

Spring 2012

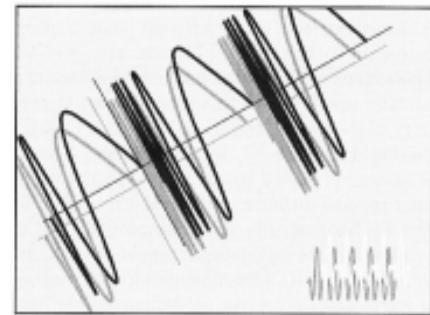
# 信號與系統 Signals and Systems

## Chapter SS-3 Fourier Series Representation of Periodic Signals

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NTU-EE

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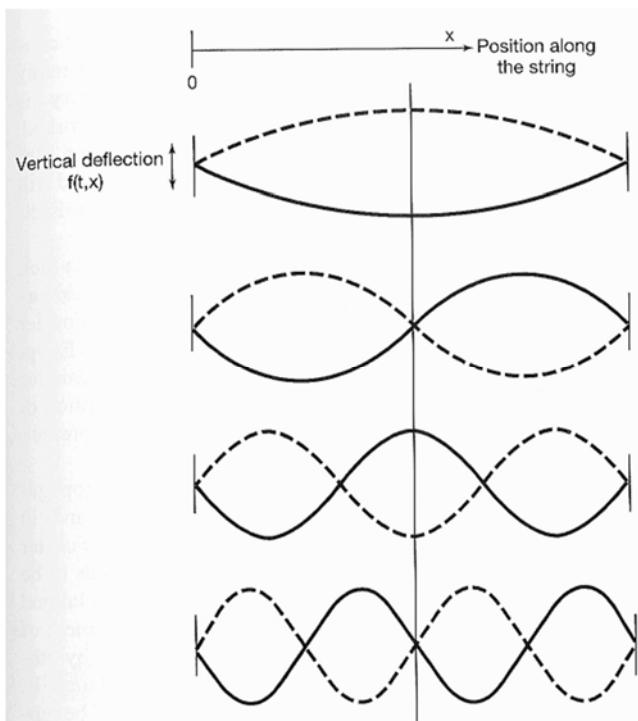
Figures and images used in these lecture notes are adopted from  
"Signals & Systems" by Alan V. Oppenheim and Alan S. Willsky, 1997

### Outline

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NTUEE-SS3-FS-2

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
- Fourier Series Representation of Discrete-Time Periodic Signals
- Properties of Discrete-Time Fourier Series
- Fourier Series & LTI Systems
- Filtering & Examples of CT & DT Filters

- L. Euler's study on the motion of a vibrating string in 1748



Leonhard Euler  
1707-1783  
Born in Switzerland  
Photo from wikipedia

- L. Euler showed (in 1748)

– The configuration of a **vibrating string** at some point in time is a **linear combination** of these **normal modes**

- D. Bernoulli argued (in 1753)

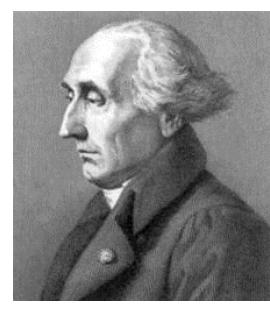
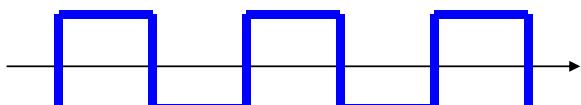
– All physical motions of a string could be represented by linear combinations of **normal modes**  
– But, he **did not** pursue this mathematically



Daniel Bernoulli  
1700-1782  
Born in Dutch  
Photo from wikipedia

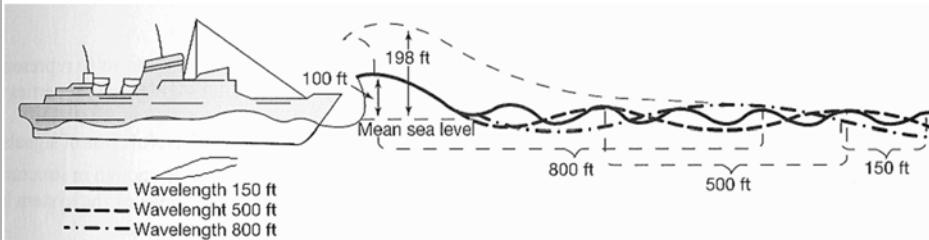
- J.L. Lagrange strongly criticized (in 1759)

– The use of **trigonometric series** in examination of **vibrating strings**  
– **Impossible** to represent signals with **corners** using **trigonometric series**



Joseph-Louis Lagrange  
1736-1813  
Born in Italy  
Photo from wikipedia

- In 1807, Jean Baptiste Joseph Fourier
  - Submitted a paper of using trigonometric series to represent “any” periodic signal
  - It is examined by S.F. Lacroix, G. Monge, P.S. de Laplace, and J.L. Lagrange,
  - But Lagrange rejected it!
- In 1822, Fourier published a book “Theorie analytique de la chaleur”
  - “The Analytical Theory of Heat”



Jean Baptiste Joseph Fourier  
1768-1830  
Born in France  
Photo from wikipedia



Figure 1.2: A medallion by David d'Angers, the only known portrait of Lacroix, made two years prior to his death. [Académie des Sciences de l'Institut de France]



Silvestre François de Lacroix

1765-1843

Born in France

Photo from

A short biography of Silvestre-François Lacroix  
In Science Networks. Historical Studies, V35,  
Lacroix and the Calculus, Birkhäuser Basel  
2008, ISBN 978-3-7643-8638-2

Gaspard Monge, Comte de Péluse

1746-1818

Born in France

Photo from wikipedia

Pierre-Simon, Marquis de Laplace

1749-1827

Born in France

Photo from wikipedia

- Fourier's main contributions:
  - Studied vibration, heat diffusion, etc.
  - Found **series** of harmonically related sinusoids to be useful in representing the temperature distribution through a body
  - Claimed that “any” periodic signal could be represented by such a series (i.e., **Fourier series** discussed in Chap 3)
  - Obtained a representation for aperiodic signals (i.e., **Fourier integral or transform** discussed in Chap 4 & 5)
  - (Fourier did not actually contribute to the mathematical theory of Fourier series)

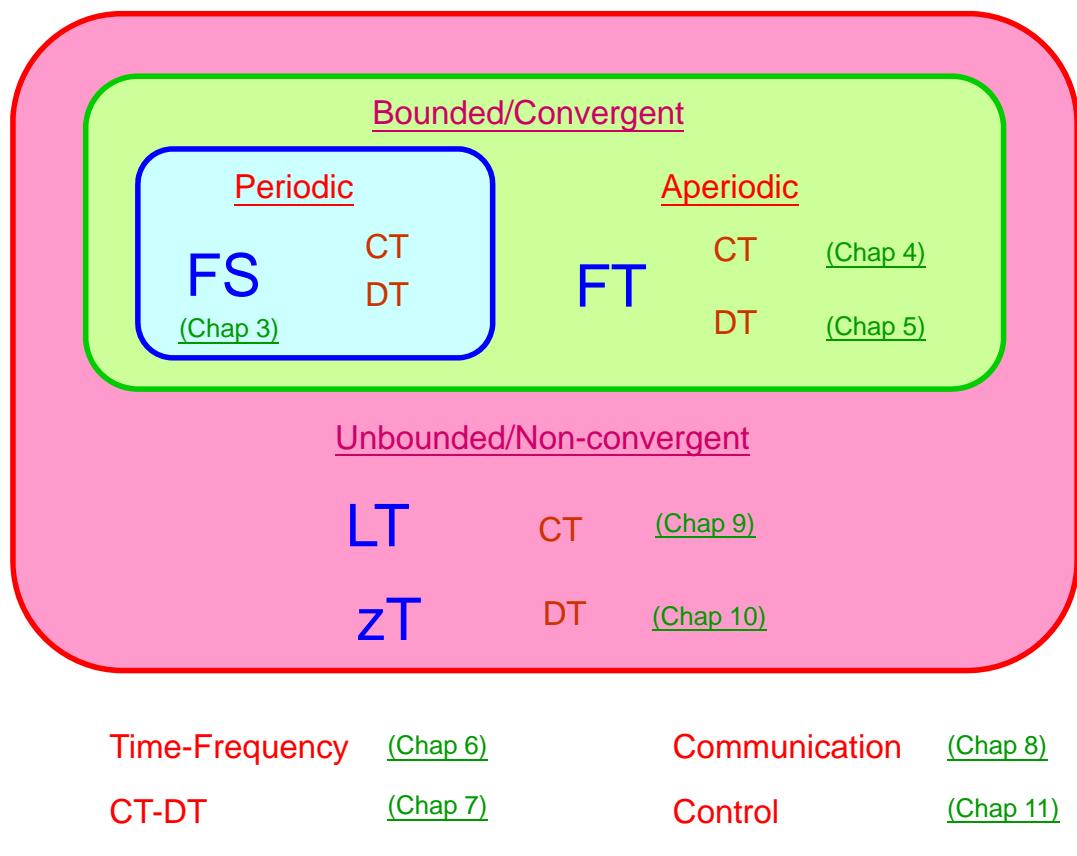


- Impact from Fourier's work:
  - Theory of integration, point-set topology, eigenfunction expansions, etc.
  - Motion of planets, periodic behavior of the earth's climate, wave in the ocean, radio & television stations
  - Harmonic time series in the 18th & 19th centuries
    - > Gauss etc. on discrete-time signals and systems
  - Faster Fourier transform (FFT) in the mid-1960s
    - > Cooley (IBM) & Tukey (Princeton) reinvented in 1965
    - > Can be found in Gauss's notebooks (in 1805)



James W. Cooley & John W. Tukey (1965):  
 "An algorithm for the machine calculation of complex Fourier series",  
 Math. Comput. 19, 297–301.

Carl Friedrich Gauss (Gauß)  
 1777-1855  
 Born in Germany  
 Photo from wikipedia

Signals & Systems [\(Chap 1\)](#)LTI & Convolution [\(Chap 2\)](#)

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
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$$\sum_{-\infty}^{\infty} \delta(t) \quad \delta[n]$$

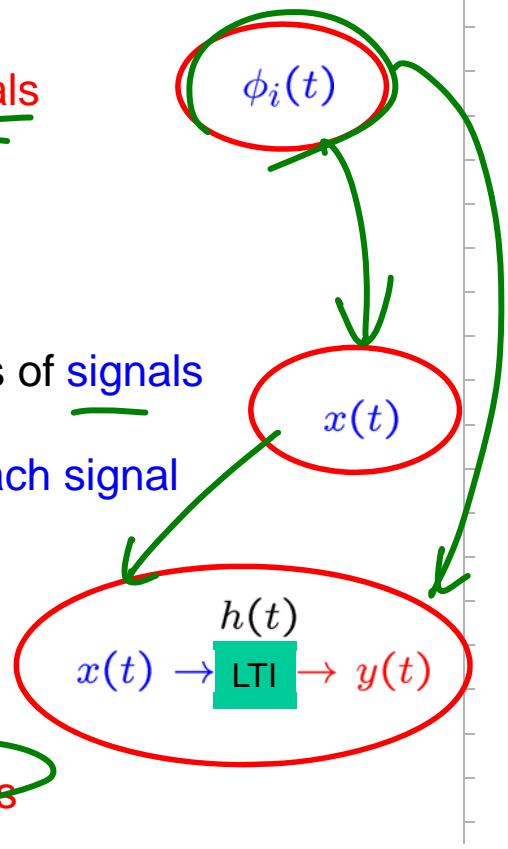
- Basic Idea:

- To represent signals as linear combinations of basic signals

- Key Properties:

1. The set of basic signals can be used to construct a broad and useful class of signals

2. The response of an LTI system to each signal should be simple enough in structure to provide us with a convenient representation for the response of the system to any signals constructed as linear combination of basic signals



- One of Choices:

- The set of complex exponential signals

$\left\{ \begin{array}{l} \text{signals of form } e^{st} \text{ in CT} \\ \text{signals of form } z^n \text{ in DT} \end{array} \right.$

$$y = A \mathcal{V} = \lambda \mathcal{V}$$

- The Response of an LTI System:

input  $\rightarrow$  LTI  $\rightarrow$  output

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) d\tau$$

$x(t)$        $h(t)$        $y(t)$

$$\left\{ \begin{array}{l} \text{CT: } e^{st} \xrightarrow{\quad} H(s) e^{st} \\ \text{DT: } z^n \xrightarrow{\quad} H(z) z^n \end{array} \right.$$

eigenfunction  
eigenvalue

Let  $x(t) = e^{st}$

$$y(t) = \int_{-\infty}^{+\infty} h(\tau)x(t-\tau)d\tau$$

$$= \int_{-\infty}^{+\infty} h(\tau)e^{s(t-\tau)}d\tau$$

$$= e^{st} \int_{-\infty}^{+\infty} h(\tau)e^{-s\tau}d\tau$$

$$\Rightarrow y(t) = H(s)x(t) = H(s)e^{st}$$

$$H(s) = \int_{-\infty}^{+\infty} h(\tau)e^{-s\tau}d\tau$$

Let  $x[n] = z^n$

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

$$= \sum_{k=-\infty}^{+\infty} h[k]z^{n-k}$$

$$= z^n \sum_{k=-\infty}^{+\infty} h[k]z^{-k}$$

$$\Rightarrow y[n] = H(z)x[n] = H(z)z^n$$

$$H(z) = \sum_{k=-\infty}^{+\infty} h[k]z^{-k}$$

- Eigenfunctions and Superposition Properties:

$$e^{s_k t} \xrightarrow{k=1,2,3} \text{LTI} \xrightarrow{} H(s_k) e^{s_k t}$$

$$a_1 e^{s_1 t} \rightarrow a_1 H(s_1) e^{s_1 t}$$

$$a_2 e^{s_2 t} \rightarrow a_2 H(s_2) e^{s_2 t}$$

$$a_3 e^{s_3 t} \rightarrow a_3 H(s_3) e^{s_3 t}$$

$$x(t) = a_1 e^{s_1 t} + a_2 e^{s_2 t} + a_3 e^{s_3 t}$$

$$y(t) = a_1 H(s_1) e^{s_1 t} + a_2 H(s_2) e^{s_2 t} + a_3 H(s_3) e^{s_3 t}$$

CT<sub>1</sub>  $\Rightarrow x(t) = \sum_k a_k e^{s_k t} \rightarrow y(t) = \sum_k a_k H(s_k) e^{s_k t}$

D.T<sub>1</sub>  $\Rightarrow x[n] = \sum_k a_k z_k^n \rightarrow y[n] = \sum_k a_k H(z_k) z_k^n$

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- **Fourier Series Representation of Continuous-Time Periodic Signals**
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
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## Fourier Series Representation of CT Periodic Signals

- Harmonically related complex exponentials

$$\phi_k(t) = e^{jk\omega_0 t} = e^{jk\left(\frac{2\pi}{T}\right)t}, \quad k = 0, \pm 1, \pm 2, \dots$$

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

- The Fourier Series Representation:

$$\begin{aligned} x(t) &= \dots + a_{-2} \phi_{-2}(t) + a_{-1} \phi_{-1}(t) - a_0 \phi_0(t) - a_1 \phi_1(t) + a_2 \phi_2(t) + \dots \\ &= \sum_{k=-\infty}^{+\infty} a_k \phi_k(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_0 t} = \sum_{k=-\infty}^{+\infty} a_k e^{jk\left(\frac{2\pi}{T}\right)t} \end{aligned}$$

$k = +1, -1$ : the **first harmonic components**  
or, the **fundamental components**

$k = +2, -2$ : the **second harmonic components**

... etc.

## ■ Example 3.2:

$$x(t) = \sum_{k=-3}^{+3} a_k e^{jk(2\pi)t}$$

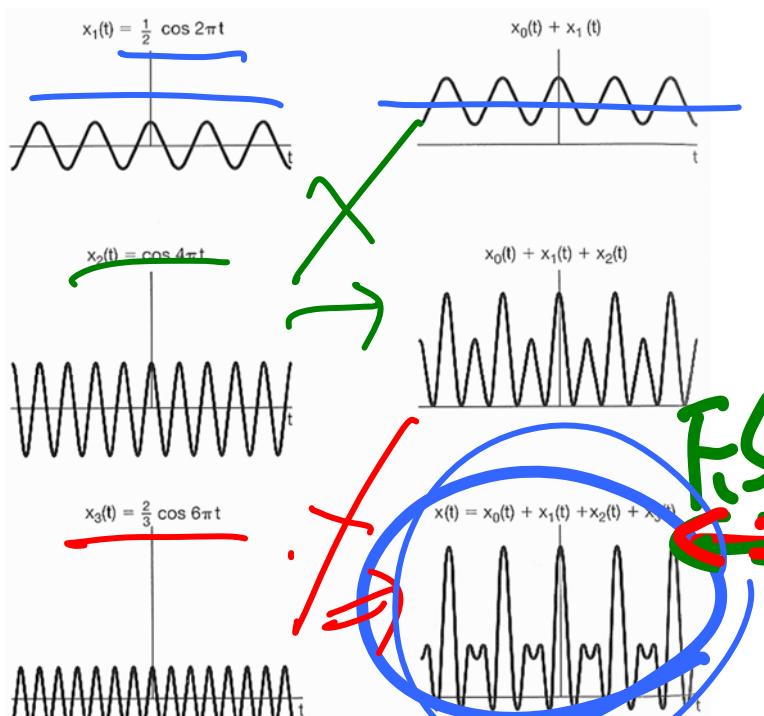
$\downarrow k=0 \quad k=1 \quad \cos 2\pi t \quad k=2 \quad (6) 4\pi t$

$$\Rightarrow x(t) = 1 + \frac{1}{4}(e^{j2\pi t} + e^{-j2\pi t}) + \frac{1}{2}(e^{j4\pi t} + e^{-j4\pi t}) + \frac{2}{3}(e^{j6\pi t} + e^{-j6\pi t})$$

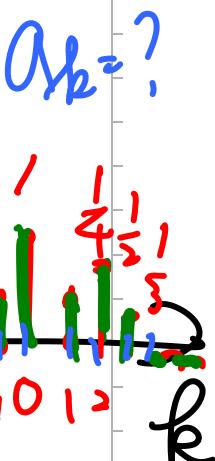
$$\Rightarrow x(t) = 1 + \frac{1}{2} \cos 2\pi t + \cos 4\pi t + \frac{2}{3} \cos 6\pi t$$

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

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



$\delta(t+1) \rightarrow$



■ Procedure of Determining the Coefficients:

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

$$\begin{aligned}
 x(t) &= \sum_{k=-\infty}^{+\infty} a_k e^{j k \omega_0 t} \\
 x(t) e^{-j n \omega_0 t} &= \sum_{k=-\infty}^{+\infty} a_k e^{j k \omega_0 t} e^{-j n \omega_0 t} \\
 \int_0^T x(t) e^{-j n \omega_0 t} dt &= \int_0^T \sum_{k=-\infty}^{+\infty} a_k e^{j k \omega_0 t} e^{-j n \omega_0 t} dt \\
 &= \sum_{k=-\infty}^{+\infty} a_k \int_0^T e^{j(k-n)\omega_0 t} dt \\
 \int_0^T e^{j(k-n)\omega_0 t} dt &= \int_0^T \cos((k-n)\omega_0 t) dt + j \int_0^T \sin((k-n)\omega_0 t) dt
 \end{aligned}$$

■ Procedure of Determining the Coefficients:

$$\begin{aligned}
 \int_0^T e^{j(k-n)\omega_0 t} dt &= \int_0^T \cos((k-n)\omega_0 t) dt + j \int_0^T \sin((k-n)\omega_0 t) dt \\
 &= \begin{cases} T, & k = n \\ 0, & k \neq n \end{cases}
 \end{aligned}$$

$$\begin{aligned}
 k \neq n &\Rightarrow 0 \\
 k = n &\Rightarrow T
 \end{aligned}$$

$$\Rightarrow \int_0^T x(t) e^{-j n \omega_0 t} dt = a_n T$$

$$a_n = \frac{1}{T} \int_0^T x(t) e^{-j n \omega_0 t} dt$$

$$a_k = \frac{1}{T} \int_0^T x(t) e^{-j k \omega_0 t} dt$$

• Furthermore,

$$\int_T^{2T} e^{j(k-n)\omega_0 t} dt = \begin{cases} T, & k = n \\ 0, & k \neq n \end{cases}$$

$$a_k = \frac{1}{T} \int_T^{2T} x(t) e^{-j k \omega_0 t} dt$$

## In Summary:

- The synthesis equation:

