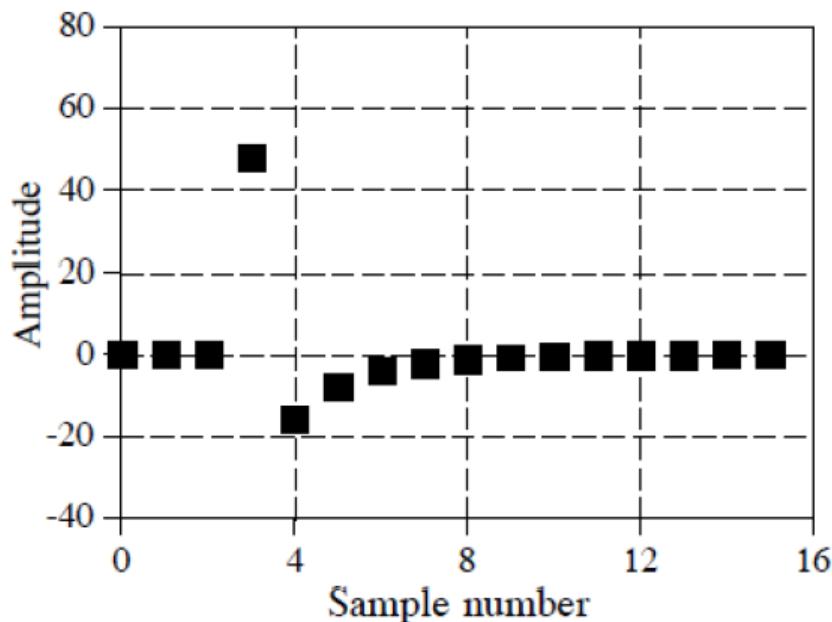
Analysis of the White Light

Analysis of the white light (sunlight) using glass prism



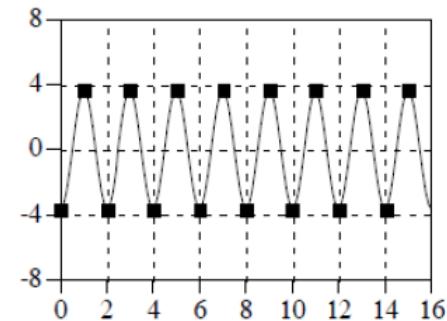
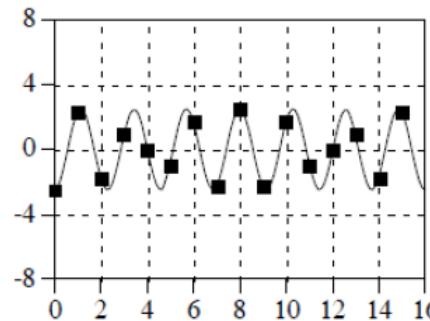
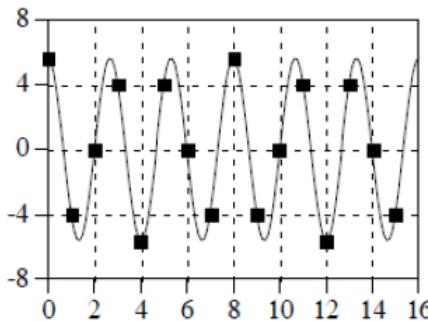
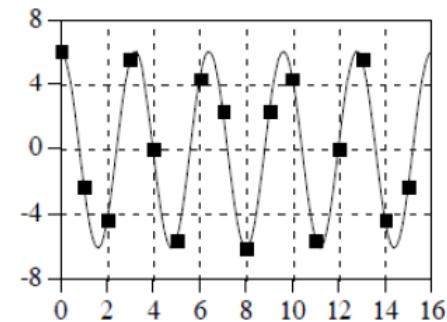
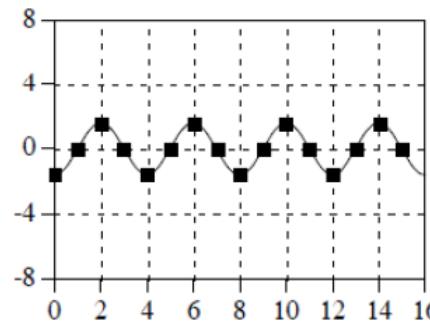
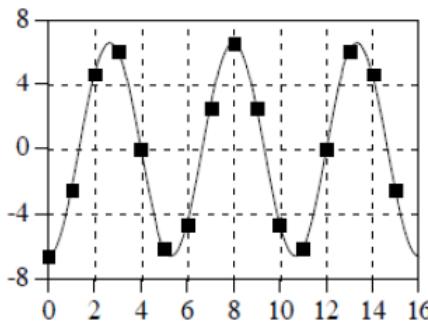
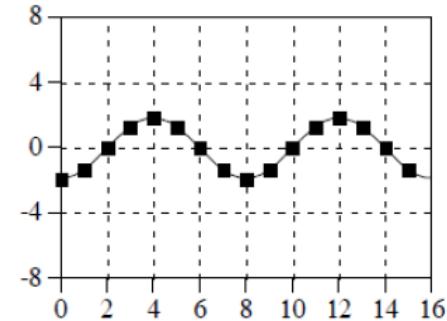
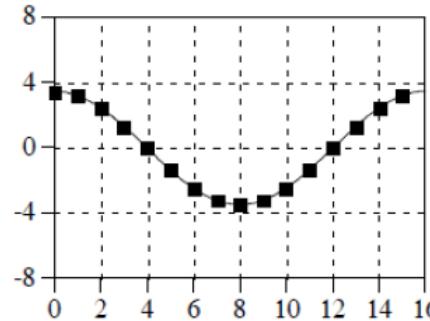
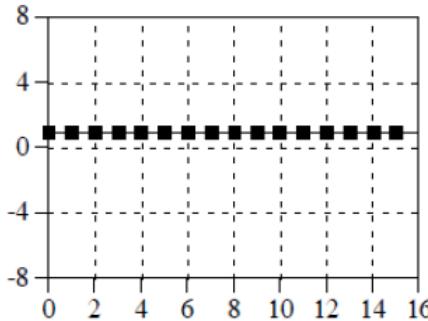
Fourier Decomposition

Example of Fourier decomposition: a 16 point signal



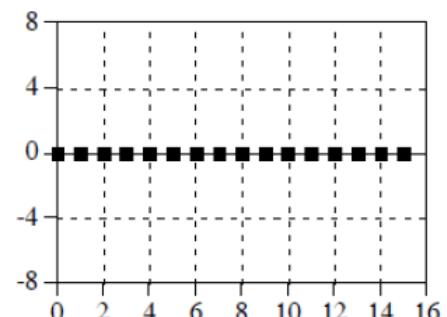
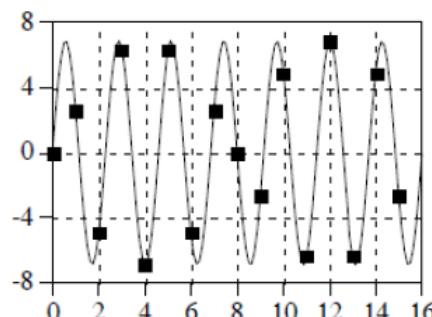
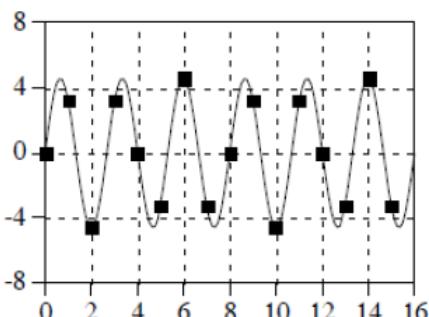
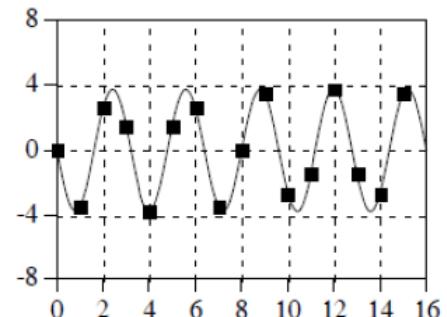
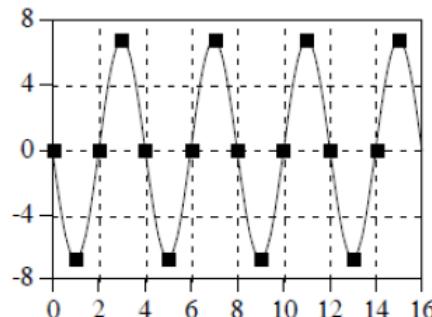
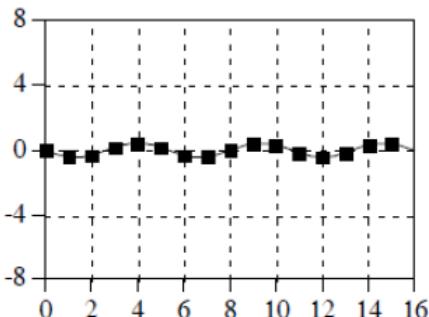
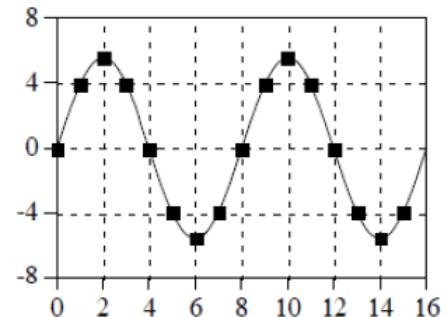
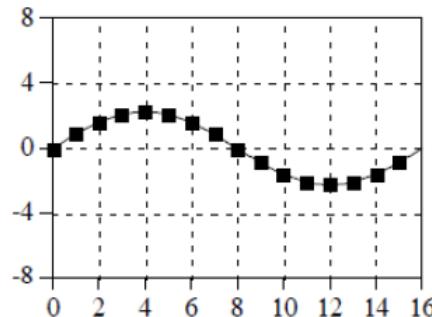
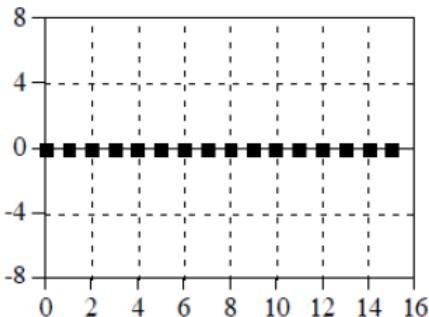
Fourier Decomposition

9 **cosine** waves, each with a different frequency and amplitude



Fourier Decomposition

9 **sine** waves, each with a different frequency and amplitude



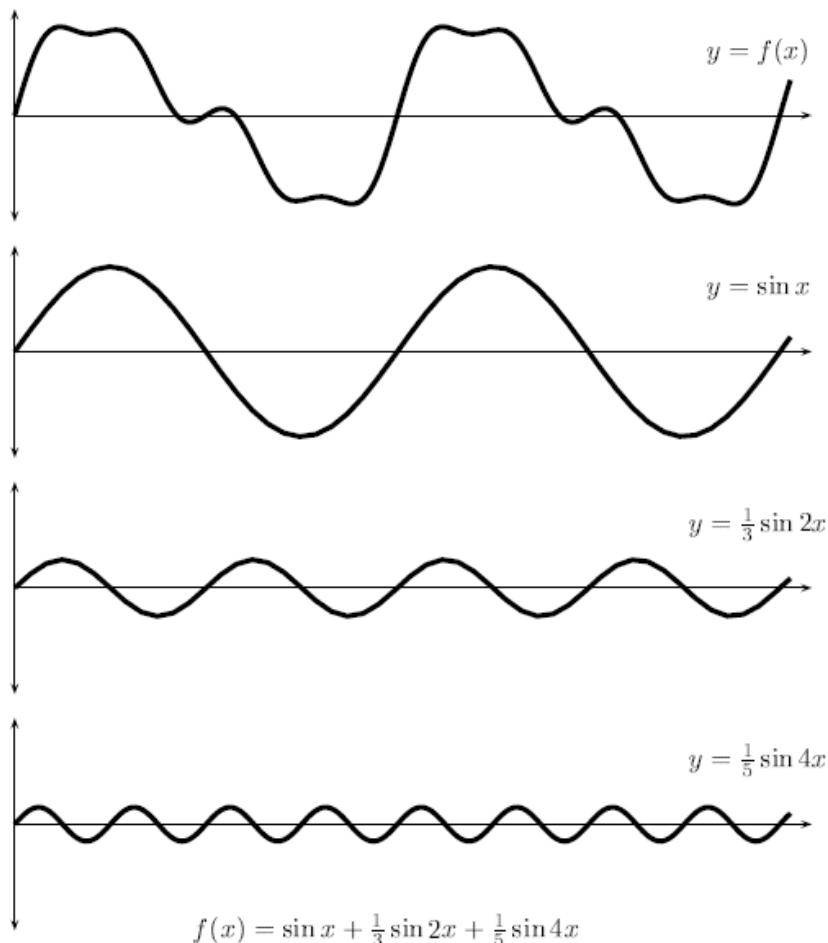
Categories of Fourier Transform

Type of Transform	Example Signal
Fourier Transform <i>signals that are continuous and aperiodic</i>	
Fourier Series <i>signals that are continuous and periodic</i>	
Discrete Time Fourier Transform <i>signals that are discrete and aperiodic</i>	
Discrete Fourier Transform <i>signals that are discrete and periodic</i>	

Fourier Series

Fourier Series Representation

■ Approximation of Periodic Signal



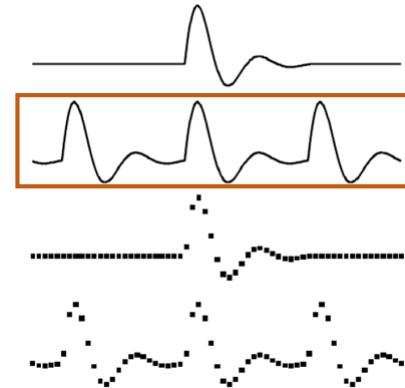
Continuous-time Fourier Series (CTFS)

■ Fourier Series Representation

→ (Almost) any periodic function $x(t)$ with fundamental frequency ω_0 can be described as a sum of sinusoids

Periodic Signal

$$x(t) = x(t + nT), \quad n \in \mathbb{Z} \quad (\text{integers})$$



Fourier Series: Infinite sum of sines and cosines

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t}, \quad \Omega_0 = 2\pi F_0 \quad \text{fundamental frequency}$$

Fourier Series Coefficients

$$C_k = \frac{1}{T} \int_{-T/2}^{T/2} x(t) e^{-jk\Omega_0 t} dt$$

Continuous-time Fourier Series (CTFS)

■ Fourier Series Representation

$$x(t) = a_0 + \sum_{k=1}^{\infty} a_k \cos(k\Omega_0 t) + \sum_{k=1}^{\infty} b_k \sin(k\Omega_0 t)$$

$$\begin{aligned}\cos\theta &= \frac{1}{2}(e^{j\theta} + e^{-j\theta}) \\ \sin\theta &= \frac{1}{2j}(e^{j\theta} - e^{-j\theta})\end{aligned}$$

$$= a_0 + \sum_{k=1}^{\infty} a_k \left(\frac{e^{jk\Omega_0 t} + e^{-jk\Omega_0 t}}{2} \right) + \sum_{k=1}^{\infty} b_k \left(\frac{e^{jk\Omega_0 t} - e^{-jk\Omega_0 t}}{2j} \right)$$

$$= a_0 + \sum_{k=1}^{\infty} a_k \left(\frac{e^{jk\Omega_0 t} + e^{-jk\Omega_0 t}}{2} \right) + \sum_{k=1}^{\infty} b_k \left(\frac{-je^{jk\Omega_0 t} + je^{-jk\Omega_0 t}}{2} \right)$$

$$= a_0 + \sum_{k=1}^{\infty} \left(\frac{a_k - jb_k}{2} \right) e^{jk\Omega_0 t} + \sum_{k=1}^{\infty} \left(\frac{a_k + jb_k}{2} \right) e^{-jk\Omega_0 t}$$

$$= \underbrace{a_0}_{C_0} + \underbrace{\sum_{k=1}^{\infty} \left(\frac{a_k - jb_k}{2} \right) e^{jk\Omega_0 t}}_{C_k \text{ for positive } k} + \underbrace{\sum_{k=-\infty}^{-1} \left(\frac{a_{-k} + jb_{-k}}{2} \right) e^{jk\Omega_0 t}}_{C_k \text{ for negative } k}, \quad k := -k$$

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t}$$

Continuous-time Fourier Series (CTFS)

■ Deriving Fourier Coefficient

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t}$$

$$x(t)e^{-jn\Omega_0 t} = \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t} e^{-jn\Omega_0 t}$$

$$\int_{-T/2}^{T/2} x(t)e^{-jn\Omega_0 t} dt = \sum_{k=-\infty}^{\infty} C_k \underbrace{\int_{-T/2}^{T/2} e^{jk\Omega_0 t} e^{-jn\Omega_0 t} dt}_{\begin{cases} 0, & k \neq n \\ T, & k = n \end{cases}}$$
$$= C_k T$$

$$C_k = \frac{1}{T} \int_{-T/2}^{T/2} x(t)e^{-jk\Omega_0 t} dt$$

Discrete-time Fourier Series (DTFS)

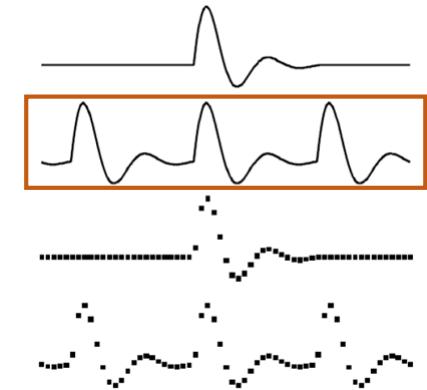
■ Continuous-time Fourier Series (CTFS)

Fourier Series

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t}, \quad \Omega_0 = 2\pi F_0 \text{ fundamental frequency}$$

Fourier Series Coefficient

$$C_k = \frac{1}{T} \int_{-T/2}^{T/2} x(t) e^{-jk\Omega_0 t} dt$$



■ Discrete-time Fourier Series (DTFS)

