

# **CAN207 Continuous and Discrete Time Signals and Systems**

## **Lecture-7 Continuous-Time Fourier Series**

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Room EE322

# Content

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- 1. Fundamentals of Fourier Analysis
  - Time and frequency analysis
  - Eigen-functions and eigen-values
- 2. CTFS for periodic signals
  - synthesis and analysis equations of CTFS
  - FS of real signals
  - Existence - Dirichlet condition
  - Signal spectrum

# Fourier Analysis

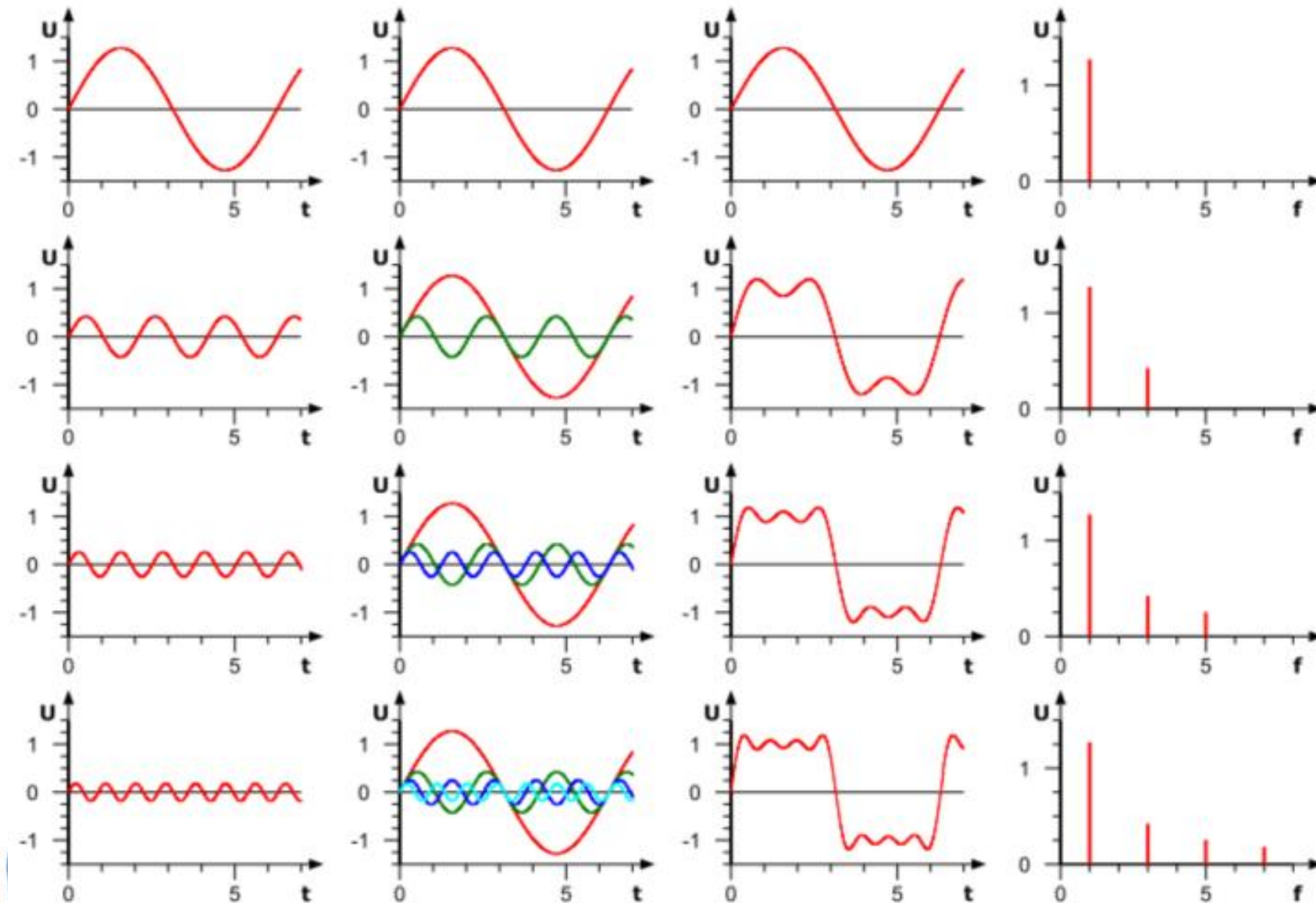


Jean Baptiste Joseph  
Fourier(1768-1830)

