



# 7

## Continuous-Time Fourier Series

In representing and analyzing linear, time-invariant systems, our basic approach has been to decompose the system inputs into a linear combination of basic signals and exploit the fact that for a linear system the response is the same linear combination of the responses to the basic inputs. The convolution sum and convolution integral grew out of a particular choice for the basic signals in terms of which we carried out the decomposition, specifically delayed unit impulses. This choice has the advantage that for systems which are time-invariant in addition to being linear, once the response to an impulse at one time position is known, then the response is known at all time positions.

In this lecture, we begin the discussion of the representation of signals in terms of a different set of basic inputs—complex exponentials with unity magnitude. For periodic signals, a decomposition in this form is referred to as the Fourier series, and for aperiodic signals it becomes the Fourier transform. In Lectures 20–22 this representation will be generalized to the Laplace transform for continuous time and the  $z$ -transform for discrete time.

Complex exponentials as basic building blocks for representing the input and output of LTI systems have a considerably different motivation than the use of impulses. Complex exponentials are eigenfunctions of LTI systems; that is, the response of an LTI system to any complex exponential signal is simply a scaled replica of that signal. Consequently, if the input to an LTI system is represented as a linear combination of complex exponentials, then the effect of the system can be described simply in terms of a weighting applied to each coefficient in that representation. This very important and elegant relationship between LTI systems and complex exponentials leads to some extremely powerful concepts and results.

Before capitalizing on this property of complex exponentials in relation to LTI systems, we must first address the question of how a signal can be represented as a linear combination of these basic signals. For periodic signals, the representation is referred to as the *Fourier series* and is the principal topic of this lecture. Specifically, we develop the Fourier series representation for periodic continuous-time signals. In Lecture 8 we extend that representation to the representation of continuous-time aperiodic signals. In Lectures 10 and 11, we develop parallel results for the discrete-time case.



The continuous-time Fourier series expresses a periodic signal as a linear combination of harmonically related complex exponentials. Alternatively, it can be expressed in the form of a linear combination of sines and cosines or sinusoids of different phase angles. In these lectures, however, we will use almost exclusively the complex exponential form. The equation describing the representation of a time function as a linear combination of complex exponentials is commonly referred to as the *Fourier synthesis equation*, and the equation specifying how the coefficients are determined in terms of the time function is referred to as the *Fourier series analysis equation*. To illustrate the Fourier series, we focus in this lecture on the Fourier series representation of a periodic square wave.

The fact that a square wave which is discontinuous can be “built” as a linear combination of sinusoids at harmonically related frequencies is somewhat astonishing. In fact, as we add terms in the Fourier series representation, we achieve an increasingly better approximation to the square wave except at the discontinuities; that is, as the number of terms becomes infinite, the Fourier series converges to the square wave at every value of  $T$  except at the discontinuities. However, for this example and more generally for periodic signals that are square-integrable, the error between the original signal and the Fourier series representation is negligible, in practical terms, in the sense that this error in the limit has zero energy. In the lecture, some of these convergence issues are touched on with the objective of developing insight into the behavior of the Fourier series rather than representing an attempt to focus formally on the mathematics.

The Fourier series for periodic signals also provides the key to representing aperiodic signals through a linear combination of complex exponentials. This representation develops out of the very clever notion of representing an aperiodic signal as a periodic signal with an increasingly large period. As the period becomes larger, the Fourier series becomes in the limit the Fourier integral or Fourier transform, which we begin to develop in the next lecture.

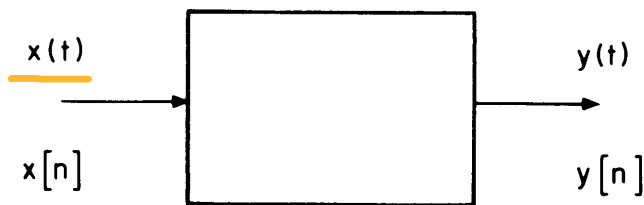
### Suggested Reading

Section 4.0, Introduction, pages 161–166

Section 4.1, The Response of Continuous-Time LTI Systems to Complex Exponentials, pages 166–168

Section 4.2, Representation of Periodic Signals: The Continuous-Time Fourier Series, pages 168–179

Section 4.3, Approximation of Periodic Signals Using Fourier Series and the Convergence of Fourier Series, pages 179–185

**TRANSPARENCY**

7.1

The principle of  
superposition for  
linear systems.