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} = \sum_{k=-\infty}^{+\infty} a_k e^{jk(2\pi/T)t}$$

- The analysis equation:

$$a_k = \frac{1}{T} \int_T x(t) e^{-jk(2\pi/T)t} dt = \frac{1}{T} \int_T x(t) e^{-jk(2\pi/T)t} dt$$

- CT Fourier series pair

- $\{a_k\}$ : the Fourier series coefficients

or the spectral coefficients of  $x(t)$

- $a_0 = \frac{1}{T} \int_T x(t) dt$ , the dc or constant component of  $x(t)$

3/2/12  
1:02 am

## Fourier Series of Real Periodic Signals:

- If  $x(t)$  is real, then  $x^*(t) = x(t)$

$$\begin{aligned} x(t) &= \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} \\ \Rightarrow x(t) &= x(t)^* = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} \\ &\quad = \sum_{k=-\infty}^{+\infty} a_k^* e^{-jkw_0 t} \quad \text{FS, } m = -k \\ &\quad = \sum_{m=-\infty}^{+\infty} a_m^* e^{jmw_0 t} \\ &\quad = \sum_{k=-\infty}^{+\infty} a_{-k}^* e^{jkw_0 t} \\ \Rightarrow a_{-k}^* &= a_k \quad \text{or, } a_k^* = a_{-k} \end{aligned}$$

$$\begin{aligned} (a+b)^* &= (a^* + b^*) \\ (a \times b)^* &= (a^* \times b^*) \end{aligned}$$

$$m = -k$$

$$k = m$$

## ■ Alternative Forms of the Fourier Series:

$$\begin{aligned}
 x(t) &= \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} \\
 \Rightarrow x(t) &= a_0 + \sum_{k=1}^{\infty} [a_k e^{jkw_0 t} + a_{-k} e^{-jkw_0 t}] \\
 &= a_0 + \sum_{k=1}^{\infty} [a_k e^{jkw_0 t} + a_k^* e^{-jkw_0 t}] \\
 a_{-k} &= a_k^* \\
 a_k e^{jkw_0 t} + a_k^* e^{-jkw_0 t} &= (R+jI)(C+jS) + (R-jI)(C-jS) \\
 &= (RC-IS) + j(RS+IC) + (RC-IS) - j(RS+IC) \\
 &= 2(RC-IS) \\
 &= a_0 + \sum_{k=1}^{\infty} 2 \operatorname{Re} \{a_k e^{jkw_0 t}\}
 \end{aligned}$$

## ■ Alternative Forms of the Fourier Series:

- If  $a_k = A_k e^{j\theta_k}$

$$\begin{aligned}
 \Rightarrow x(t) &= a_0 + \sum_{k=1}^{\infty} 2 \operatorname{Re} \{A_k e^{j\theta_k} e^{jkw_0 t}\} \\
 &= a_0 + \sum_{k=1}^{\infty} 2 \operatorname{Re} \{A_k e^{j(kw_0 t + \theta_k)}\} \\
 &= a_0 + 2 \sum_{k=1}^{\infty} A_k \cos(kw_0 t + \theta_k)
 \end{aligned}$$

$e^{j\theta} = \cos(\theta) + j\sin(\theta)$

- If  $a_k = B_k + j C_k$

$$\begin{aligned}
 \Rightarrow x(t) &= a_0 + \sum_{k=1}^{\infty} 2 \operatorname{Re} \{(B_k + j C_k) e^{jkw_0 t}\} \\
 &= a_0 + 2 \sum_{k=1}^{\infty} [B_k \cos(kw_0 t) - C_k \sin(kw_0 t)]
 \end{aligned}$$

$(a+jb)(c+jd) = (ac-bd) + j(ad+bc)$   
 $C(a+b) = C(a)C(b) - S(a)S(b)$

## ■ Example 3.4:

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk w_0 t} = \sum_{k=-\infty}^{+\infty} a_k e^{jk(2\pi/T)t}$$

e<sup>jst</sup>

$$e^{j\theta} = \cos(\theta) + j \sin(\theta); \quad \cos(\theta) = \frac{1}{2}(e^{j\theta} + e^{-j\theta}); \quad \sin(\theta) = \frac{1}{2j}(e^{j\theta} - e^{-j\theta})$$

$$x(t) = 1 + \sin w_0 t + 2 \cos w_0 t + \cos(2w_0 t + \frac{\pi}{4}) \rightarrow a_k = \frac{1}{T} \int \dots$$

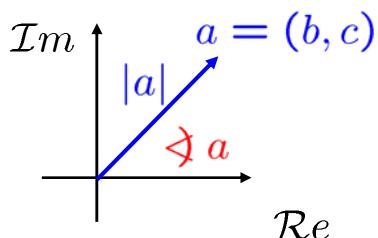
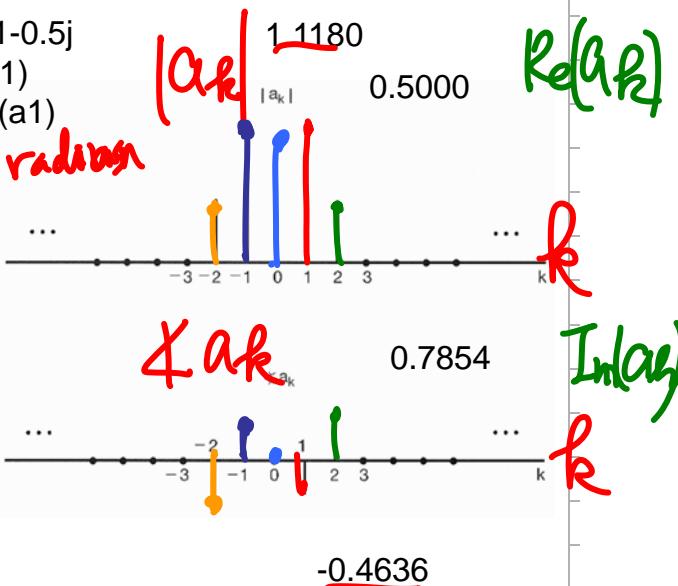
$$\Rightarrow x(t) = 1 + \frac{1}{2j} [e^{jw_0 t} - e^{-jw_0 t}] + [e^{jw_0 t} + e^{-jw_0 t}] + \frac{1}{2} [e^{j(2w_0 t + \pi/4)} + e^{-j(2w_0 t + \pi/4)}]$$

$$\Rightarrow x(t) = 1 + a_0 + a_1 e^{jw_0 t} + a_{-1} e^{-jw_0 t} + a_2 e^{j2w_0 t} + a_{-2} e^{-j2w_0 t}$$

## ■ Example 3.4:

$$\begin{cases} a_0 = 1, \\ a_1 = (1 + \frac{1}{2j}) = 1 - \frac{1}{2}j, \\ a_{-1} = (1 - \frac{1}{2j}) = 1 + \frac{1}{2}j, \\ a_2 = \frac{1}{2}e^{j(\pi/4)} = \frac{\sqrt{2}}{4}(1 + j), \\ a_{-2} = \frac{1}{2}e^{-j(\pi/4)} = \frac{\sqrt{2}}{4}(1 - j), \\ a_k = 0, \end{cases}$$

>> a1 = 1-0.5j  
>> abs(a1)  
>> angle(a1)

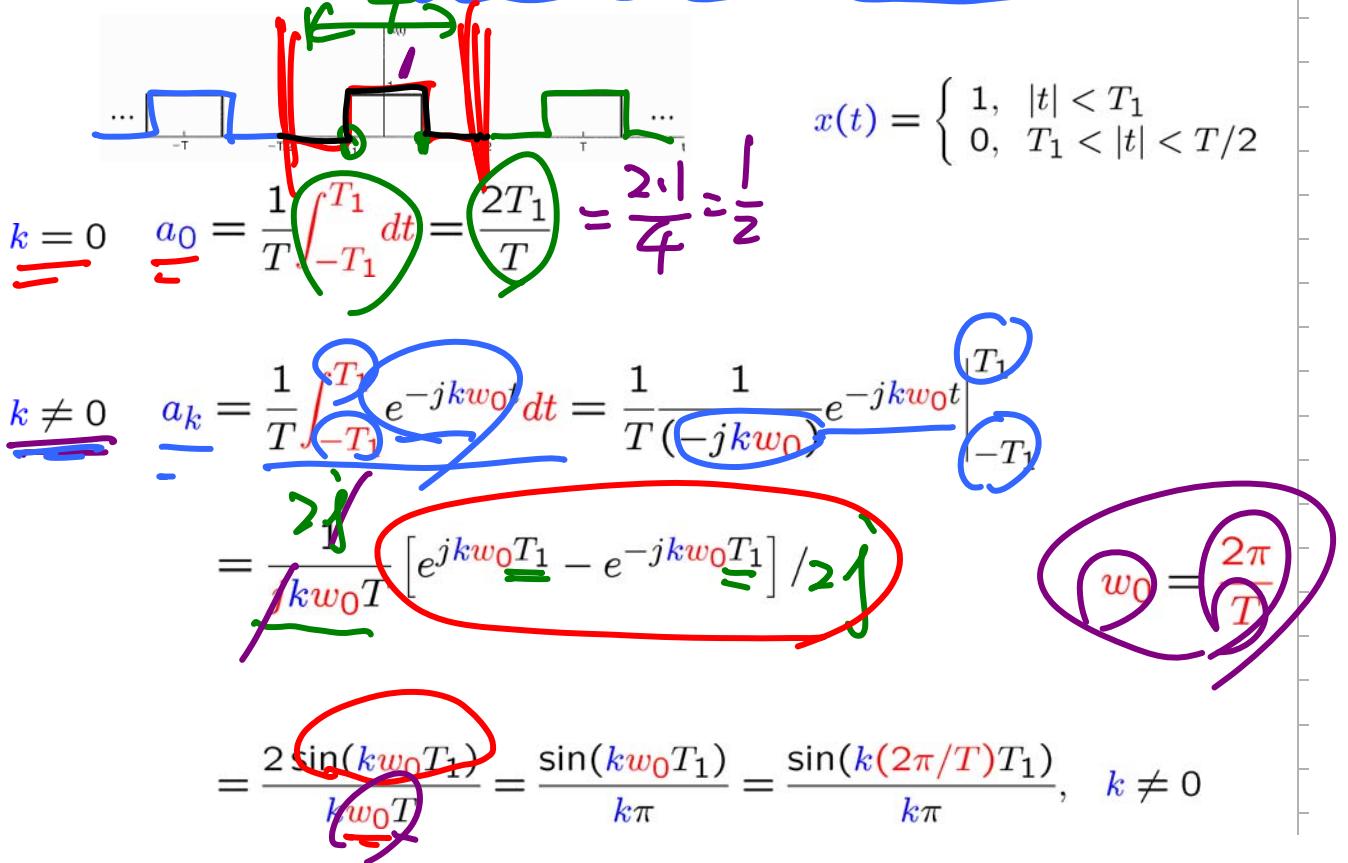


$$a = |a| e^{j\theta_a}$$

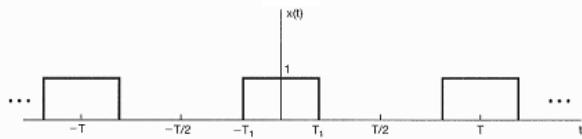
$$a = |a| [\cos(\theta_a) + j \sin(\theta_a)]$$

$$a = b + jc = \sqrt{b^2+c^2} \left[ \frac{b}{\sqrt{b^2+c^2}} + j \frac{c}{\sqrt{b^2+c^2}} \right]$$

■ Example 3.5:  $a_k = \frac{1}{T} \int_T x(t) e^{-jkw_0 t} dt = \frac{1}{T} \int_T x(t) e^{-jk(2\pi/T)t} dt$



■ Example 3.5:  $T = 4T_1$



$$a_k = \frac{\sin(k2\pi \frac{T_1}{T})}{k\pi}$$

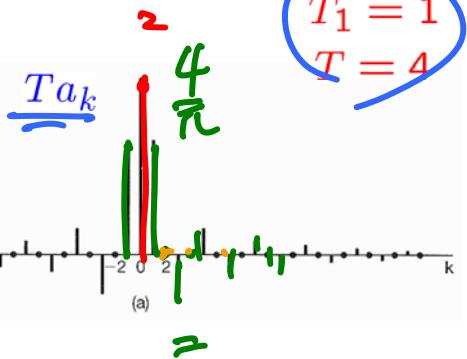
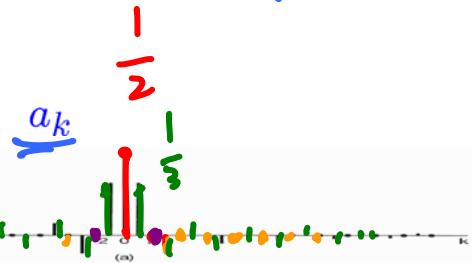
$$= \frac{\sin(k\pi/2)}{k\pi}$$

$$\frac{T}{a_k} = T \frac{\sin(k2\pi \frac{T_1}{T})}{k\pi}$$

$$= T \frac{\sin(k\pi/2)}{k\pi}$$

$$a_0 = \frac{1}{4} = \frac{1}{2}$$

$$a_1 = \frac{\sin(1 - \frac{1}{2})}{1 \cdot \pi} = \frac{1}{\pi}$$



## Fourier Series Representation of CT Periodic Signals

### ■ Example 3.5:

$$T = 4T_1$$



$$a_k = \frac{\sin(k2\pi\frac{T_1}{T})}{k\pi}$$

$$a_k = \frac{\sin(k\frac{\pi}{2})}{k\pi} \quad \frac{1}{4} \cdot 2\pi$$

$$a_0 = \frac{2T_1}{T}$$

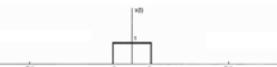
$$T_1 = 1$$

$$T = 4$$

$$\frac{2}{4} = \frac{1}{2}$$

(a)

$$T = 8T_1$$



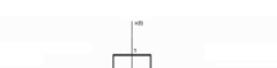
$$a_k = \frac{\sin(k\frac{\pi}{2})}{k\pi} \quad \frac{1}{8} \cdot 2\pi$$

$$a_0 = \frac{1}{4}$$

$$\frac{2}{8} > \frac{1}{4}$$

(b)

$$T = 16T_1$$



$$a_k = \frac{\sin(k\frac{\pi}{8})}{k\pi} \quad \frac{1}{16} \cdot 2\pi$$

$$a_0 = \frac{1}{8}$$

$$\frac{2}{16} = \frac{1}{8}$$

(c)

## Fourier Series Representation of CT Periodic Signals

### ■ Example 3.5: $T a_k = T \frac{\sin(k2\pi\frac{T_1}{T})}{k\pi}$

$$T = 4T_1$$



$$Ta_k = T \frac{\sin(k\frac{\pi}{2})}{k\pi}$$

$$2$$

$$Ta_k$$

(a)

$$T = 8T_1$$



$$Ta_k = T \frac{\sin(k\frac{\pi}{4})}{k\pi}$$

$$2$$

$$Ta_k$$

(b)

$$T = 16T_1$$



$$Ta_k = T \frac{\sin(k\frac{\pi}{8})}{k\pi}$$

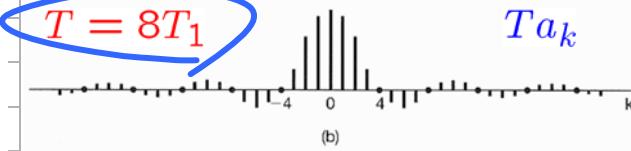
$$2$$

$$Ta_k$$

(c)

## ■ Example 3.5:

$$T = 8T_1$$



$$T a_k = T \frac{\sin(k\pi)}{k\pi}$$

$$= \frac{1}{4} T \frac{\sin(\pi \frac{k}{4})}{\pi \frac{k}{4}}$$

$$= \frac{1}{4} T \text{sinc}\left(\frac{k}{4}\right)$$

$$K = \theta = 1, 2, 3, 4$$

$$\text{sinc}(\theta) = \frac{\sin(\pi\theta)}{\pi\theta}$$

$$\frac{\sin(\pi\theta)}{\pi\theta}$$

$$w_0 = \frac{2\pi}{T} \quad T a_k = \frac{2 \sin(wT_1)}{w}$$

$$w = kw_0 \quad wT_1 = k \left(\frac{2\pi}{T}\right) \cdot T_1$$

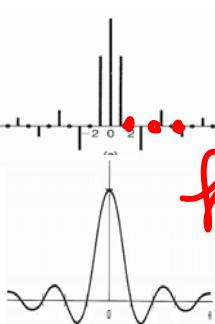
$$= \frac{2k\pi}{A}$$

## ■ Example 3.5:

$$T = 4T_1$$

$$T a_k = T \frac{\sin(k\pi)}{k\pi}$$

$$= \frac{1}{2} T \text{sinc}\left(\frac{k}{2}\right) \quad \theta = \frac{k}{2}$$



$$k = \pm 2, 4, 6 \dots$$

$$T a_k = T \frac{\sin(k2\pi)}{k\pi}$$

$$T a_k = T \frac{\sin(k\pi)}{k\pi}$$

$$= \frac{1}{8} T \text{sinc}\left(\frac{k}{8}\right)$$

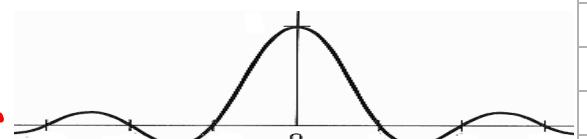
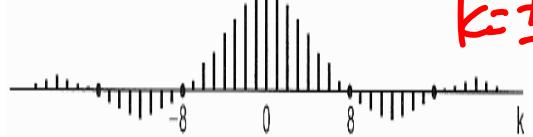
$$k = \pm 8, \pm 16 \dots$$

$$T = 8T_1$$

$$T a_k = T \frac{\sin(k\pi)}{k\pi}$$

$$= \frac{1}{4} T \text{sinc}\left(\frac{k}{4}\right)$$

$$k = \pm 4, \pm 8, \dots$$



■ Example 3.5:

$$T a_k = T \frac{2 \sin(k w_0 T_1)}{k w_0 T}$$

$$w_0 = \frac{2\pi}{T}$$

$$= T_1 \frac{2 \sin(k w_0 T_1)}{k w_0 T_1}$$

$$= \frac{1}{2} T \text{sinc}\left(\frac{k}{2}\right)$$

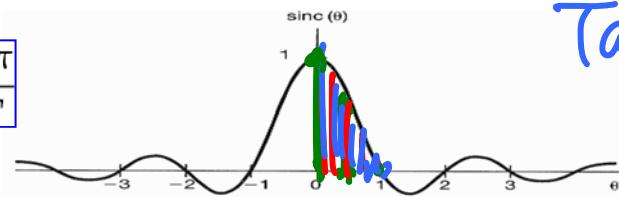
$$= \frac{1}{4} T \text{sinc}\left(\frac{k}{4}\right)$$

$$= \frac{1}{8} T \text{sinc}\left(\frac{k}{8}\right)$$

$$T = 4T_1$$

$$T = 8T_1$$

$$T = 16T_1$$



Tak

Outline

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- **Convergence of the Fourier Series**
- Properties of Continuous-Time Fourier Series
- Fourier Series Representation of Discrete-Time Periodic Signals
- Properties of Discrete-Time Fourier Series
- Fourier Series & LTI Systems
- Filtering & Examples of CT & DT Filters

- Fourier maintained that "any" periodic signal could be represented by a Fourier series
- The truth is that Fourier series can be used to represent an extremely large class of periodic signals
- The question is that when a periodic signal  $x(t)$  does in fact have a Fourier series representation?