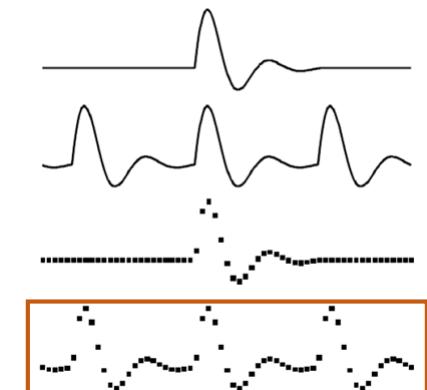
Fourier Series

$$x[n] = \sum_{k=-N}^{N-1} C_k e^{jkn\frac{2\pi}{N}}$$

Fourier Series Coefficients

$$C_k = \frac{1}{T} \int_0^T x(t) e^{-j\frac{2\pi}{T}kt} dt \xrightarrow[T=NT_s]{t=nT_s} \frac{1}{NT_s} \sum_{n=0}^{N-1} \tilde{x}[n] e^{-j\frac{2\pi}{NT_s}knT_s}$$

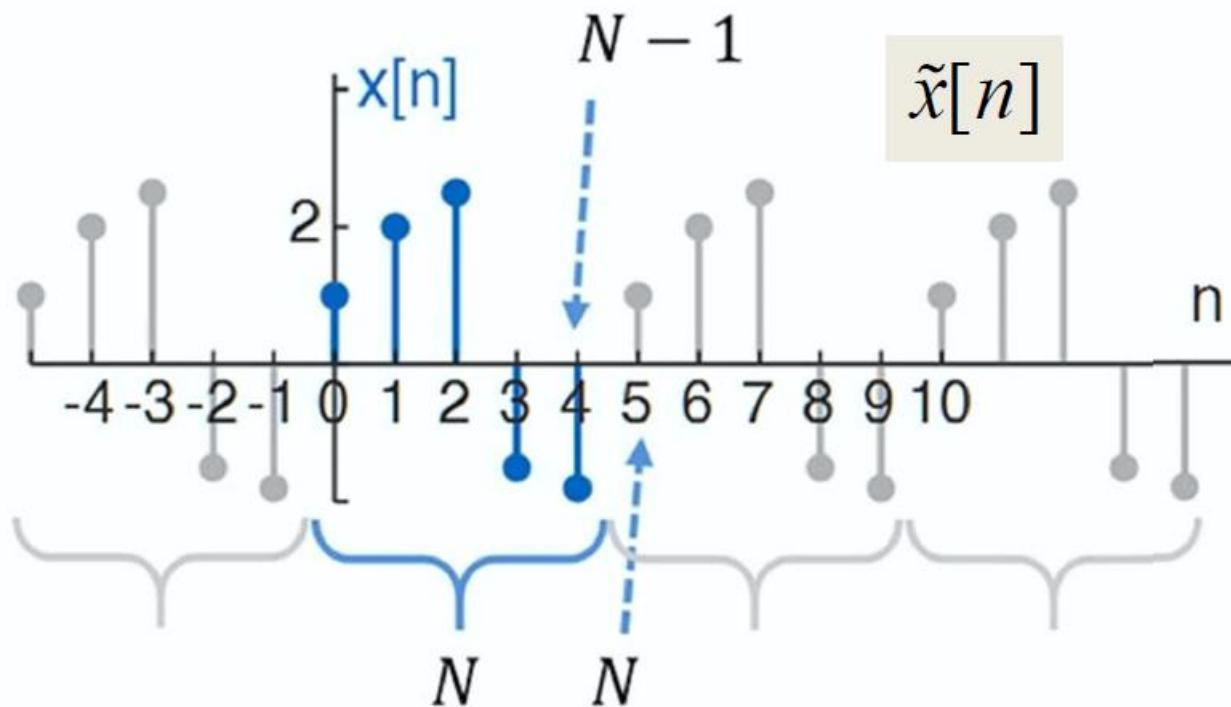
$$C_k = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jkn\frac{2\pi}{N}}$$



Discrete-time Fourier Series (DTFS)

■ Aperiodic and Periodic Signals

$$x[n] \rightarrow \tilde{x}[n]$$



Fourier Transform

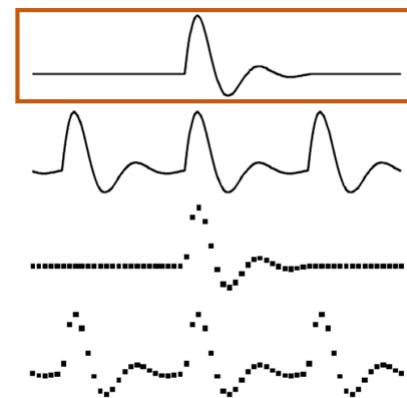
Continuous-time Fourier Transform (CTFT)

■ Fourier Transform

Aperiodic Signal

$$x(t) = x(t + nT), \quad n \in \mathbb{Z} \quad (\text{integers})$$

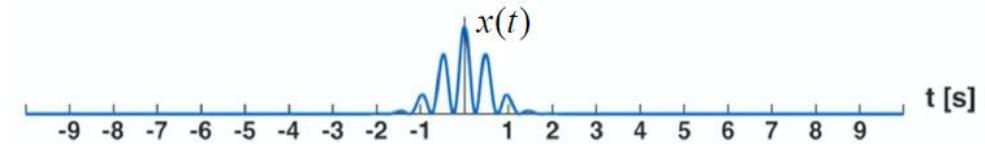
$$T \rightarrow \infty$$



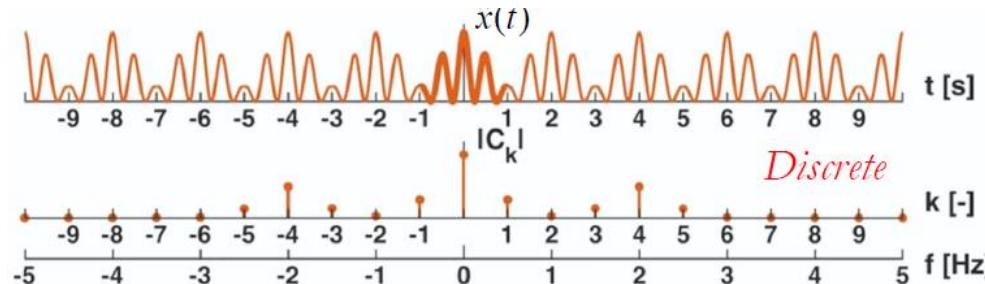
Fourier Transform

$$X(j\Omega) = \int_{-\infty}^{\infty} x(t) e^{-j\Omega t} dt$$

Aperiodic & Periodic Signals

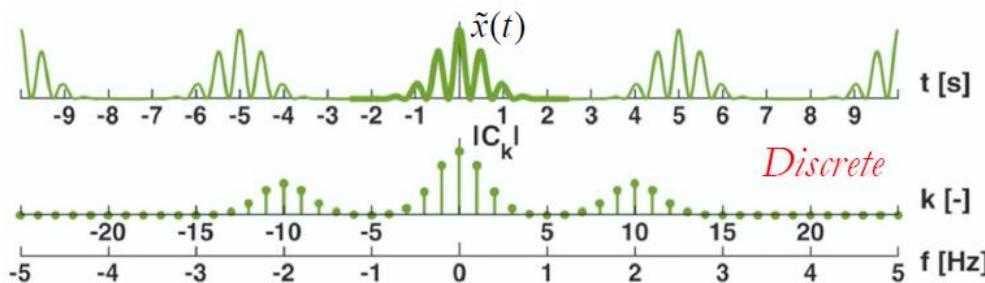


*a*periodic



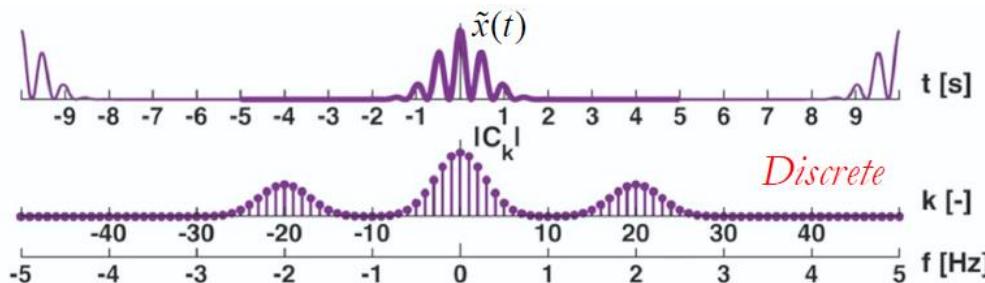
$$T = 2s$$

$$F_0 = 0.5 \text{ Hz}$$



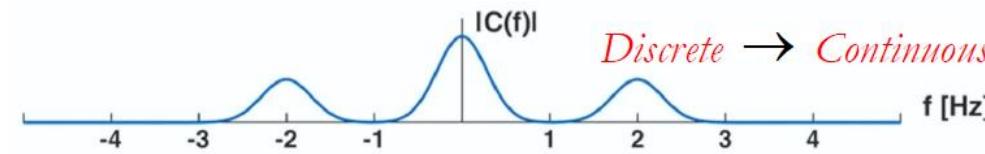
$$T = 5s$$

$$F_0 = 0.2 \text{ Hz}$$



$$T = 10s$$

$$F_0 = 0.1 \text{ Hz}$$



$$T \rightarrow \infty$$

$$\Delta F \rightarrow 0$$

Continuous-time Fourier Transform (CTFT)

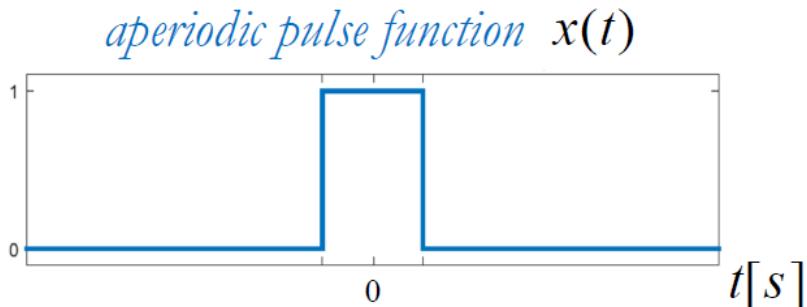
$$\begin{aligned}
 \tilde{x}(t) &= \sum_{k=-\infty}^{\infty} C_k e^{jk\Omega_0 t}, \quad C_k = \frac{1}{T} \int_{-T/2}^{T/2} \tilde{x}(t) e^{-jk\Omega_0 t} dt \quad \text{discrete spectra} \\
 &= \sum_{k=-\infty}^{\infty} \left[\frac{1}{T} \int_{-T/2}^{T/2} \tilde{x}(\tau) e^{-jk\Omega_0 \tau} d\tau \right] e^{jk\Omega_0 t}, \quad \Omega_0 = \frac{2\pi}{T} \rightarrow \frac{1}{T} = \frac{\Omega_0}{2\pi} \\
 &= \frac{1}{2\pi} \sum_{k=-\infty}^{\infty} \left[\int_{-T/2}^{T/2} \tilde{x}(\tau) e^{-jk\Omega_0 \tau} d\tau \right] \Omega_0 e^{jk\Omega_0 t} \\
 &= \frac{1}{2\pi} \sum_{k=-\infty}^{\infty} \left[\int_{-T/2}^{T/2} \tilde{x}(\tau) e^{-jk\Omega_0 \tau} d\tau \right] e^{jk\Omega_0 t} \Delta\Omega, \quad \Delta\Omega = (k+1)\Omega - k\Omega = \Omega_0 \\
 &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} \tilde{x}(\tau) e^{-j\Omega\tau} d\tau \right] e^{j\Omega t} d\Omega, \quad T \rightarrow \infty
 \end{aligned}$$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \underbrace{\left[\int_{-\infty}^{\infty} x(\tau) e^{-j\Omega\tau} d\tau \right]}_{X(j\Omega)} e^{j\Omega t} d\Omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) e^{j\Omega t} d\Omega$$

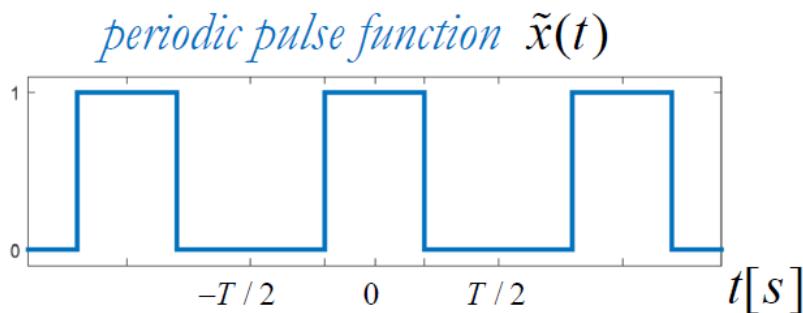
X(j\Omega) continuous spectra

Continuous-time Fourier Transform (CTFT)

■ Relationship between FS and CTFT



$$X(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt = \int_{-T/2}^{T/2} x(t)e^{-j\Omega t} dt$$



$$C_k = \frac{1}{T} \int_{-T/2}^{T/2} \tilde{x}(t)e^{-jk\Omega_0 t} dt = \frac{1}{T} \int_{-T/2}^{T/2} x(t)e^{-jk\Omega_0 t} dt$$

Discrete-time Fourier Transform (DTFT)

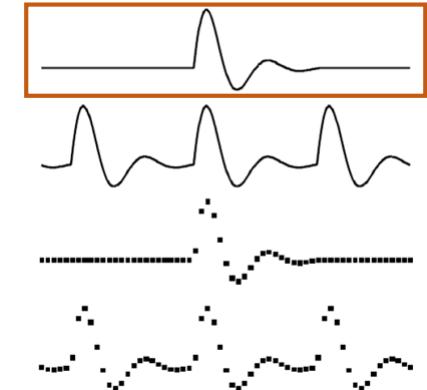
■ Continuous-time Fourier Transform (CTFT)

Aperiodic Signal

$$x(t) = x(t + nT), \quad n \in \mathbb{Z} \quad (\text{integers})$$
$$T \rightarrow \infty$$

Fourier Transform

$$X(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$



■ Discrete-time Fourier Transform (DTFT)

Aperiodic Signal

$$t = nT_s, \quad n \in \mathbb{Z}$$

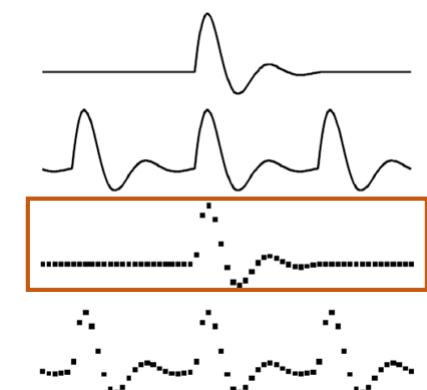
$$x(t) \xrightarrow{t=nT_s} x(nT_s) = x[n]$$

Fourier Transform

$$\int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt \xrightarrow{t=nT_s} \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega nT_s}$$

$$\Omega T_s = 2\pi \frac{F}{F_s} = \omega$$

normalized frequency



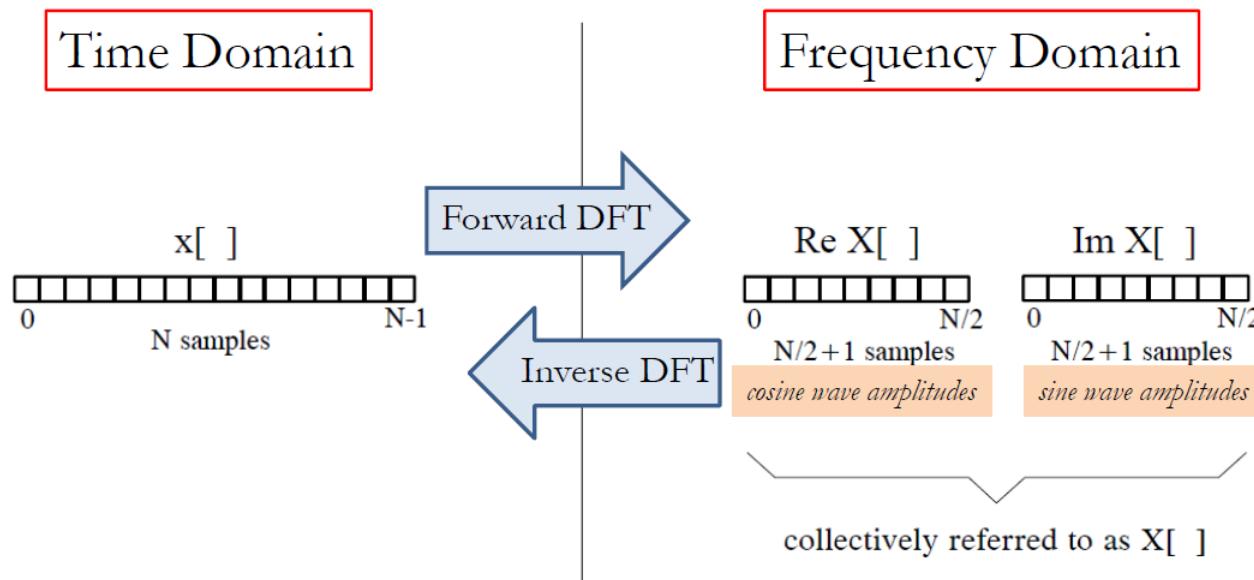
$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n}$$

Discrete Fourier Transform (DFT)

Discrete Fourier Transform (DFT)

■ Discrete Fourier Transform

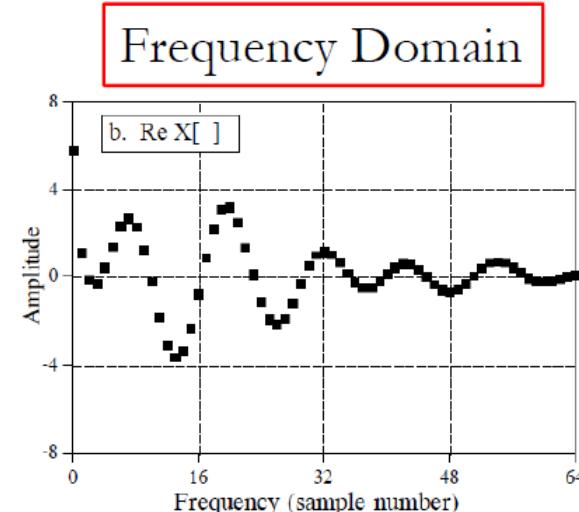
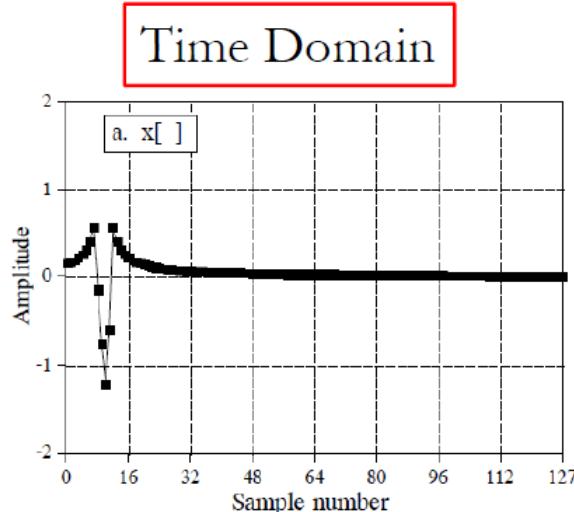
- Changes an N point input signal into two $\frac{N}{2} - 1$ point output signals
- The input signal contains the signal being decomposed, while the two output signals contain the amplitudes of the component sine and cosine waves