If:  $\underline{x(t)} = a_1 \phi_1(t) + a_2 \phi_2(t) + \dots$

$$\phi_k(t) \longrightarrow \psi_k(t)$$

and system is linear

Then:  $y(t) = a_1 \psi_1(t) + a_2 \psi_2(t) + \dots$

Identical for discrete-time

If:  $x = a_1 \phi_1 + a_2 \phi_2 + \dots$

Then:  $y = a_1 \psi_1 + a_2 \psi_2 + \dots$

Choose  $\phi_k(t)$  or  $\phi_k[n]$  so that:

- a broad class of signals can be constructed as a linear combination of  $\phi_k$ 's
- response to  $\phi_k$ 's easy to compute

**TRANSPARENCY**

7.2

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

**TRANSPARENCY**

7.3

Choice for the basic signals that led to the convolution integral and convolution sum.

**LTI SYSTEMS:**

• **C-T:**  $\phi_k(t) = \delta(t - k\Delta)$

$\psi_k(t) = h(t - k\Delta)$

$\Rightarrow$  **Convolution Integral**

• **D-T:**  $\phi_k[n] = \delta[n - k]$

$\psi_k[n] = h[n - k]$

$\Rightarrow$  **Convolution Sum**

**TRANSPARENCY**

7.4

Complex exponentials as a set of basic signals.

将信号分解成复指数的线性组合,  $s_k$ 是复指数

$\phi_k(t) = e^{s_k t}$        $s_k$  complex

$\phi_k[n] = z_k^n$        $z_k$  complex

**Fourier Analysis:**

• **C-T:**  $s_k = j\omega_k$        $\phi_k(t) = e^{j\omega_k t}$

• **D-T:**  $|z_k| = 1$        $\phi_k[n] = e^{j\Omega_k n}$

$s_k$  complex  $\Rightarrow$  Laplace transforms

$z_k$  complex  $\Rightarrow$  z-transforms



## MARKERBOARD

7.1

$$\phi_k(t) = e^{j\omega_k t}$$

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

"Proof"

$$e^{j\omega_k t} \rightarrow \int_{-\infty}^{\infty} R(\tau) e^{j\omega_k(t-\tau)} d\tau$$

$$e^{j\omega_k t} e^{-j\omega_k \tau}$$

$$e^{j\omega_k t} \rightarrow e^{j\omega_k t} \int_{-\infty}^{\infty} h(\tau) e^{-j\omega_k \tau} d\tau$$

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

特征函数  $e^{-j\omega_k \tau}$   $e^{-j\omega_k \tau}$  特征值

Periodic Signals

- Fourier Series

Aperiodic Signals

- Fourier Transform

## C-T Fourier Series

$$x(t) = x(t+T_0) \quad \text{period}$$

$$\omega_0 = \frac{2\pi}{T_0} = 2\pi f_0$$

$$e^{j\omega_0 t} \quad T_0 = \frac{2\pi}{\omega_0} \rightarrow f_0$$

$$e^{jk\omega_0 t} \quad \frac{T_0}{k} = \frac{2\pi}{k\omega_0} \rightarrow f_0/k$$