First published in 1807



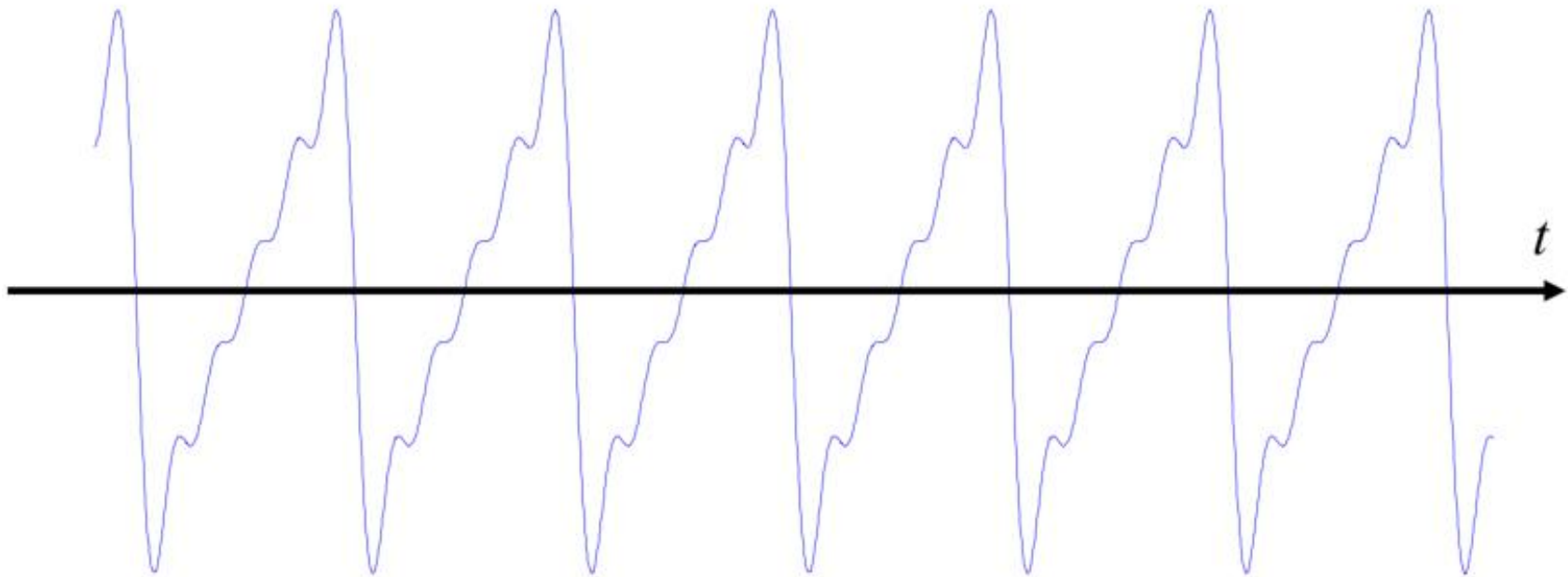
# Synthesis of square wave



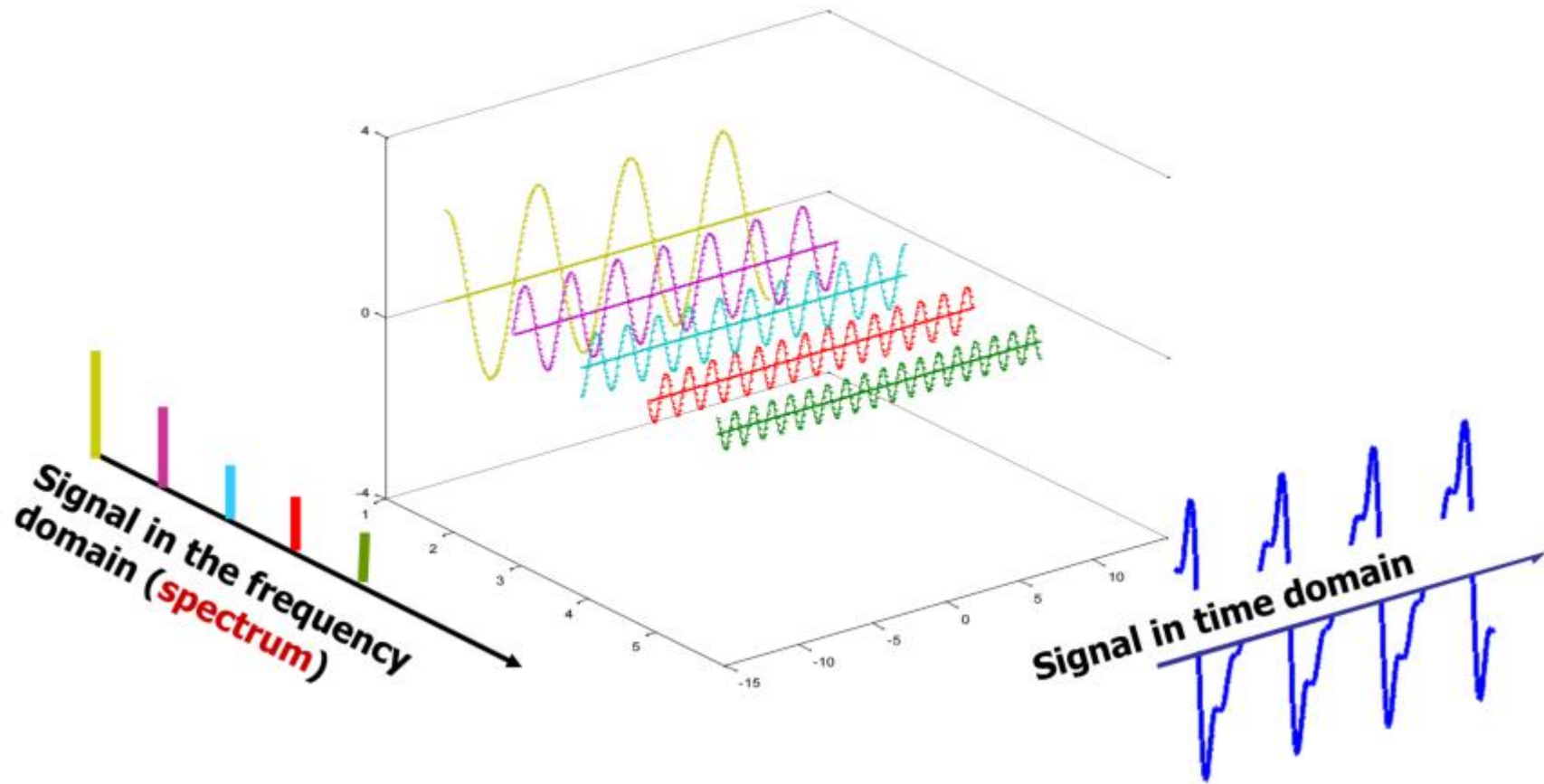
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# *Time and frequency*

- What is the Fourier series of the following periodic signal (“nearly sawtooth” signal)?

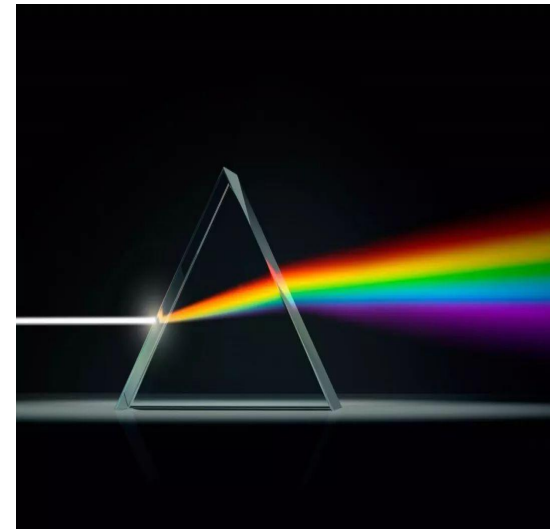
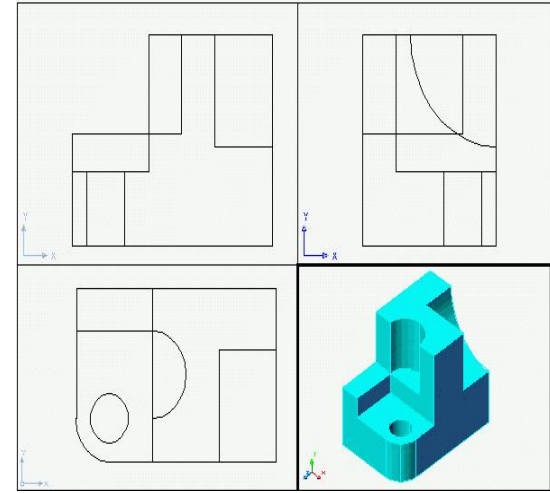


# Relationship between time and frequency



# Relationship between time and frequency

- The Fourier series may be thought of as a tool for looking at a signal from a different perspective.
- Fourier series is also called “Mathematical prism”, cause it can decompose a signal down to fundamental building blocks, like what prism does for lights.



