Frequency Domain Representation of Signals and LTI Systems: Continuous Time Case

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**Frequency Response of LTI Systems** 

#### Frequency Response of Continuous-Time Systems

Consider a continuous-time LTI system with impulse response h(t).

• Suppose a **complex sinusoid** is applied as input. i.e.,  $x(t) = e^{j\omega t}$ .

• Then, the otput y(t) is given by

$$y(t) = \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau = \int_{-\infty}^{\infty} h(\tau)e^{j\omega(t-\tau)}d\tau$$
$$= e^{j\omega t} \int_{-\infty}^{\infty} h(\tau)e^{-j\omega\tau}d\tau = H(j\omega)e^{j\omega t}, \text{ where}$$

$$H(j\omega) = \int_{-\infty}^{\infty} h(\tau)e^{-j\omega\tau}d\tau$$
 = Frequency Response.

#### Frequency Response of Continuous-Time Systems

In summary, the output of a continuous-time LTI system for a complex sinusoid input is a complex sinusoid of the same frequency, multiplied by a complex number  $H(j\omega)$ , called the frequency response of the system, which is a function of only the frequency  $\omega$  and not the time t.

Therefore, we say that the complex sinusoid  $\psi(t)=e^{j\omega t}$  is an **eigenfunction** of the LTI system with an associated **eigenvalue**  $\lambda=H(j\omega)$ .

#### Frequency Response of Continuous-Time Systems

Substituting  $H(j\omega)=|H(j\omega)|e^{j\arg\{H(j\omega)\}}$ , we obtain the output corresponding to a complex sinusoid  $e^{j\omega t}$  as

$$y(t) = H(j\omega)e^{j\omega t} = |H(j\omega)|e^{j(\omega t + \arg\{H(j\omega)\})}$$

which implies that **in steady state** the LTI system modifies the magnitude of the input by a factor  $|H(j\omega)|$  and modifies the phase by a shift of  $arg\{H(j\omega)\}$ .

Therefore,  $|H(j\omega)|$  is called the **magnitude response** and  $arg\{H(j\omega)\}$  is called the **phase response** of the LTI system.

#### **Example: RC Circuit**

The impulse response is  $h(t) = \frac{1}{RC}e^{-\frac{t}{RC}}u(t)$ . Then, we have

$$H(j\omega) = \frac{\frac{1}{RC}}{j\omega + \frac{1}{RC}}.$$

$$\Rightarrow |H(j\omega)| = \frac{1/RC}{\sqrt{\omega^2 + (1/RC)^2}} \text{ and } \arg\{H(j\omega)\} = -\tan^{-1}(\omega RC).$$

Taking  $x(t) = e^{j\omega t}u(t)$ , it can be shown that

$$y(t) = H(j\omega) \left( e^{j\omega t} - e^{-\frac{t}{RC}} \right) \longrightarrow H(j\omega) e^{j\omega t}$$
 in steady state.

#### Frequency Response of Discrete-Time Systems

Consider a discrete-time LTI system with impulse response h[n].

• Suppose a complex sinusoid is applied as input. i.e.,

$$x[n] = e^{j\Omega n}$$

• Then, the otput y[n] is given by

$$y[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k] = \sum_{k=-\infty}^{\infty} h[k]e^{j\Omega(n-k)}$$
$$= e^{j\Omega n} \sum_{k=-\infty}^{\infty} h[k]e^{-j\Omega k} = H(e^{j\Omega})e^{j\Omega n}, \text{ where}$$

$$H(e^{j\Omega}) = \sum_{k=-\infty}^{\infty} h[k]e^{-j\Omega k}$$
 = Frequency Response.

#### Frequency Response of Discrete-Time Systems

In summary, the output of a discrete-time LTI system for a complex sinusoid input is a complex sinusoid of the same frequency, multiplied by a complex number  $H(e^{i\Omega})$ , called the frequency response of the system, which is a function of only the frequency  $\Omega$  and not the time n.