 $x(t)$ 

$$x_{FS}(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk(2\pi/T)t}$$

$$a_k = \frac{1}{T} \int_T x(t) e^{-jk(2\pi/T)t} dt$$

$$x_N(t) = \sum_{k=-N}^{+N} a_k e^{jk(2\pi/T)t}$$

- One class of periodic signals:**
  - Which have finite energy over a single period:

$$\int_T |x(t)|^2 dt < \infty \Rightarrow a_k = \frac{1}{T} \int_T x(t) e^{-jkw_0 t} dt < \infty$$

$$x_N(t) = \sum_{k=-N}^{+N} a_k e^{jkw_0 t}$$

$$e_N(t) = x(t) - x_N(t)$$

$$E_N(t) = \int_T |e_N(t)|^2 dt$$

$\rightarrow 0$  as  $N \rightarrow \infty$

$$e(t) = x(t) - \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} \rightarrow 0$$

$$E(t) = \int_T |e(t)|^2 dt = 0$$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t}, \quad \forall t ???$$

- The other class of periodic signals:

- Which satisfy **Dirichlet conditions**:

- Condition 1:**

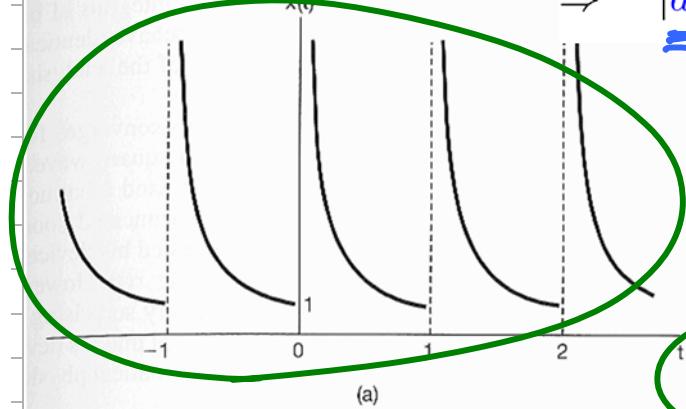
- Over any period,

$x(t)$  must be absolutely integrable,  
i.e.,

$$\int_T |x(t)| dt < \infty$$



Johann Peter Gustav Lejeune Dirichlet  
1805-1859  
Born in Germany  
Photo from wikipedia



$$\Rightarrow |a_k| \leq \frac{1}{T} \int_T |x(t) e^{-jk\omega_0 t}| dt = \frac{1}{T} \int_T |x(t)| dt < \infty$$

$$x(t) = \frac{1}{t}, \quad 0 < t \leq 1$$

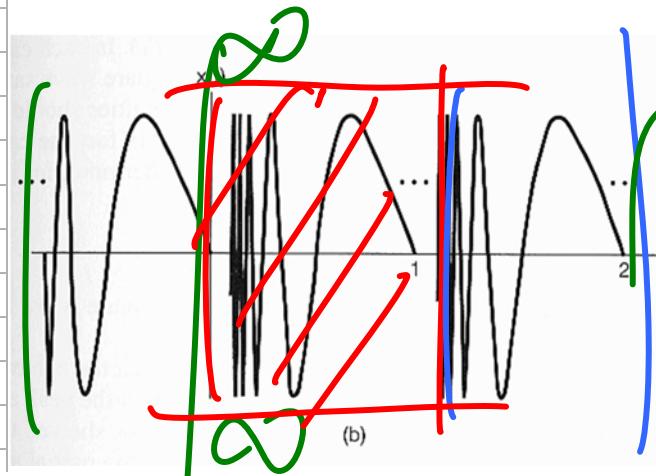
- The other class of periodic signals:

- Which satisfy **Dirichlet conditions**:

- Condition 2:**

- In any finite interval,  $x(t)$  is of bounded variation; i.e.,

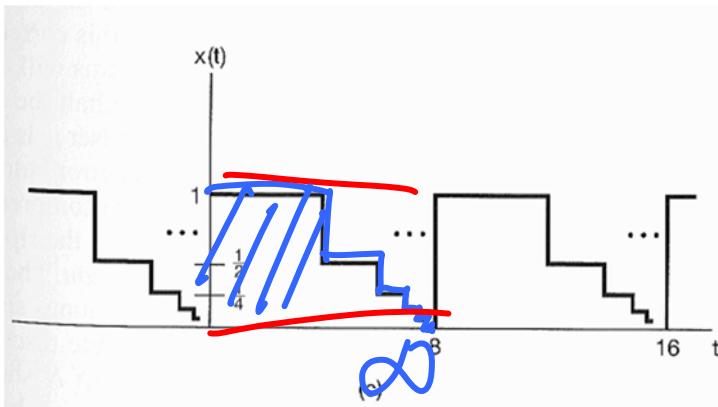
- There are **no more than a finite number of maxima and minima** during any single period of the signal



$$x(t) = \sin(\frac{2\pi}{t}) \quad 0 < t \leq 1$$

$$\int_0^1 |x(t)| dt < 1$$

- The other class of periodic signals:
  - Which satisfy **Dirichlet conditions**:
  - **Condition 3:**
    - In any finite interval,  
 $x(t)$  has only **finite number of discontinuities**.
    - Furthermore, each of these discontinuities is **finite**



- How the Fourier series converges for a periodic signal with discontinuities

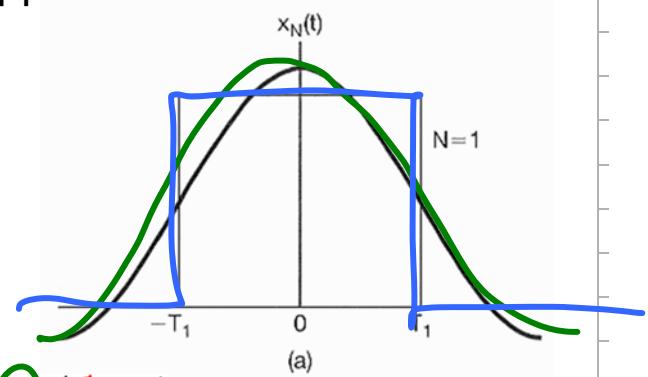
- In 1898,  
Albert Michelson (an American physicist) used his harmonic analyzer to compute the truncated Fourier series approximation for the square wave



Albert Abraham Michelson  
1852-1931  
Polish-born German-American  
Photo from wikipedia

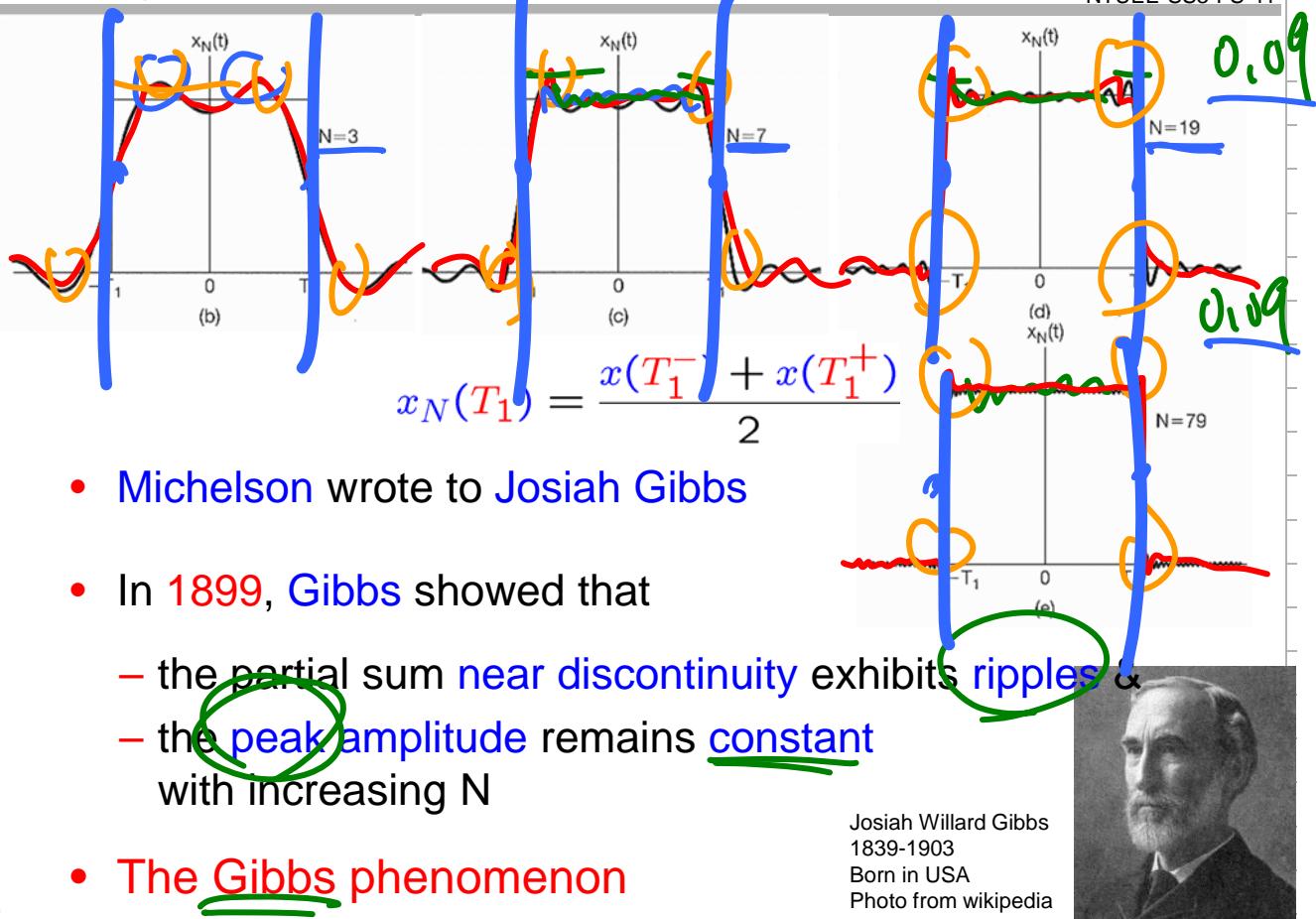
$$x_N(t) = \sum_{k=-N}^{+N} a_k e^{j k w_0 t}$$

$$x_1(t) = a_{-1} e^{-j \cdot 1 \cdot w_0 t} + a_0 + a_1 e^{j \cdot 1 \cdot w_0 t} =$$



## Convergence of the Fourier Series

$$x(t) = \sum \dots$$

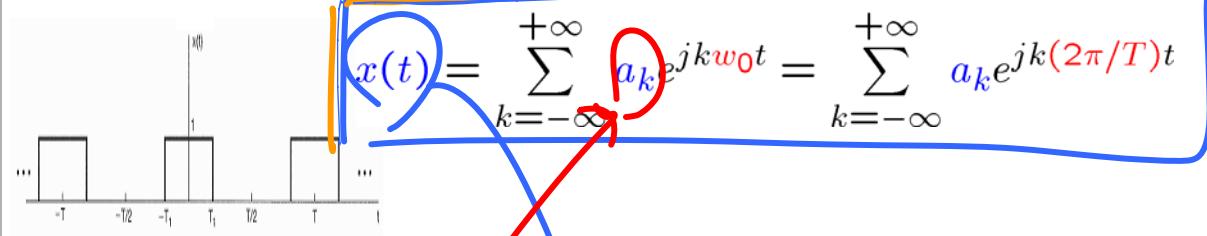


## Outline

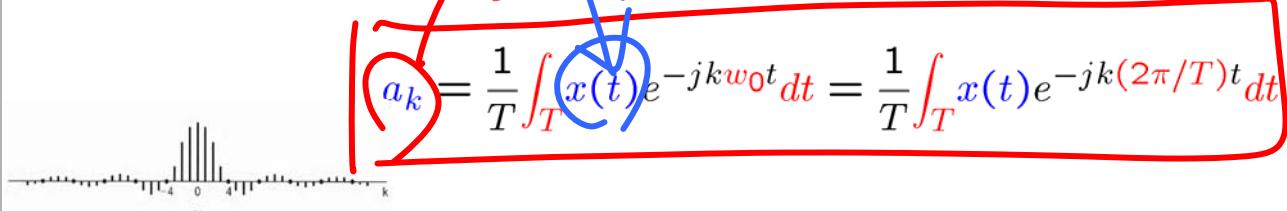
- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
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## ▪ CT Fourier Series Representation:

- The **synthesis** equation:



- The **analysis** equation:



- $x(t) \xrightarrow{\mathcal{FS}} a_k$ : Fourier series pair

## Outline

Section	Property
3.5.1	Linearity
3.5.2	Time Shifting
	Frequency Shifting
3.5.6	Conjugation
3.5.3	Time Reversal
3.5.4	Time Scaling
	Periodic Convolution
3.5.5	Multiplication
	Differentiation
	Integration
3.5.6	Conjugate Symmetry for Real Signals
3.5.6	Symmetry for Real and Even Signals
3.5.6	Symmetry for Real and Odd Signals
	Even-Odd Decomposition for Real Signals
3.5.7	Parseval's Relation for Periodic Signals

■ Linearity:  $X(t+T) = X(t)$

- $x(t), y(t)$ : periodic signals with period  $T$

$$x(t) \xleftrightarrow{\mathcal{FS}} a_k$$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t}$$

$$y(t) \xleftrightarrow{\mathcal{FS}} b_k$$

$$y(t) = \sum_{m=-\infty}^{+\infty} b_m e^{jmw_0 t}$$

$$\Rightarrow z(t) = Ax(t) + By(t) \xleftrightarrow{\mathcal{FS}} c_k = Aa_k + Bb_k$$

$$z(t) = \sum_{k=-\infty}^{+\infty} c_k e^{jkw_0 t}$$

$$= \frac{1}{T} \int_T \left[ A X(t) e^{-jkw_0 t} dt \right] + \frac{1}{T} \int_T \left[ B Y(t) e^{-jkw_0 t} dt \right]$$

$$c_k = \frac{1}{T} \int_T z(t) e^{-jkw_0 t} dt = \frac{1}{T} \int_T (A X(t) + B Y(t)) e^{-jkw_0 t} dt$$

Add

$$A a_k + B b_k$$

■ Time Shifting:

- $x(t)$ : periodic signal with period  $T$

$$x(t) \xleftrightarrow{\mathcal{FS}} a_k$$

$$\Rightarrow x(t - t_0) \xleftrightarrow{\mathcal{FS}} b_k = e^{-jkw_0 t_0} a_k = e^{-jk\left(\frac{2\pi}{T}\right)t_0} a_k$$

$$b/c \quad b_k = \frac{1}{T} \int_T x(t - t_0) e^{-jkw_0 t} dt$$

$$t - t_0 = \tau$$

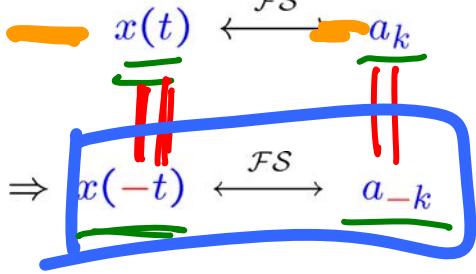
$$t = \tau + t_0$$

$$dt = d\tau$$

$$= \frac{1}{T} \int_T x(\tau) e^{-jkw_0 (\tau + t_0)} d\tau$$

$$= e^{-jkw_0 t_0} \left[ \frac{1}{T} \int_T x(\tau) e^{-jkw_0 \tau} d\tau \right] = a_k$$

### ▪ Time Reversal:



- If  $x(t)$  is even, i.e.,  $x(-t) = x(t)$

$\Rightarrow$   $a_k$  is even, i.e.,  $a_{-k} = a_k$

- If  $x(t)$  is odd, i.e.,  $x(-t) = -x(t)$

$\Rightarrow$   $a_k$  is odd, i.e.,  $a_{-k} = -a_k$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\left(\frac{2\pi}{T}\right)t}$$

$$x(-t) = \sum_{k=-\infty}^{+\infty} a_k e^{-jk\left(\frac{2\pi}{T}\right)t}$$

$$= \sum_{m=-\infty}^{+\infty} a_{-m} e^{jm\left(\frac{2\pi}{T}\right)t}$$

$$\text{---} \quad -k = m$$

### ▪ Time Scaling:

$\chi(2t)$

- $x(t)$ : periodic signals with period  $T$  and fundamental frequency  $w_0$

- $x(\alpha t)$ : periodic signals with period  $\frac{T}{\alpha}$  and fundamental frequency  $\alpha w_0$

$\frac{T}{\alpha}$   
 $\frac{1}{2w_0}$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t} = \sum_{k=-\infty}^{+\infty} a_k e^{jk\left(\frac{2\pi}{T}\right)t}$$

$$x(\alpha t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk w_0 (\alpha t)} = \sum_{k=-\infty}^{+\infty} a_k e^{j k \alpha \left(\frac{2\pi}{T}\right) t}$$

$$= \sum_{k=-\infty}^{+\infty} a_k e^{j k \left(\frac{2\pi}{T}\right) t}$$

## Multiplication:

•  $\underline{x(t)}, \underline{y(t)}$ : periodic signals with period  $\underline{T}$

$$x(t) \xleftrightarrow{\mathcal{FS}} a_k$$

$$y(t) \xleftrightarrow{\mathcal{FS}} b_k$$

$$\Rightarrow \underline{x(t)y(t)} \text{ also periodic with } \underline{T}$$

$$z(t) = x(t)y(t) \xleftrightarrow{\mathcal{FS}} c_k = \sum_{k=-\infty}^{+\infty} a_l b_{k-l}$$

$$= \sum_l a_l e^{jkw_0 t} \sum_m b_m e^{jmw_0 t}$$

$$= \sum_l \left( \sum_m a_l b_m \right) e^{j(l+m)w_0 t}$$

$$= \sum_k \left( \sum_m a_k b_m \right) e^{jkw_0 t}$$

$$(a+b+c)(d+e+f)$$
  

$$= ad + ae + af$$
  

$$+ bd + be + bf$$
  

$$+ cd + ce + cf$$

$k = l+m$   
 $l = k-m$

Add

## Differentiation:

- $\underline{x(t)}$ : periodic signals with period  $\underline{T}$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t}$$

$$\frac{d}{dt} x(t) \xleftrightarrow{\mathcal{FS}} jk w_0 a_k$$

$$\frac{d}{dt} x(t) = \sum_{k=-\infty}^{+\infty} a_k \frac{d}{dt} e^{jkw_0 t} = \sum_{k=-\infty}^{+\infty} a_k jk w_0 e^{jkw_0 t}$$

## ■ Integration:

- $x(t)$ : periodic signals with period  $T$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{j k w_0 t}$$

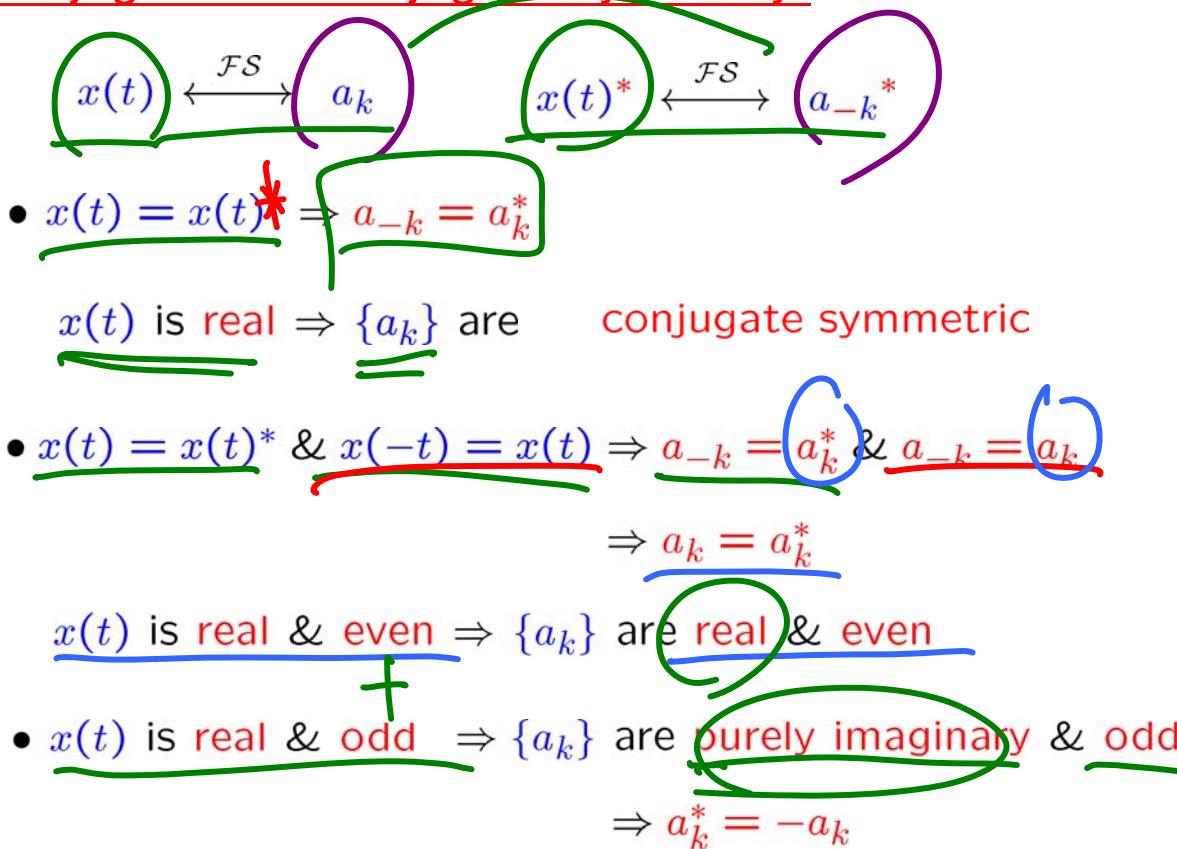
only if  $a_0 = 0$ , it is finite valued and periodic