# Recall Lect.6\_p.4

- LTI systems possess the *superposition property*.
  - Input (linearly combined)  $\rightarrow$  Output (linearly combined)
- Strategy:
  - Decompose input signal into a linear combination of *basic signals*;
  - Choose basic signals so that responses are easy to compute.
- Basic signals?

delayed impulses  $\longleftrightarrow$  convolution in Time Domain

complex exponentials  $\longleftrightarrow$  Fourier analyses in Frequency Domain





# 1.1 Basic signals

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- Criteria for choosing a set of *basic signals* in terms of which to decompose the input to a linear system:

$$\text{if: } x = a_1\phi_1 + a_2\phi_2 + \dots$$

$$\text{then: } y = a_1\psi_1 + a_2\psi_2 + \dots$$

- choose the *basic signal*  $\phi_k(t)$  or  $\phi_k[n]$  so that:
  - a broad class of signals can be constructed as a linear combination of  $\phi_k$ s
  - responses to  $\phi_k$ s are easy to compute
- choose *complex exponentials* as a set of *basic signals*:
  - CT:  $\phi_k(t) = e^{s_k t} \Rightarrow \phi_k(t) = e^{j\Omega_k t}$  when  $s_k = j\Omega_k$  is pure imaginary.
    - $s_k$  is complex  $\Rightarrow$  Laplace transform
  - DT:  $\phi_k[n] = z_k^n \Rightarrow \phi_k[n] = e^{j\omega_k n}$  when  $z_k = e^{j\omega_k}$  is on unit circle.
    - $z_k$  is complex  $\Rightarrow$  Z-transform

# 1.2 Eigen-function and eigen-values

- The response of an LTI system to a complex exponential input is the same complex exponential with only a change in amplitude

$$\phi_k(t) = e^{j\Omega_k t} \rightarrow H(\Omega_k)e^{j\Omega_k t}$$

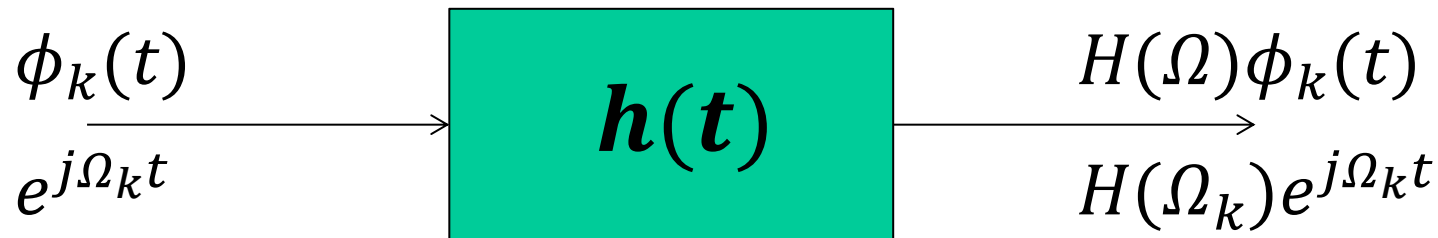
- where the complex amplitude factor  $H(\Omega_k)$  is a function of the frequency  $\Omega_k$ .
- Proof:

$$e^{j\Omega_k t} \rightarrow \int_{-\infty}^{\infty} h(\tau)e^{j\Omega_k(t-\tau)} d\tau = \underbrace{e^{j\Omega_k t}}_{\text{eigen-function}} \underbrace{\int_{-\infty}^{\infty} h(\tau)e^{-j\Omega_k \tau} d\tau}_{H(\Omega_k) \text{ eigen-value}}$$

- A signal for which the system output is a (possibly complex) constant times the input is referred to as an **eigen-function** of the system, and the amplitude factor is referred to as the system's **eigen-value**.

## 1.2 Eigen-function and eigen-values

- An ***eigenfunction*** of a system (or mathematical equation) is a function which you put it through the system, comes out looking exactly the same, except for its change in amplitude.
- The changing amplitude is the ***eigenvalue***.



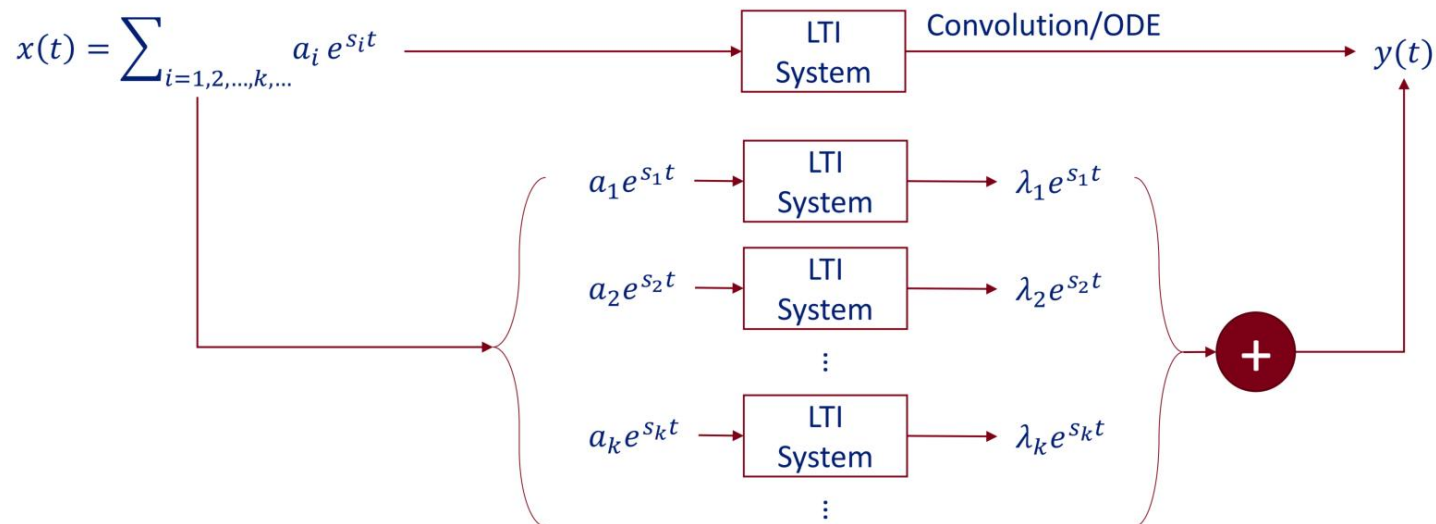
- $e^{j\Omega_k t}$  is a set of convenient building blocks  
    └─ going through the system, get the same  $e^{j\Omega_k t}$  but scaled

# 1.2 Eigen-function and eigen-values

- $e^{j\Omega_k t}$  (or  $e^{st}$ ) is an eigenfunction of LTI systems, such that



- If the input signal of an LTI system can be written as a linear combination of **complex exponential** functions, then:



- the output is also a linear combination of the **weighted complex exponential** functions.