Similarly, we can write 
$$H(e^{j\Omega})=|H(e^{j\Omega})|e^{j\arg\{H(e^{j\Omega})\}}$$
 where

- $|H(e^{j\Omega})|$  is called the **magnitude response** and
- $arg\{H(e^{j\Omega})\}$  is called the **phase response** of the LTI system.

Fourier Representation of Signals

#### **Motivation for Fourier Representations**

Suppose we can decompose a general input signal x(t) as a weighted sum of complex sinusoids as

$$x(t) = \sum_{k} a_k e^{j\omega_k t}$$

Then, the output y(t) is also a weighted sum of exponentials, given by

$$y(t) = \sum_{k} a_k H(j\omega_k) e^{j\omega_k t}$$

Note that the operation of convolution, h(t) \* x(t), has been replaced by multiplication,  $a_k H(j\omega_k)$ , because x(t) has been expressed as a sum of eigenfunctions; analogous relationships hold in the discrete time as well.

# Four Classes of Fourier Representations

Time Property	Periodic	Nonperiodic
Continuous (t)	Fourier Series (FS) [Chapter 3.5]	Fourier Transform (FT) [Chapter 3.7]
Discrete (n)	Discrete Time Fourier Series (DTFS) [Chapter 3.4]	Discrete Time Fourier Transform (DTFT) [Chapter 3.6]

# Fourier Series Representation of Continuous-Time Periodic Signals

### Fundamental (Angular) Frequency and Harmonics

A sinusoid whose (angular) frequency  $\omega$  is an integer multiple of a fundamental (angular) frequency  $\omega_0$  is said to be a **harmonic** of the sinusoid of the fundamental frequency.

• The sinusoids  $e^{\pm jk\omega_0}$  are the k-th harmonics of  $e^{j\omega_0}$ .

Consider representing a continuous-time periodic signal x(t) with fundamental period  $T_0$  as a weighted sum of complex sinusoids as

$$x(t) = \sum_{k} a_k e^{j\omega_k t}$$

How do we choose the frequencies  $\omega_k$  and coefficients  $a_k$ ?

### Fundamental (Angular) Frequency and Harmonics

The weighted sum must have the same fundamental period  $T_0$ .

- Therefore, each complex sinusoid in the weighted sum must have a fundamental period  $T_0/n$ , n=1,2,3,..., etc., so that the LCM of the fundamenatal periods of the constituent complex sinusoids is  $T_0$ .
- Equivalently, the frequency of each constituent complex sinusoid must be an integer multiple of the signal x(t)'s fundamental frequency  $\omega_0 = \frac{2\pi}{T_0}$ .

This implies that a continuous-time periodic signal x(t) with frequency  $\omega_0$  can be written as a weighted sum of the harmonics of  $e^{j\omega_0}$  as

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

#### **Fourier Series Coefficients**

The weighted sum representation of a continuous-time periodic signal x(t) with fundamental period  $T_0$  can be re-written as

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

where any coefficient  $X[\ell]$  is given by (why?)

$$X[\mathscr{E}] = \frac{1}{T_0} \int_0^{T_0} x(t) e^{-j\mathscr{E}\omega_0 t} dt.$$

The FS coefficients X[k] are known as a **frequency-domain representation** of x(t) because each FS coefficient is associated with a complex sinusoid of a different frequency.

We say that x(t) and X[k] are a **FS pair** and denote this relationship as

$$x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} X[k].$$

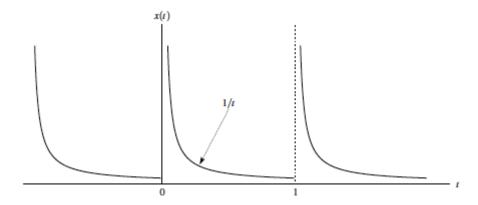
The foregoing equations for weighted sum decomposition of x(t) and finding the corresponding Fourier Series coefficients X[k] are valid only if the **Dirichlet conditions** are satisfied:

• x(t) is bounded or absolutely integrable over one time period, i.e., if

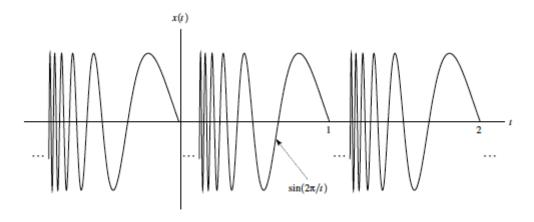
$$\int_{T_0} |x(t)| dt \le M_x < \infty.$$

- x(t) has a finite number of maxima and minima in one period.
- x(t) has a finite number of discontinuities in one period.