其他谐波相关的复指数  
Harmonically Related

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

Period:

$$T_0 = \frac{2\pi}{\omega_0}$$

fundamental frequency:

$$\omega_0 = \frac{2\pi}{T_0}$$

傅立叶级数的复指数形式

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

傅立叶提出：普通的周期信号可以表示成谐波相关复指数的线性组合

Complex Exponential Form

$$a_k = A_k e^{j\theta_k} \quad \text{极坐标形式}$$

$$= B_k + jC_k \quad \text{直角坐标形式}$$

$$e^{jk\omega_0 t} = \cos k\omega_0 t + j \sin k\omega_0 t$$

傅立叶级数的三角函数形式

$$x(t) = a_0 + 2 \sum_{k=1}^{\infty} A_k \cos(k\omega_0 t + \theta_k)$$

$$a_0 + 2 \sum_{k=1}^{\infty} [B_k \cos k\omega_0 t - C_k \sin k\omega_0 t]$$

Trigonometric Form

The minus sign at the beginning of the exponent at the bottom of column 1 is incorrect;  $dt$  should be included at the end of the analysis integral at the bottom of column 2.

## MARKERBOARD

7.2

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

$$\int_{T_0} e^{jm\omega_0 t} dt = \begin{cases} T_0 & m=0 \\ 0 & m \neq 0 \end{cases}$$

$$= \int_{T_0} \cos m\omega_0 t dt + j \int_{T_0} \sin m\omega_0 t dt$$

$$m \neq 0 \quad 0 \quad 0$$

$$m=0 \quad T_0 \quad 0$$

$$\int_{T_0} x(t) e^{-jn\omega_0 t} dt =$$

$$\int_{T_0} e^{-jn\omega_0 t} \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} dt$$

$$= \sum_{k=-\infty}^{\infty} a_k \int_{T_0} e^{j(k-n)\omega_0 t} dt$$

$$\int_{T_0} x(t) e^{-jn\omega_0 t} dt =$$

$$\sum_{k=-\infty}^{\infty} a_k \int_{T_0} e^{j(k-n)\omega_0 t} dt$$

$$T_0 \text{ if } k=n$$

$$0 \text{ if } k \neq n$$

$$\int_{T_0} x(t) e^{-jn\omega_0 t} dt = a_n(T_0)$$

Synthesis

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

Analysis

$$a_k = \frac{1}{T_0} \int_{T_0} x(t) e^{-jk\omega_0 t} dt$$

Cool!

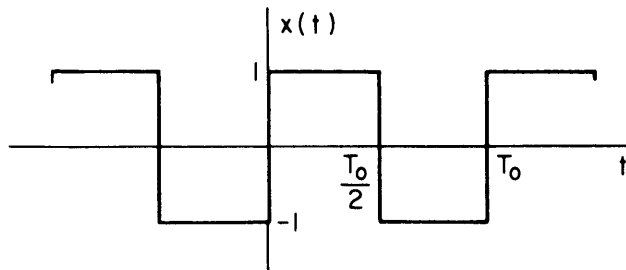


## TRANSPARENCY

7.5

Determination of the Fourier series coefficients for an antisymmetric periodic square wave.

## ANTISYMMETRIC PERIODIC SQUARE WAVE



$$a_k = \frac{1}{T_o} \int_{-T_o/2}^0 (-1) e^{-jk\omega_o t} dt + \frac{1}{T_o} \int_0^{T_o/2} (+1) e^{-jk\omega_o t} dt$$

$$\vdots$$

$$= \frac{1}{j\pi k} \{1 - (-1)^k\} \quad k \neq 0$$

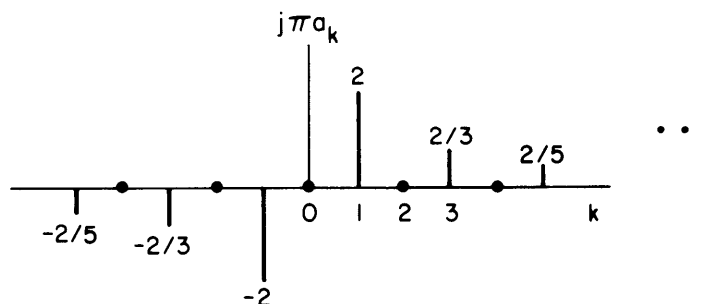
$$a_o = \frac{1}{T_o} \int_{T_o} x(t) e^{-jk\omega_o t} dt = \frac{1}{T_o} \int_{T_o} x(t) dt = 0$$

## TRANSPARENCY

7.6

The Fourier series coefficients for an antisymmetric periodic square wave.