# 1.2 Eigen-function and eigen-values

- Let the input signal  $x(t)$  correspond to a linear combination of several complex exponentials:

$$x(t) = \sum_k a_k e^{j\Omega_k t}$$

- From the eigen-function property, the response to each separately is

$$a_k e^{j\Omega_k t} \rightarrow a_k H(j\Omega_k) e^{j\Omega_k t}$$

- and from the superposition property, the response is:

$$y(t) = \sum_k a_k H(j\Omega_k) e^{j\Omega_k t}$$

- This can be extended to the complex exponentials  $e^{s_k t}$  for CT and  $z_k^n$  for DT:

- CT:  $\sum_k a_k e^{s_k t} \rightarrow \sum_k a_k H(s_k) e^{s_k t}$

- DT:  $\sum_k a_k z_k^n \rightarrow \sum_k a_k H(z_k) z_k^n$



# 1.2 Orthogonality for Exponential functions

- Exponential functions

- The set of complex periodic basis functions is expressed as:  $\phi_k = e^{jk\Omega_0 t}$ ,  $k$  is integer

- with  $\Omega_0$  as the fundamental frequency

- $T_0 = \frac{2\pi}{\Omega_0}$  as the corresponding period.

- The basis function set is orthogonal in the sense

$$\int_0^{T_0} \phi_m(t) \phi_n^*(t) dt = \begin{cases} 0, & n \neq m \\ T_0, & n = m \end{cases}$$

$$\Rightarrow \int_0^{T_0} e^{jm\Omega_0 t} e^{-jn\Omega_0 t} dt = \begin{cases} 0, & n \neq m \\ T_0, & n = m \end{cases}$$

# Example 1

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- Consider an LTI system for which the input  $x(t)$  and output  $y(t)$  are related by a time shift of 3:

$$y(t) = x(t - 3)$$

- If the input to this system is the complex exponential signal  $x(t) = e^{j2t}$ , determine the eigen-function and corresponding eigen-value of this system.

## 2.1 CT periodic signals

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- To express signals as a linear combination of complex exponentials, we have 2 questions:
  - Does this expression always exist?
  - How broad can this expression be used?

Complex exponentials are periodic



Linear combination of them should also be periodic?  
=> Let's start with PERIODIC signals, in CT.



## 2.1 CT periodic signals

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- $x(t)$  is *periodic* if, for some positive value of  $T$ :

$$x(t) = x(t + T) \quad \text{for all } t$$

- for all  $t$ , no exception;
- minimum positive, nonzero value of  $T$  is the *fundamental period*;
- $\Omega_0 = \frac{2\pi}{T}$  is referred to as the *fundamental frequency*.
- For example:

sinusoidal signal:  $x(t) = \cos \Omega_0 t$

complex exponential:  $x(t) = e^{j\Omega_0 t}$

- fundamental frequencies are both  $\Omega_0$
- fundamental periods are  $T = \frac{2\pi}{\Omega_0}$

## 2.1 Synthesis equation

- Harmonically related complex exponential:

$$\phi_k(t) = e^{jk\Omega_0 t} = e^{j\frac{2\pi k}{T}t}, \quad k = 0, \pm 1, \pm 2, \dots$$

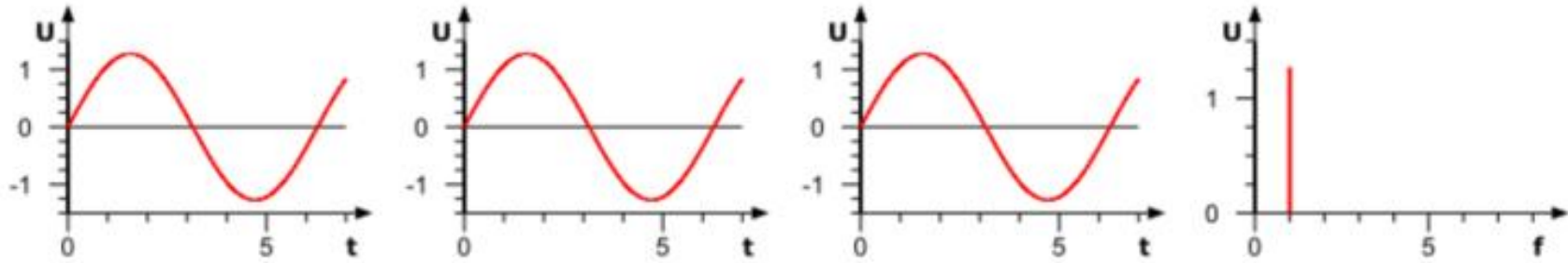
- Each of these signals has a fundamental frequency  $k\Omega_0$  that is a multiple of  $\Omega_0$ , and therefore, each is periodic with period T
  - although for  $|k| \geq 2$ , the fundamental period is a fraction of T.
- The linear combination of harmonically related complex exponentials:

$$a_0 + a_1 e^{j\Omega_0 t} + a_2 e^{j2\Omega_0 t} + \dots = \sum_{k=-\infty}^{\infty} a_k e^{jk\Omega_0 t} = x(t)$$