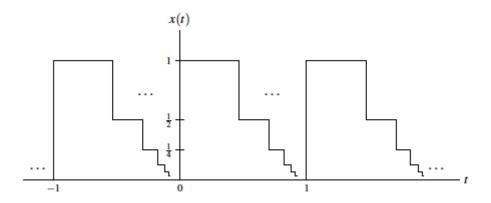
Here, x(t) is **unbounded**.



Here, x(t) has an **infinite number of maxima and minima in one** time period.



Here, x(t) has an **infinite number of jump discontinuities in one** time period.



Find the Fourier Series (FS) expansion of the following signals: (i)  $x(t) = cos(\omega_0 t)$ , (ii)  $x(t) = sin(\omega_0 t)$ , (iii)  $x(t) = cos(2t + \frac{\pi}{4})$ , (iv) x(t) = cos(4t) + sin(6t), (v)  $x(t) = sin^2(t)$ .

**Solution:** (i) We have  $cos(\omega_0 t) = \frac{1}{2} \left( e^{j\omega_0 t} + e^{-j\omega_0 t} \right)$ , which implies that  $X[1] = X[-1] = \frac{1}{2}$  and X[k] = 0 for all  $k \neq \pm 1$ .

(ii) We have  $sin(\omega_0 t) = \frac{1}{2j} \left( e^{j\omega_0 t} - e^{-j\omega_0 t} \right)$ , which implies that  $X[1] = \frac{1}{2j}, X[-1] = -\frac{1}{2j}$  and X[k] = 0 for all  $k \neq \pm 1$ .

Find the Fourier Series (FS) expansion of the following signals: (i)  $x(t) = cos(\omega_0 t)$ , (ii)  $x(t) = sin(\omega_0 t)$ , (iii)  $x(t) = cos(2t + \frac{\pi}{4})$ , (iv) x(t) = cos(4t) + sin(6t), (v)  $x(t) = sin^2(t)$ .

**Solution:** (iii) We have  $\omega_0 = 2$ , and

$$\cos(2t + \frac{\pi}{4}) = \frac{1}{2} \left( e^{j(2t + \frac{\pi}{4})} + e^{-j(2t + \frac{\pi}{4})} \right)$$
$$= \frac{1}{2} e^{j\frac{\pi}{4}} e^{j2t} + \frac{1}{2} e^{-j\frac{\pi}{4}} e^{-j2t},$$

which implies that  $X[1] = \frac{1}{2}e^{j\frac{\pi}{4}}$ ,  $X[-1] = \frac{1}{2}e^{-j\frac{\pi}{4}}$  and X[k] = 0 for all  $k \neq \pm 1$ .

Find the Fourier Series (FS) expansion of the following signals: (i)  $x(t) = cos(\omega_0 t)$ , (ii)  $x(t) = sin(\omega_0 t)$ , (iii)  $x(t) = cos(2t + \frac{\pi}{4})$ , (iv) x(t) = cos(4t) + sin(6t), (v)  $x(t) = sin^2(t)$ .

Solution: (iv) We have

$$T_0 = LCM(T^1, T^2) = LCM\left(\frac{2\pi}{\omega^1}, \frac{2\pi}{\omega^2}\right)$$

$$= LCM\left(\frac{2\pi}{4}, \frac{2\pi}{6}\right) = \pi,$$
which implies that  $\omega_0 = \frac{2\pi}{T_0} = 2$ , and

$$\cos(4t) + \sin(6t) = \frac{1}{2}e^{j4t} + \frac{1}{2}e^{-j4t} + \frac{1}{2j}e^{j6t} - \frac{1}{2j}e^{-j6t}$$
$$= \frac{1}{2}e^{j2\omega_0 t} + \frac{1}{2}e^{-j2\omega_0 t} + \frac{1}{2j}e^{j3\omega_0 t} - \frac{1}{2j}e^{-j3\omega_0 t},$$

which implies that 
$$X[2] = X[-2] = \frac{1}{2}$$
,  $X[3] = \frac{1}{2j}$ ,  $X[-3] = -\frac{1}{2j}$  and  $X[k] = 0$  for all  $k \neq \pm 2, \pm 3$ .

