

Chapter 5. Fourier Analysis for Discrete-Time Signals and Systems

Chapter Objectives

1. Learn techniques for representing *discrete-time periodic* signals using *orthogonal* sets of *periodic basis* functions.
2. Study properties of *exponential, trigonometric* and *compact Fourier series*, and conditions for their existence.
3. Learn the *Fourier transform* for *non-periodic* signal as an extension of Fourier series for periodic signals
4. Study the *properties* of the Fourier transform. Understand the concepts of *energy* and *power spectral density*.

5.2 Exponential Fourier Series (EFS)

Continue-Time Fourier Series

Synthesis equation:

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

Analysis equation:

$$c_k = \frac{1}{T_0} \int_{t_0}^{t_0+T_0} \tilde{x}(t) e^{-jk\omega_0 t} dt$$

Discrete-Time Fourier Series

Synthesis equation:

$$\tilde{x}[n] = \sum_{k=0}^{N-1} c_k e^{j(2\pi/N)kn}$$

Analysis equation:

$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn}$$

5.2.7 Properties of Fourier Series

Linearity

Continue-Time Fourier Series

$$x(t) \xleftrightarrow{\mathfrak{F}} c_k$$

$$y(t) \xleftrightarrow{\mathfrak{F}} d_k$$

$$a_1 x(t) + a_2 y(t) \xleftrightarrow{\mathfrak{F}} a_1 c_k + a_2 d_k$$

Discrete-Time Fourier Series

$$x[n] \xleftrightarrow{\mathfrak{F}} c_k$$

$$y[n] \xleftrightarrow{\mathfrak{F}} d_k$$

$$a_1 x[n] + a_2 y[n] \xleftrightarrow{\mathfrak{F}} a_1 c_k + a_2 d_k$$

Where a_1 and a_2 are any two constants

5.2.7 Properties of Fourier Series

Time shift

Continue-Time Fourier Series

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

$$\tilde{x}(t - \tau) = \sum_{k=-\infty}^{\infty} [c_k e^{-jk\omega_0 \tau}] e^{jk\omega_0 t}$$

Discrete-Time Fourier Series

$$\tilde{x}[n] = \sum_{k=0}^{N-1} c_k e^{j(2\pi/N)kn}$$

$$\tilde{x}[n - m] = \sum_{k=0}^{N-1} c_k e^{j(2\pi/N)kn} e^{-j(2\pi/N)km}$$

5.3 Analysis of Non-periodic Continuous-Time Signals

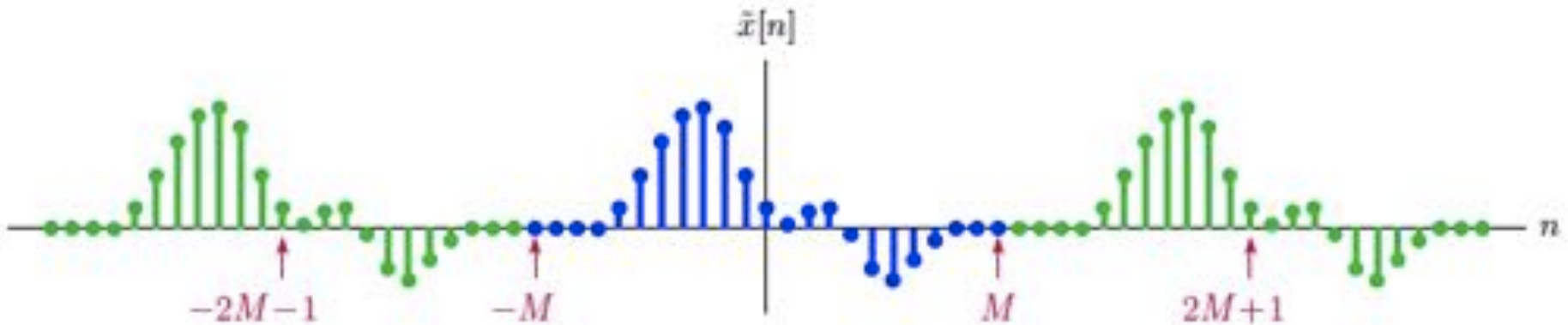
Discrete-Time Fourier Transform

Synthesis equation (inverse):

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

Analysis equation (forward):

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



$$\frac{2\pi k}{2M+1} \rightarrow \Omega$$

5.3 Analysis of Non-periodic Continuous-Time Signals

Continue-Time Fourier Transform Discrete-Time Fourier Transform

Synthesis equation (inverse):

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

Analysis equation (forward):

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

Synthesis equation (inverse):

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

Analysis equation (forward):

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

5.3.2 Existence of Fourier Transform

Is it always possible to determine the Fourier series coefficients?