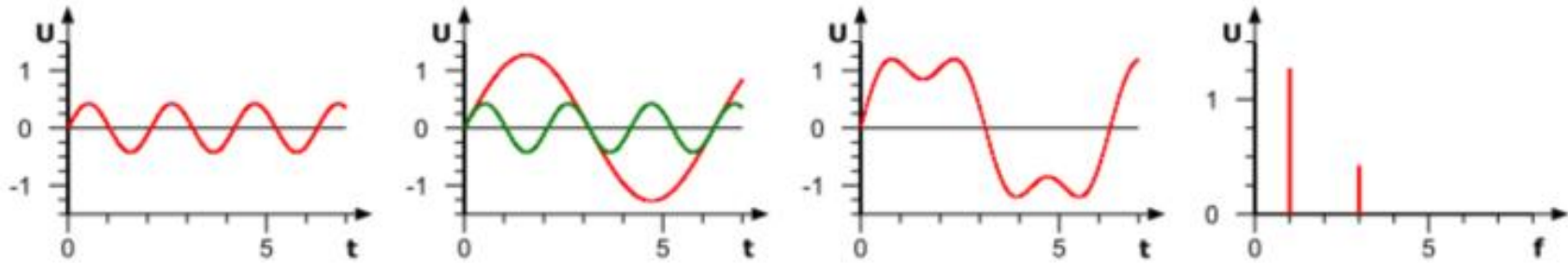
- is also periodic with period T;
  - the components for  $k = +N$  and  $k = -N$  are referred to as the  $N^{\text{th}}$  harmonic components.
- This representation is referred to as the *synthesis equation* of **Fourier Series**.

# *(P5 again) Synthesis of square wave*

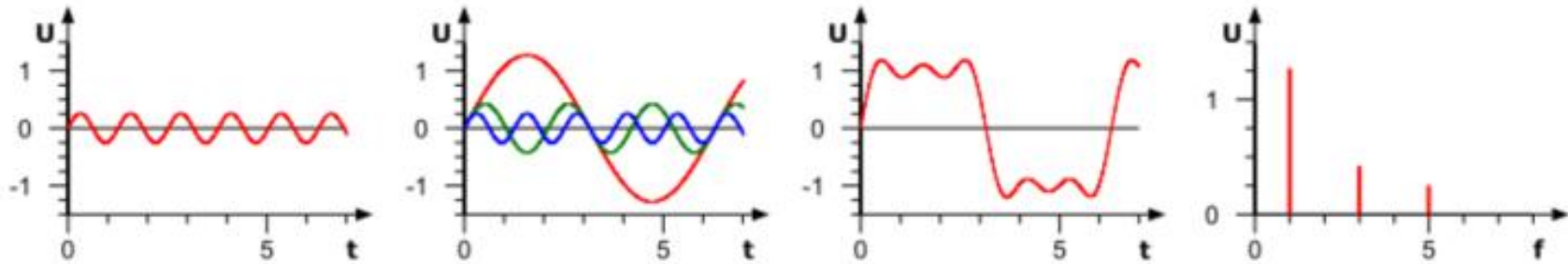
Fundamental



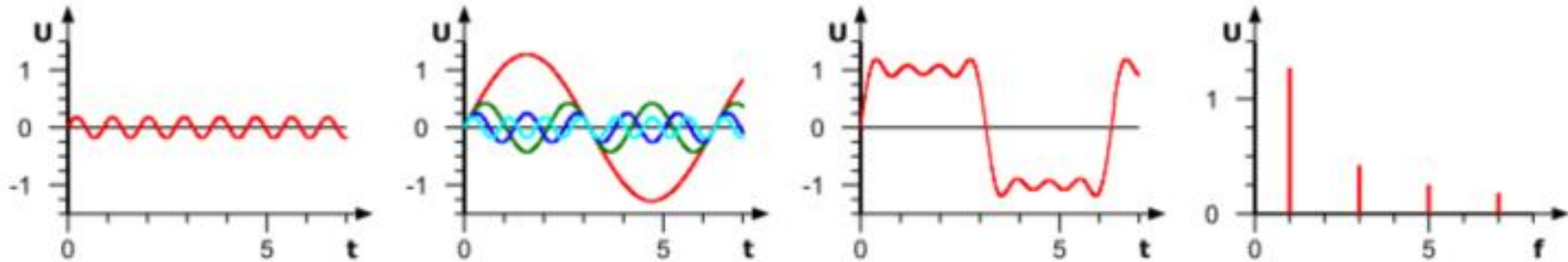
3rd harmonic



5th harmonic



7th harmonic



## 2.2 Fourier series coefficients

- Review the property of complex exponentials:

$$\int_T e^{jm\Omega_0 t} dt = \begin{cases} T & m = 0 \\ 0 & m \neq 0 \end{cases}$$

- Modify the original Fourier series by:

$$\begin{aligned} \int_T x(t) e^{-jn\Omega_0 t} dt &= \int_T \sum_{k=-\infty}^{\infty} a_k e^{jk\Omega_0 t} e^{-jn\Omega_0 t} dt \\ &= \sum_{k=-\infty}^{\infty} a_k \int_T e^{jk\Omega_0 t} e^{-jn\Omega_0 t} dt = \sum_{k=-\infty}^{\infty} a_k \int_T e^{j(k-n)\Omega_0 t} dt \end{aligned}$$

$$\Rightarrow \int_T x(t) e^{-jn\Omega_0 t} dt = a_n T \Rightarrow a_n = \frac{1}{T} \int_T x(t) e^{-jn\Omega_0 t} dt$$



## Example 2

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- Consider the signal

$$x(t) = \sin(\omega_0 t)$$

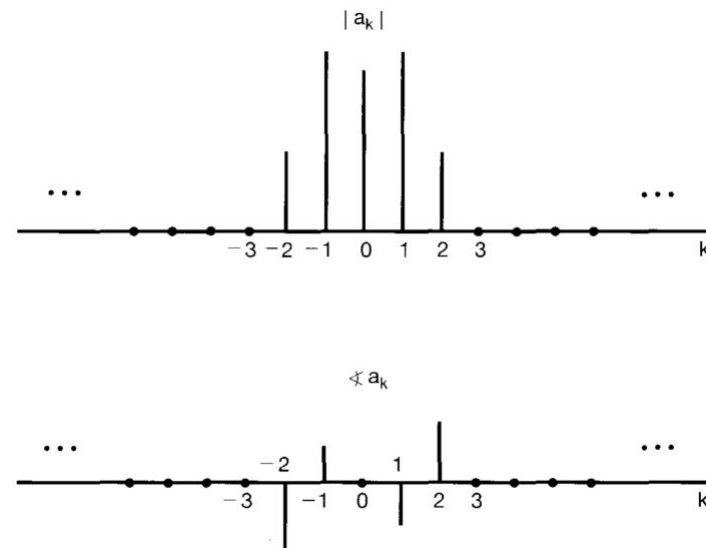
- Determine the Fourier series coefficients  $a_k$  for this signal.