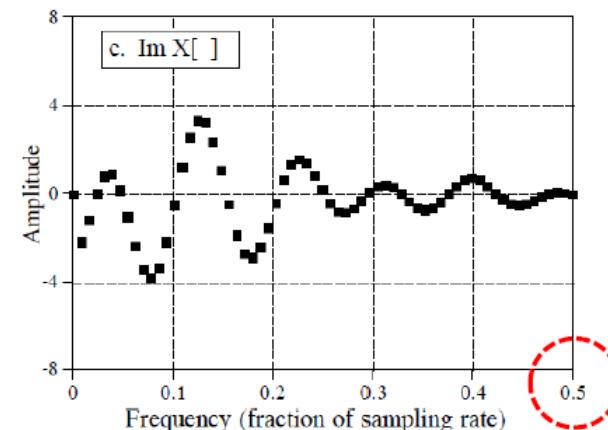
- decomposition, analysis, *the* forward DFT, *the* DFT
- synthesis, *the* inverse DFT

Independent Variables (Frequency Domain)

■ Horizontal Axis of the Frequency Domain



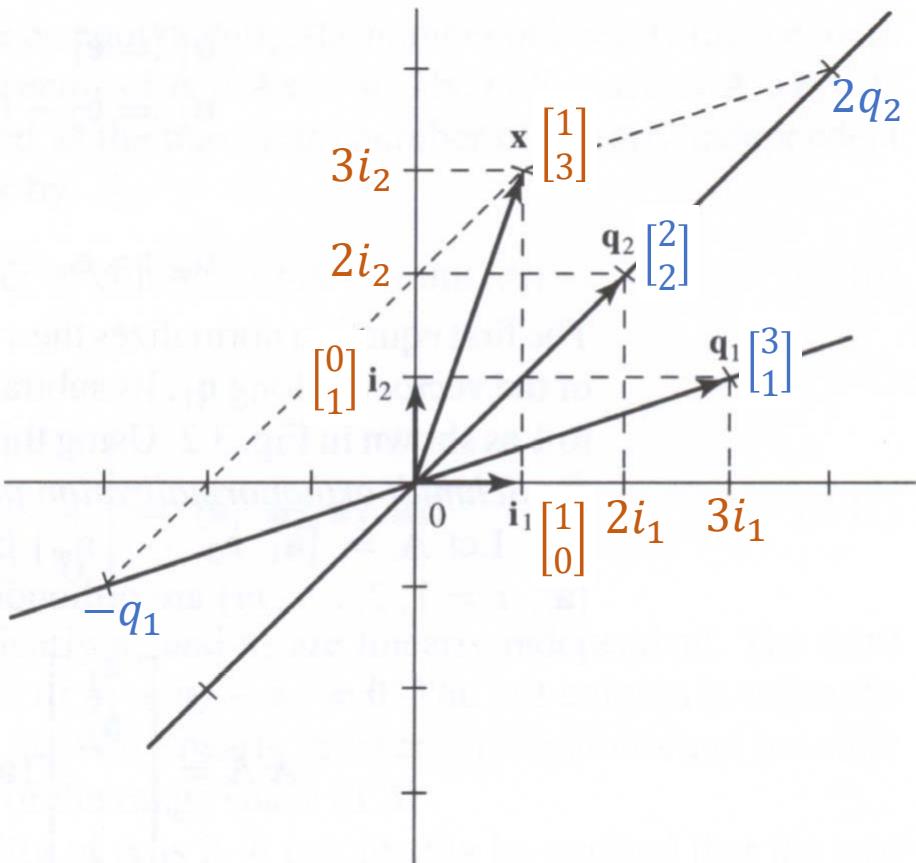
1. Samples
 $c[n] = \cos(2\pi kn / N)$
2. fraction of sampling rate
 $c[n] = \cos(2\pi fn)$
3. natural frequency [rad]
 $c[n] = \cos(\omega n)$
4. analog frequency



Basis Function

■ Review:

→ Different representation of vector x



$$x = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

Representation of x
with respect to the basis $\{i_1, i_2\}$

$$x = Q \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

Representation of x
with respect to the basis $\{q_1, q_2\}$

$$x = Q \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

Find representation of x with basis $\{q_2, i_2\}$

DFT Basis Functions

■ Basis Functions:

→ A set of **sine** and **cosine** waves with unity amplitude

$$c_k[i] = \cos(2\pi k i / N)$$

← Re $X[k]$

$$s_k[i] = \sin(2\pi k i / N)$$

← Im $X[k]$

→ The output of the DFT is a set of numbers that represent amplitudes

Assign each amplitude (the frequency domain) to the proper sine or cosine waves (the basis functions)



A set of scaled sine and cosine waves that can be added to form the time domain signal.

DFT Basis Functions

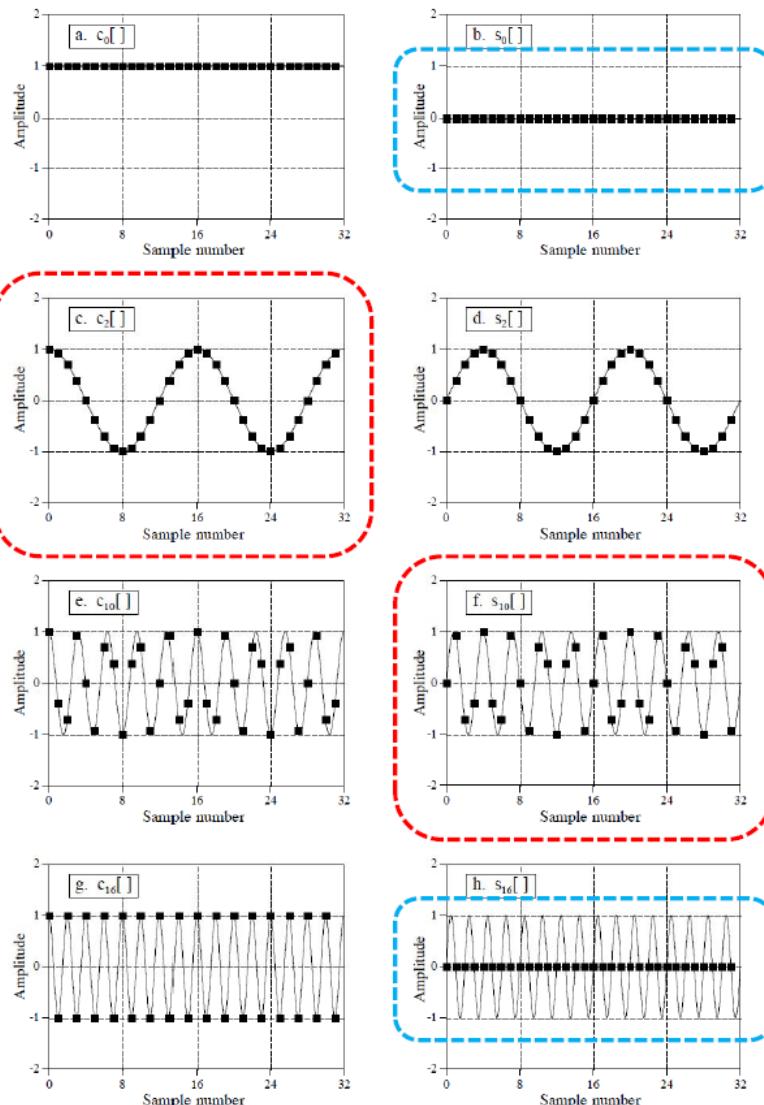
■ Example: 32-point DFT

→ Has 17 discrete cosine & 17 discrete sine waves for its basis functions

$$c_2[] \rightarrow$$

*The cosine wave
that makes
two complete cycles
in N points*

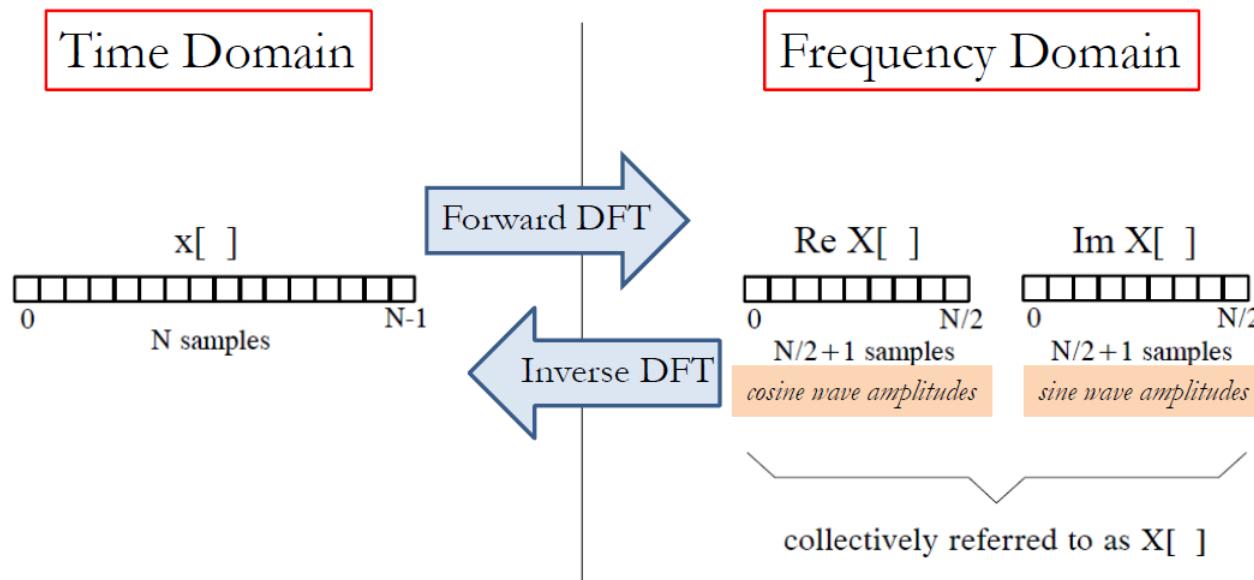
$$\text{Re } X[2]$$



Discrete Fourier Transform (DFT)

■ Discrete Fourier Transform

- Changes an N point input signal into two $\frac{N}{2} + 1$ point output signals
- The input signal contains the signal being decomposed, while the two output signals contain the amplitudes of the component sine and cosine waves



- decomposition, analysis, *the* forward DFT, *the* DFT
- synthesis, *the* inverse DFT

Analysis

■ Analysis: Calculating the DFT

→ Three different ways:

1. Simultaneous Equation

- ✓ Given N values from the time domain, calculate the N values of the frequency domain
- ✓ Basic algebra: to solve for N unknowns, we need N linearly independent equations
- ✓ Too inefficient to be of practical use

2. Correlations

- ✓ Correlating the input signal with each basis function
- ✓ When a particular sine wave is present, it produces the value. Otherwise, it will be zero.

$$ReX[k] = \sum_{i=0}^{N-1} x[i] \cos(2\pi k i / N)$$
$$ImX[k] = - \sum_{i=0}^{N-1} x[i] \sin(2\pi k i / N)$$

3. Fast Fourier Transform (FFT)

- ✓ An ingenious algorithm that decomposes a DFT with N points, into N DFTs each with a single point.

Analysis

Analysis

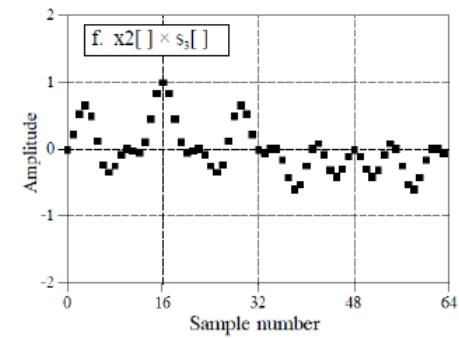
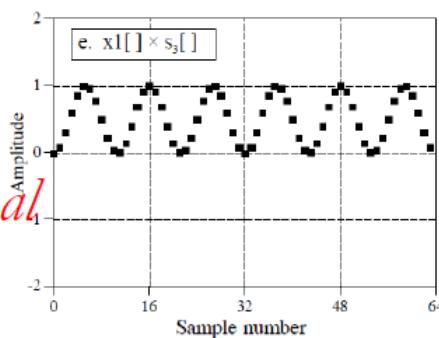
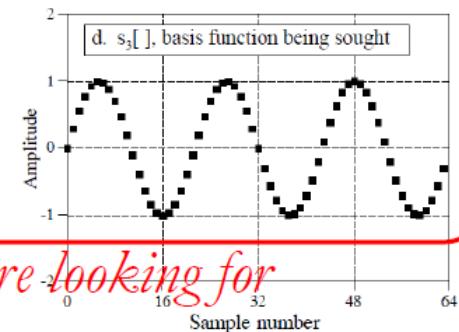
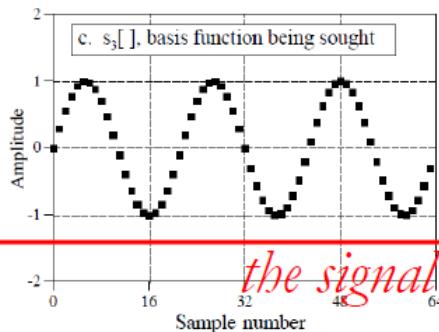
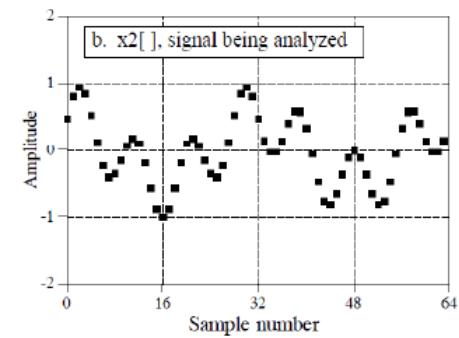
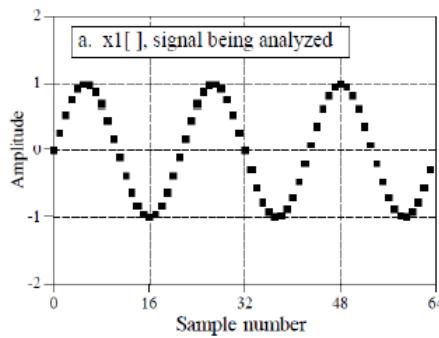
$$\operatorname{Re} X[k] =$$

$$\sum_{i=0}^{N-1} x[i] \cos(2\pi k i / N)$$

$$\operatorname{Im} X[k] =$$

$$-\sum_{i=0}^{N-1} x[i] \sin(2\pi k i / N)$$

*Correlating the input signal
with each basis function!*



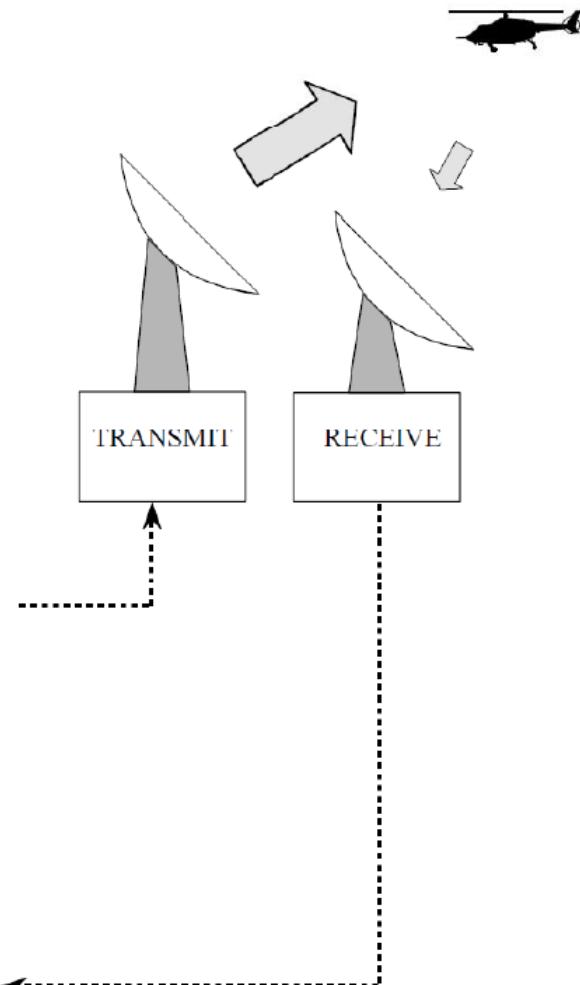
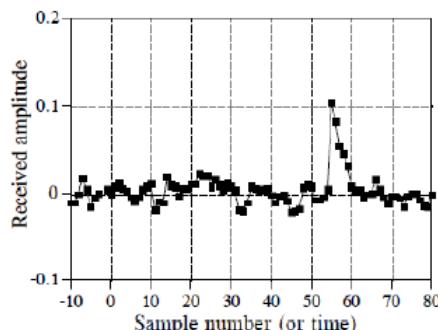
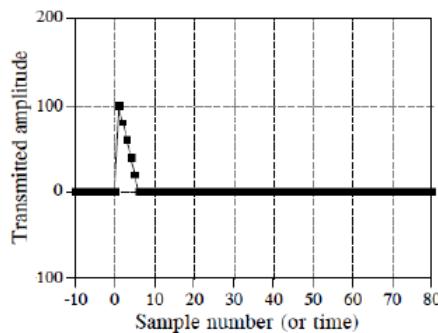
the signal we're looking for

Detect a known waveform contained in another signal.