✧ Absolute summable: $\sum_{n=-\infty}^{\infty} |x[n]| < \infty$

✧ Square - summable: $\sum_{n=-\infty}^{\infty} |x[n]|^2 < \infty$

5.3.5 Properties of Fourier Transform

Linearity:

$$x_1[n] \xleftrightarrow{\mathfrak{F}} X_1(\Omega) \quad \text{and} \quad x_2[n] \xleftrightarrow{\mathfrak{F}} X_2(\Omega)$$

$$\alpha_1 x_1[n] + \alpha_2 x_2[n] \xleftrightarrow{\mathfrak{F}} \alpha_1 X_1(\Omega) + \alpha_2 X_2(\Omega)$$

Where α_1 and α_2 are any two constants

Periodicity:

$$X(\Omega + 2\pi r) = X(\Omega)$$

for all integers r

5.3.5 Properties of Fourier Transform

Time Shifting:

$$x[n] \xleftrightarrow{\mathfrak{F}} X(\Omega) \quad \Rightarrow \quad x[n-m] \xleftrightarrow{\mathfrak{F}} X(\Omega)e^{-j\Omega m}$$

Frequency Shifting:

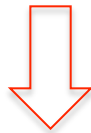
$$x[n] \xleftrightarrow{\mathfrak{F}} X(\Omega) \quad \Rightarrow \quad x[n]e^{-j\Omega_0 n} \xleftrightarrow{\mathfrak{F}} X(\Omega - \Omega_0)$$

5.3.5 Properties of Fourier Transform

Convolution Property:

$$x_1[n] \xleftrightarrow{\mathfrak{F}} X_1(\Omega)$$

$$x_2[n] \xleftrightarrow{\mathfrak{F}} X_2(\Omega)$$



$$x_1[n] * x_2[n] \xleftrightarrow{\mathfrak{F}} X_1(\Omega) X_2(\Omega)$$

$$X_1(\Omega) * X_2(\Omega) \xleftrightarrow{\mathfrak{F}} x_1[n] x_2[n]$$

5.4 Energy and Power in Frequency Domain

Parseval's Theorem:

For a periodic power signal $x(t)$

Continue-Time

$$\frac{1}{T_0} \int_{t_0}^{t_0+T_0} |x(t)|^2 dt = \sum_{k=-\infty}^{\infty} |c_k|^2$$

Discrete-Time

$$\frac{1}{N} \sum_{k=0}^{N-1} |x[n]|^2 = \sum_{k=0}^{N-1} |c_k|^2$$

For a non-periodic power signal

Continue-Time

$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$$

Discrete-Time

$$\sum_{k=0}^{N-1} |x[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\Omega)|^2 d\Omega$$

5.4 Energy and Power in Frequency Domain

Power Spectral Density:

$$S_x(\Omega) = 2\pi \sum_{k=-\infty}^{\infty} |c_k|^2 \delta(\Omega - k\Omega_0)$$

5.4 Energy and Power in Frequency Domain

Autocorrelation Function:

For a energy signal $x(t)$ the autocorrelation function is

$$r_{xx}[m] = \sum_{n=-\infty}^{\infty} x[n]x[n+m]$$

5.5 System Function Concept

System function (frequency response)

Impulse response ($h[n]$) $\xleftrightarrow{\text{Fourier Transform}}$ System function ($H(\Omega)$)

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

In general, $H(\omega)$ is a complex function of ω , and can be written in polar form as:

$$H(\Omega) = |H(\Omega)|e^{j\Theta(\Omega)}$$

5.8 Discrete Fourier Transform

DTFS

DTFT

DFT

Synthesis equation (inverse):

$$x[n] = \sum_{k=0}^{N-1} c_k e^{j(2\pi/N)kn} \quad x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\Omega) e^{j\Omega n} d\Omega \quad x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j(2\pi/N)kn}$$

$n = 0, 1, \dots, N-1$

Analysis equation (forward):

$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn} \quad X(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n} \quad X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn}$$

$k = 0, 1, \dots, N-1$

5.8 Discrete Fourier Transform

Relationship of the DFT to the DTFT

DTFT

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

DFT

$$X[k] = \sum_{n=0}^{N-1} x[n]e^{-j(2\pi/N)kn}$$

The DFT of a length- N signal is equal to its DTFT evaluated at a set of N angular frequencies equally spaced in the interval $[0, 2\pi)$. Let an indexed set of angular frequencies be defined as

$$\Omega_k = \frac{2\pi k}{N}, \quad k = 0, 1, \dots, N-1$$

$$X[k] = X(\Omega) = \sum_{n=0}^{N-1} x[n]e^{-j(2\pi/N)kn}$$

5.8 Discrete Fourier Transform

Why do we need DFT?

- ✧ The signal $x[n]$ and its DFT $X[k]$ each have N samples, making the discrete Fourier transform practical for computer implementation.
- ✧ Fast and efficient algorithm, known as fast Fourier transforms (FFTs), are available for the computation of the DFT.
- ✧ DFT can be used for approximating other forms of Fourier series and transforms for both continuous-time and discrete-time systems.
- ✧ Dedicated processors are available for fast and efficient computation of the DFT with minimal or no programming needed.