# Quiz 1

- Consider the signal

$$x(t) = 1 + \sin(\omega_0 t) + 2\cos(\omega_0 t) + \cos(2\omega_0 t + \frac{\pi}{4})$$

- Determine the Fourier series coefficients  $a_k$  for this signal.

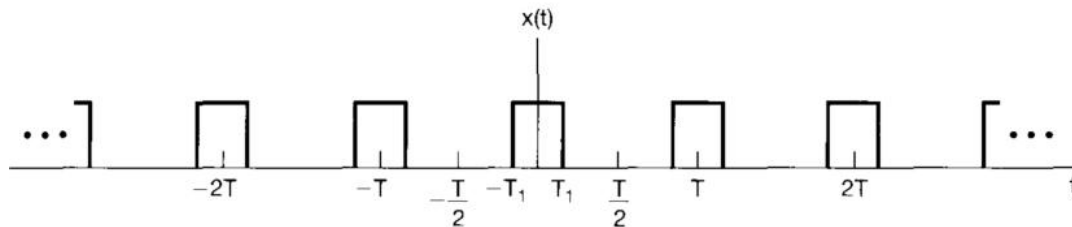


## Example 3

- Consider the signal

$$x(t) = \begin{cases} 1, & |t| < T_1 \\ 0, & T_1 < |t| < T/2 \end{cases}$$

$x(t)$  is periodic with fundamental period  $T$ .



- Determine the Fourier series coefficients  $a_k$  for this signal.

# Example 3

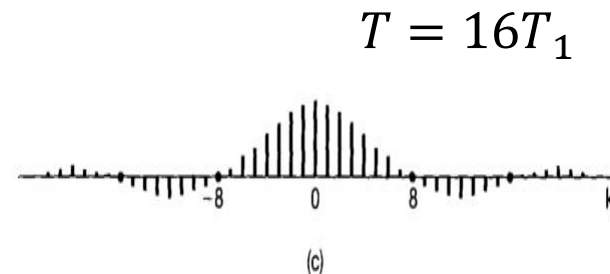
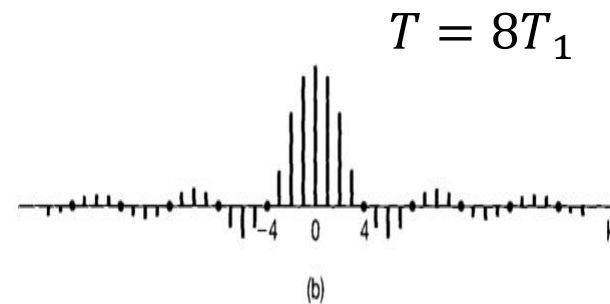
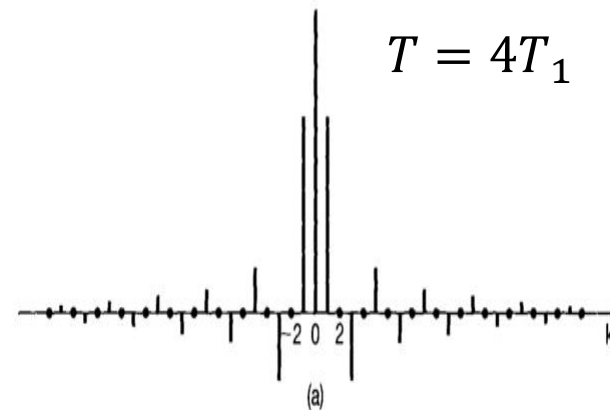
- Solution:

- For  $k = 0$ :

$$a_0 = \frac{1}{T} \int_{-T_1}^{T_1} dt = \frac{2T_1}{T}$$

- For  $k \neq 0$ :

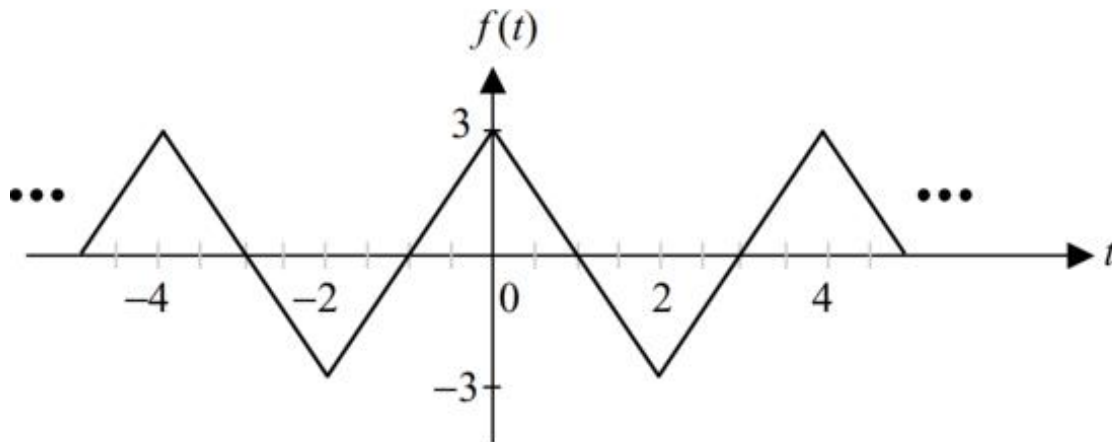
$$\begin{aligned} a_k &= \frac{1}{T} \int_{-T_1}^{T_1} e^{-jk\Omega_0 t} dt = -\frac{1}{jk\Omega_0 T} e^{-jk\Omega_0 t} \Big|_{-T_1}^{T_1} \\ &= \frac{2}{k\Omega_0 T} \left[ \frac{e^{jk\Omega_0 T_1} - e^{-jk\Omega_0 T_1}}{2j} \right] = \frac{2 \sin(k\Omega_0 T_1)}{k\Omega_0 T} \\ &= \frac{\sin(k\Omega_0 T_1)}{k\pi} = \frac{2T_1}{T} \text{sinc}(k\Omega_0 T_1) \end{aligned}$$





## Quiz 2

- Consider a signal as shown, determine the Fourier series of it.