Find the Fourier Series (FS) expansion of the following signals: (i)  $x(t) = cos(\omega_0 t)$ , (ii)  $x(t) = sin(\omega_0 t)$ , (iii)  $x(t) = cos(2t + \frac{\pi}{4})$ , (iv) x(t) = cos(4t) + sin(6t), (v)  $x(t) = sin^2(t)$ .

**Solution:** (v) We have  $sin^2(t) = \frac{1}{2}(1 - cos(2t))$ , which implies that  $\omega_0 = 2$ , and

$$\frac{1}{2}(1-\cos(2t)) = \frac{1}{2} - \frac{1}{4}(e^{j2t} + e^{-j2t})$$

which implies that  $X[0] = \frac{1}{2}$ ,  $X[1] = X[-1] = -\frac{1}{4}$ , and X[k] = 0 for all  $k \neq 0$ ,  $\pm 1$ .

#### **Exponential and Trigonometric Forms**

The decomposition of a continuous-time periodic signal x(t) with period  $T_0$  as a weighted sum of complex exponentials, as discussed earlier, and given by

$$x(t) = \sum_{k=-\infty}^{\infty} X[k]e^{jk\omega_0 t},$$

is called the **exponential form** of FS representation.

Applying Euler's identity, we get the **trigonometric form** of FS representation as

$$x(t) = X[0] + \sum_{k=1}^{\infty} \left( X[k]e^{jk\omega_0 t} + X[-k]e^{-jk\omega_0 t} \right)$$

$$= X[0] + \sum_{k=1}^{\infty} B[k] \cos(k\omega_0 t) + A[k] \sin(k\omega_0 t)$$

#### **Exponential and Trigonometric Forms**

where for any x(t) (real- or complex-valued) we have

$$X[0] = \frac{1}{T_0} \int_0^{T_0} x(t)dt = \text{time-averaged value of } x(t)$$

$$B[k] = (X[k] + X[-k]) = \frac{2}{T_0} \int_0^{T_0} x(t) \cos(k\omega_0 t) dt$$

$$A[k] = j(X[k] - X[-k]) = \frac{2}{T_0} \int_0^{T_0} x(t) \sin(k\omega_0 t) dt$$

Recall that any FS coefficient  $X[\ell]$  is given by

$$X[\mathscr{E}] = \frac{1}{T_0} \int_0^{T_0} x(t) e^{-j\mathscr{E}\omega_0 t} dt.$$

#### **Exponential and Trigonometric Forms**

From the previous equation, it is easy to see that if x(t) is real-valued, then  $X[-\ell] = X^*[\ell]$ , where  $X^*[\ell]$  denotes the complex conjugate of  $X[\ell]$ .

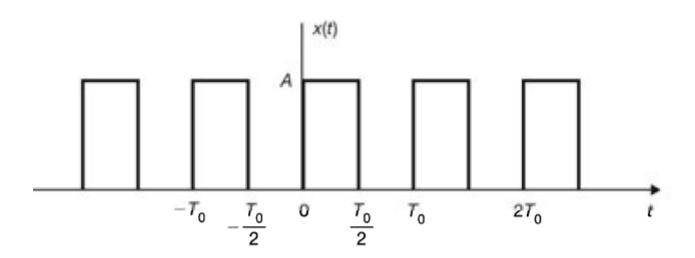
Substituting  $X[-\ell] = X^*[\ell]$ , we get the coefficients of the **trigonometric form** of FS representation for a real-valued signal x(t) as

$$B[k] = (X[k] + X^*[k])$$
 and  $A[k] = j(X[k] - X^*[k])$ 

If x(t) is an even (resp. odd) signal, then the trigonometric form of FS expansion will consist of only cosine (resp. sine) terms.