## ANTISYMMETRIC PERIODIC SQUARE WAVE



$$a_o = 0; \quad a_k = \frac{1}{j\pi k} \{1 - (-1)^k\} \quad k \neq 0$$

• odd harmonic

•  $a_k$  imaginary  $a_k$ 是虚数•  $a_k = -a_{-k}$  (antisymmetric)
$$\left. \begin{array}{l} \bullet \text{ odd harmonic} \\ \bullet a_k \text{ imaginary } a_k \text{是虚数} \\ \bullet a_k = -a_{-k} \text{ (antisymmetric)} \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} \text{sine series} \\ x(t) = a_o + \sum_{k=1}^{\infty} 2j a_k \sin k\omega_o t \end{array} \right.$$

sine series

$$x(t) = a_o + \sum_{k=1}^{\infty} 2j a_k \sin k\omega_o t$$

非对称周期正方波的傅立叶级数实际是一个正弦级数

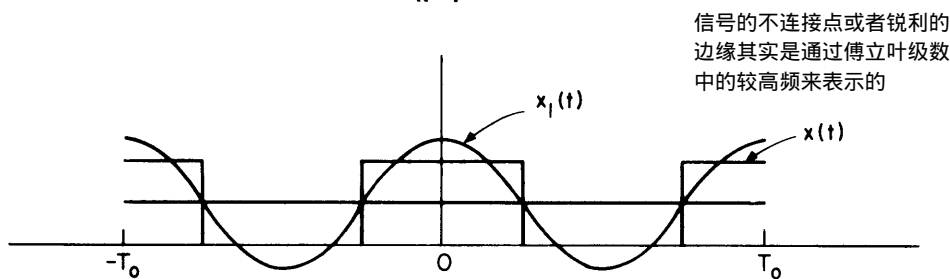
实数

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_0 t} \quad \omega_0 = \frac{2\pi}{T_0}$$

$$x_N(t) \triangleq \sum_{k=-N}^N a_k e^{jk\omega_0 t}$$

Symmetric square wave:  $a_k = a_{-k}$

$$x_N(t) = \frac{1}{2} + \sum_{k=1}^N 2a_k \cos k\omega_0 t$$



信号的不连接点或者锐利的边缘其实是通过傅立叶级数中的较高频来表示的

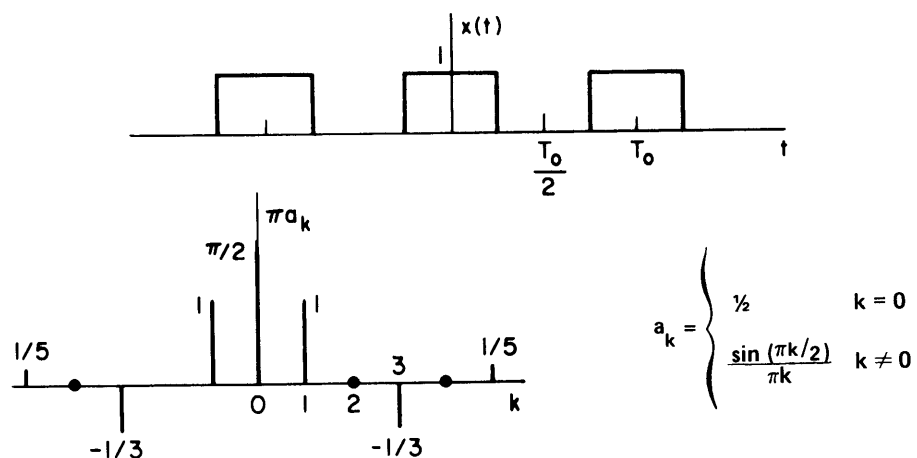
低频的项表现了长期内的特性，高频则用来构建时域中的急剧转变

## TRANSPARENCY 7.7

Fourier series coefficients for a symmetric periodic square wave.