## 2.2 Fourier Series Properties

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- TD: Complex signal
  - $\mathcal{Re}\{x(t)\}$  is odd, an  $\mathcal{Im}\{x(t)\}$  is even  
 $\Rightarrow$  FD: Imaginary only
  - $\mathcal{Re}\{x(t)\}$  is even, an  $\mathcal{Im}\{x(t)\}$  is odd  
 $\Rightarrow$  FD: Real only
- TD: Real signal
  - $x(t)$  is odd  
 $\Rightarrow$  FD: Sine only
  - $x(t)$  is even  
 $\Rightarrow$  FD: Cosine only



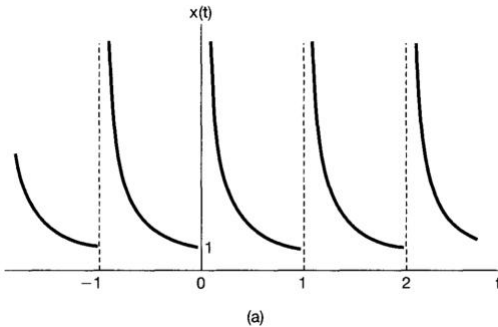
## 2.3 Existence of the Fourier Series

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- Not every periodic signal has a Fourier series
- The Dirichlet conditions are:
  - $x(t)$  must be absolutely integrable;
  - $x(t)$  has a finite number of maxima and minima in any single period;
  - $x(t)$  has a finite number of discontinuities during any time interval.
- The Dirichlet conditions, if met, guarantees that  $x(t)$  equals its Fourier series representation except at isolated values of  $t$  for which  $x(t)$  is discontinuous.
- At a mismatch point, the infinite series converges to the average of the values on either side of the discontinuity.
- Because the isolated points of discontinuity has no effect on an integration operation, the Fourier series representation is as good as the original signal for LTI system analysis.

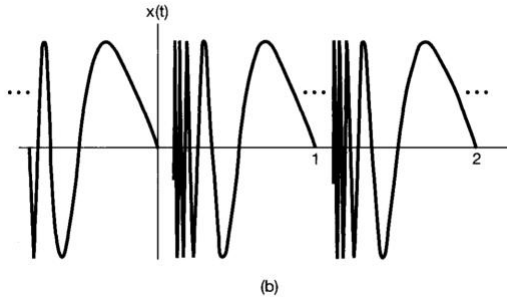
## 2.3 Existence of the Fourier Series

- Some examples (violates the Dirichlet conditions:



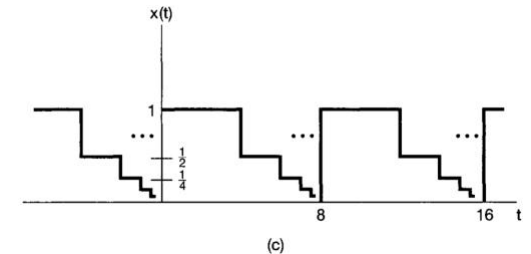
(a) the signal  $x(t) = 1/t$  for  $0 < t \leq 1$ , a periodic signal with period 1;

this signal violates the first Dirichlet condition



(b) the periodic signal of  $x(t) = \sin(2\pi/t)$ ;

It violates the second Dirichlet condition

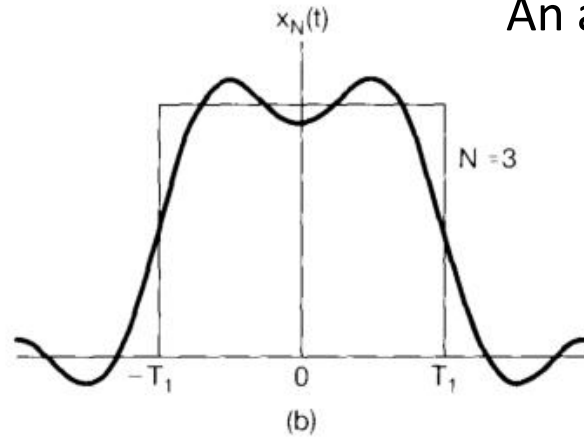
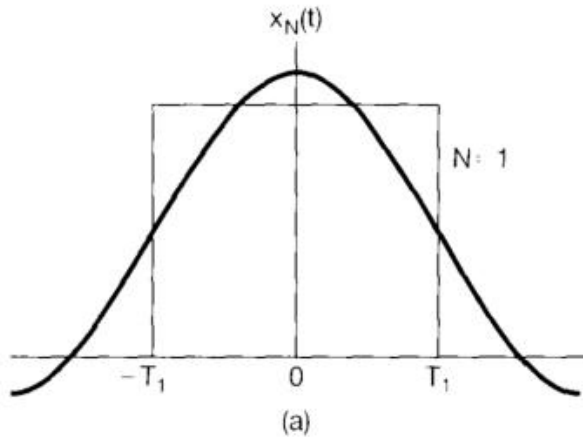


(c) a signal periodic with period 8 [for  $0 < t < 8$ , the value of  $x(t)$  decreases by a factor of 2 whenever the distance from  $t$  to 8 decreases by a factor of 2];

It violates the third Dirichlet condition

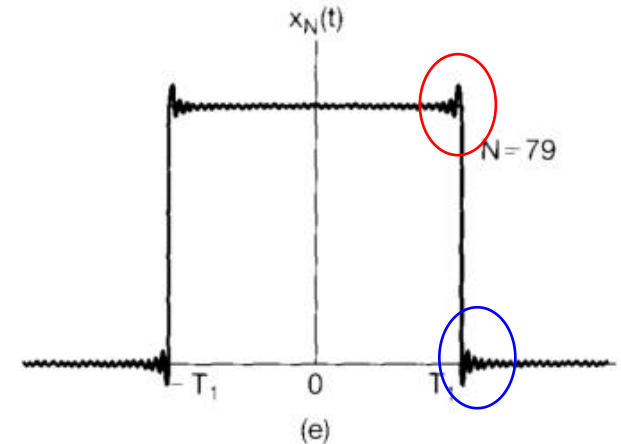
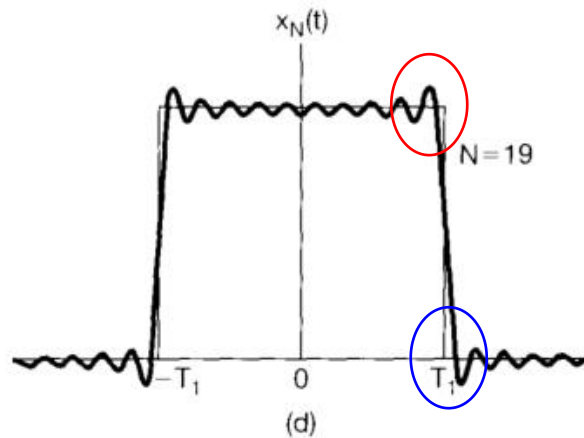
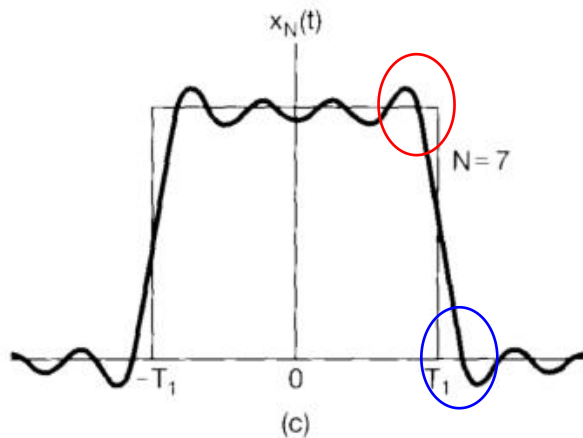
## 2.3 Gibbs phenomenon