Analysis

Radar (RAdio Detection And Ranging)

Detection of a known waveform in a noisy signal is the fundamental problem in echo location: correlation



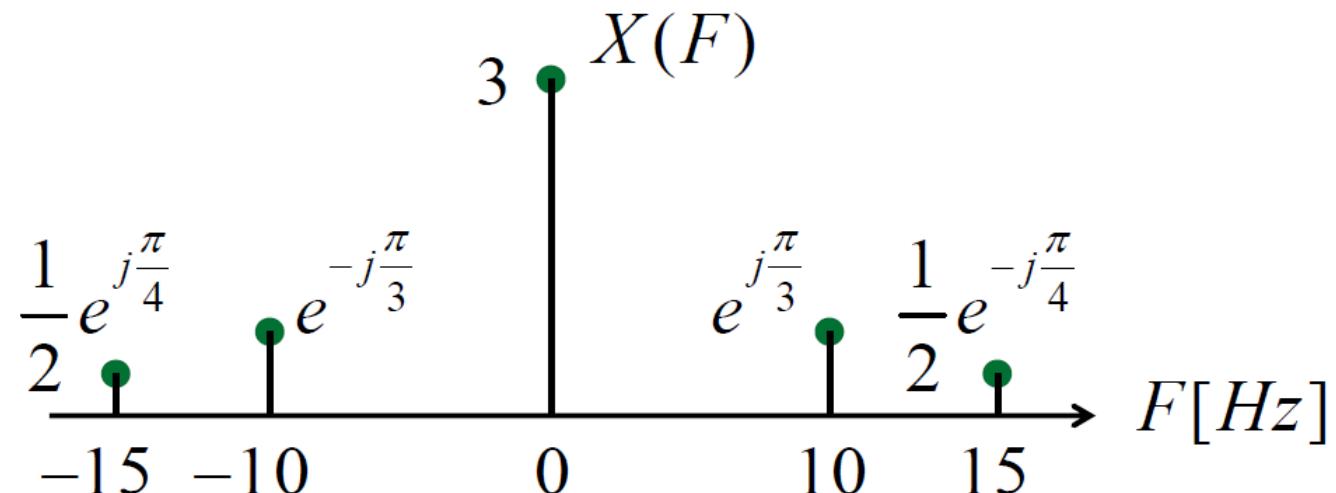
Example

Frequency Spectrum of a Signal

$$x(t) = 3 + 2 \cos\left(20\pi t + \frac{\pi}{3}\right) + \cos\left(30\pi t - \frac{\pi}{4}\right)$$

$$= 3 + \frac{2}{2} e^{j\frac{\pi}{3}} e^{j2\pi 10t} + \frac{2}{2} e^{-j\frac{\pi}{3}} e^{-j2\pi 10t}$$

$$+ \frac{1}{2} e^{-j\frac{\pi}{4}} e^{j2\pi 15t} + \frac{1}{2} e^{j\frac{\pi}{4}} e^{-j2\pi 15t}$$



Synthesis

■ Synthesis: Calculating the Inverse DFT (IDFT)

- Any N point signal, $x[i]$, can be created by adding $\frac{n}{2} + 1$ cosine & $\frac{n}{2} + 1$ sine waves.
- The amplitudes of the cosine & sine waves are held in the arrays $\text{Re}\bar{X}[k]$ & $\text{Im}\bar{X}[k]$

$$x[i] = \sum_{k=0}^{N/2} \text{Re } \bar{X}[k] \cos(2\pi ki / N) + \sum_{k=0}^{N/2} \text{Im } \bar{X}[k] \sin(2\pi ki / N)$$

$$\text{Re } \bar{X}[k] = \frac{\text{Re } X[k]}{N/2}$$

$$\text{Im } \bar{X}[k] = -\frac{\text{Im } X[k]}{N/2}$$

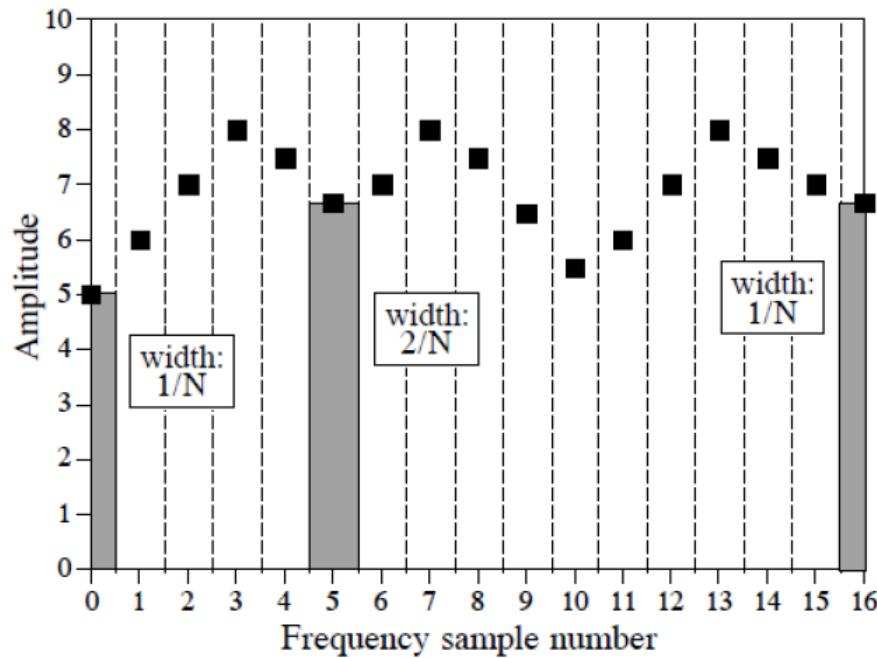
$$\text{Re } \bar{X}[0] = \frac{\text{Re } X[0]}{N}$$

$$\text{Im } \bar{X}[N/2] = -\frac{\text{Im } X[N/2]}{N}$$

Synthesis

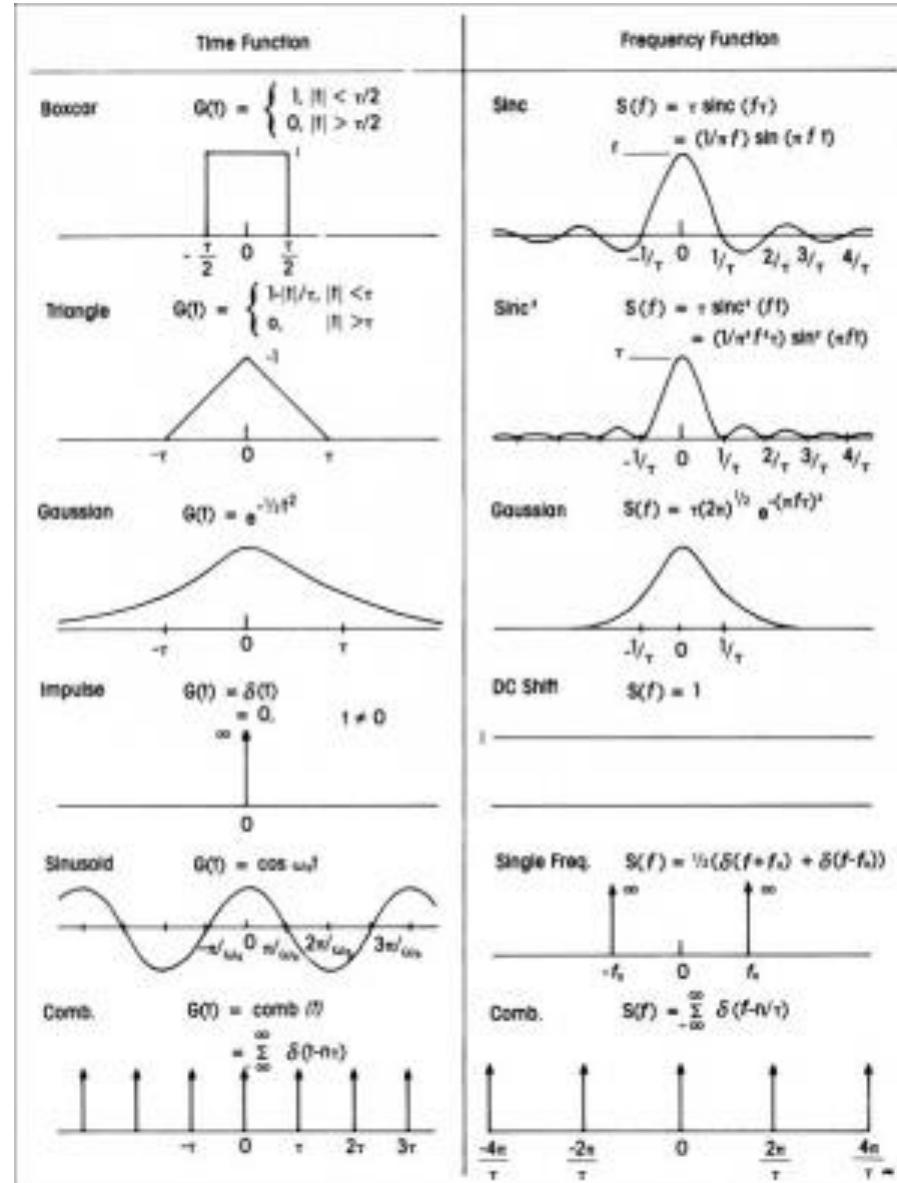
The frequency domain is defined as a *spectral density*:

How much signal (amplitude) is present *per unit of bandwidth*?



Each sample in the frequency domain can be thought of being contained in the frequency band of width $2/N$.

Fourier Transform Pairs



Discrete-time Fourier Transform

Discrete-time Fourier Transform (DTFT)

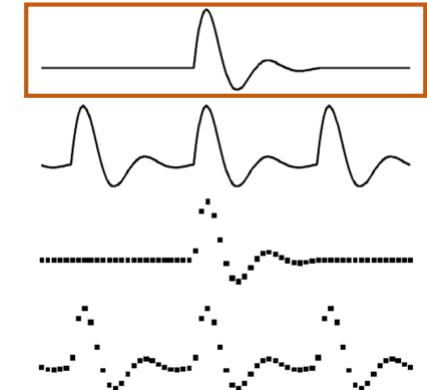
■ Continuous-time Fourier Transform (CTFT)

Aperiodic Signal

$$x(t) = x(t + nT), \quad n \in \mathbb{Z} \quad (\text{integers})$$
$$T \rightarrow \infty$$

Fourier Transform

$$X(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$



■ Discrete-time Fourier Transform (DTFT)

Aperiodic Signal

$$t = nT_s, \quad n \in \mathbb{Z}$$

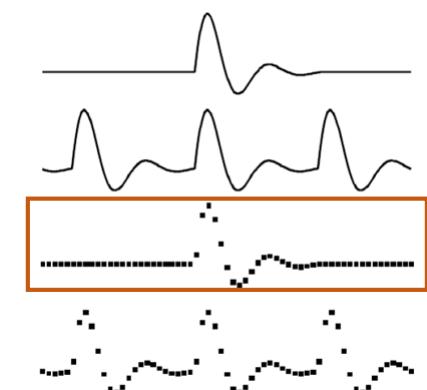
$$x(t) \xrightarrow{t=nT_s} x(nT_s) = x[n]$$

Fourier Transform

$$\int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt \xrightarrow{t=nT_s} \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega nT_s}$$

$$\Omega T_s = 2\pi \frac{F}{F_s} = \omega$$

normalized frequency



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

Discrete-time Fourier Transform (DTFT)

If $x(k)$ is absolutely summable, that is, $\sum_{k=-\infty}^{\infty} |x(k)| < \infty$, then its DTFT is given by

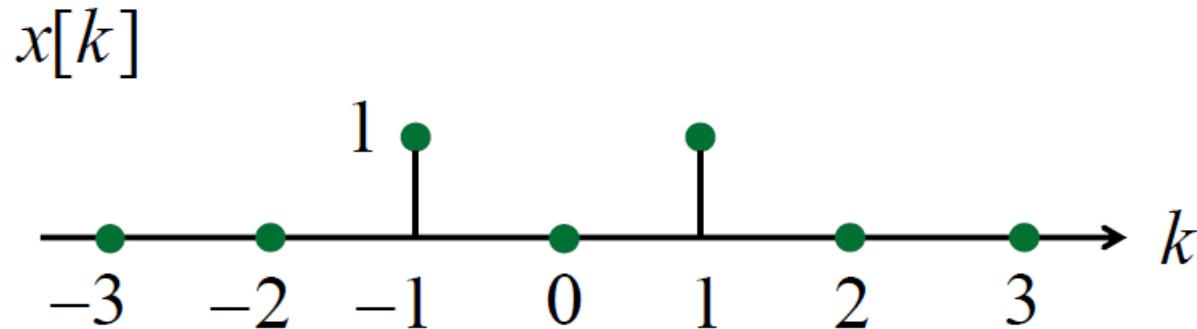
$$X(e^{j\omega}) \equiv F[x(k)] = \sum_{k=-\infty}^{\infty} x(k) e^{-j\omega k}$$

The inverse discrete-time Fourier transform (IDTFT) of $X(e^{j\omega})$ is given by

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

Example

- DTFT



$$x[k] = \delta[k - 1] + \delta[k + 1]$$

$$\begin{aligned} X(e^{j\omega}) &= 1 \cdot e^{-j(-1)\omega} + 1 \cdot e^{-j(1)\omega} \\ &= e^{j\omega} + e^{-j\omega} \\ &= 2 \cos \omega \end{aligned}$$

Example

■ Inverse DTFT (IDTFT)

$$X(e^{j\omega}) = 2 \cos \omega$$

$$x[k] = \frac{1}{2\pi} \int_{-\pi}^{\pi} 2 \cos \omega e^{j\omega k} d\omega$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} [2 \cos \omega \cos \omega k + 2 \cos \omega j \sin \omega k] d\omega$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \{ [\cos(1-k)\omega + \cos(1+k)\omega] + j[\sin(1+k)\omega - \sin(1-k)\omega] \} d\omega$$

$$= \begin{cases} \frac{1}{2\pi} \cdot 2\pi = 1, & k = 1 \\ \frac{1}{2\pi} \cdot 2\pi = 1, & k = -1 \end{cases}$$

Product-to-sum identities

$$2 \cos \theta \cos \varphi = \cos(\theta - \varphi) + \cos(\theta + \varphi)$$

$$2 \cos \theta \sin \varphi = \sin(\theta + \varphi) - \sin(\theta - \varphi)$$

Example

■ DTFT: Finite & Infinite Duration

infinite duration

Determine the discrete-time Fourier transform of step function

$$x(k) = (0.5)^k u(k).$$

$$\begin{aligned} X(e^{j\omega}) &= \sum_{-\infty}^{\infty} x(k) e^{-j\omega k} = \sum_0^{\infty} (0.5)^k e^{-j\omega k} \\ &= \sum_0^{\infty} (0.5e^{-j\omega})^k = \frac{1}{1 - 0.5e^{-j\omega}} = \frac{e^{j\omega}}{e^{j\omega} - 0.5} \end{aligned}$$

finite duration

Determine the discrete-time Fourier transform of the following finite-duration sequence: $x(k) = \{1, 2, 3, 4, 5\}$

$$X(e^{j\omega}) = \sum_{-\infty}^{\infty} x(k) e^{-j\omega k} = e^{j\omega} + 2 + 3e^{-j\omega} + 4e^{-j2\omega} + 5e^{-j3\omega}$$

Some Common DTFT Pairs

$$x[k] \quad X(e^{j\omega}), -\pi \leq \omega \leq \pi$$

$$\delta[k] \quad 1$$

$$1 \quad 2\pi\delta(\omega)$$

$$u[k] \quad \frac{1}{1-e^{-j\omega}} + \pi\delta(\omega)$$

$$\alpha^k u[k] \quad \frac{1}{1-\alpha e^{-j\omega}}$$

$$e^{j\omega_0 k} \quad 2\pi\delta(\omega - \omega_0)$$

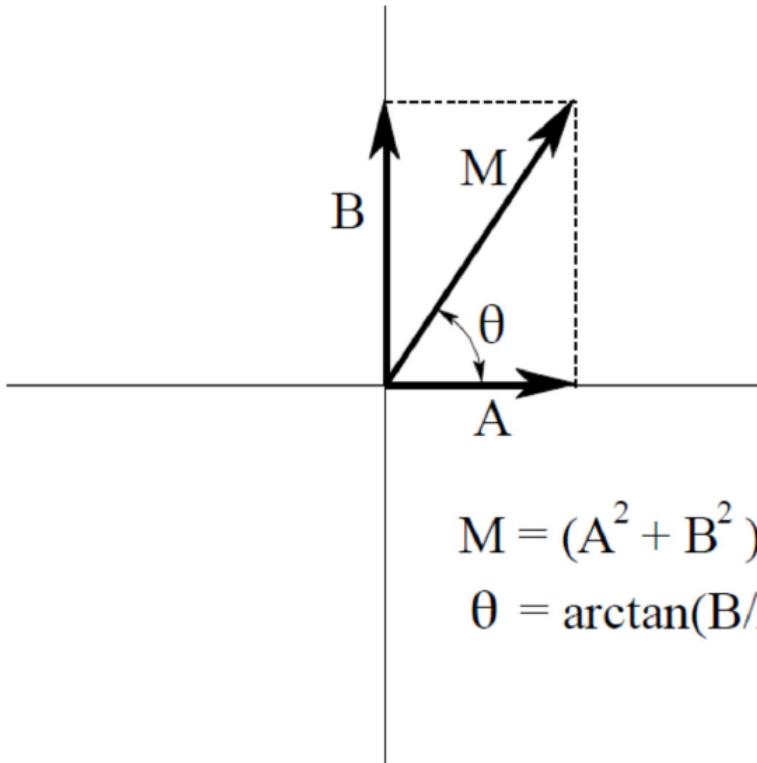
$$\cos[\omega_0 k] \quad \pi [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$$

$$\sin[\omega_0 k] \quad j\pi [\delta(\omega + \omega_0) - \delta(\omega - \omega_0)]$$

$$\alpha^{|k|} u[k] \quad \frac{1-\alpha^2}{1-2\alpha \cos(\omega) + \alpha^2}$$

Rectangular & Polar Form

■ Rectangular-to-Polar Conversion



$$M = (A^2 + B^2)^{1/2}$$

$$\theta = \arctan(B/A)$$

$$MagX[k] = (\operatorname{Re} X[k]^2 + \operatorname{Im} X[k]^2)^{1/2}$$

$$\operatorname{Phase} X[k] = \arctan\left(\frac{\operatorname{Im} X[k]}{\operatorname{Re} X[k]}\right)$$

$$\operatorname{Re} X[k] = MagX[k] \cos(\operatorname{Phase} X[k])$$

$$\operatorname{Im} X[k] = MagX[k] \sin(\operatorname{Phase} X[k])$$

Rectangular & Polar Form

■ Rectangular Form

Rectangular coordinates: rarely used in signal processing

$$X(\omega) = X_R(\omega) + j X_I(\omega)$$

where $X_R(\omega), X_I(\omega) \in \mathbb{R}$.