Square Wave Example 1  

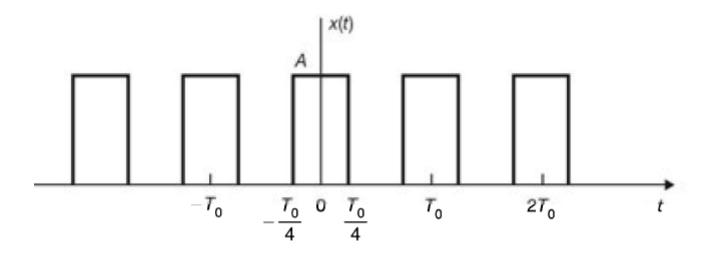
$$X[0] = A/2, X[2m] = 0, X[2m + 1] = A/[j(2m + 1)\pi]$$
  
 $A[k] = ?B[k] = ?$ 



Square Wave Example 2  

$$X[0] = A/2, X[2m] = 0, X[2m + 1] = (-1)^m A/[(2m + 1) \pi]$$
  
 $A[k] = ?B[k] = ?$ 

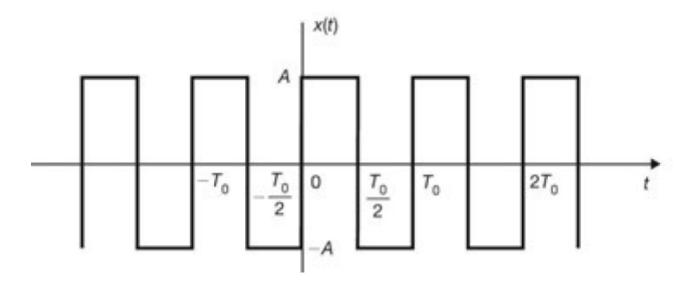
Can we get the coefficients directly from Example 1?



Square Wave Example 3  

$$X[0] = 0, X[2m] = 0, X[2m + 1] = 2A/[j(2m + 1)\pi]$$
  
 $A[k] = ?B[k] = ?$ 

Can we get the coefficients directly from Example 1?



### Amplitude and Phase Spectra of a CT Periodic Signal

The FS coefficients can be written in polar form as

$$X[k] = |X[k]|e^{j\theta_k},$$

where |X[k]| and  $\theta_k$  as functions of discrete frequency  $k\omega_0$  are called the **magnitude spectrum** and **phase spectrum**, respectively.

Recall that for a real-valued signal x(t) we have

$$X[-k] = X^*[k]$$
, which implies that

# Amplitude and Phase Spectra of a CT Periodic Signal

- (i)  $|X[-k]| = |X^*[k]| = |X[k]|$ , i.e., the magnitude spectrum is an even function of the discrete index k.
- spectrum is an even function of the discrete index k.

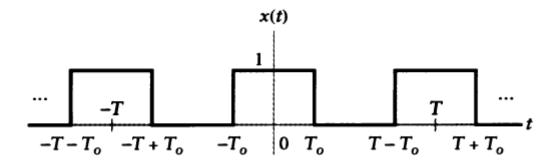
   (ii)  $\theta_{-k} = arg\{X^*[k]\} = -arg\{X[k]\} = -\theta_k$ , i.e., the phase spectrum is an odd function of the discrete index k.

For the above reason of symmetry, when dealing with real-valued signals x(t), one can ignore the negative frequencies altogether.

• If the signal x(t) is complex-valued, then the above symmetry will not hold and one has to consider both positive and negative frequencies separately.