An approximation of the square wave



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

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



*These ripples near discontinuity never go away and has a maximum value of **1.09**.*

## 2.4 Signal spectrum

- Signal spectrum is the graphical representation of  $a_k$

$$a_k = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) e^{-j\frac{2\pi kt}{T}} dt$$

where  $a_k$  is a complex function of the variable  $k$ .

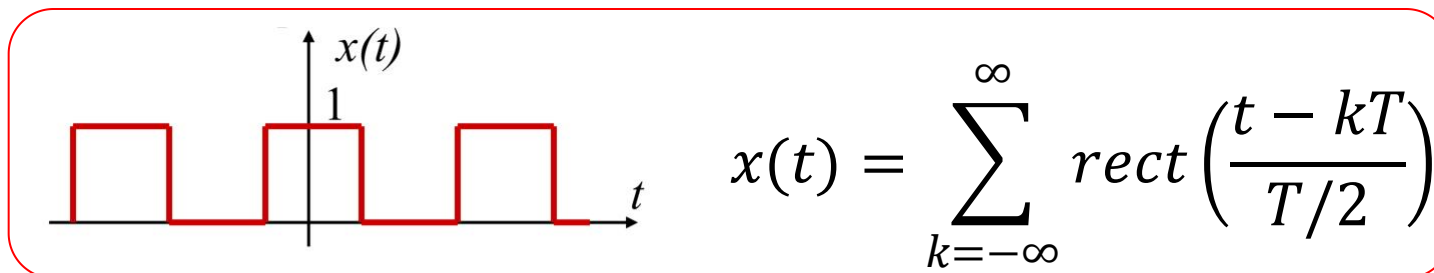
- How could we draw it (complex function)?

$$a_k = |a_k| e^{j\angle a_k}$$

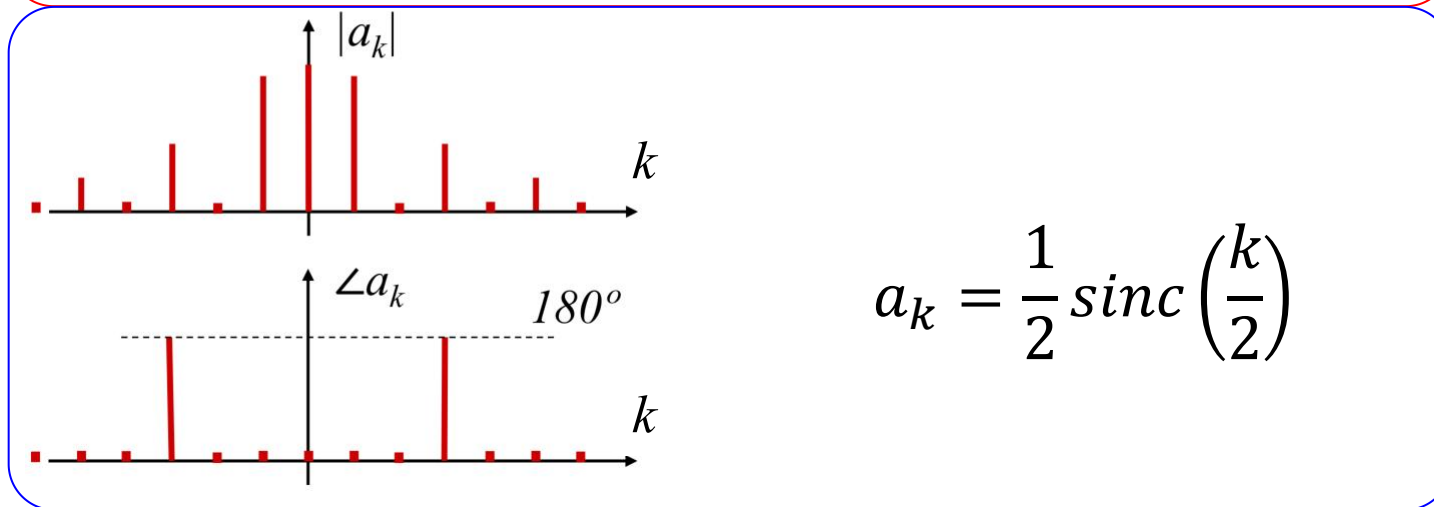
- Magnitude  $|a_k|$
- Phase  $\angle a_k$

# Example

- For the square wave



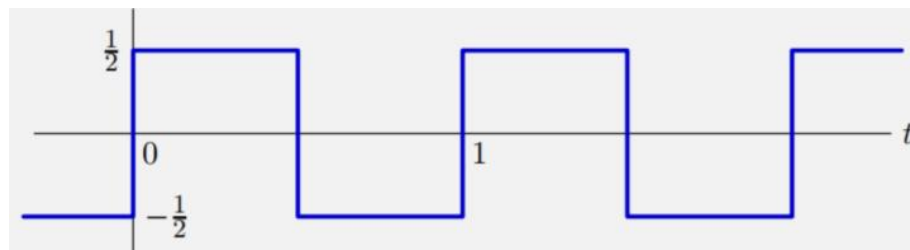
Time  
Domain



Frequency  
Domain

## Quiz 3

- Let  $a_k$  represent the Fourier series coefficients of the following square wave:



- How many of the following statements are true?
  - $a_k = 0$  if  $k$  is even;
  - $a_k$  is real-valued;
  - $|a_k|$  decreases with  $k^2$ ;
  - there are an infinite number of non-zero  $a_k$ ;
  - all of the above.



# Next ...

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- Continuous Time Fourier Transform (CTFT)
  - Fundamentals of CTFT
    - From CTFS to CTFT
  - Fourier transform pairs
    - $e^{j\omega_0 t}$ ,  $\delta(t)$ , etc.
  - Fourier transform properties
  - Frequency response of a system