## ■ Conjugation & Conjugate Symmetry:

$$\begin{aligned}
 x(t) &\xleftrightarrow{\mathcal{F}S} a_k \\
 x(t)^* &\xleftrightarrow{\mathcal{F}S} a_{-k}^* \\
 x(t) &= \sum_{k=-\infty}^{+\infty} a_k e^{j k w_0 t} \\
 &= \sum_{k=-\infty}^{+\infty} a_k^* e^{j -k w_0 t} \\
 &= \sum_{k=-\infty}^{+\infty} a_{-k}^* e^{j k w_0 t}
 \end{aligned}$$

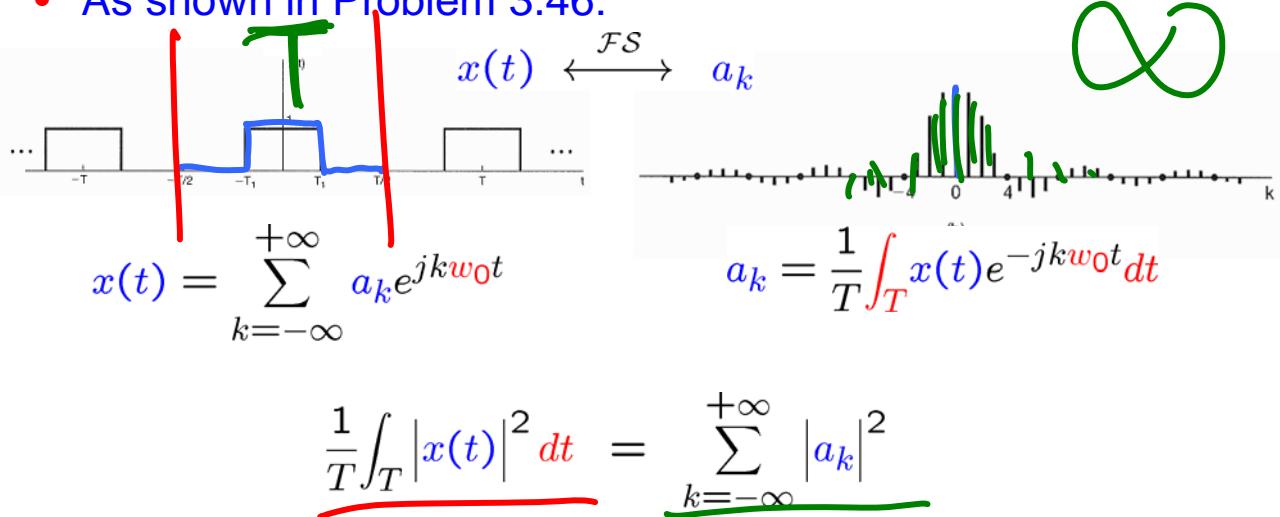
$k \rightarrow m$   
 $m \rightarrow k$

## Conjugation & Conjugate Symmetry:



## Parseval's relation for CT periodic signals:

- As shown in Problem 3.46:



- Parseval's relation states that the total average power in a periodic signal equals the sum of the average powers in all of its harmonic components

TABLE 3.1 PROPERTIES OF CONTINUOUS-TIME FOURIER SERIES

Property	Section	Periodic Signal	Fourier Series Coefficients
		$x(t)$ } Periodic with period T and $y(t)$ } fundamental frequency $\omega_0 = 2\pi/T$	$a_k$ $b_k$
Linearity	3.5.1	$Ax(t) + By(t)$	$Aa_k + Bb_k$
Time Shifting	3.5.2	$x(t - t_0)$	$a_k e^{-jkt_0\omega_0} = a_k e^{-jk(2\pi/T)t_0}$
Frequency Shifting		$e^{jM\omega_0 t} = e^{jM(2\pi/T)t} x(t)$	$a_{k-M}$
Conjugation	3.5.6	$x^*(t)$	$a_{-k}$
Time Reversal	3.5.3	$x(-t)$	$a_{-k}$
Time Scaling	3.5.4	$x(\alpha t), \alpha > 0$ (periodic with period $T/\alpha$ )	$a_k$
Periodic Convolution		$\int_T x(\tau)y(t-\tau)d\tau$	$Ta_k b_k$
Multiplication	3.5.5	$x(t)y(t)$	$\sum_{l=-\infty}^{+\infty} a_l b_{k-l}$
Differentiation		$\frac{dx(t)}{dt}$	$jk\omega_0 a_k = jk \frac{2\pi}{T} a_k$
Integration		$\int_{-\infty}^t x(t) dt$ (finite valued and periodic only if $a_0 = 0$ )	$\left(\frac{1}{jk\omega_0}\right)a_k = \left(\frac{1}{jk(2\pi/T)}\right)a_k$
Conjugate Symmetry for Real Signals	3.5.6	$x(t)$ real	$\begin{cases} a_k = a_{-k} \\ \Re\{a_k\} = \Re\{a_{-k}\} \\ \Im\{a_k\} = -\Im\{a_{-k}\} \\  a_k  =  a_{-k}  \\ \angle a_k = -\angle a_{-k} \end{cases}$
Real and Even Signals	3.5.6	$x(t)$ real and even	$a_k$ real and even
Real and Odd Signals	3.5.6	$x(t)$ real and odd	$a_k$ purely imaginary and odd
Even-Odd Decomposition of Real Signals	3.5.6	$\begin{cases} x_e(t) = \Re\{x(t)\} \quad [x(t) \text{ real}] \\ x_o(t) = \Im\{x(t)\} \quad [x(t) \text{ real}] \end{cases}$	$\Re\{a_k\}$ $j\Im\{a_k\}$

Parseval's Relation for Periodic Signals

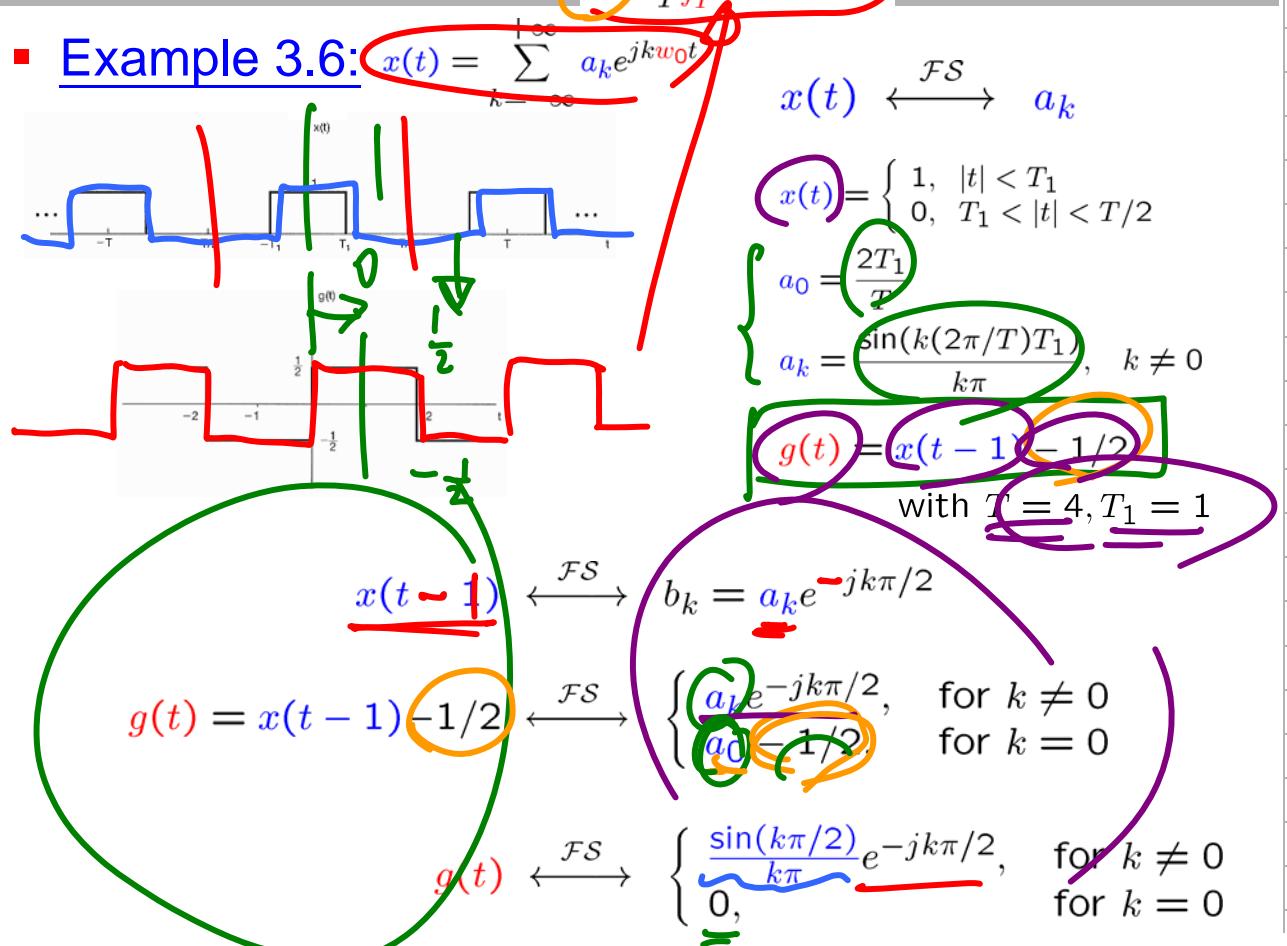
$$\frac{1}{T} \int_T |x(t)|^2 dt = \sum_{k=-\infty}^{+\infty} |a_k|^2$$

$$\int u v' = (uv)' - \int u' v$$

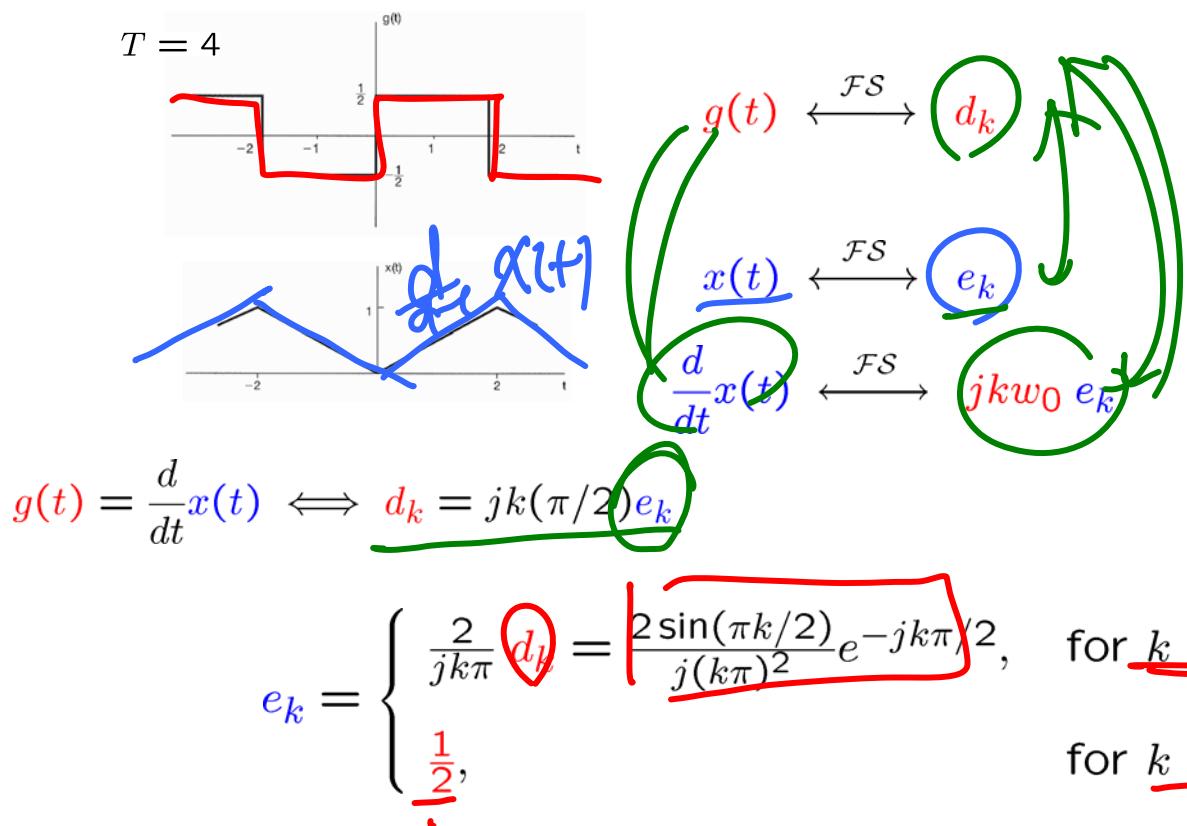
3/15/2  
3=14pm

## Properties of CT Fourier Series

## Example 3.6:

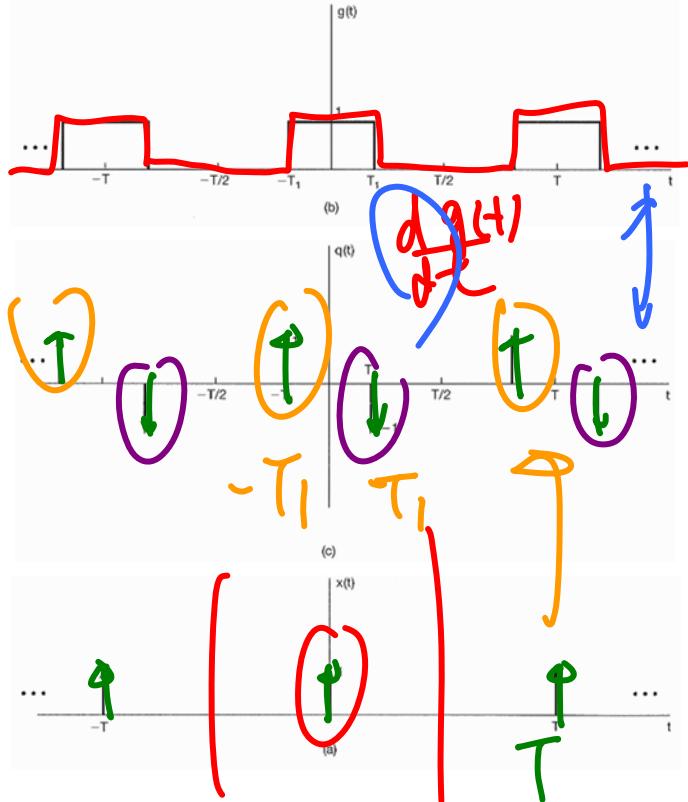


## ■ Example 3.7:



## ■ Example 3.8:

$$a_k = \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt \Big|_{t=0} = \frac{1}{T}$$



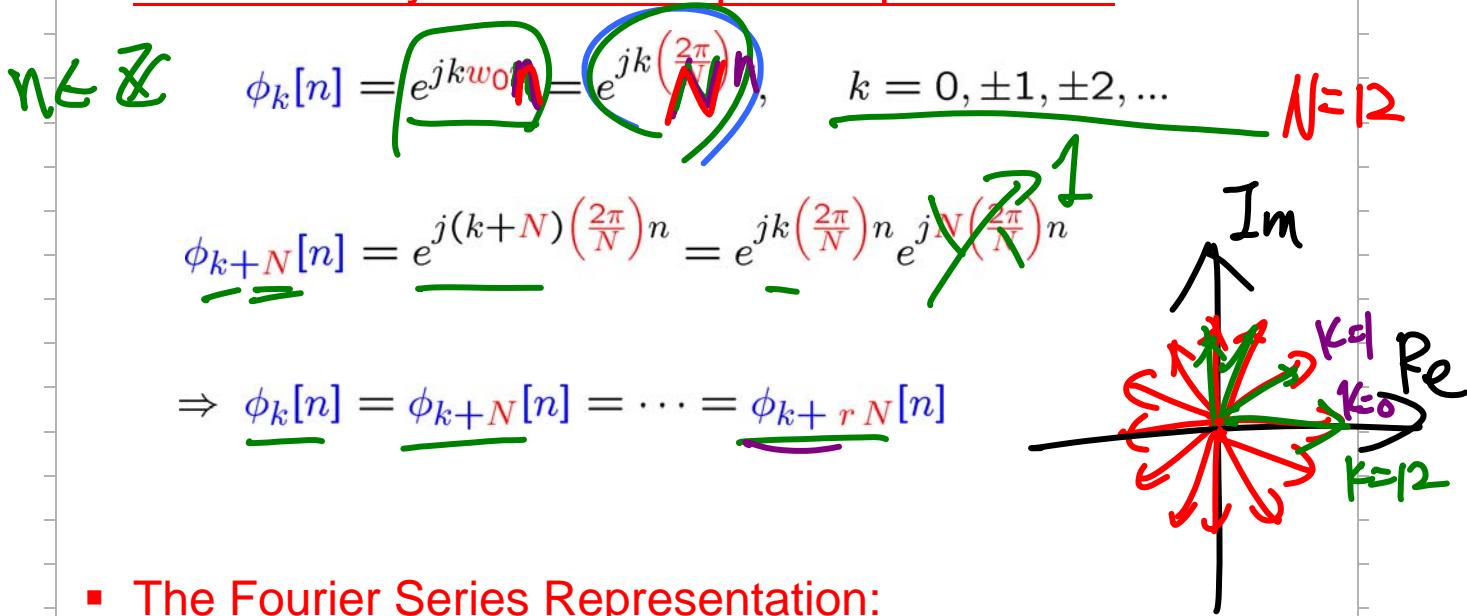
- Example 3.8:

$$\begin{aligned}
 b_k &= e^{jk\omega_0 T_1} \underline{a_k} - e^{-jk\omega_0 T_1} \underline{\bar{a}_k} \quad | \\
 &\stackrel{2j}{=} \frac{1}{T} [e^{jk\omega_0 T_1} - e^{-jk\omega_0 T_1}] \quad | 2j \\
 &= \frac{2j \sin(k\omega_0 T_1)}{T} \\
 b_k &= jk\omega_0 \underline{\underline{c_k}} \\
 \left. \begin{array}{l} c_k = \frac{b_k}{jk\omega_0} = \frac{2j \sin(k\omega_0 T_1)}{jk\omega_0 T} = \frac{\sin(k\omega_0 T_1)}{k\pi} \\ c_0 = \frac{2T_1}{T} \end{array} \right\} k \neq 0
 \end{aligned}$$

## Outline

- A Historical Perspective
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- Fourier Series & LTI Systems
- Filtering & Examples of CT & DT Filters

- Harmonically related complex exponentials



- The Fourier Series Representation:

$$x[n] = \sum_{k=-N}^N a_k \phi_k[n] = \sum_{k=-N}^N a_k e^{jkw_0 n} = \sum_{k=-N}^N a_k e^{jk\left(\frac{2\pi}{N}\right)n}$$