### **Example: Periodic Rectangular Pulses**

For 
$$x(t)$$
 shown below, we have  $\omega_0 = \frac{2\pi}{T}$ 



#### **Example: Periodic Rectangular Pulses**

It can be shown that 
$$X[0] = \frac{2T_0}{T}$$
 and for all  $k \neq 0$ , we have

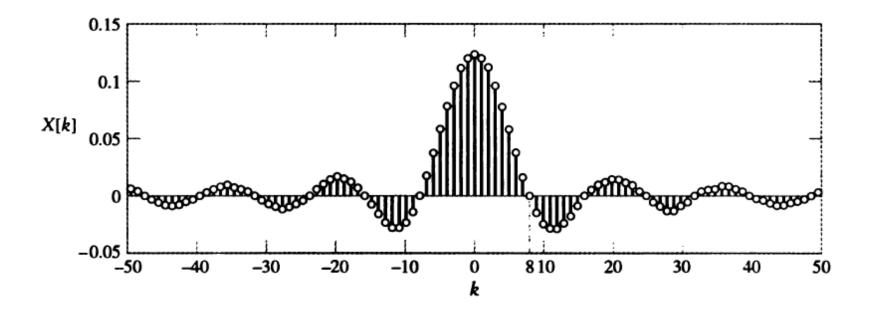
$$X[k] = \frac{2\sin(k\omega_0 T_0)}{k\omega_0 T} = \frac{2T_0}{T} \frac{\sin(k\omega_0 T_0)}{k\omega_0 T_0}$$
$$= \frac{2T_0}{T} \frac{\sin(k2\pi T_0/T)}{k2\pi T_0/T} = \frac{2T_0}{T} \operatorname{sinc}\left(k\frac{2T_0}{T}\right),$$

where\_

$$sinc(x) = \frac{sin(\pi x)}{\pi x}$$

# **Example: Periodic Rectangular Pulses**

Can you sketch the amplitude and phase spectra?



### Output Computation of an LTI System using FS

Suppose a continuous-time periodic signal x(t) is applied as input to an LTI system with impulse response h(t).

From time-domain analysis, we know that the output y(t) can be obtained by taking a convolution of the input x(t) and impulse response h(t).

An alternate method that does not require convolution:

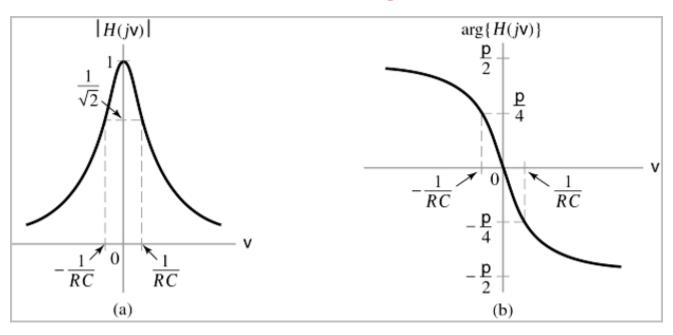
- Obtain the FS spectrum of x(t). Let  $x(t) \overset{CTFS}{\longleftrightarrow} X[k]$ .
- Obtain the frequency response of the LTI system  $H(j\omega)$ .
- Obtain the FS spectrum of y(t) from that of x(t) as  $Y[k] = H(jk\omega_0)X[k]$ .
- Obtain the output  $Y[k] \stackrel{CTFS}{\longleftrightarrow} y(t)$ .

Recall that for the RC circuit, the frequency response is given by

$$H(j\omega) = \frac{\frac{1}{RC}}{j\omega + \frac{1}{RC}}.$$

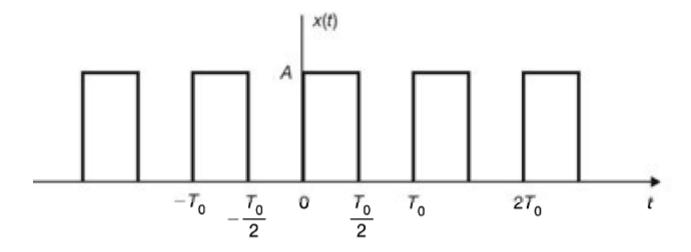
The magnitude and phase responses are given by (plotted on the next slide)

$$|H(j\omega)| = \frac{1/RC}{\sqrt{\omega^2 + (1/RC)^2}}$$
 and  $\arg\{H(j\omega)\} = -\tan^{-1}(\omega RC)$ .