■ Polar Form

Polar coordinates: more intuitive way to represent frequency content

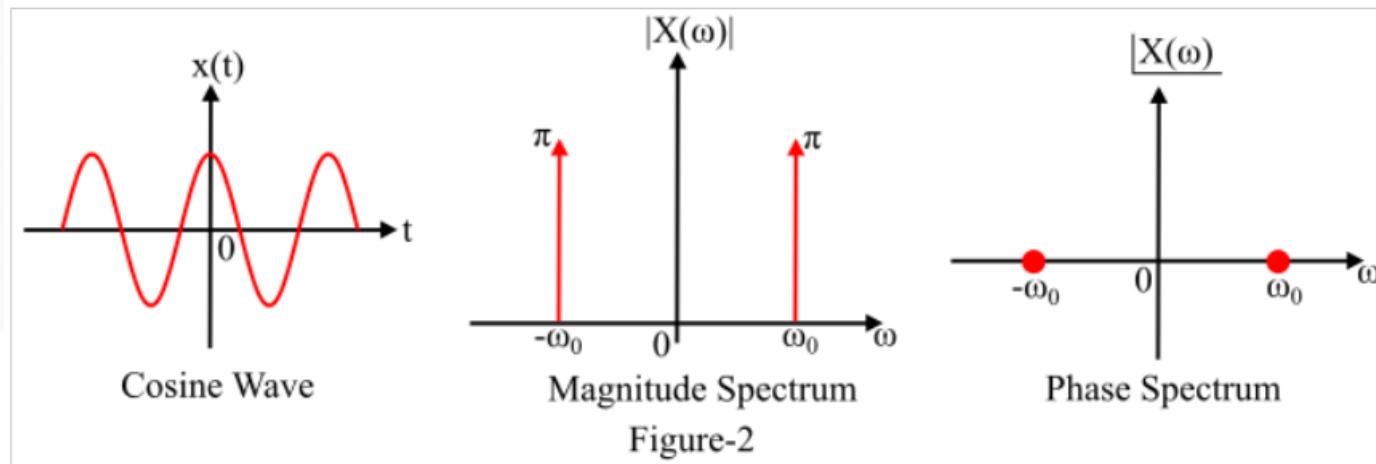
$$X(\omega) = |X(\omega)| e^{j\Theta(\omega)}$$

where $|X(\omega)|, \Theta(\omega) = \angle X(\omega) \in \mathbb{R}$.

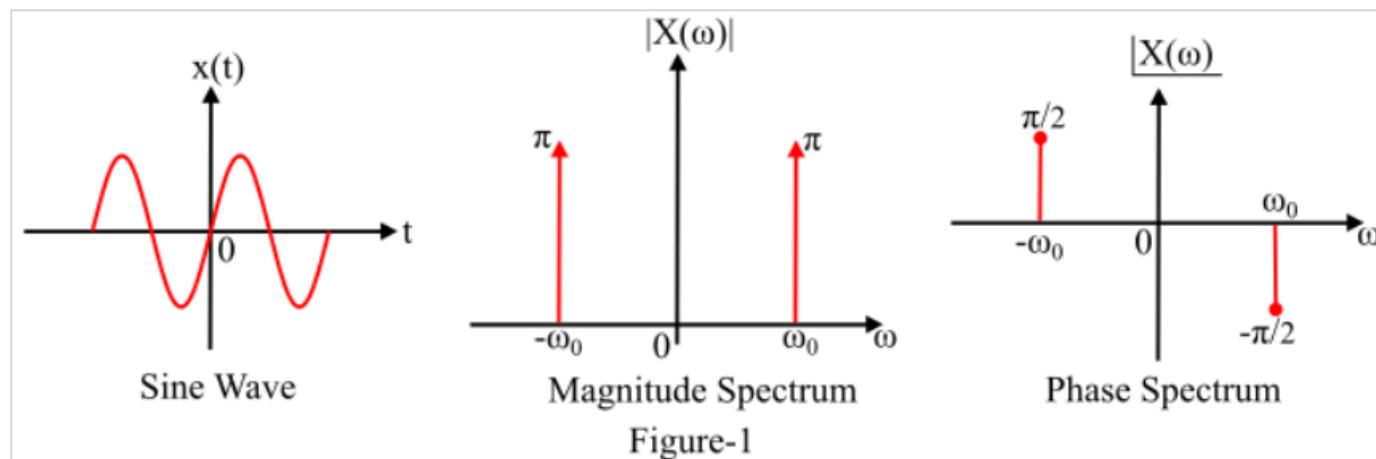
- ▶ $|X(\omega)|$: determines the relative presence of a sinusoid $e^{j\omega n}$ in $x(n)$
- ▶ $\Theta(\omega) = \angle X(\omega)$: determines how the sinusoids line up relative to one another to form $x(n)$

Polar Form: Cosine & Sine Wave

■ Cosine Waves



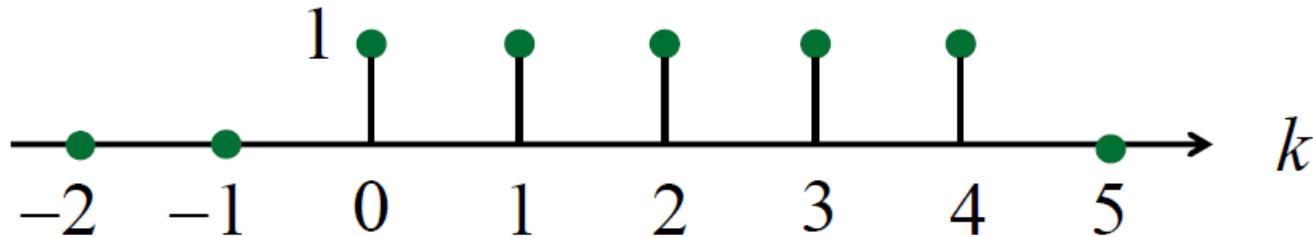
■ Sine Waves



Example (1)

■ Rectangular Form

$x[k]$



$$\begin{aligned} X(e^{j\omega}) &= \sum_{k=0}^4 x[k] e^{-jk\omega} \\ &= 1 + e^{-j\omega} + e^{-2j\omega} + e^{-3j\omega} + e^{-4j\omega} \end{aligned}$$

Example (1)

■ Polar Form

$$|X(e^{j\omega})|, \angle X(e^{j\omega})$$

$$X(e^{j\omega}) = 1 + e^{-j\omega} + e^{-2j\omega} + e^{-3j\omega} + e^{-4j\omega}$$

$$= \sum_{k=0}^4 e^{-j\omega k}$$

$$\sum_{k=0}^{N-1} \alpha^k = \frac{1 - \alpha^N}{1 - \alpha}, \quad \forall \alpha$$

$$= \frac{1 - e^{-j\omega(5)}}{1 - e^{-j\omega}}$$

$$\alpha = e^{-j\omega}, \quad N = 5$$

$$= \frac{e^{-j5\omega/2} [e^{j5\omega/2} - e^{-j5\omega/2}]}{e^{-j\omega/2} [e^{j\omega/2} - e^{-j\omega/2}]}$$

$$= e^{-j2\omega} \frac{\sin 5\omega/2}{\sin \omega/2}$$

$$|X(e^{j\omega})|$$

$$\angle X(e^{j\omega}) = -2\omega$$

Example (2)

$$x[k] = a^{|k|}, \quad |a| < 1$$

$$\begin{aligned} x[k] &= a^k, \quad k \geq 0 \\ &= a^{-k}, \quad k < 0 \end{aligned}$$

$$\sum_{k=0}^{\infty} \alpha^k \rightarrow \frac{1}{1-\alpha}, \quad |\alpha| < 1$$

$$\begin{aligned} X_1(e^{j\omega}) &= \sum_{k=0}^{\infty} x[k] e^{-j\omega k} = \sum_{k=0}^{\infty} a^k e^{-j\omega k} = \sum_{k=0}^{\infty} [ae^{-j\omega}]^k \\ &= \frac{1}{1 - ae^{-j\omega}} \end{aligned}$$

$$\begin{aligned} X_2(e^{j\omega}) &= \sum_{-\infty}^{-1} a^{-k} e^{-j\omega k} = \sum_{1}^{\infty} a^k e^{j\omega k} = \sum_{1}^{\infty} [ae^{j\omega}]^k \\ &= \sum_{0}^{\infty} [ae^{j\omega}]^k - 1 = \frac{1}{1 - ae^{j\omega}} - 1 = \frac{ae^{j\omega}}{1 - ae^{j\omega}} \end{aligned}$$

Example (2)

$$\begin{aligned} X(e^{j\omega}) &= X_1(e^{j\omega}) + X_2(e^{j\omega}) \\ &= \frac{1}{1 - ae^{-j\omega}} + \frac{ae^{j\omega}}{1 - ae^{j\omega}} \\ &= \frac{1 - a^2}{1 - 2a \cos \omega + a^2} \end{aligned}$$

$$\begin{aligned} \cos \omega &= \frac{e^{j\omega} + e^{-j\omega}}{2} \\ \sin \omega &= \frac{e^{j\omega} - e^{-j\omega}}{2j} \end{aligned}$$

$$|X(e^{j\omega})|$$

$$\angle X(e^{j\omega}) = 0$$

Example (3): CTFT

$$x(t) = e^{-at}u(t), \quad a > 0$$

$$X(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$

$$= \int_{-\infty}^{\infty} e^{-at}u(t)e^{-j\Omega t} dt$$

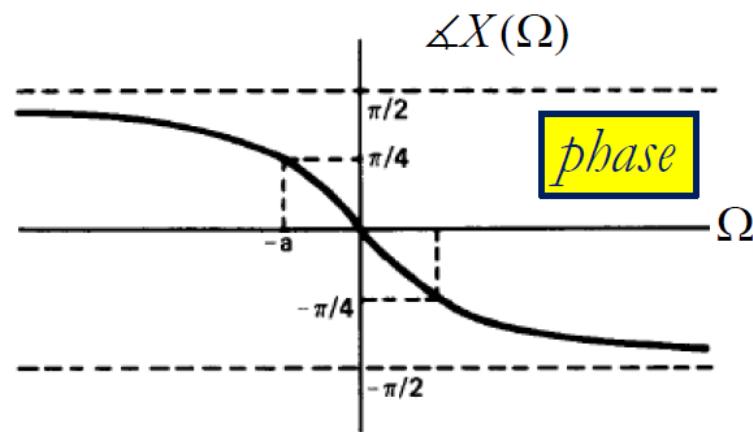
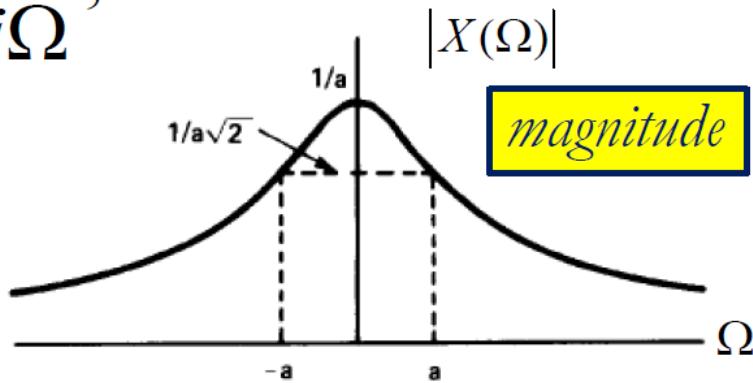
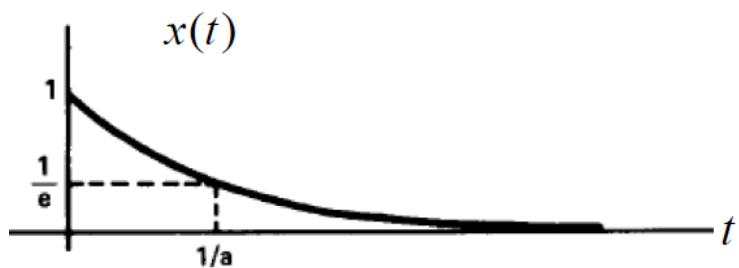
$$= \int_0^{\infty} e^{-at}e^{-j\Omega t} dt$$

$$= \int_0^{\infty} e^{-(a+j\Omega)t} dt$$

$$= \frac{-1}{a+j\Omega} e^{-(a+j\Omega)t} \Big|_0^{\infty} = \frac{-1}{a+j\Omega} (0 - 1) = \frac{1}{a+j\Omega}; \quad a > 0$$

Example (3): CTFT

$$e^{-at} u(t) \leftrightarrow \frac{1}{a + j\Omega}; \quad a > 0$$

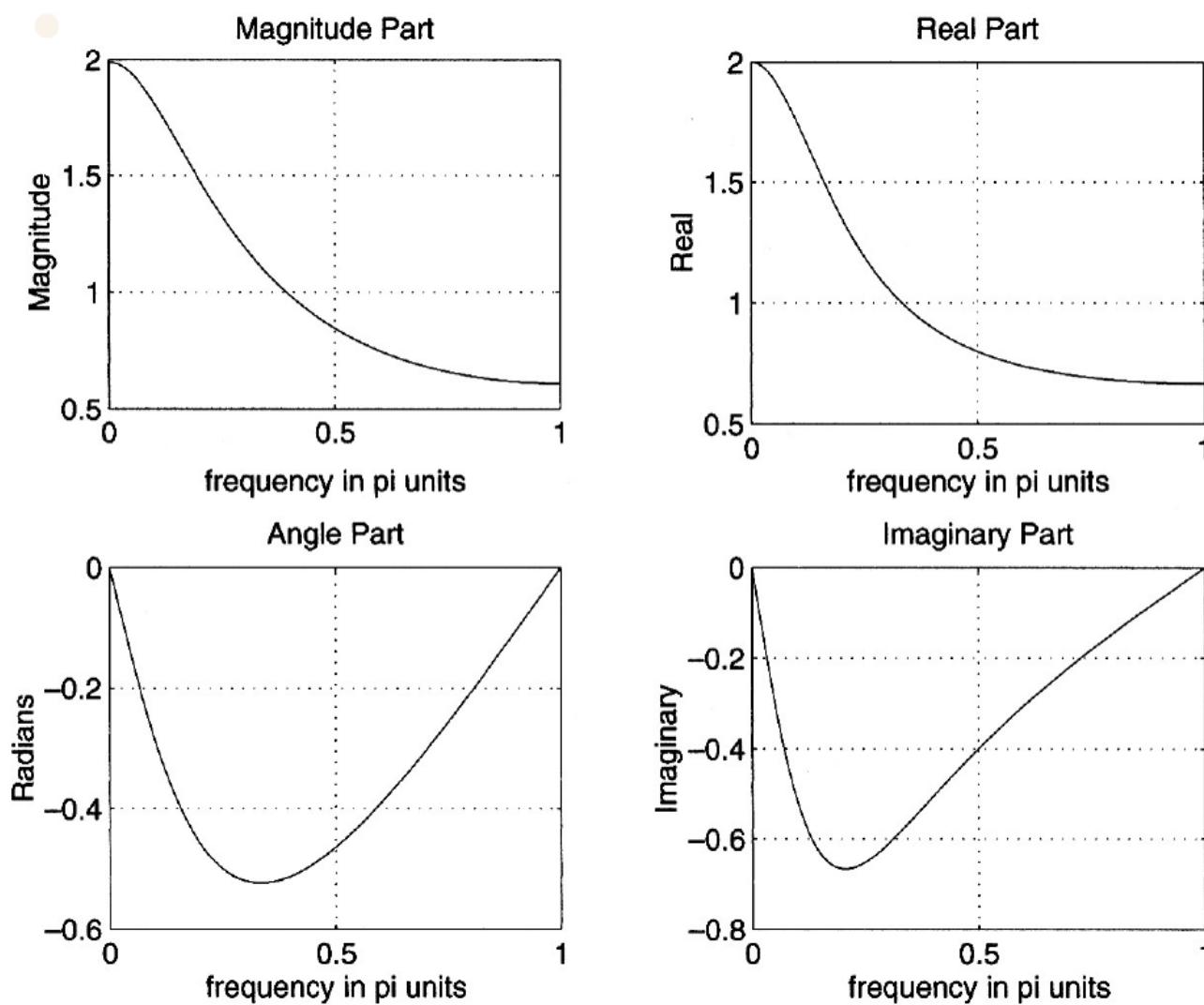


Example using MATLAB (1)

Example: Evaluate $X(e^{j\omega}) = \frac{e^{j\omega}}{e^{j\omega} - 0.5}$ at 501 equispaced points between $[0, \pi]$.

```
>> w = [0:1:500]*pi/500; % [0, pi] axis divided into 501 points.  
>> X = exp(j*w) ./ (exp(j*w) - 0.5*ones(1,501));  
>> magX = abs(X); angX = angle(X); realX = real(X); imagX = imag(X);  
>> subplot(2,2,1); plot(w/pi,magX); grid  
>> xlabel('frequency in pi units'); title('Magnitude part'); ylabel('Magnitude')  
>> subplot(2,2,2); plot(w/pi,realX); grid  
>> xlabel('frequency in pi units'); title('Real part'); ylabel('Real')  
>> subplot(2,2,3); plot(w/pi, angX); grid  
>> xlabel('frequency in pi units'); title('Angle part'); ylabel('Radians')  
>> subplot(2,2,4); plot(w/pi,imagX); grid  
>> xlabel('frequency in pi units'); title('Imaginary part'); ylabel('Imaginary')
```

Example using MATLAB (1)