## SYMMETRIC PERIODIC SQUARE WAVE

Example 4.5:



- odd harmonic
- $a_k$  real
- $a_k = a_{-k}$  (symmetric)

$$\left. \begin{array}{l} \text{cosine series} \\ \Rightarrow \end{array} \right\} x(t) = a_0 + \sum_{k=1}^{\infty} 2a_k \cos k\omega_0 t$$

## TRANSPARENCY 7.8

Illustration of the superposition of terms in the Fourier series representation for a symmetric periodic square wave. [Example 4.5 from the text.]



## TRANSPARENCY

7.9

Partial sum  
incorporating  
(2N + 1) terms in the  
Fourier series.

[The analysis equation  
should read  $a_k =$

$$1/T \int_{T_0} x(t) e^{-jk\omega_0 t} dt.]$$

$$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_0 t} \quad \text{synthesis}$$

$$a_k = \frac{1}{T_0} \int_{T_0} x(t) e^{-jk\omega_0 t} dt \quad \text{analysis}$$

$$x_N(t) \triangleq \sum_{k=-N}^N a_k e^{jk\omega_0 t}$$

$$e_N(t) \triangleq x(t) - x_N(t)$$

Does  $e_N(t)$  decrease as N increases?

## CONVERGENCE OF FOURIER SERIES

## TRANSPARENCY

7.10

Conditions for  
convergence of the  
Fourier series.

•  $x(t)$  square integrable:

$$\text{if } \int_{T_0} |x(t)|^2 dt < \infty \quad \text{如果可积}$$

$$\text{then } \int_{T_0} |e_N(t)|^2 dt \rightarrow 0 \text{ as } N \rightarrow \infty$$

• Dirichlet conditions.

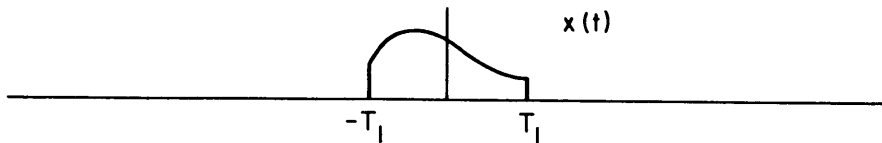
$$\text{if } \int_{T_0} |x(t)| dt < \infty \text{ and } x(t) \text{ "well behaved"}$$

$$\text{then } e_N(t) \rightarrow 0 \text{ as } N \rightarrow \infty$$

except at discontinuities



## FOURIER REPRESENTATION OF APERIODIC SIGNALS



$$\tilde{x}(t) = x(t) \quad |t| < \frac{T_o}{2}$$

$$\text{As } T_o \rightarrow \infty \quad \tilde{x}(t) \rightarrow x(t)$$

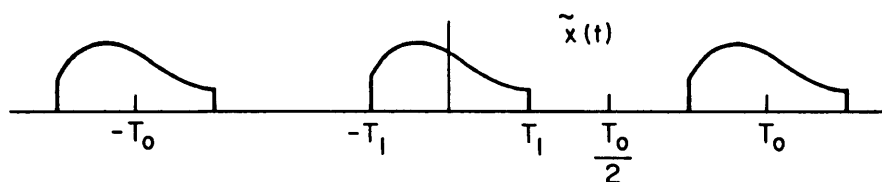


- use Fourier series to represent  $\tilde{x}(t)$
- let  $T_o \rightarrow \infty$  to represent  $x(t)$

### TRANSPARENCY 7.11

An aperiodic signal to be represented as a linear combination of complex exponentials.

## FOURIER REPRESENTATION OF APERIODIC SIGNALS



$$\tilde{x}(t) = x(t) \quad |t| < \frac{T_o}{2}$$

$$\text{As } T_o \rightarrow \infty \quad \tilde{x}(t) \rightarrow x(t)$$



- use Fourier series to represent  $\tilde{x}(t)$
- let  $T_o \rightarrow \infty$  to represent  $x(t)$

### TRANSPARENCY 7.12

Representation of an aperiodic signal as the limiting form of a periodic signal with the period increasing to infinity.



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