- Procedure of Determining the Coefficients:

$$x[0] = \sum_{k=-N}^N a_k$$

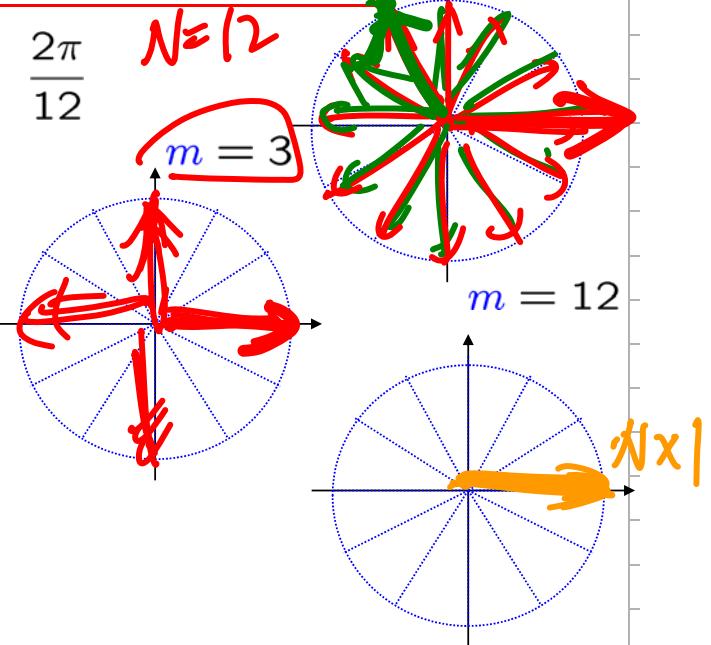
$$x[1] = \sum_{k=-N}^N a_k e^{jk\left(\frac{2\pi}{N}\right)}$$

$$x[2] = \sum_{k=-N}^N a_k e^{jk2\left(\frac{2\pi}{N}\right)}$$

⋮

$$x[N-1] = \sum_{k=-N}^N a_k e^{jk(N-1)\left(\frac{2\pi}{N}\right)}$$

$$x[N] = \sum_{k=-N}^N a_k e^{jk(N)\left(\frac{2\pi}{N}\right)}$$



and  $\sum_{n=-N}^N e^{jm\left(\frac{2\pi}{N}\right)n} = \begin{cases} N, & m = 0, \pm N, \pm 2N, \dots \\ 0, & \text{otherwise} \end{cases}$

## Procedure of Determining the Coefficients.

$$x[n] = \sum_{k=-N}^{N-1} a_k e^{jk\left(\frac{2\pi}{N}\right)n}$$

$$\sum_{n=-N}^{N-1} x[n] e^{-jr\left(\frac{2\pi}{N}\right)n} = \sum_{n=-N}^{N-1} \sum_{k=-N}^{N-1} a_k e^{j(k-r)\left(\frac{2\pi}{N}\right)n} = 0$$

$$\sum_{n=-N}^{N-1} x[n] e^{-jr\left(\frac{2\pi}{N}\right)n} = \sum_{k=r}^{N-1} a_k \sum_{n=-N}^{N-1} e^{j(k-r)\left(\frac{2\pi}{N}\right)n} \quad \text{if } k \neq r$$

$$= a_r N$$

$\cancel{k=r}$   
 $\cancel{k=r+N}$   
 $\cancel{k=r+2N}$

$m$   
 $r$   
 $k$

$$\Rightarrow a_r = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jr\left(\frac{2\pi}{N}\right)n}$$

## In Summary:

- The **synthesis** equation:

$$x[n] = \sum_{k=-N}^{N-1} a_k e^{jk\omega_0 n} = \sum_{k=-N}^{N-1} a_k e^{jk\left(\frac{2\pi}{N}\right)n}$$

- The **analysis** equation:

$$a_k = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jk\omega_0 n} = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jk\left(\frac{2\pi}{N}\right)n}$$

$$a_k = \underline{a_k + N}$$

- $x[n] \xleftrightarrow{\mathcal{FS}} a_k$ : DT Fourier series pair
- $\{a_k\}$ : the Fourier series coefficients or the spectral coefficients of  $x[n]$

## ■ Example 3.11:

$$\begin{aligned}
 x[n] &= 1 + \sin\left(\frac{2\pi}{N}n\right) + 3\cos\left(\frac{2\pi}{N}n\right) + \cos\left(\frac{4\pi}{N}n + \frac{\pi}{2}\right) \\
 \Rightarrow x[n] &= 1 + \frac{1}{2j} \left[ e^{j\left(\frac{2\pi}{N}n\right)} - e^{-j\left(\frac{2\pi}{N}n\right)} \right] + \frac{3}{2} \left[ e^{j\left(\frac{2\pi}{N}n\right)} + e^{-j\left(\frac{2\pi}{N}n\right)} \right] \\
 &\quad + \frac{1}{2} \left[ e^{j\left(\frac{4\pi}{N}n + \frac{\pi}{2}\right)} + e^{-j\left(\frac{4\pi}{N}n + \frac{\pi}{2}\right)} \right] \\
 \Rightarrow x[n] &= a_0 + \left( \frac{3}{2} + \frac{1}{2j} \right) e^{j\left(\frac{2\pi}{N}n\right)} + \left( \frac{3}{2} - \frac{1}{2j} \right) e^{-j\left(\frac{2\pi}{N}n\right)} \\
 &\quad + \frac{1}{2} e^{j\left(\frac{\pi}{2}\right)} e^{j2\left(\frac{2\pi}{N}n\right)} + \frac{1}{2} e^{-j\left(\frac{\pi}{2}\right)} e^{-j2\left(\frac{2\pi}{N}n\right)}
 \end{aligned}$$

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

$a_0$        $a_1$        $a_{-1}$   
                  $a_2$        $a_{-2}$

## ■ Example 3.11:

$$\Rightarrow \begin{cases} a_0 = 1 \\ a_1 = \left(\frac{3}{2} + \frac{1}{2j}\right) \\ a_{-1} = \left(\frac{3}{2} - \frac{1}{2j}\right) \\ a_2 = 0 \\ a_{-2} = 0 \\ a_k = 0, \text{ others in } \langle N \rangle \end{cases}$$

$$a = |a| e^{j\arg a}$$

$$a = |a| [\cos(\arg a) + j \sin(\arg a)]$$

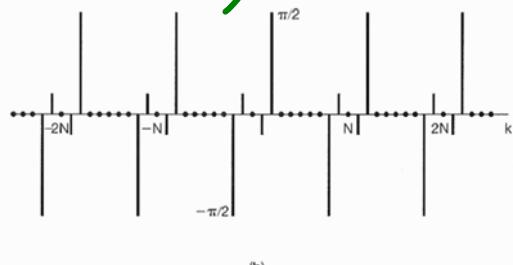
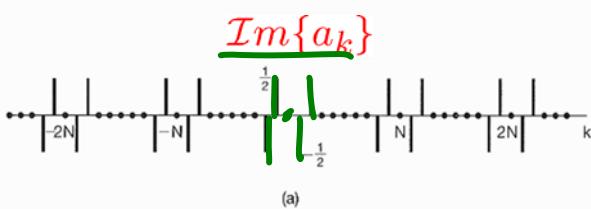
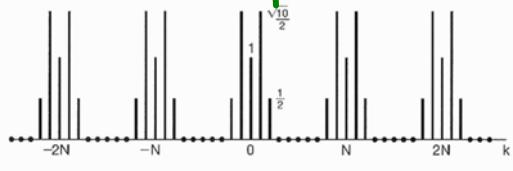
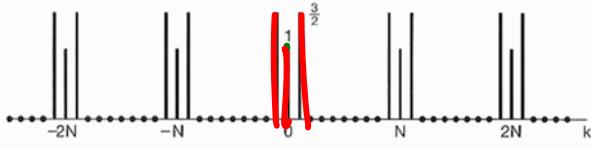
$$a = b + jc = \sqrt{b^2+c^2} \left[ \frac{b}{\sqrt{b^2+c^2}} + j \frac{c}{\sqrt{b^2+c^2}} \right]$$

$|a|$        $|a_k|$

$$a_{k+N} = a_k$$

$$\underline{\operatorname{Re}\{a_k\}}$$

$$N = 10$$



## ■ Example 3.12:

$a_k = \frac{1}{N} \sum_{n=-N_1}^{N_1} x[n] e^{-jk(\frac{2\pi}{N})n}$

$x[n] = \begin{cases} 1, & -N_1 \leq n \leq N_1 \\ 0, & \text{others in } \langle N \rangle \end{cases}$

$$\begin{aligned} a_k &= \frac{1}{N} \sum_{n=-N_1}^{N_1} 1 \cdot e^{-jk(\frac{2\pi}{N})n} \\ &= \frac{1}{N} \sum_{n=-N_1}^{N_1} (e^{-jk(\frac{2\pi}{N})})^n \\ &= \frac{1}{N} [(\cdot)^{-N_1} + (\cdot)^{-N_1+1} + \dots + (\cdot)^{N_1}] \\ &= \frac{1}{N} (\cdot)^{-N_1} \left[ \frac{1 - (\cdot)^{(2N_1+1)}}{1 - (\cdot)} \right] \quad (\cdot) \neq 1 \\ &= \frac{1}{N} (\cdot)^{-N_1} [1 + (\cdot)^1 + \dots + (\cdot)^{2N_1}] \end{aligned}$$

$a_k = \frac{1}{N} \sum_{m=0}^{2N_1} e^{-jk(\frac{2\pi}{N})(m-N_1)} = \frac{1}{N} e^{jk(\frac{2\pi}{N})N_1} \sum_{m=0}^{2N_1} e^{-jk(\frac{2\pi}{N})m}$

## ■ Example 3.12:

- $k = 0, \pm N, \pm 2N, \dots$

$$a_k = \frac{2N_1 + 1}{N}$$

- $k \neq 0, \pm N, \pm 2N, \dots$

$$\begin{aligned} &1 - e^{-j\theta} \\ &= e^{-j\theta/2} e^{j\theta/2} - e^{-j\theta/2} e^{-j\theta/2} \\ &= e^{-j\theta/2} (e^{j\theta/2} - e^{-j\theta/2}) \end{aligned}$$

$$\begin{aligned} a_k &= \frac{1}{N} e^{jk(\frac{2\pi}{N})N_1} \left( \frac{1 - e^{-jk(\frac{2\pi}{N})(2N_1+1)}}{1 - e^{-jk(\frac{2\pi}{N})}} \right) \\ &= \frac{1}{N} e^{-jk(\frac{2\pi}{2N})} \left[ e^{jk(\frac{2\pi}{2N})(2N_1+1)} - e^{-jk(\frac{2\pi}{2N})(2N_1+1)} \right] \\ &\quad e^{-jk(\frac{2\pi}{2N})} \left[ e^{jk(\frac{2\pi}{2N})} - e^{-jk(\frac{2\pi}{2N})} \right] \\ &= \frac{1}{N} \frac{\sin[(\frac{2\pi}{N})k(N_1 + \frac{1}{2})]}{\sin[(\frac{\pi}{N})k]} \end{aligned}$$

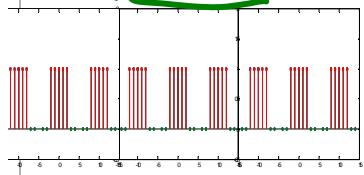
## Fourier Series Representation of DT Periodic Signals

5  
2/2

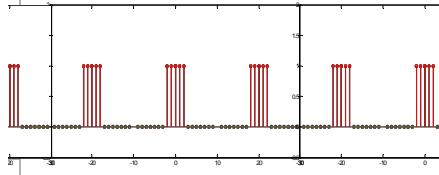
### ■ Example 3.12:

- $2N_1 + 1 = 5$

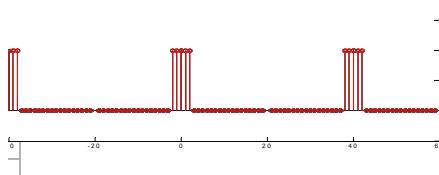
- $N = 10$



- $N = 20$



- $N = 40$



$a_k$

$$a_k = \frac{1}{N} \frac{\sin[(\frac{2\pi}{N})k(N_1 + \frac{1}{2})]}{\sin[(\frac{\pi}{N})k]}$$

$k=10$

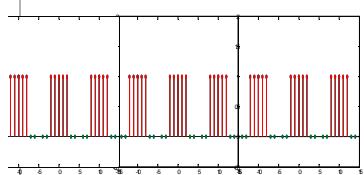
$> 0$

40

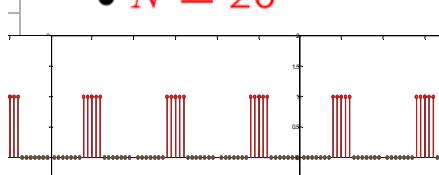
### ■ Example 3.12:

- $2N_1 + 1 = 5$

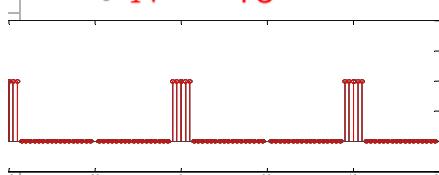
- $N = 10$



- $N = 20$



- $N = 40$



$Na_k$

$Na_k$

$Na_k$

$Na_k$

$Na_k$

$Na_k$

$Na_k$

$Na_k$

$Na_k$

$$a_k = \frac{1}{N} \frac{\sin[(\frac{2\pi}{N})k(N_1 + \frac{1}{2})]}{\sin[(\frac{\pi}{N})k]}$$

$k=10$

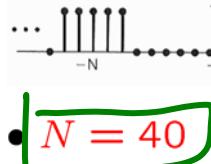
$> 0$

40

## Fourier Series Representation of DT Periodic Signals

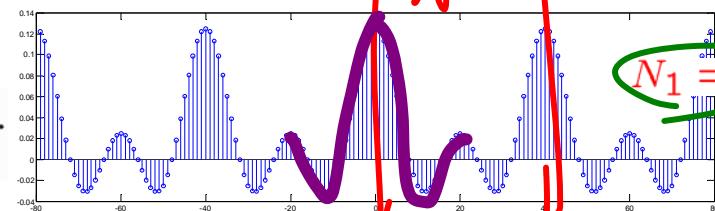
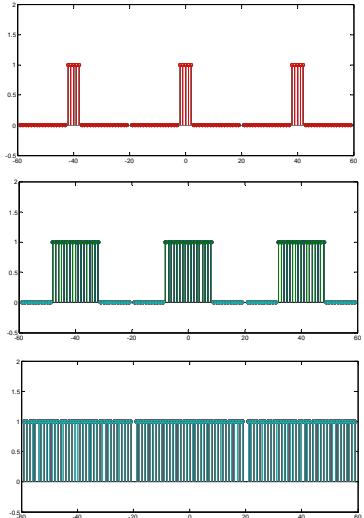
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NTUEE-SS3-FS-71

### Example 3.12:



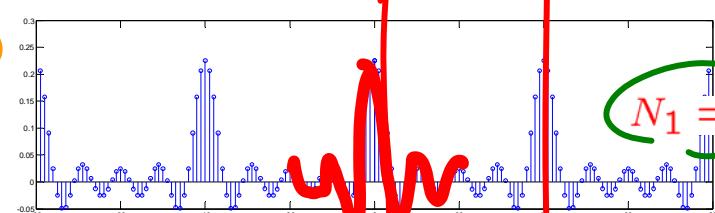
$\bullet N = 40$

$$a_k = \frac{1}{N} \sin \left[ \left( \frac{2\pi}{N} \right) k \left( N_1 + \frac{1}{2} \right) \right] / \sin \left[ \left( \frac{\pi}{N} \right) k \right]$$



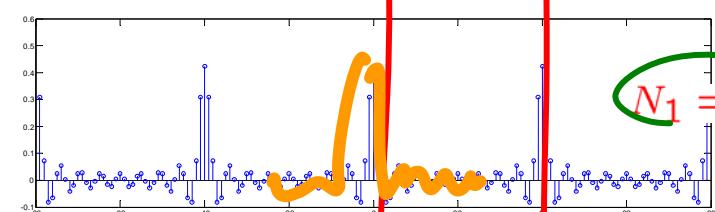
$N_1 = 2$

5



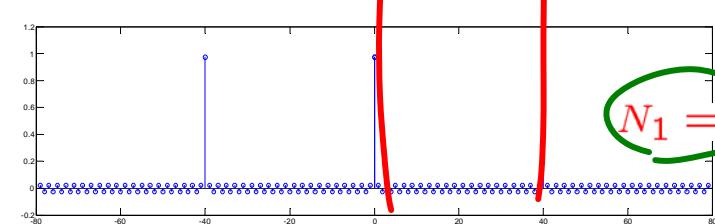
$N_1 = 4$

9



$N_1 = 8$

16



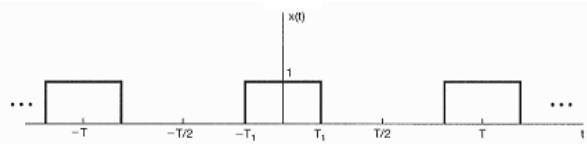
$N_1 = 19$

38

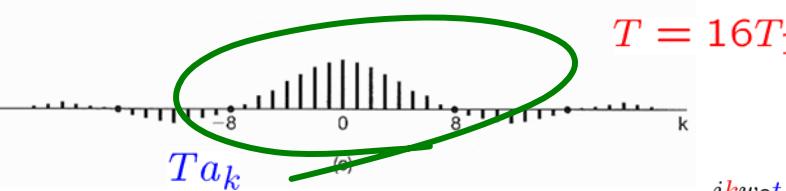
## Fourier Series Representation of CT Periodic Signals

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### Examples 3.5 (CT) & 3.12 (DT):



$$T a_k = T \frac{\sin(k\pi/8)}{k\pi}$$



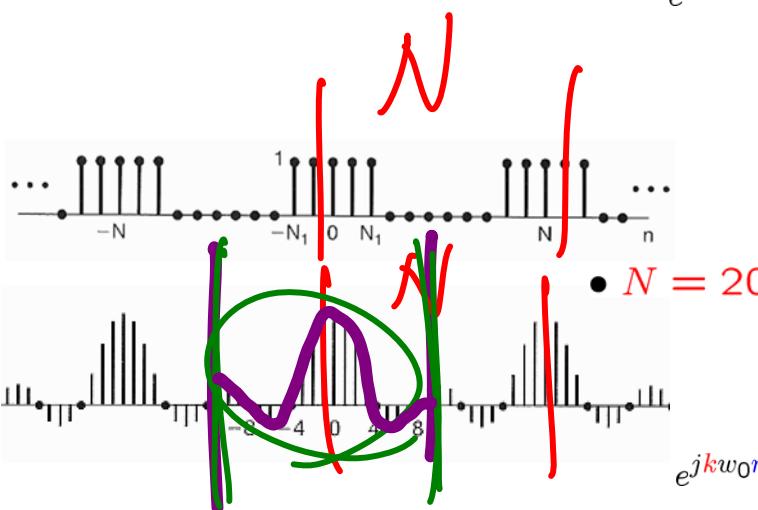
$T = 16T_1$

$T a_k$

$e^{jk\omega_0 t}$

$$a_k = \frac{1}{N} \sin \left[ \left( \frac{2\pi}{N} \right) k \left( N_1 + \frac{1}{2} \right) \right] / \sin \left[ \left( \frac{\pi}{N} \right) k \right]$$

$$a_k = \frac{2N_1 + 1}{N}$$



$N = 20$

$e^{jk\omega_0 n}$

- Partial Sum:

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

 $N = 9$  $M = 1$ 

3

2M+1

- If  $N$  is odd

$$\hat{x}[n] = \sum_{k=-M}^{M} a_k e^{jk(\frac{2\pi}{N})n}$$

 $M = 2$ 

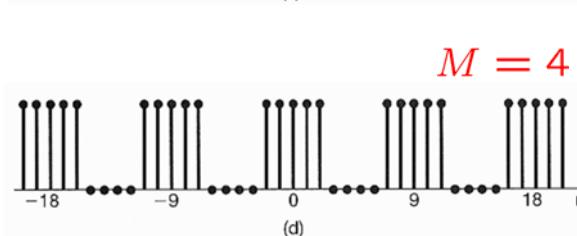
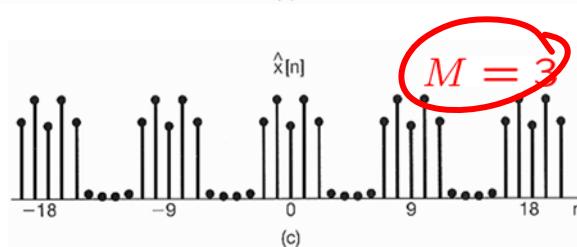
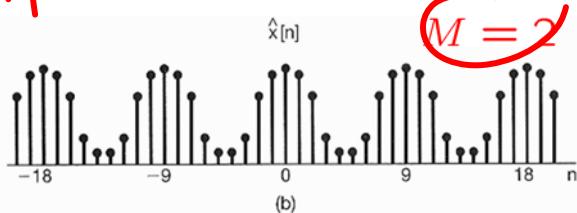
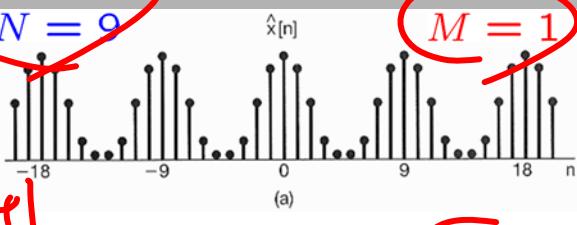
6

- If  $N$  is even

$$\hat{x}[n] = \sum_{k=-M+1}^{M} a_k e^{jk(\frac{2\pi}{N})n}$$

 $M = 3$ 

7

3/19/12  
10:22 am

## Outline

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
- Fourier Series Representation of Discrete-Time Periodic Signals
- Properties of Discrete-Time Fourier Series
- Fourier Series & LTI Systems
- Filtering & Examples of CT & DT Filters

Section	Property
	Linearity
	Time Shifting
	Frequency Shifting
	Conjugation
	Time Reversal
	Time Scaling
	Periodic Convolution
3.7.1	Multiplication ✓
3.7.2	First Difference ✓
	Running Sum
	Conjugate Symmetry for Real Signals
	Symmetry for Real and Even Signals
	Symmetry for Real and Odd Signals
	Even-Odd Decomposition for Real Signals
3.7.3	Parseval's Relation for Periodic Signals ✓

## Properties of DT Fourier Series

TABLE 3.2 PROPERTIES OF DISCRETE-TIME FOURIER SERIES

Property	Periodic Signal	Fourier Series Coefficients
	$x[n]$ } Periodic with period $N$ and $y[n]$ } fundamental frequency $\omega_0 = 2\pi/N$	$a_k$ } Periodic with $b_k$ } period $N$
Linearity	$Ax[n] + By[n]$	$Aa_k + Bb_k$
Time Shifting	$x[n - n_0]$	$a_k e^{-j k(2\pi/N)n_0}$
Frequency Shifting	$e^{jM(2\pi/N)n} x[n]$	$a_{k-M}$
Conjugation	$x^*[n]$	$a_k^*$
Time Reversal	$x[-n]$	$a_{-k}$
Time Scaling	$x_{(m)}[n] = \begin{cases} x[n/m], & \text{if } n \text{ is a multiple of } m \\ 0, & \text{if } n \text{ is not a multiple of } m \end{cases}$ (periodic with period $mN$ )	$\frac{1}{m} a_k$ (viewed as periodic) (with period $mN$ )
Periodic Convolution	$\sum_{r=0}^{N-1} x[r]y[n-r]$	$Na_k b_k$
Multiplication	$x[n]y[n]$	$\sum_{l=0}^{N-1} a_l b_{k-l}$
First Difference	$x[n] - x[n-1]$	$(1 - e^{-jk(2\pi/N)})a_k$
Running Sum	$\sum_{k=-\infty}^n x[k]$ (finite valued and periodic only)	$\left( \frac{1}{(1 - e^{-jk(2\pi/N)})} \right) a_k$
Conjugate Symmetry for Real Signals	$x[n]$ real	$\begin{cases} a_k = a_{-k}^* \\ \Re\{a_k\} = \Re\{a_{-k}\} \\ \Im\{a_k\} = -\Im\{a_{-k}\} \\  a_k  =  a_{-k}  \\ \angle a_k = -\angle a_{-k} \end{cases}$
Real and Even Signals	$x[n]$ real and even	$a_k$ real and even
Real and Odd Signals	$x[n]$ real and odd	$a_k$ purely imaginary and odd
Even-Odd Decomposition of Real Signals	$\begin{cases} x_e[n] = \Re\{x[n]\} & [x[n] \text{ real}] \\ x_o[n] = \Im\{x[n]\} & [x[n] \text{ real}] \end{cases}$	$\begin{cases} \Re\{a_k\} \\ j\Im\{a_k\} \end{cases}$
Parseval's Relation for Periodic Signals		
$\frac{1}{N} \sum_{n=0}^{N-1}  x[n] ^2 = \sum_{k=-\infty}^{\infty}  a_k ^2$		

## In Summary:

- The synthesis equation:

$$x[n] = \sum_{k=-N}^{N-1} a_k e^{jk\omega_0 n} = \sum_{k=-N}^{N-1} a_k e^{jk\left(\frac{2\pi}{N}\right)n}$$

$x[n+N] \in X[n]$

- The analysis equation:

$$a_k = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jk\omega_0 n} = \frac{1}{N} \sum_{n=-N}^{N-1} x[n] e^{-jk\left(\frac{2\pi}{N}\right)n}$$

$a_k = a_k + N$

- $x[n] \xleftrightarrow{\mathcal{FS}} a_k$ : DT Fourier series pair

## Properties of DT Fourier Series

### Linearity:

- $x[n], y[n]$ : periodic signals with period  $N$

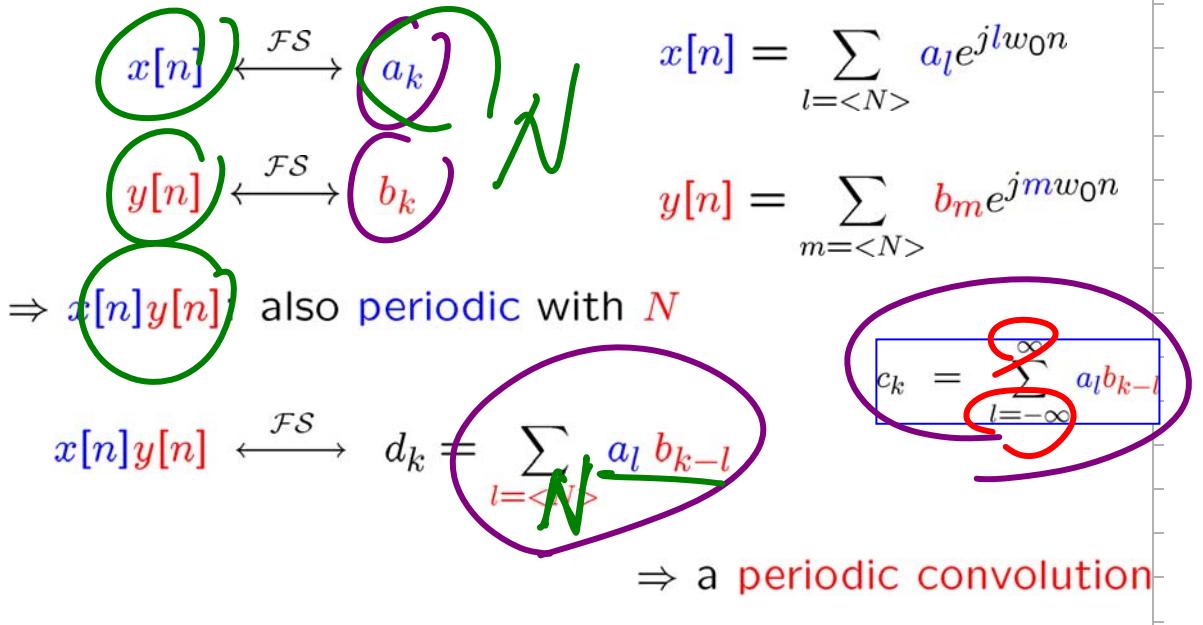
$$\begin{aligned} x[n] &\xleftrightarrow{\mathcal{FS}} a_k \\ y[n] &\xleftrightarrow{\mathcal{FS}} b_k \\ \Rightarrow z[n] = Ax[n] + By[n] &\xleftrightarrow{\mathcal{FS}} c_k = Aa_k + Bb_k \end{aligned}$$

### Time Shifting:

$$\begin{aligned} x[n] &\xleftrightarrow{\mathcal{FS}} a_k \\ \Rightarrow x[n - n_0] &\xleftrightarrow{\mathcal{FS}} e^{-jk\omega_0 n_0} a_k = e^{-jk\left(\frac{2\pi}{N}\right)n_0} a_k \end{aligned}$$

## ■ Multiplication:

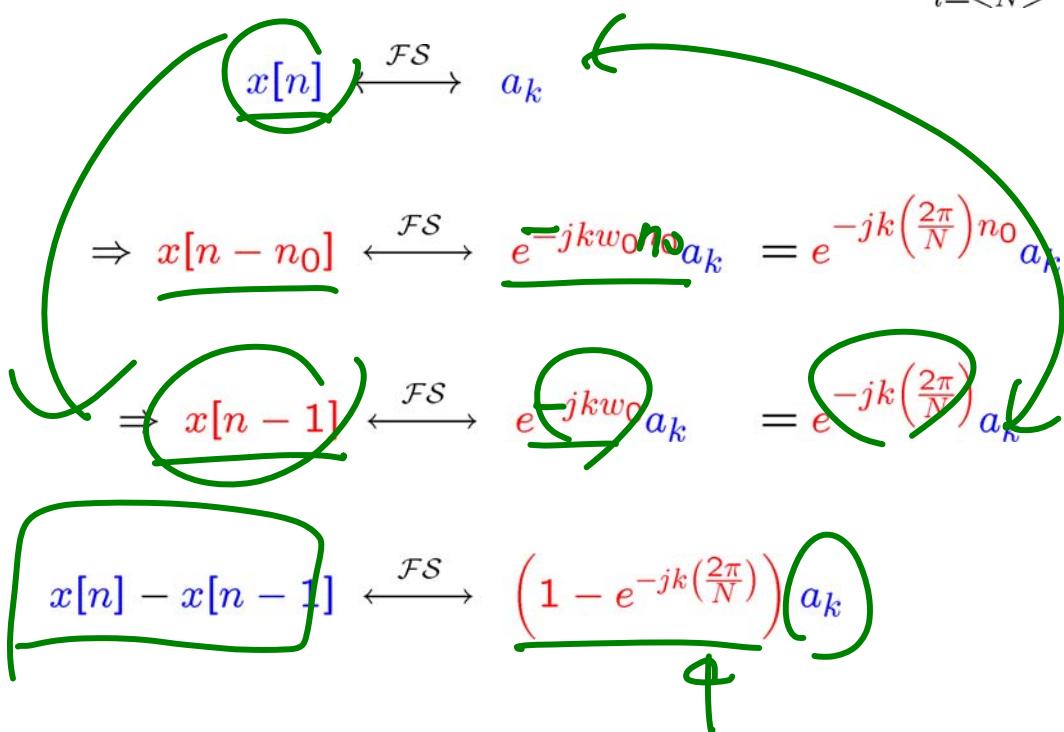
- $x[n], y[n]$ : periodic signals with period  $N$



Add

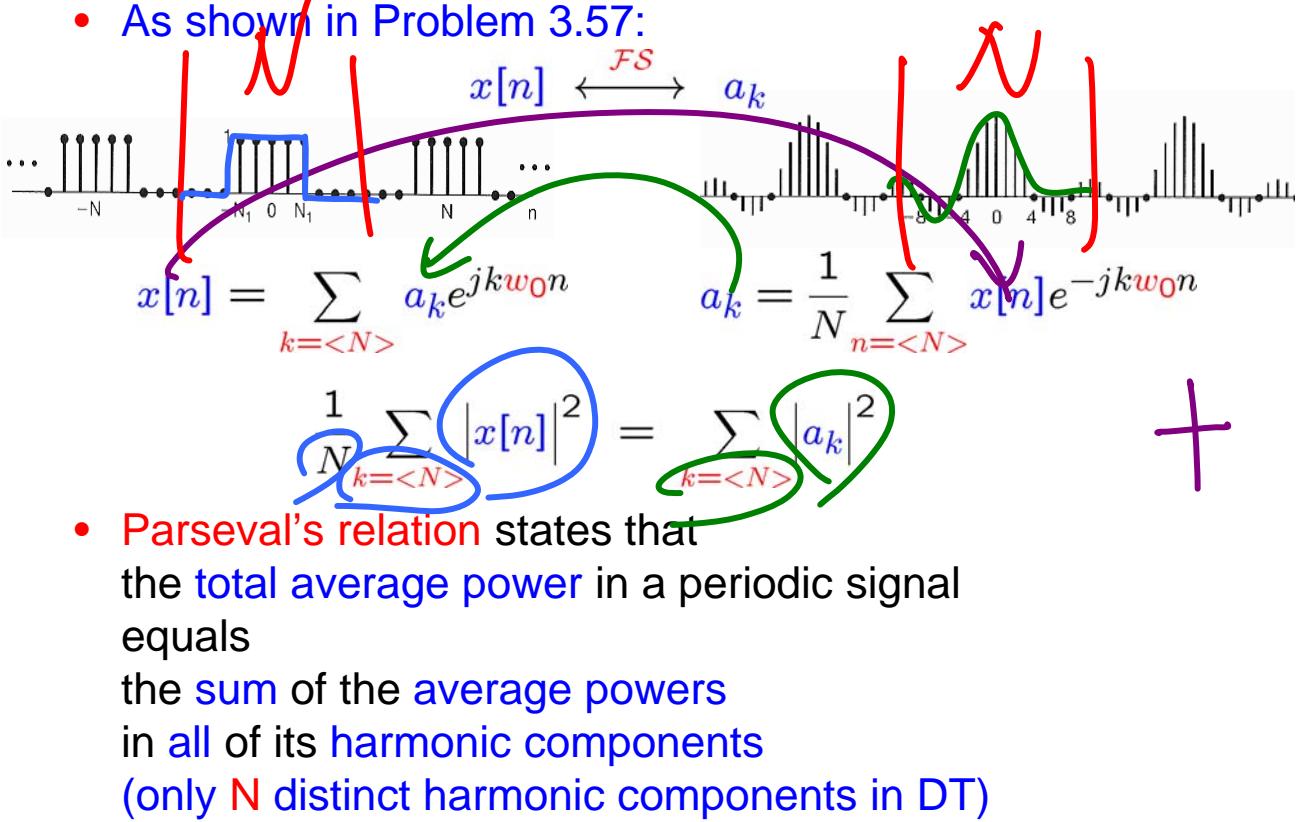
## ■ First Difference:

$$x[n] = \sum_{l=-N}^{N} a_k e^{jk w_0 n}$$

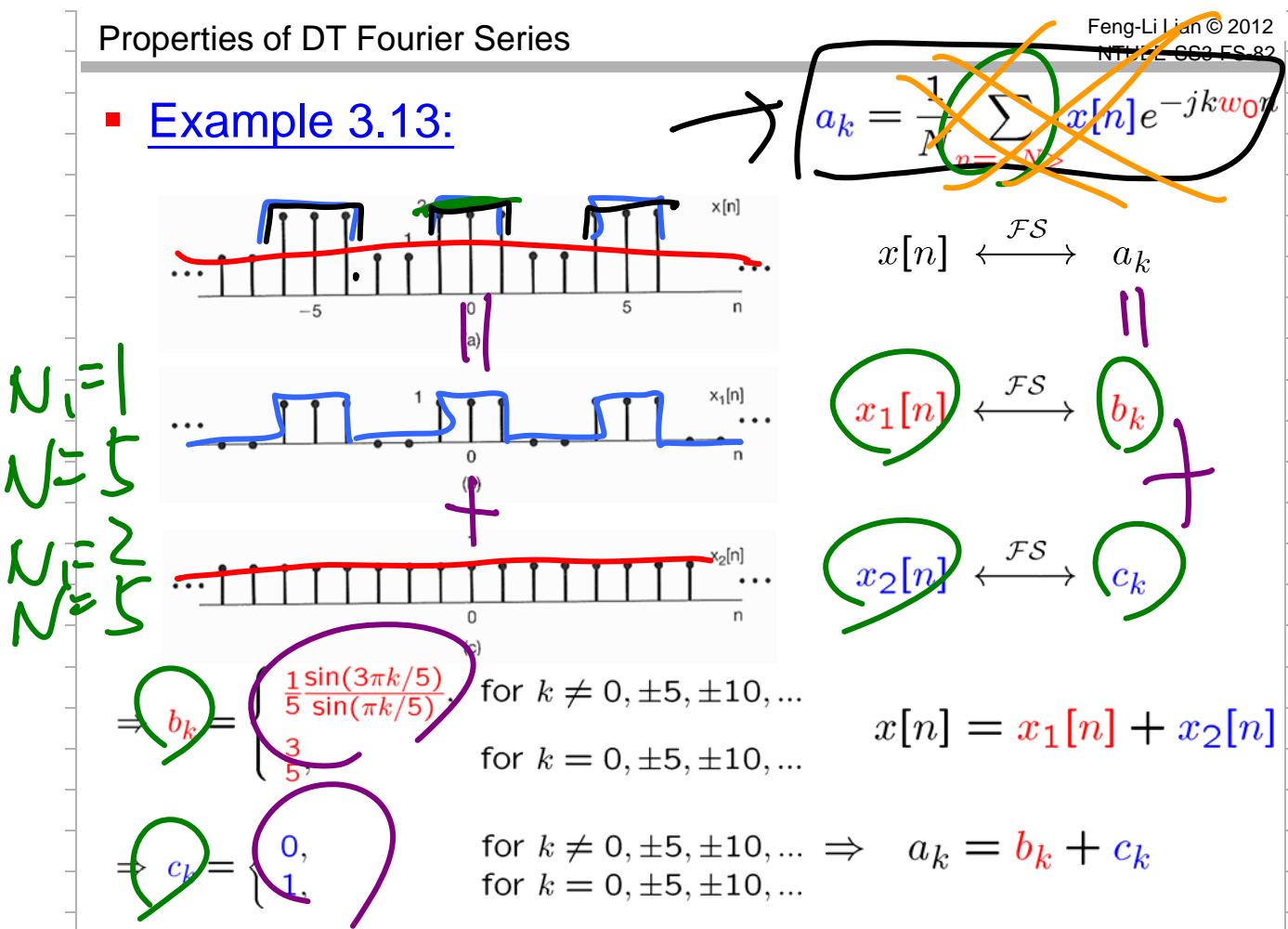


- Parseval's relation for DT periodic signals:

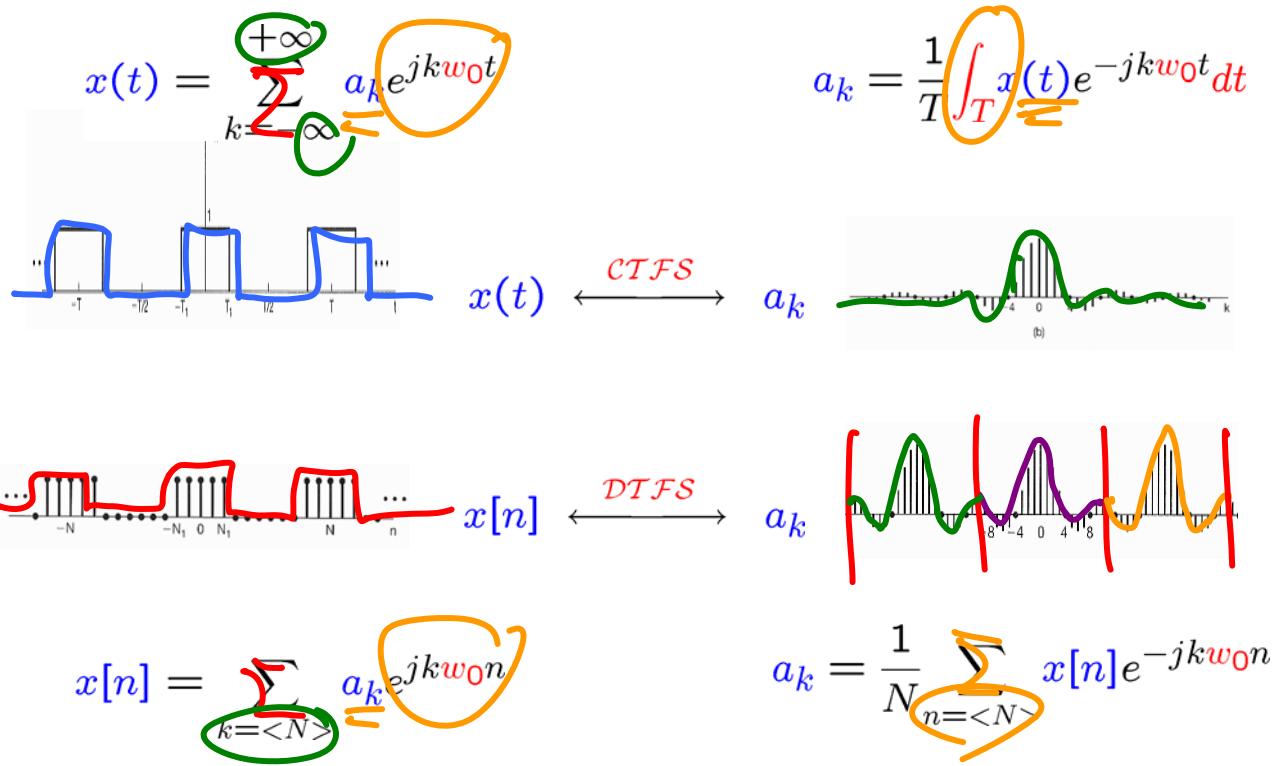
- As shown in Problem 3.57:



- Example 3.13:



- CT & DT Fourier Series Representation:



## Outline

- A Historical Perspective
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- **Fourier Series & LTI Systems**
- Filtering & Examples of CT & DT Filters

## The Response of an LTI System:

$$\text{in} \rightarrow \text{LTI} \rightarrow \text{out}$$

$$\delta(t) \quad h(t)$$

$$\left\{ \begin{array}{l} \text{CT: } e^{st} \rightarrow H(s)e^{st} \\ \text{DT: } z^n \rightarrow H(z)z^n \end{array} \right.$$

$$H(s) = \int_{-\infty}^{+\infty} h(t) e^{-st} dt$$

$$H(z) = \sum_{k=-\infty}^{+\infty} h[k] z^{-k}$$

⇒ the impulse response

⇒ the system functions

- If  $s = jw$  or  $z = e^{jw}$ :

$$H(jw) = \int_{-\infty}^{+\infty} h(t) e^{-jwt} dt$$

$$H(e^{jw}) = \sum_{n=-\infty}^{+\infty} h[n] e^{-jwn}$$

$w = kw_0$

$w = kw_0$

$w = kw_0$

⇒ the frequency response

## In Summary:

$$a = |a|e^{j\vartheta_a}$$

$$H = |H|e^{j\vartheta_H}$$

$$\text{in} \rightarrow \text{LTI} \rightarrow \text{out}$$

$$H(s/z/w)$$

$$(s_i = jw_i \text{ or } z_i = e^{jw_i})$$

$$\left\{ \begin{array}{l} \text{CT: } e^{s_i t} \rightarrow H(s_i)e^{s_i t} \\ \text{DT: } z_i^n \rightarrow H(z_i)z_i^n \end{array} \right.$$

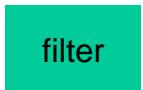
$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jkw_0 t}$$

$$\rightarrow y(t) = \sum_{k=-\infty}^{+\infty} a_k H(jkw_0) e^{jkw_0 t}$$

$$x[n] = \sum_{k=-N}^{N} a_k e^{jk(\frac{2\pi}{N})n}$$

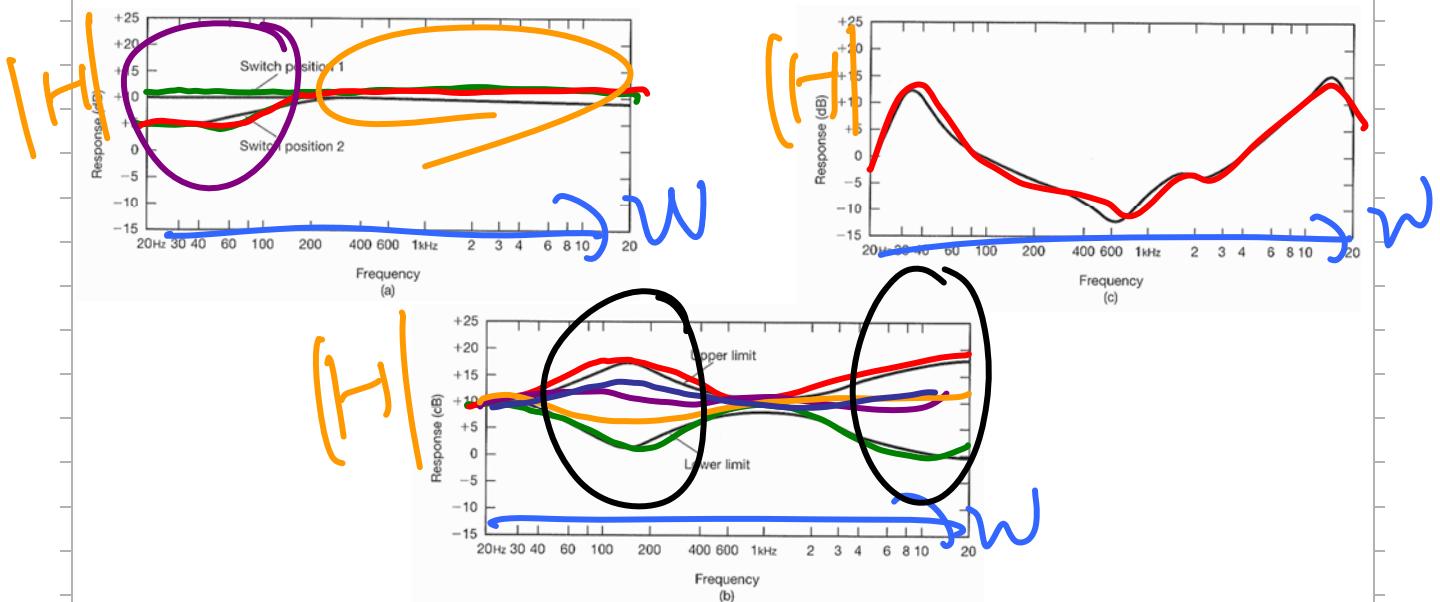
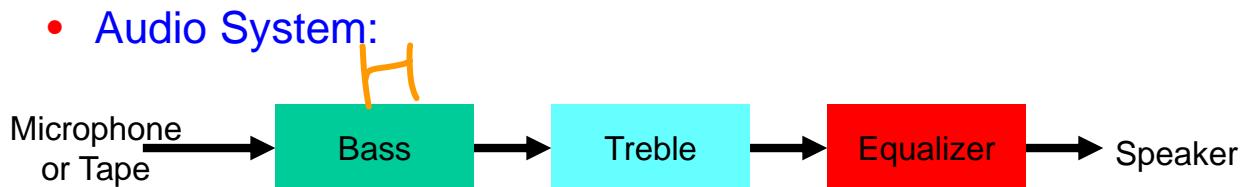
$$\rightarrow y[n] = \sum_{k=-N}^{N} a_k H(e^{j(\frac{2\pi}{N})k}) e^{jk(\frac{2\pi}{N})n}$$

- A Historical Perspective
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- **Filtering & Examples of CT & DT Filters**

- **Filtering:**  $in \rightarrow$    $\rightarrow out$
- Change the relative amplitudes of the frequency components in a signal,
  - Frequency-shaping filters
- OR, significantly attenuate or eliminate some frequency components entirely
  - Frequency-selective filters

- Frequency-Shaping Filters:

- Audio System:

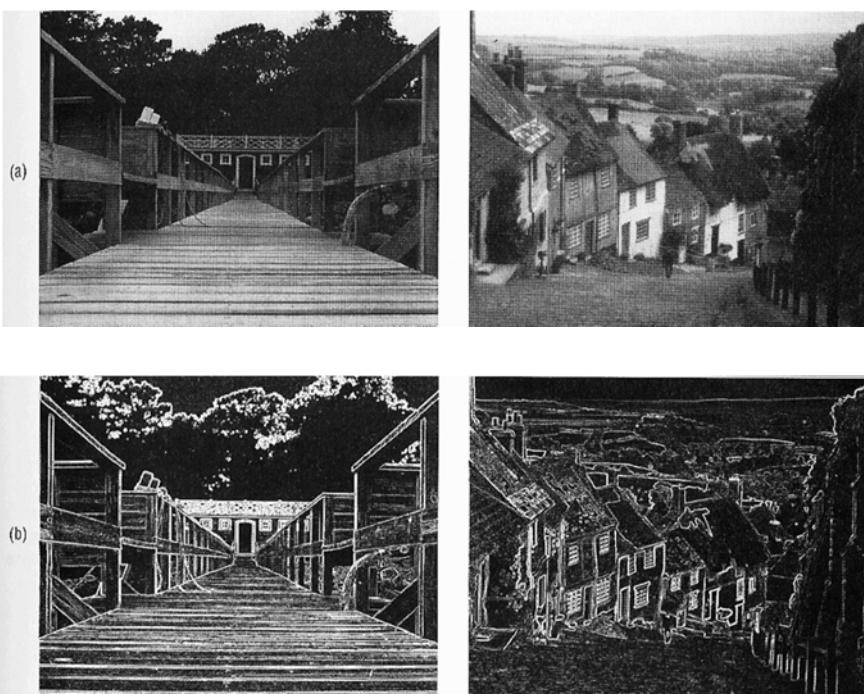


- Frequency-Shaping Filters:

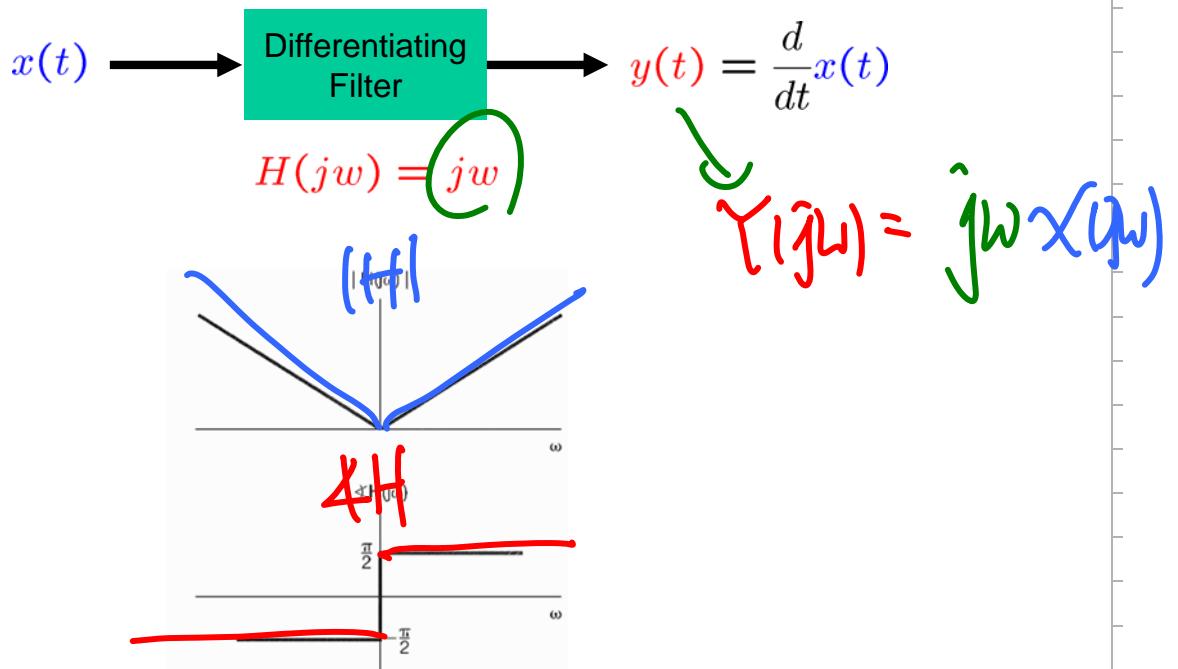
- Differentiating filter on enhancing edges:

$$x(t) \rightarrow \text{Differentiating Filter} \rightarrow y(t) = \frac{d}{dt}x(t)$$

$$H(jw) = jw$$



- Frequency-Shaping Filters:
  - Differentiating filter:



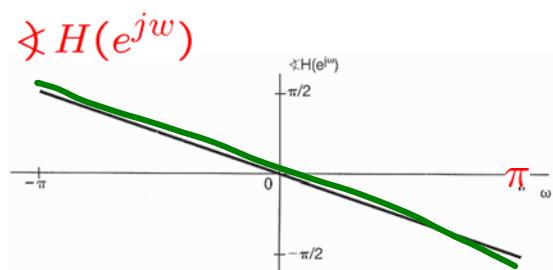
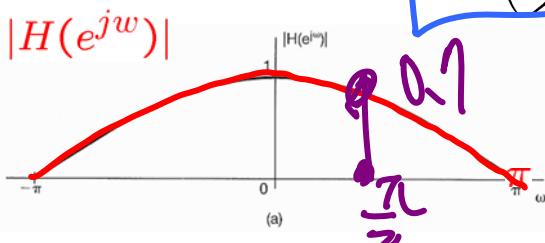
- Frequency-Shaping Filters:

$$1 \pm e^{-j\theta} = e^{-j\theta/2} (e^{j\theta/2} \pm e^{-j\theta/2})$$

- A simple DT filter: Two-point average

$$y[n] = \frac{1}{2} (x[n] + x[n-1]) = \frac{1}{2} (1 + e^{-jw}) x[n] = H(e^{jw}) x[n]$$

$$\begin{aligned} \Rightarrow H(e^{jw}) &= \frac{1}{2} [1 + e^{-jw}] = \frac{1}{2} e^{-j(\frac{w}{2})} [e^{j(\frac{w}{2})} + e^{-j(\frac{w}{2})}] \\ &= e^{-j(\frac{w}{2})} \cos\left(\frac{w}{2}\right) \end{aligned}$$

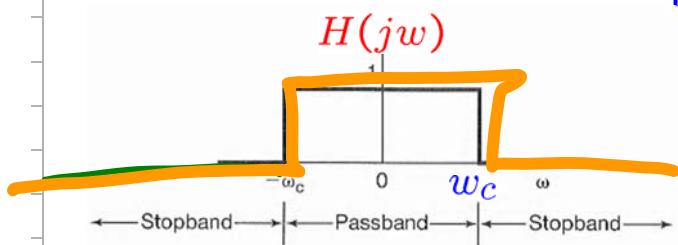


$$\text{if } x[n] = K e^{j(\frac{\pi}{2}) \cdot n} \quad w = \frac{\pi}{2}$$

$$\text{then } y[n] = H\left(e^{j(\frac{\pi}{2})}\right) K e^{j(\frac{\pi}{2}) \cdot n}$$

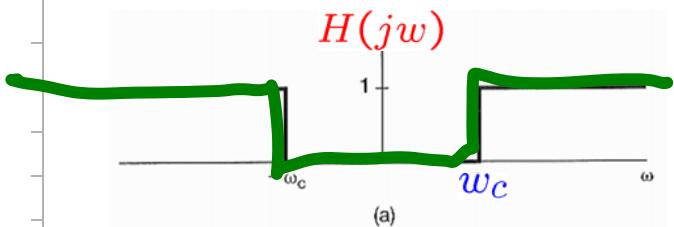
- Frequency-Selective Filters:

- Select some bands of frequencies and reject others



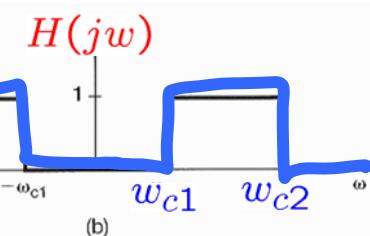
CT ideal lowpass filter

$$H(jw) = \begin{cases} 1, & |w| \leq w_c \\ 0, & |w| > w_c \end{cases}$$



CT ideal highpass filter

$$H(jw) = \begin{cases} 0, & |w| < w_c \\ 1, & |w| \geq w_c \end{cases}$$

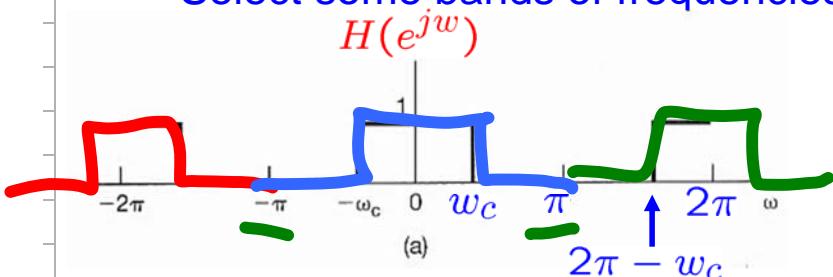


CT ideal bandpass filter

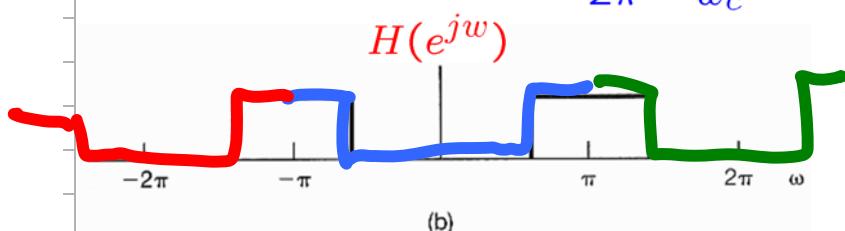
$$H(jw) = \begin{cases} 1, & w_{c1} \leq |w| \leq w_{c2} \\ 0, & \text{otherwise} \end{cases}$$

- Frequency-Selective Filters:

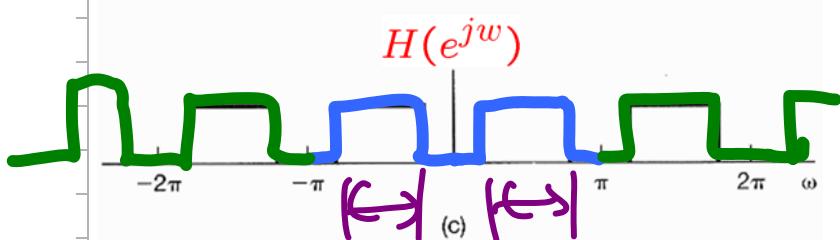
- Select some bands of frequencies and reject others



DT ideal lowpass filter



DT ideal highpass filter



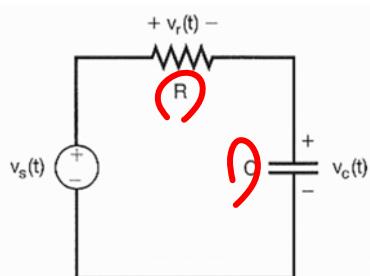
DT ideal bandpass filter

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- Fourier Series Representation of Continuous-Time Periodic Signals
- Convergence of the Fourier Series
- Properties of Continuous-Time Fourier Series
- Fourier Series Representation of Discrete-Time Periodic Signals
- Properties of Discrete-Time Fourier Series
- Fourier Series & LTI Systems
- Filtering & Examples of CT & DT Filters

## CT Filters by Differential Equations

■ A Simple RC Lowpass Filter:

Input signal:  
 $v_s(t) = e^{j\omega t}$   
 $\delta(t)$   
 $u(t)$



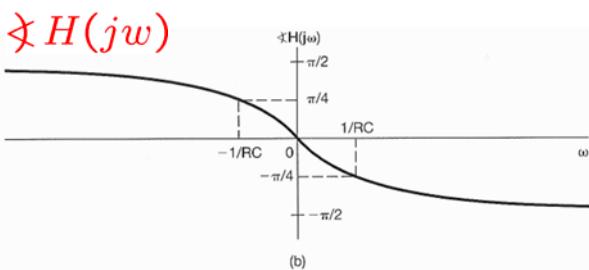
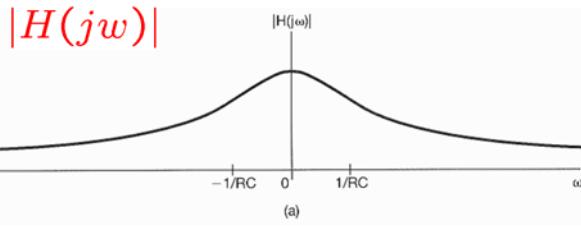
Output signal:  
 $v_c(t) = H(j\omega)e^{j\omega t}$   
 $h(t)$   
 $s(t)$

$$\begin{aligned}
 \mathcal{J} \rightarrow & [RC \frac{d}{dt} v_c(t) + v_c(t) = v_s(t)] \\
 \Rightarrow & RC \frac{d}{dt} [H(j\omega)e^{j\omega t}] + H(j\omega)e^{j\omega t} = e^{j\omega t} \\
 \Rightarrow & RC(j\omega H(j\omega)e^{j\omega t}) + H(j\omega)e^{j\omega t} = e^{j\omega t} \\
 \Rightarrow & H(j\omega)e^{j\omega t} = \frac{1}{1 + RCj\omega}e^{j\omega t}
 \end{aligned}$$

- A Simple RC Lowpass Filter:  $H(jw) = \int_{-\infty}^{+\infty} h(t) e^{-jwt} dt$

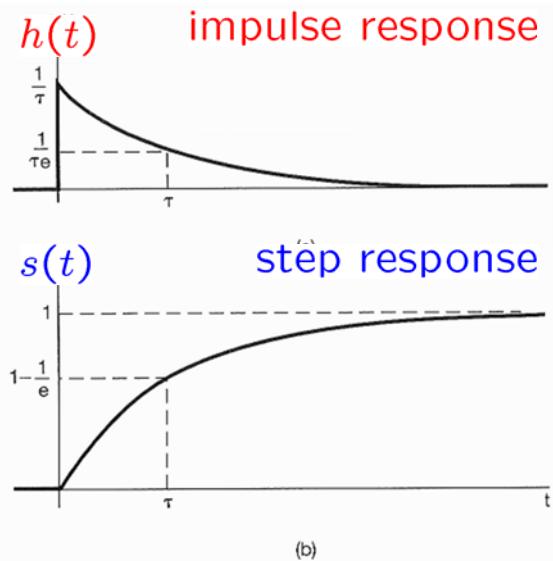
$$\Rightarrow H(jw) = \frac{1}{1 + RCjw}$$

$$H = |H| e^{j\varphi_H}$$



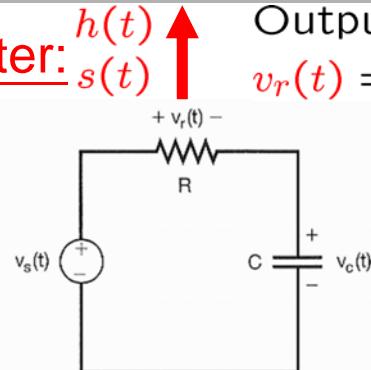
$$\Rightarrow h(t) = \frac{1}{RC} e^{-t/RC} u(t)$$

$$\Rightarrow s(t) = [1 - e^{-t/RC}] u(t)$$



- A Simple RC Highpass Filter:  $\frac{h(t)}{s(t)} = \frac{v_r(t)}{v_s(t)}$  Output signal:  $v_r(t) = G(jw)e^{jwt}$

Input signal:  
 $v_s(t) = e^{jwt}$   
 $\delta(t)$   
 $u(t)$



$$\Rightarrow RC \frac{d}{dt} v_r(t) + v_r(t) = RC \frac{d}{dt} v_s(t)$$

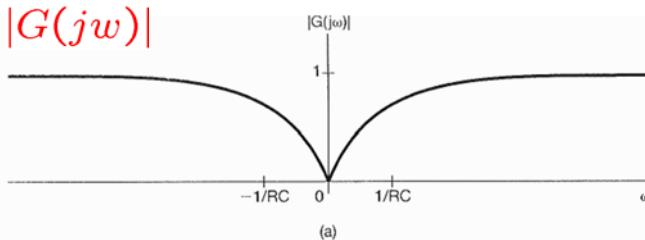
$$\Rightarrow RC \frac{d}{dt} [G(jw)e^{jwt}] + G(jw)e^{jwt} = RC \frac{d}{dt} e^{jwt}$$

$$\Rightarrow RC \cancel{jw} G(jw)e^{jwt} + G(jw)e^{jwt} = RC \cancel{jw} e^{jwt}$$

$$\Rightarrow \underline{\underline{G(jw)e^{jwt}}} = \frac{jw RC}{1 + jw RC} e^{jwt}$$

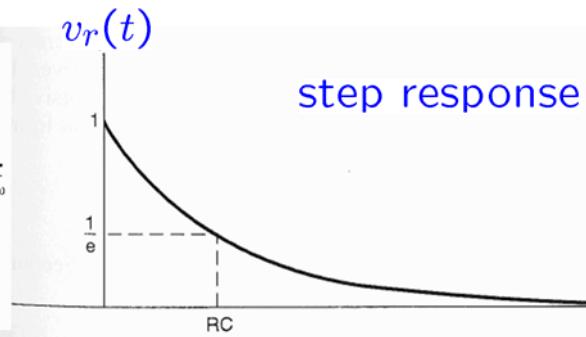
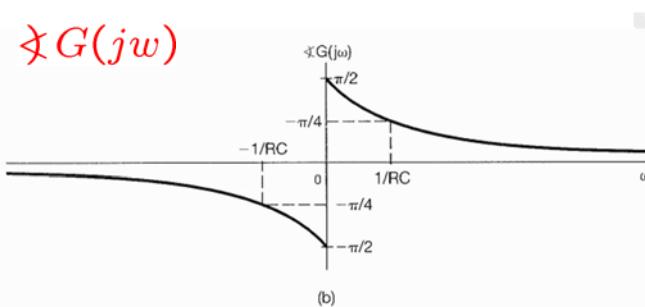
■ A Simple RC Highpass Filter:

$$\Rightarrow G(jw) = \frac{jw RC}{1 + jw RC}$$



$$v_r(t) = v_s(t) - v_c(t)$$

$$\Rightarrow v_r(t) = e^{-t/RC} u(t)$$



■ First-Order Recursive DT Filters:

$$y[n] - ay[n-1] = x[n]$$

- If  $x[n] = e^{jwn}$ , then  $y[n] = H(e^{jw})e^{jwn}$

where  $H(e^{jw})$ : the frequency response

$$\Rightarrow \underbrace{H(e^{jw}) e^{jwn}}_{\text{frequency response}} - a \underbrace{H(e^{jw}) e^{jw(n-1)}}_{\text{previous output}} = e^{jwn}$$

$$\Rightarrow [1 - a e^{-jw}] H(e^{jw}) e^{jwn} = e^{jwn}$$

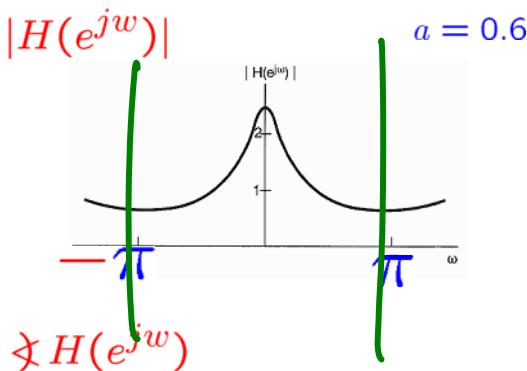
$$\Rightarrow \underbrace{H(e^{jw})}_{\text{frequency response}} = \frac{1}{1 - a e^{-jw}}$$

■ First-Order Recursive DT Filters:

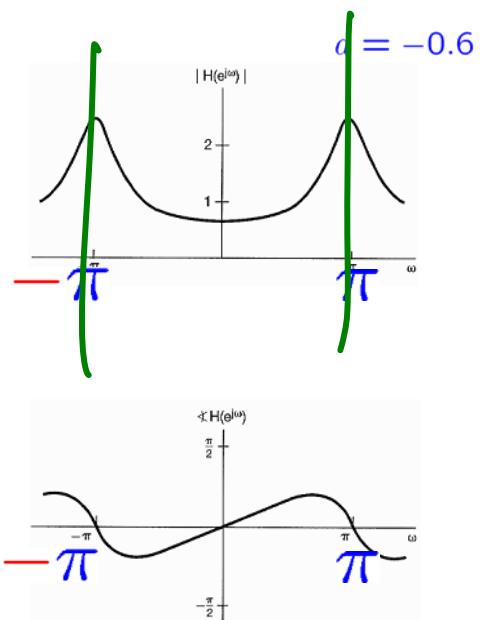
$$H(e^{j\omega}) = \frac{1}{1 - a e^{-j\omega}}$$

$$y[n] = ay[n-1] + x[n]$$

lowpass filter:  $0 < a < 1$



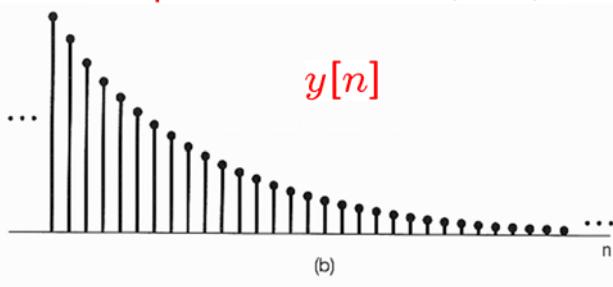
highpass filter:  $-1 < a < 0$



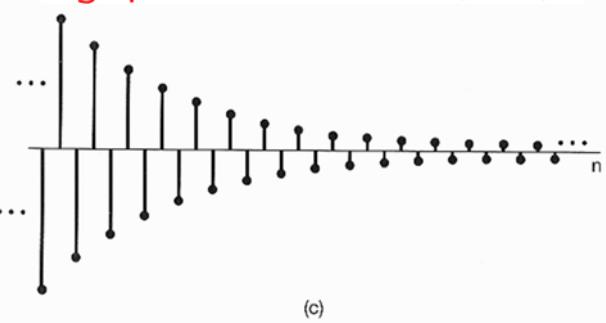
■ First-Order Recursive DT Filters:

$$y[n] = ay[n-1] + x[n]$$

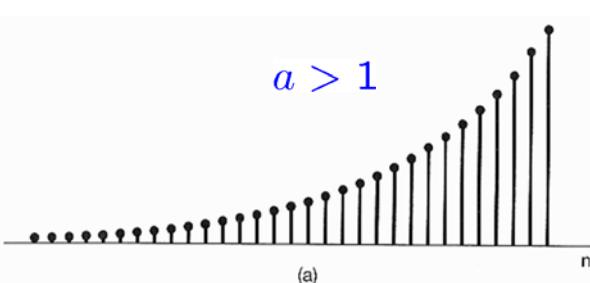
lowpass filter:  $0 < a < 1$



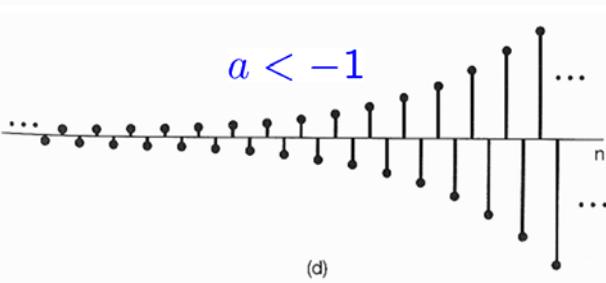
highpass filter:  $-1 < a < 0$



$a > 1$



$a < -1$



- Nonrecursive DT Filters:

- An FIR nonrecursive difference equation:

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

$N+M+1$

$$= b_{-N} x[n + N] + b_{-N+1} x[n + N - 1] + \dots$$

$+ b_{-1} x[n+1]$

$$+ b_0 x[n] + b_1 x[n - 1] + \dots + b_M x[n - M]$$

$$b_k = \frac{1}{N+M+1}$$

$$b_k = \frac{\sin \omega_k}{\pi k}$$

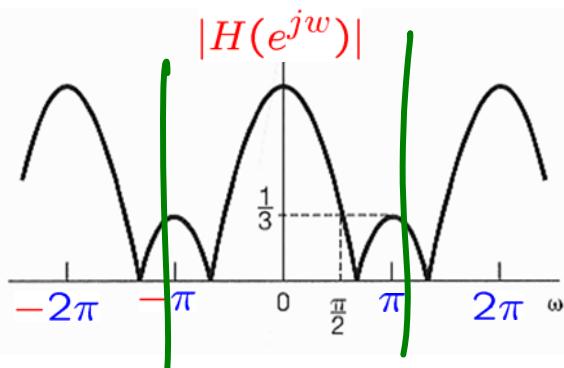
- Nonrecursive DT Filters:

- Three-point moving average (lowpass) filter:

$$y[n] = \frac{1}{3} (x[n + 1] + x[n] + x[n - 1])$$

$$\Rightarrow h[n] = \frac{1}{3} (\delta[n + 1] + \delta[n] + \delta[n - 1])$$

$$\Rightarrow H(e^{jw}) = \frac{1}{3} (e^{jw} + 1 + e^{-jw}) = \frac{1}{3} (1 + 2 \cos w)$$



- Nonrecursive DT Filters:

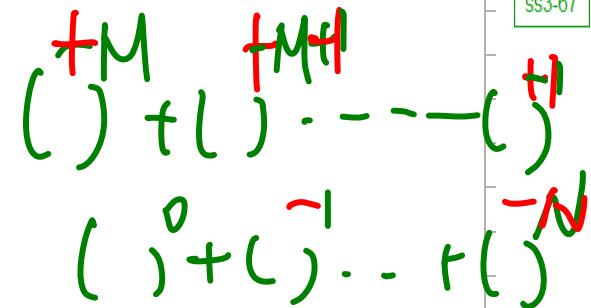
- N+M+1 moving average (lowpass) filter:

$$y[n] = \frac{1}{N + M + 1} \sum_{k=-N}^M x[n - k]$$

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

ss3-67

$$\Rightarrow H(e^{jw}) = \frac{1}{N + M + 1} \sum_{k=-N}^M e^{-jkw}$$

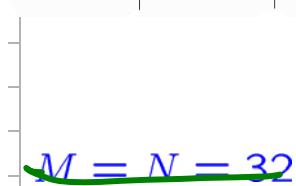
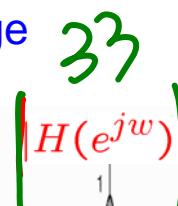
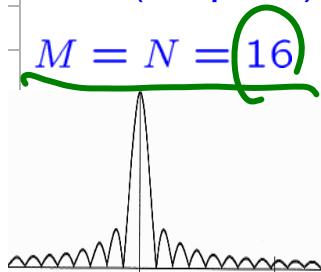


$$\Rightarrow H(e^{jw}) = \frac{1}{N + M + 1} e^{jw\left(\frac{N-M}{2}\right)} \frac{\sin((M+N+1)\frac{w}{2})}{\sin(\frac{w}{2})}$$

$$\frac{1 - e^{-ja}}{1 - e^{-jb}} = \frac{e^{-ja/2}(e^{ja/2} - e^{-ja/2})}{e^{-jb/2}(e^{jb/2} - e^{-jb/2})}$$

- Nonrecursive DT Filters:

- N+M+1 moving average (lowpass) filter:



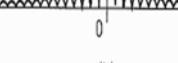
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(a)

(b)

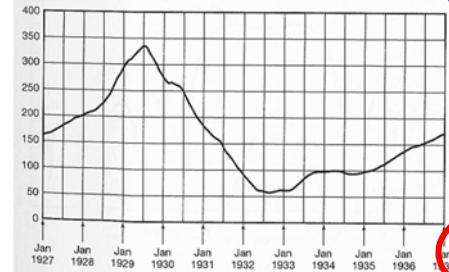
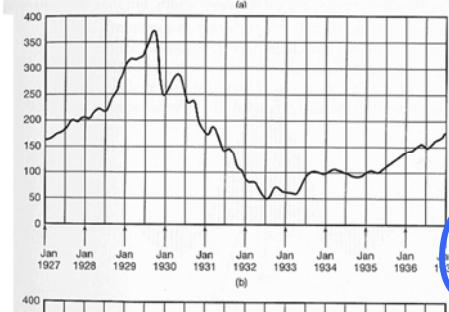
65



(a)

(b)

- Lowpass Filtering on Dow Jones Weekly Stock Market Index:



- Nonrecursive DT Filters:

- Highpass filters:

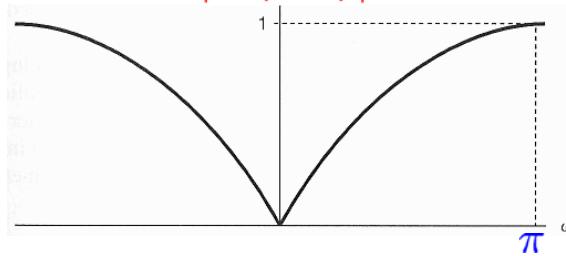
$$y[n] = \frac{x[n] - x[n-1]}{2}$$

$$\Rightarrow h[n] = \frac{1}{2} \{ \delta[n] - \delta[n-1] \}$$

$$\begin{aligned} \Rightarrow H(e^{jw}) &= \frac{1}{2} [1 - e^{-jw}] = \frac{1}{2} e^{-j(\frac{w}{2})} \left[ e^{j(\frac{w}{2})} - e^{-j(\frac{w}{2})} \right] \\ &= j e^{-j(\frac{w}{2})} \sin\left(\frac{w}{2}\right) \end{aligned}$$

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$$|H(e^{jw})|$$



- On page 235, Eq. 3.139

$$1 \pm e^{-j\theta} = e^{-j\theta/2} (e^{j\theta/2} \pm e^{-j\theta/2})$$

$$\begin{aligned}\Rightarrow H(e^{jw}) &= \frac{1}{2} [1 + e^{-jw}] = \frac{1}{2} e^{-j(\frac{w}{2})} \left[ e^{j(\frac{w}{2})} + e^{-j(\frac{w}{2})} \right] \\ &= e^{-j(\frac{w}{2})} \cos\left(\frac{w}{2}\right)\end{aligned}$$

- On page 249, Eq. 3.164

$$\begin{aligned}\Rightarrow H(e^{jw}) &= \frac{1}{2} [1 - e^{-jw}] = \frac{1}{2} e^{-j(\frac{w}{2})} \left[ e^{j(\frac{w}{2})} - e^{-j(\frac{w}{2})} \right] \\ &= j e^{-j(\frac{w}{2})} \sin\left(\frac{w}{2}\right)\end{aligned}$$

- A Historical Perspective
- The Response of LTI Systems to Complex Exponentials
- FS Representation of CT Periodic Signals
- Convergence of the FS
- Properties of CT FS
  - Linearity
  - Time Reversal
  - Differentiation
  - Symmetry for Real and Even Signals
  - Even-Odd Decomposition for Real Signals
  - Time Shifting
  - Time Scaling
  - Integration
  - Frequency Shifting
  - Periodic Convolution
  - Conjugate Symmetry for Real Signals
  - Symmetry for Real and Odd Signals
  - Parseval's Relation for Periodic Signals
  - Conjugation
  - Multiplication
  - Running Sum
- FS Representation of DT Periodic Signals
- Properties of DT FS
  - Multiplication
  - First Difference
  - Running Sum
- FS & LTI Systems
- Filtering
  - Frequency-shaping filters & Frequency-selective filters
- Examples of CT & DT Filters

Signals & Systems [\(Chap 1\)](#)LTI & Convolution [\(Chap 2\)](#)Bounded/ConvergentPeriodic**FS**[\(Chap 3\)](#)

CT

DT

Aperiodic**FT**

CT

[\(Chap 4\)](#)

DT

[\(Chap 5\)](#)Unbounded/Non-convergent**LT**

CT

[\(Chap 9\)](#)**zT**

DT

[\(Chap 10\)](#)Time-Frequency [\(Chap 6\)](#)

CT-DT

[\(Chap 7\)](#)Communication [\(Chap 8\)](#)

Control

[\(Chap 11\)](#)