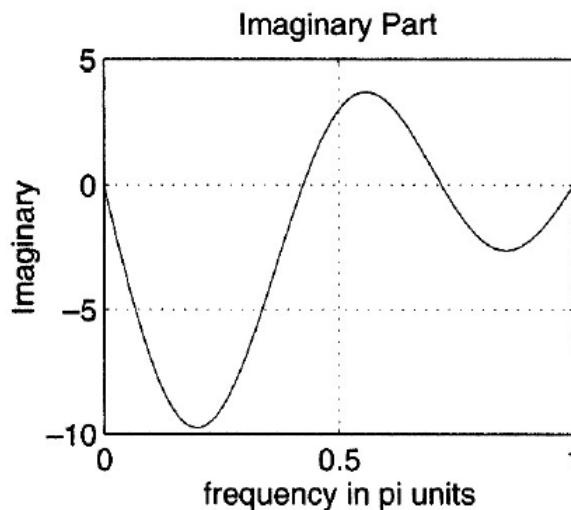
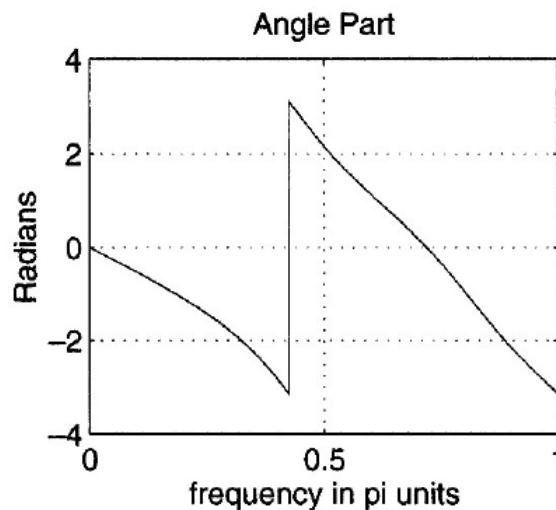
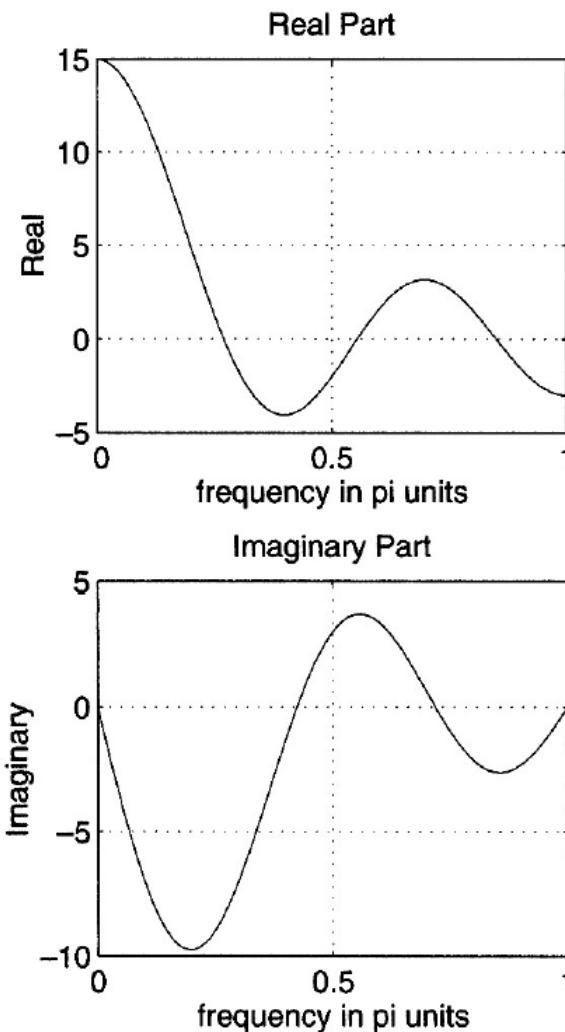
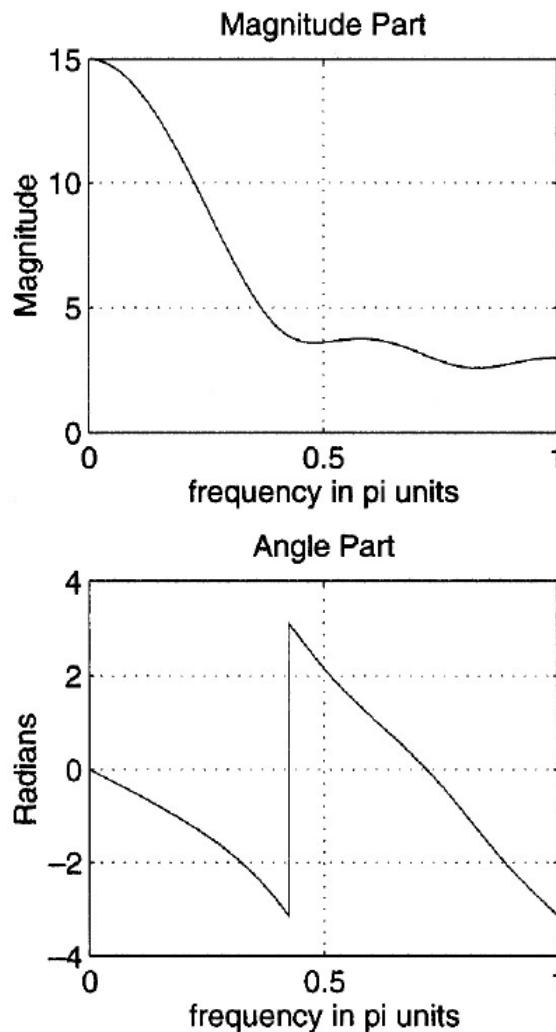


Example using MATLAB (2)

Example: Compute the DTFT of $x(k) = \{1, \underset{\uparrow}{2}, 3, 4, 5\}$ at 501 equispaced frequencies between $[0, \pi]$.

```
>> k = -1:3; x = 1:5; i = 0:500; w = (pi/500)*i;
>> X = x * exp(-j*pi/500)) .^ (k'*i);
>> magX = abs(X); angX = angle(X); realX = real(X); imagX = imag(X);
>> subplot(2,2,1); plot(n/500,magX); grid
>> xlabel('frequency in pi units'); title('Magnitude part'); ylabel('Magnitude')
>> subplot(2,2,2); plot(n/500,realX); grid
>> xlabel('frequency in pi units'); title('Real part'); ylabel('Real')
>> subplot(2,2,3); plot(n/500, angX); grid
>> xlabel('frequency in pi units'); title('Angle part'); ylabel('Radians')
>> subplot(2,2,4); plot(n/500,imagX); grid
>> xlabel('frequency in pi units'); title('Imaginary part'); ylabel('Imaginary')
```

Example using MATLAB (2)



Periodicity & Symmetry

ω is a real variable between $-\infty$ and ∞

Periodicity

The DTFT $X(e^{j\omega})$ is periodic in ω with period 2π .

$$X(e^{j\omega}) = X(e^{j[\omega+2\pi]}) \quad \omega \in [0, 2\pi], [-\pi, \pi], \text{etc.}$$

Symmetry

For real-valued $x(k)$, $X(e^{j\omega})$ is conjugate symmetric.

$$X(e^{-j\omega}) = X^*(e^{j\omega}) \quad \omega \in [0, \pi]$$

$$\operatorname{Re}[X(e^{-j\omega})] = \operatorname{Re}[X(e^{j\omega})], \quad \operatorname{Im}[X(e^{-j\omega})] = -\operatorname{Im}[X(e^{j\omega})]$$

$$|X(e^{-j\omega})| = |X(e^{j\omega})|, \quad \angle X(e^{-j\omega}) = -\angle X(e^{j\omega})$$

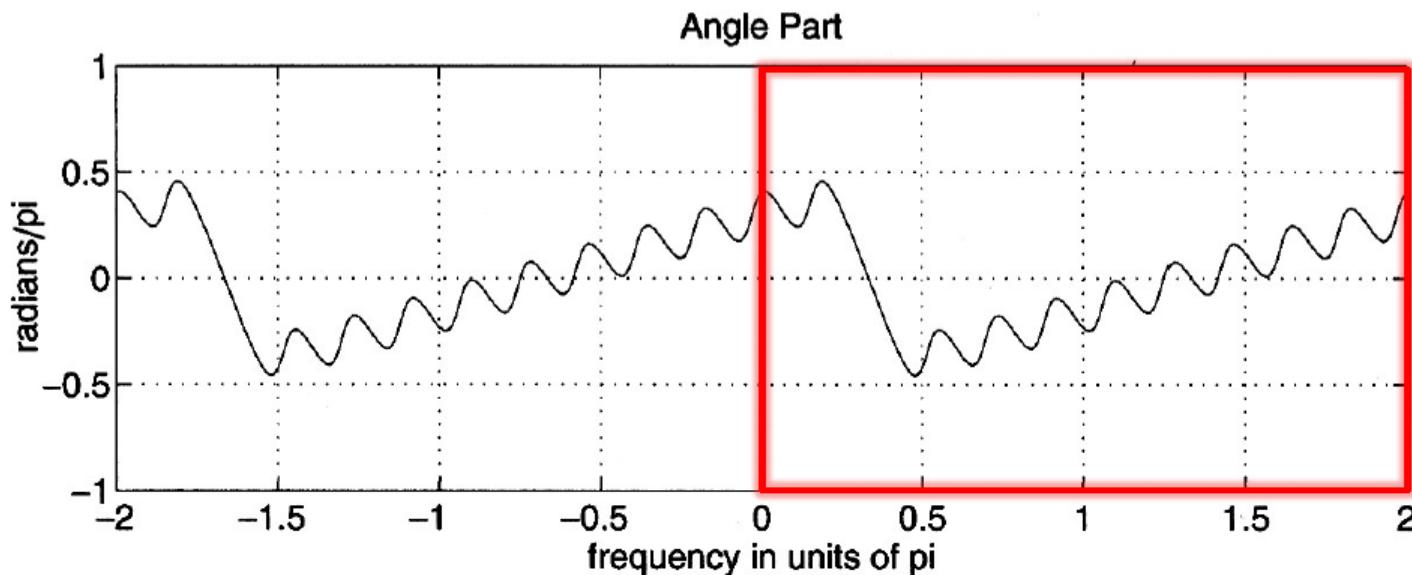
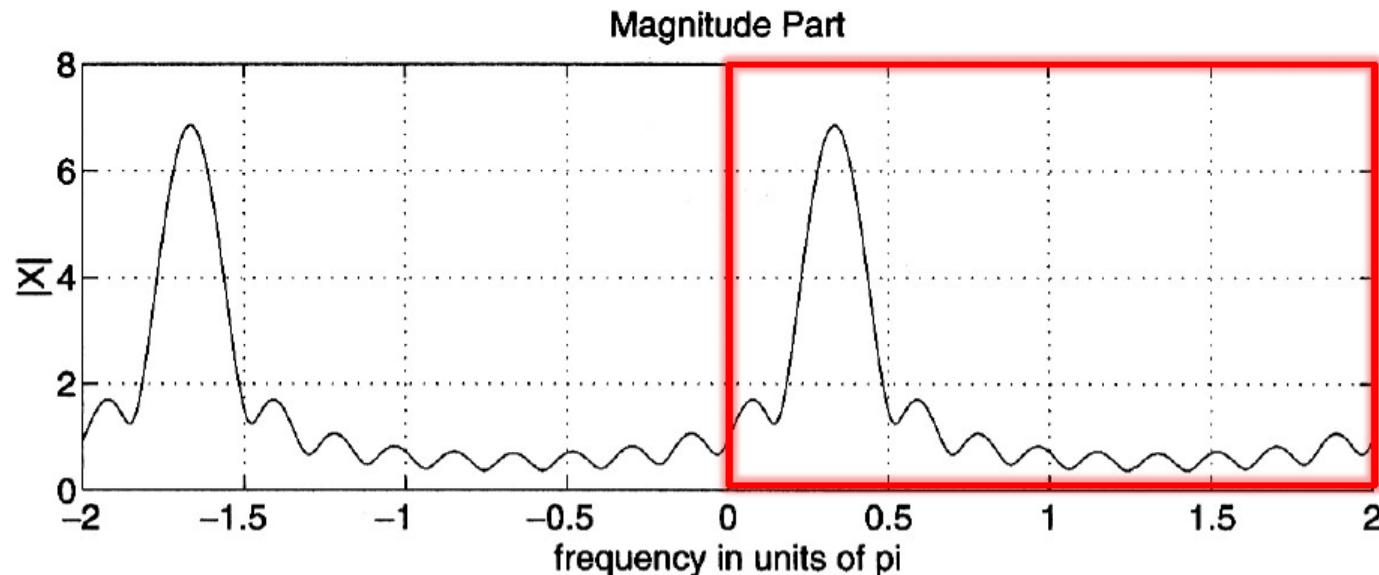
Example of Periodicity

Example: $x(k) = (0.9 \exp(j\pi/3))^k, \quad 0 \leq k \leq 10$

Determine $X(e^{j\omega})$ and investigate its periodicity.

```
>> k = 0:10; x = (0.9*exp(j*pi/3)) .^ k;
>> i = -200:200; w = (pi/100)*i; ← [-2π, 2π]
>> X = x * (exp(-i*pi/100)) .^ (k'*i);
>> magX = abs(X); angX = angle(X);
>> subplot(2,1,1); plot(w/pi,magX); grid
>> xlabel('frequency in pi units'); ylabel('|X|')
>> title('Magnitude part')
>> subplot(2,1,2); plot(w/pi, angX/pi); grid
>> xlabel('frequency in pi units'); ylabel('radians/pi')
>> title('Angle part')
```

Example of Periodicity



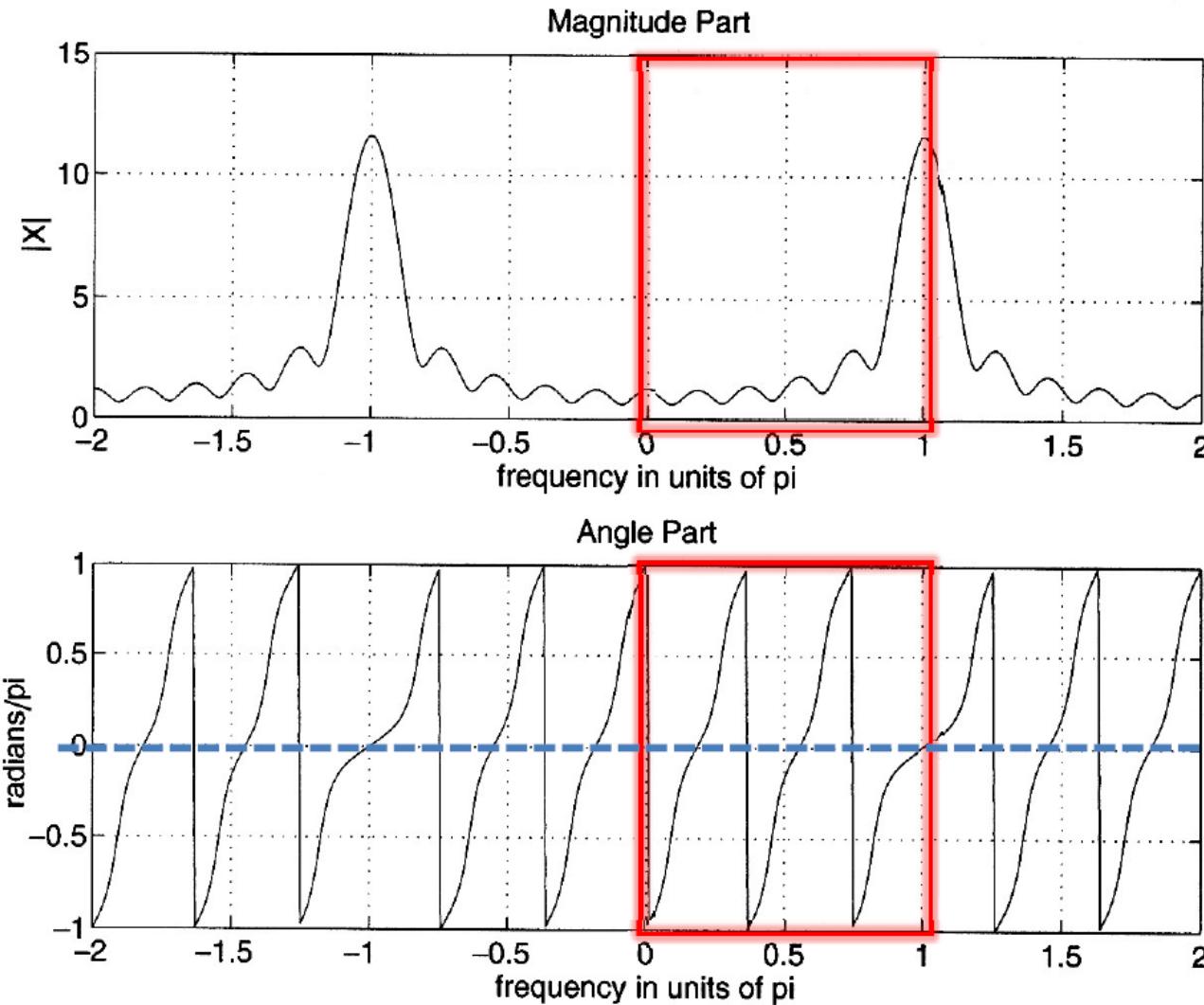
Example of Symmetry

Example: $x(k) = (-0.9)^k, -5 \leq k \leq 5$

Investigate the conjugate-symmetry property of its DTFT.

```
>> k = -5:5; x = (-0.9).^k;
>> i = -200:200; w = (pi/100)*i; X = x * (exp(-i*pi/100)).^(n.*i);
>> magX = abs(X); angX = angle(X);
>> subplot(2,1,1); plot(w/pi,magX); grid; axis([-2,2,0,15])
>> xlabel('frequency in pi units'); ylabel('|X|')
>> title('Magnitude part')
>> subplot(2,1,2); plot(w/pi, angX/pi); grid; axis([-2,2,-1,1])
>> xlabel('frequency in pi units'); ylabel('radians/pi')
>> title('Angle part')
```

Example of Symmetry



Not only periodic, but also conjugate-symmetric

Properties of DTFT

Properties of DTFT

1. Linearity

$$F[\alpha x_1(k) + \beta x_2(k)] = \alpha F[x_1(k)] + \beta F[x_2(k)]$$

2. Time shifting

$$F[x(k - m)] = X(e^{j\omega})e^{-j\omega m}$$

3. Frequency shifting

$$F[x(k)e^{j\omega_0 k}] = X(e^{j(\omega - \omega_0)})$$

4. Conjugation

$$F[x^*(k)] = X^*(e^{-j\omega})$$

Properties of DTFT

5. Folding

$$F[x(-k)] = X(e^{-j\omega})$$

6. Symmetries in real sequence

$$x(k) = x_e(k) + x_o(k)$$

$$F[x_e(k)] = \text{Re}[X(e^{j\omega})]$$

$$F[x_o(k)] = j \text{Im}[X(e^{j\omega})]$$

7. Convolution

$$F[x_1(k) * x_2(k)] = F[x_1(k)]F[x_2(k)] = X_1(e^{j\omega})X_2(e^{j\omega})$$

Properties of DTFT

8. Multiplication

$$\begin{aligned} F[x_1(k) \cdot x_2(k)] &= F[x_1(k)] \circledast F[x_2(k)] \\ &\equiv \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(e^{j\omega}) X_2(e^{j(\omega-\theta)}) d\theta \end{aligned}$$

9. Energy

$$\begin{aligned} E_x &= \sum_{-\infty}^{\infty} |x(k)|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega \\ &= \int_0^{\pi} \frac{|X(e^{j\omega})|^2}{\pi} d\omega \end{aligned}$$

$\Phi_x(\omega)$ Energy density spectrum of $x(k)$

$$\boxed{\begin{array}{c} [\omega_1, \omega_2] \\ \downarrow \\ \int_{\omega_1}^{\omega_2} \Phi_x(\omega) d\omega, \\ 0 \leq \omega_1 < \omega_2 \leq \pi \end{array}}$$

Properties of DTFT (1): Linearity

$$F[\alpha x_1(k) + \beta x_2(k)] = \alpha F[x_1(k)] + \beta F[x_2(k)]$$

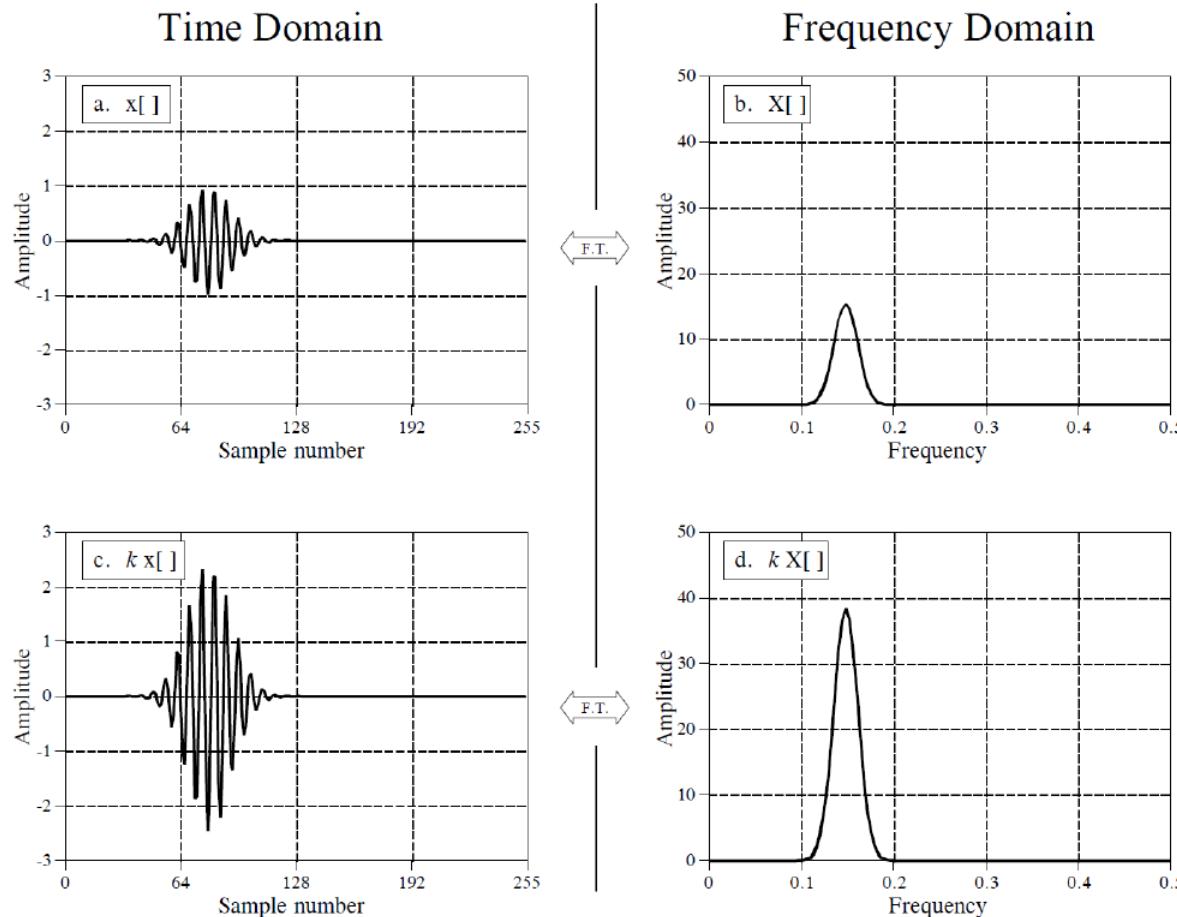
$$X(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x[k] e^{-j\omega k}$$

$$x[k] = \alpha_1 x_1[k] + \alpha_2 x_2[k]$$

$$\begin{aligned} X(e^{j\omega}) &= \sum_{k=-\infty}^{\infty} [\alpha_1 x_1[k] + \alpha_2 x_2[k]] e^{-j\omega k} \\ &= \sum_{k=-\infty}^{\infty} \alpha_1 x_1[k] e^{-j\omega k} + \sum_{k=-\infty}^{\infty} \alpha_2 x_2[k] e^{-j\omega k} \\ &= \alpha_1 \sum_{k=-\infty}^{\infty} x_1[k] e^{-j\omega k} + \alpha_2 \sum_{k=-\infty}^{\infty} x_2[k] e^{-j\omega k} \\ &= \alpha_1 X_1(e^{j\omega}) + \alpha_2 X_2(e^{j\omega}) \end{aligned}$$

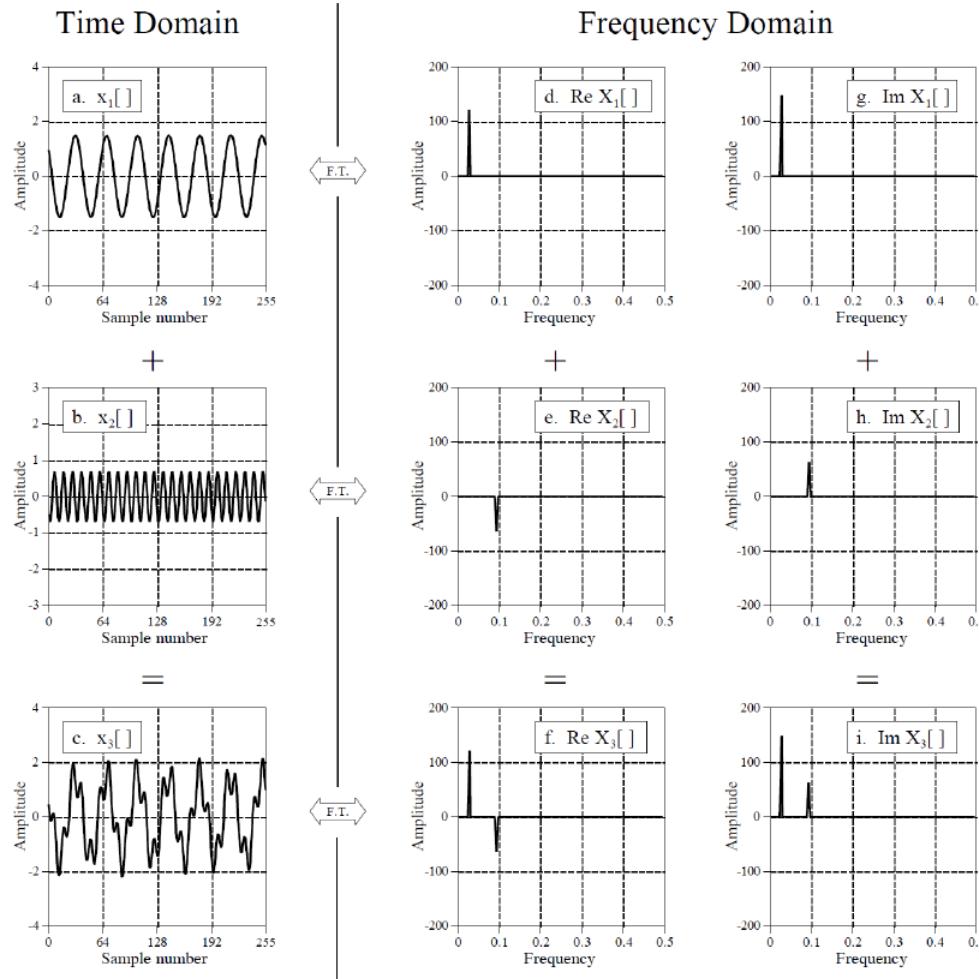
Linearity: Homogeneity & Additivity

Homogeneity of the Fourier transform



Linearity: Homogeneity & Additivity

Additivity of the Fourier transform



Example: MATLAB

Verify the *linearity* property (*using real-valued finite duration sequences*)

$x_1(k), x_2(k)$: random sequences uniformly distributed between $[0,1]$ over $0 \leq k \leq 10$.

```
>> x1 = rand(1,11); x2 = rand(1,11); k = 0:10;
>> alpha = 2; beta = 3; i = 0:500; w = (pi/500)*i;
>> X1 = x1 * (exp(-j*pi/500)).^(k'*i); % DTFT of x1
>> X2 = x2 * (exp(-j*pi/500)).^(k'*i); % DTFT of x2
>> x = alpha*x1 + beta*x2; % Linear combination of x1 & x2
>> X = x * (exp(-j*pi/500)).^(k'*i); % DTFT of x
>> % verification
>> X_check = alpha*X1 + beta*X2; % Linear combination of X1 & X2
>> error = max(abs(X-X_check)) % Difference
error =
7.1054e-015
```

Properties of DTFT (2): Time Shifting

$$F[x(k - m)] = X(e^{j\omega})e^{-j\omega m}$$

$$x[k] = x[k - m]$$

$$\sum_{k=-\infty}^{\infty} x[k - m]e^{-j\omega k}, \quad k - m = n$$

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

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

$$= X(e^{j\omega})e^{-j\omega m}$$

Example: MATLAB

Verify the *sample shift* property:

$x(k)$: a random sequence uniformly distributed between $[0,1]$ over $0 \leq k \leq 10$.

$$y(k) = x(k - 2)$$

```
>> x = rand(1,11); k = 0:10;
>> i = 0:500; w = (pi/500)*i;
>> X = x * (exp(-j*pi/500)).^(k'*i); % DTFT of x
>> % signal shifted by two samples
>> y = x; n = k+2;
>> Y = y * (exp(-j*pi/500)).^(n'*i); % DTFT of y
>> % verification
>> Y_check = (exp(-j*2).^w).*X;      % Multiplication by exp(-j2w)
>> error = max(abs(Y-Y_check))        % Difference
error =
5.7737e-015
```

Properties of DTFT (3): Frequency Shifting

$$F[x(k)e^{j\omega_0 k}] = X(e^{j(\omega - \omega_0)})$$

$$x[k] = x[k]e^{j\omega_0 k}$$

$$\sum_{k=-\infty}^{\infty} x[k]e^{j\omega_0 k} e^{-j\omega k}$$

$$= \sum_{k=-\infty}^{\infty} x[k]e^{-j(\omega - \omega_0)k}$$

$$= X(e^{j(\omega - \omega_0)})$$

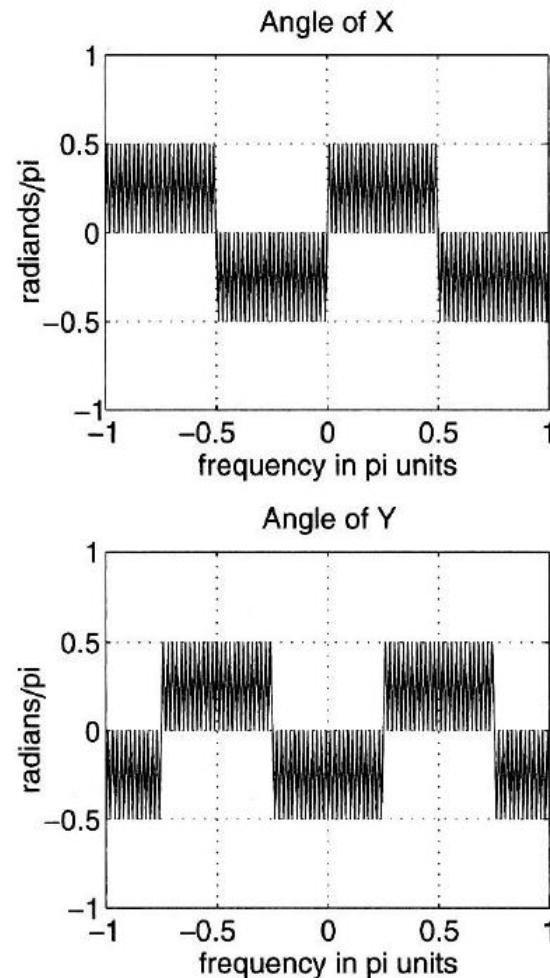
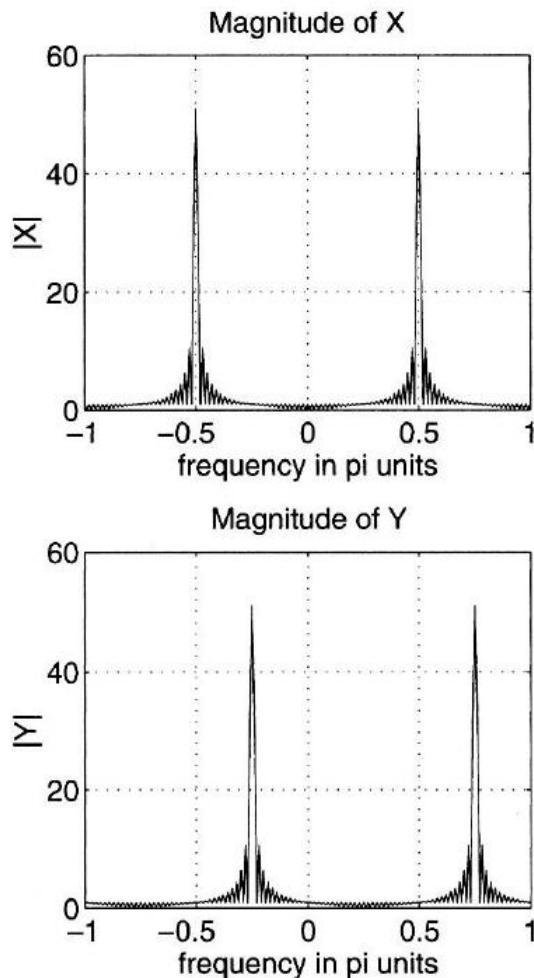
Example: MATLAB

Verify the *frequency shift* property:

$$x(k) = \cos(\pi k / 2), \quad 0 \leq k \leq 100, \quad y(k) = e^{j\pi k/4} x(k)$$

```
>> k = 0:100; x = cos(pi*k/2);
>> i = -100:100; w = (pi/100)*i;      % frequency between -pi and +pi
>> X = x * (exp(-j*pi/100)).^(k'*i); % DTFT of x
%
>> y = exp(j*pi*k/4).*x;              % signal multiplied by exp(j*pi*k/4)
>> Y = y * (exp(-j*pi/100)).^(k'*i); % DTFT of y
% Graphical verification
>> subplot(2,2,1); plot(w/pi,abs(X)); grid; axis([-1,1,0,60])
>> subplot(2,2,2); plot(w/pi,angle(X)/pi); grid; axis([-1,1,-1,1])
>> subplot(2,2,3); plot(w/pi,abs(Y)); grid; axis([-1,1,0,60])
>> subplot(2,2,4); plot(w/pi,angle(Y)/pi); grid; axis([-1,1,-1,1])
```

Example: MATLAB



$X(e^{j\omega})$ is indeed shifted by $\pi / 4$ in both magnitude and angle.

Properties of DTFT (4): Conjugation

$$F[x^*(k)] = X^*(e^{-j\omega})$$

$$x[k] \Leftrightarrow X(e^{j\omega}), \quad x^*[k] \Leftrightarrow X^*(e^{-j\omega})$$

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

$$x^*[k] = x[k] \quad \text{real}$$

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

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

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

$$\begin{aligned} &= \sum_{k=-\infty}^{\infty} x[k] e^{-j(-\omega)k} \\ &= X(e^{-j\omega}) \end{aligned}$$

Example: MATLAB

Verify the *conjugation* property:

$x(k)$: a complex-valued random sequence over $-5 \leq k \leq 10$ with real and imaginary parts uniformly distributed between $[0,1]$.

```
>> k = -5:10; x = rand(1,length(k)) + j*rand(1,length(k));
>> i = -100:100; w = (pi/100)*i;      % frequency between -pi and +pi
>> X = x * (exp(-j*pi/100)).^(k'*i); % DTFT of x
% conjugation property
>> y = conj(x));                      % signal conjugation
>> Y = y * (exp(-j*pi/100)).^(k'*i); % DTFT of y
% verification
>> Y_check = conj(fliplr(X));          % conj(X(-w))
>> error = max(abs(Y-Y_check))        % Difference
error =
0
```

Properties of DTFT (5): Folding

$$F[x(-k)] = X(e^{-j\omega})$$

$$x[k] = x[-k]$$

$$\sum_{k=-\infty}^{\infty} x[-k]e^{-j\omega k}, \quad -k = m$$

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

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

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

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

Example: MATLAB

Verify the *folding* property:

$x(k)$: a random sequence over $-5 \leq k \leq 10$ uniformly distributed between $[0,1]$.

```
>> k = -5:10; x = rand(1,length(k));
>> i = -100:100; w = (pi/100)*i;      % frequency between -pi and +pi
>> X = x * (exp(-j*pi/100)).^(k'*i); % DTFT of x
% folding property
>> y = fliplr(x); n = -fliplr(k);      % signal folding
>> Y = y * (exp(-j*pi/100)).^(n'*i); % DTFT of y
% verification
>> Y_check = fliplr(X);                % X(-w)
>> error = max(abs(Y-Y_check))        % Difference
error =
    0
```

Properties of DTFT (6): Symmetry in Real Sequence

$$x(k) = x_e(k) + x_o(k)$$

$$F[x_e(k)] = \text{Re}[X(e^{j\omega})]$$

$$F[x_o(k)] = j \text{Im}[X(e^{j\omega})]$$

Example: MATLAB

Verify the *symmetry* property of real signals.

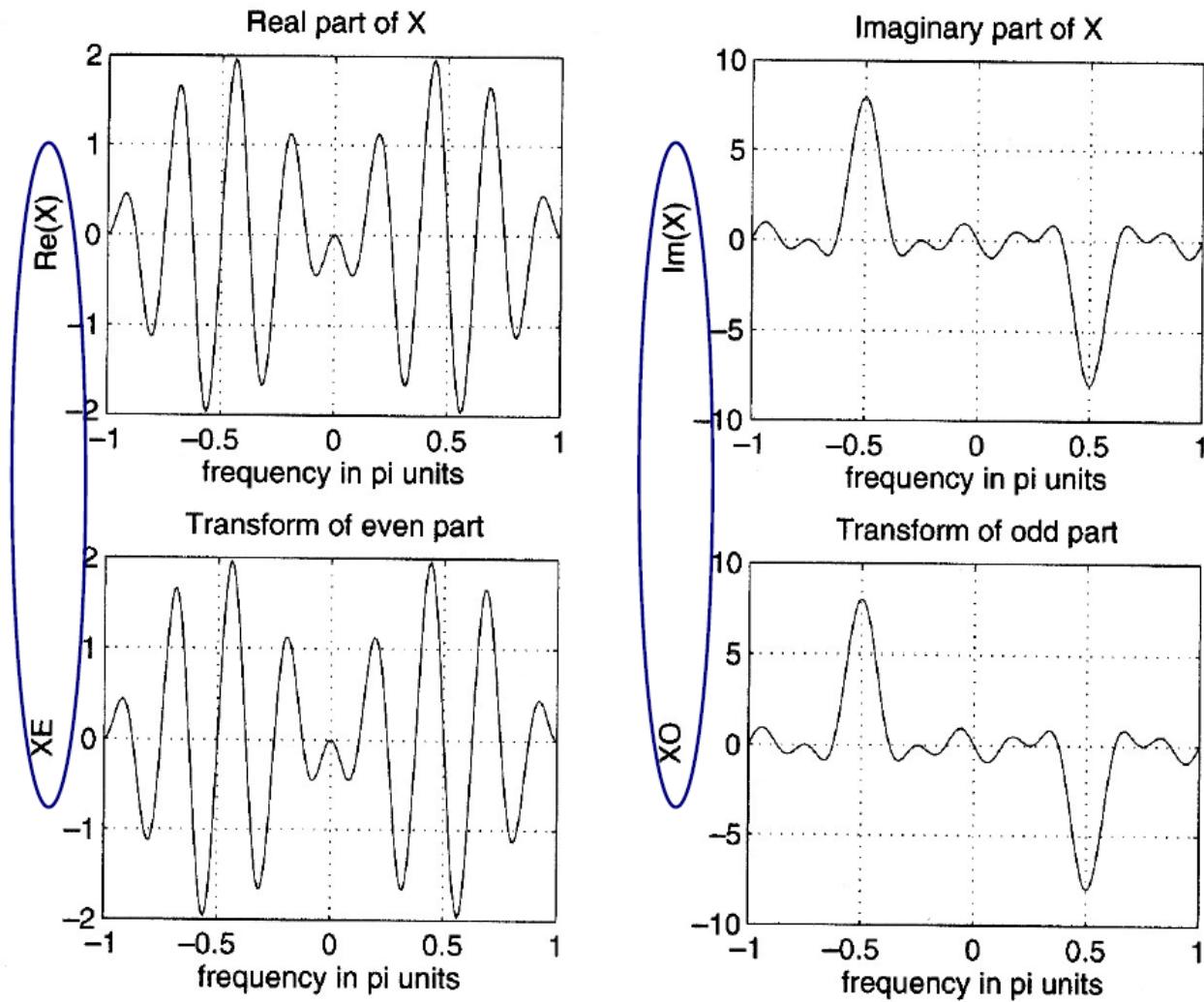
$$x(k) = \sin(\pi k / 2), \quad -5 \leq k \leq 10$$

```
>> k = -5:10; x = sin(pi*k/2);
>> i = -100:100; w = (pi/100)*i;      % frequency between -pi and +pi
>> X = x * (exp(-j*pi/100)).^(k'*i); % DTFT of x
% signal decomposition
>> [xe,xo,m] = evenodd(x,k);          % even and odd parts
>> XE = xe * (exp(-j*pi/100)).^(m'*i); % DTFT of xe
>> XO = xo * (exp(-j*pi/100)).^(m'*i); % DTFT of xo
% verification
>> XR = real(X);                      % real part of X
>> error1 = max(abs(XE-XR))           % Difference
error1 =
    1.8974e-019
>> XI = imag(X);                     % imag part of X
>> error2 = max(abs(XO-j*XI))         % Difference
error2 =
    1.8033e-019
```

Example: MATLAB

```
% graphical verification
>> subplot(2,2,1); plot(w/pi,XR); grid; axis([-1,1,-2,2])
>> xlabel('frequency in pi units'); ylabel('Re(X)');
>> title('Real part of X')
>> subplot(2,2,2); plot(w/pi,XI); grid; axis([-1,1,-10,10])
>> xlabel('frequency in pi units'); ylabel('Im(X)');
>> title('Imaginary par of X')
>> subplot(2,2,3); plot(w/pi,real(XE)); grid; axis([-1,1,-2,2])
>> xlabel('frequency in pi units'); ylabel('XE');
>> title('Transform of even part')
>> subplot(2,2,4); plot(w/pi,imag(XO)); grid; axis([-1,1,-10,10])
>> xlabel('frequency in pi units'); ylabel('XO');
>> title('Transform of odd part')
```

Example: MATLAB



Properties of DTFT (7): Convolution

$$F[x_1(k) * x_2(k)] = F[x_1(k)]F[x_2(k)] = X_1(e^{j\omega})X_2(e^{j\omega})$$

$$x[k] = x_1[k] * x_2[k]$$

$$\sum_{k=-\infty}^{\infty} (x_1[k] * x_2[k]) e^{-j\omega k}$$

$$= \sum_{k=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} x_1[n] x_2[k-n] e^{-j\omega k}$$

$$= \sum_{n=-\infty}^{\infty} x_1[n] \sum_{k=-\infty}^{\infty} x_2[k-n] e^{-j\omega k}, \quad k-n = m$$

$$= \sum_{n=-\infty}^{\infty} x_1[n] \sum_{m=-\infty}^{\infty} x_2[m] e^{-j\omega(m+n)}$$

$$= \sum_{n=-\infty}^{\infty} x_1[n] e^{-j\omega n} \sum_{m=-\infty}^{\infty} x_2[m] e^{-j\omega m} = X_1(e^{j\omega})X_2(e^{j\omega})$$

Properties of DTFT (8): Multiplication

$$F[x_1(k) \cdot x_2(k)] = F[x_1(k)] \circledast F[x_2(k)]$$

$$x_1[k] \rightleftharpoons X_1(e^{j\omega}), \quad x_2[k] \rightleftharpoons X_2(e^{j\omega})$$

$$\begin{aligned} & \sum_{k=-\infty}^{\infty} (x_1[k] \cdot x_2[k]) e^{-j\omega k} \\ &= \sum_{k=-\infty}^{\infty} \left(\frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(e^{j\theta}) e^{j\theta k} d\theta \right) x_2[k] e^{-j\omega k} \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(e^{j\theta}) d\theta \left(\sum_{k=-\infty}^{\infty} x_2[k] e^{-j(\omega-\theta)k} \right) \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \end{aligned}$$

Properties of DTFT (9): Energy

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

Parseval's relation for discrete-time aperiodic signals with finite energy

$$\varepsilon_x = \sum_{k=-\infty}^{\infty} x(k)x^*(k) = \sum_{k=-\infty}^{\infty} x(k) \left[\frac{1}{2\pi} \int_{-\pi}^{\pi} X^*(e^{j\omega}) e^{-j\omega k} d\omega \right]$$

energy in temporal space

$$\begin{aligned} \varepsilon_x &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X^*(e^{j\omega}) \left[\sum_{k=-\infty}^{\infty} x(k) e^{-j\omega k} \right] d\omega \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(e^{j\omega})|^2 d\omega \\ &= \int_0^\pi \frac{|X(e^{j\omega})|^2}{\pi} d\omega \quad (\text{for real sequences using even symmetry}) \end{aligned}$$

energy in spectral space

Properties of DTFT (9): Energy

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

Parseval's relation for periodic signals with finite power

$$P_x = \frac{1}{T} \int_0^T |x(t)|^2 dt, \quad |x(t)|^2 = x(t)x^*(t)$$

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{jk\omega_0 t}, \quad x^*(t) = \sum_{k=-\infty}^{\infty} C_k^* e^{-jk\omega_0 t}$$

$$P_x = \frac{1}{T} \int_0^T x(t) \left[\sum_{k=-\infty}^{\infty} C_k^* e^{-jk\omega_0 t} \right] dt$$

$$= \sum_{k=-\infty}^{\infty} C_k^* \left[\frac{1}{T} \int_0^T x(t) e^{-jk\omega_0 t} dt \right]$$

$$= \sum_{k=-\infty}^{\infty} C_k^* C_k = \sum_{k=-\infty}^{\infty} |C_k|^2$$

$$x = a + jb, \quad x^* = a - jb$$

$$\begin{aligned} xx^* &= a^2 - j^2 b^2 \\ &= a^2 + b^2 \end{aligned}$$

$$|x| = \sqrt{a^2 + b^2}$$

$$|x|^2 = a^2 + b^2$$

Frequency Domain Representation of LTI Systems

Review: Eigenvector and Eigenvalue

■ Definition

A *real or complex number* λ is called an *eigenvalue* of the $n \times n$ real matrix A if there exists a *nonzero* vector x such that $Ax = \lambda x$. Any *nonzero* vector x satisfying $Ax = \lambda x$ is called a (right) *eigenvector* of A associated with eigenvalue λ .

■ How to find eigenvector and eigenvalue?

$$Ax = y \longrightarrow Ax = \lambda x$$

$$Ax = \lambda Ix \longrightarrow (A - \lambda I)x = 0$$

$\Delta(\lambda) = \det(A - \lambda I)$ *The characteristic polynomial of A*
Every root of $\Delta(\lambda)$ is eigenvalue

*The null space of $A - \lambda I$ is called the *eigenspace* of A associated with eigenvalue λ .*

Eigenfunction of LTI System

- For a LTI system \mathcal{H} with impulse response $h(t)$,

$$e^{j\omega t} \xrightarrow{\mathcal{H}} H(j\omega)e^{j\omega t},$$

where ω is a real constant and

$$H(j\omega) = \int_{-\infty}^{\infty} h(t)e^{-j\omega t} dt.$$

- That is, $e^{j\omega t}$ is an *eigenfunction* of a LTI system and $H(j\omega)$ is the corresponding *eigenvalue*.
- We refer to $H(j\omega)$ as the **frequency response** of the system \mathcal{H} .

Frequency Domain Representation of LTI System

Response to a complex exponential $e^{j\omega_0 k}$

$$e^{j\omega_0 k} \longrightarrow \boxed{g(k)} \longrightarrow g(k) * e^{j\omega_0 k}$$

$$\begin{aligned} y(k) &= g(k) * e^{j\omega_0 k} = \sum_{-\infty}^{\infty} g(m) e^{j\omega_0 (k-m)} \\ &= \left[\sum_{-\infty}^{\infty} g(m) e^{-j\omega_0 m} \right] e^{j\omega_0 k} \\ &= \left[F[g(k)] \Big|_{\omega=\omega_0} \right] e^{j\omega_0 k} \end{aligned}$$

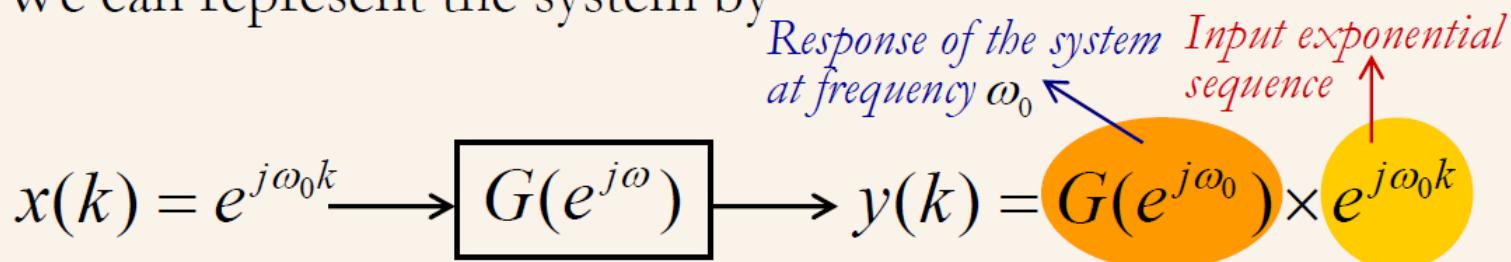
The Fourier transform representation is the most useful signal representation for LTI systems.

Frequency Response

The DTFT of an impulse response is called the *Frequency Response* (or *Transfer Function*) of an LTI system and is denoted by

$$G(e^{j\omega}) \equiv \sum_{-\infty}^{\infty} g(k) e^{-j\omega k}$$

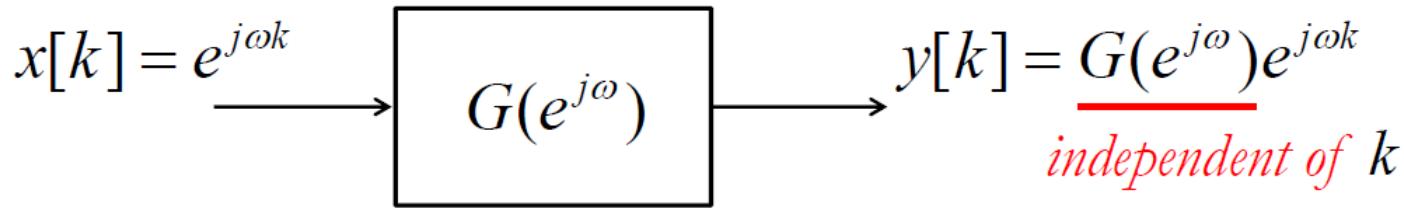
We can represent the system by



$$\sum_i A_i e^{j\omega_i k}$$

$$\sum_i A_i G(e^{j\omega_i}) e^{j\omega_i k}$$

Frequency Response



$$G(e^{j\omega}) = |G(e^{j\omega})| e^{j\angle G(e^{j\omega})} \quad : polar\ form$$

↓

$$\begin{aligned} y[k] &= |G(e^{j\omega})| e^{j\angle G(e^{j\omega})} e^{j\omega k} \\ &= |G(e^{j\omega})| e^{j(\omega k + \angle G(e^{j\omega}))} \end{aligned}$$

Amplitude of the input sinusoid changed by $|G(e^{j\omega})|$: magnitude response

Phase of the input sinusoid changed by $\angle G(e^{j\omega})$: phase response

Frequency Response

$$x[k] = \alpha_1 e^{j\omega_1 k} + \alpha_2 e^{j\omega_2 k} \Rightarrow y[k] = \alpha_1 G(e^{j\omega_1}) e^{j\omega_1 k} + \alpha_2 G(e^{j\omega_2}) e^{j\omega_2 k}$$

Relationship to impulse response

$$\begin{aligned} y[k] &= \sum_{n=-\infty}^{\infty} g[n] x[k-n] = \sum_{n=-\infty}^{\infty} g[n] e^{j\omega(k-n)} \\ &= e^{j\omega k} \underbrace{\sum_{n=-\infty}^{\infty} g[n] e^{-j\omega n}}_{G(e^{j\omega})} \end{aligned}$$

$$g[k] \xleftrightarrow{DTFT} G(e^{j\omega})$$

Example:

$$g[k] = \delta[k] + 2\delta[k-1] + 3\delta[k-2],$$

$$x[k] = -2 + 3 \cos\left(\frac{\pi}{4}k + \frac{\pi}{3}\right) + 10 \cos\left(\frac{3\pi}{4}k - \frac{\pi}{5}\right)$$

$\omega = 0$

Determine the frequency response $G(e^{j\omega})$ of this system.

$$G(e^{j\omega}) = \sum_{-\infty}^{\infty} g[k]e^{-j\omega k} = 1 + 2e^{-j\omega} + 3e^{-j2\omega}$$

Determine the steady-state output $y[k]$.

Example:

$$G(e^{j\omega}) = 1 + 2e^{-j\omega} + 3e^{-j2\omega}$$

$$G(e^{j\omega}) \Big|_{\omega=0} = 1 + 2 + 3 = 6$$

$$\begin{aligned} G(e^{j\omega}) \Big|_{\omega=\frac{\pi}{4}} &= 1 + 2e^{-j\frac{\pi}{4}} + 3e^{-j\frac{\pi}{2}} = 1 + 2(0.7071 - j0.7071) + 3(-j) \\ &= 2.414 - j4.414 = 5.031e^{j(-1.070)} \end{aligned}$$

$$\begin{aligned} G(e^{j\omega}) \Big|_{\omega=\frac{3\pi}{4}} &= 1 + 2e^{-j\frac{3\pi}{4}} + 3e^{-j\frac{3\pi}{2}} = 1 + 2(-0.7071 - j0.7071) + 3(j) \\ &= -0.4142 - j1.586 = 1.639e^{j1.826} \end{aligned}$$

$$y[k] = -2 \cdot 6 + 3 \cdot 5.031 \cos\left(\frac{\pi}{4}k + \frac{\pi}{3} - 1.070\right)$$

$$+ 10 \cdot 1.639 \cos\left(\frac{3\pi}{4}k - \frac{\pi}{5} + 1.826\right)$$

Lab #6.1 (1 pt.): Due Jan 17

Determine the frequency response $G(e^{j\omega})$ of a system characterized by $g(k) = (0.\textcolor{brown}{A})^k u(k)$. Plot the magnitude and phase response.

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Frequency Response of Difference Equations

$$y(k) + \sum_{m=1}^N a_m y(k-m) = \sum_{n=0}^M b_n x(k-n)$$

$$G(e^{j\omega})e^{j\omega k} + \sum_{m=1}^N a_m G(e^{j\omega})e^{j\omega(k-m)} = \sum_{n=0}^M b_n e^{j\omega(k-n)}$$

$$G(e^{j\omega}) = \frac{\sum_{n=0}^M b_n e^{-j\omega n}}{1 + \sum_{m=1}^N a_m e^{-j\omega m}}$$

Example

$$y[k] + \frac{1}{2}y[k-1] = x[k], \quad x[k] = \left(\frac{1}{2}\right)^k u[k]$$

Determine the frequency response $G(e^{j\omega})$ of this system.

$$Y(e^{j\omega}) + \frac{1}{2}e^{-j\omega}Y(e^{j\omega}) = X(e^{j\omega})$$

$$G(e^{j\omega}) = \frac{Y(e^{j\omega})}{X(e^{j\omega})} = \frac{1}{1 + \frac{1}{2}e^{-j\omega}}$$

Example

What is the response of the system?

$$X(e^{j\omega}) = \sum_{-\infty}^{\infty} x[k]e^{-j\omega k} = \sum_0^{\infty} \left(\frac{1}{2}\right)^k e^{-j\omega k} = \sum_0^{\infty} \left(\frac{1}{2}e^{-j\omega}\right)^k$$
$$= \frac{1}{1 - \frac{1}{2}e^{-j\omega}}$$

$$Y(e^{j\omega}) = G(e^{j\omega})X(e^{j\omega}) = \frac{1}{1 + \frac{1}{2}e^{-j\omega}} \cdot \frac{1}{1 - \frac{1}{2}e^{-j\omega}}$$

$$= \frac{1}{2} \frac{1}{1 + \frac{1}{2}e^{-j\omega}} + \frac{1}{2} \frac{1}{1 - \frac{1}{2}e^{-j\omega}}$$

$$y[k] = \frac{1}{2} \left(-\frac{1}{2}\right)^k u[k] + \frac{1}{2} \left(\frac{1}{2}\right)^k u[k]$$

Lab #6.2 (1 pt.): Due Jan 17

An LTI system is specified by the difference equation

$$y(k) = (0.5)y(k-1) + x(k)$$

- Determine $G(e^{j\omega})$.
- Calculate and plot the steady-state response $y_{ss}(k)$ to

$$x(k) = \cos(0.05\pi k)u(k).$$

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Thank you