Suppose a square wave as shown below is applied as input x(t).

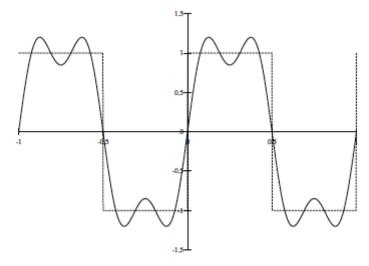
• We had obtained the FS coefficients X[k] earlier.



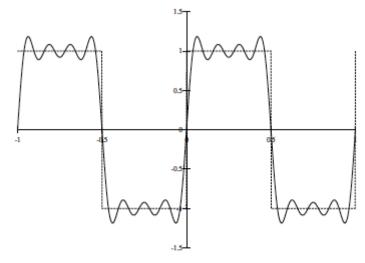
The output is then given by

$$y(t) = \sum_{k=-\infty}^{\infty} Y[k]e^{j\omega_0 t} = \sum_{k=-\infty}^{\infty} H(jk\omega_0)X[k]e^{j\omega_0 t}$$
$$\approx \sum_{k=-K}^{K} H(jk\omega_0)X[k]e^{j\omega_0 t}$$

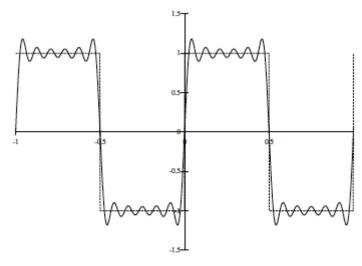
Can you sketch the output (at least approximately)?



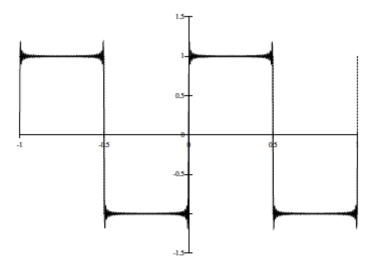
Fourier series truncated after the 3rd harmonic components



Fourier series truncated after the 7th harmonic components



Fourier series truncated after the 11th harmonic components



Fourier series truncated after the 101st harmonic components

# **CTFS Properties**

### Linearity

Let x and y be two periodic functions with the same period. If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} a_k$  and  $y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} b_k$ , then

$$\alpha x(t) + \beta y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} \alpha a_k + \beta b_k$$

where  $\alpha$  and  $\beta$  are complex constants.

That is, a linear combination of functions produces the same linear combination of their Fourier series coefficients.

# Time Shifting (Translation)

Let x denote a periodic function with period T and the corresponding frequency  $\omega_0 = 2\pi/T$ . If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_k$ , then

$$x(t-t_0) \stackrel{\text{CTFS}}{\longleftrightarrow} e^{-jk\omega_0 t_0} c_k = e^{-jk(2\pi/T)t_0} c_k,$$

where  $t_0$  is a real constant.

In other words, time shifting a periodic function changes the argument (but not magnitude) of its Fourier series coefficients.

# Frequency Shifting (Modulation)

Let x denote a periodic function with period T and the corresponding frequency  $\omega_0 = 2\pi/T$ . If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_k$ , then

$$e^{jM(2\pi/T)t}x(t) = e^{jM\omega_0t}x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_{k-M},$$

where M is an integer constant.

In other words, multiplying a periodic function by  $e^{jM\omega_0t}$  shifts the Fourier-series coefficient sequence.

### Time Reversal (Reflection)

Let x denote a periodic function with period T and the corresponding frequency  $\omega_0 = 2\pi/T$ . If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_k$ , then

$$x(-t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_{-k}$$
.

That is, time reversal of a function results in a time reversal of its Fourier series coefficients.

### Conjugation

■ For a *T*-periodic function *x* with Fourier series coefficient sequence *c*, the following property holds:

$$x^*(t) \stackrel{\text{CTFS}}{\longleftrightarrow} c_{-k}^*$$

In other words, conjugating a function has the effect of time reversing and conjugating the Fourier series coefficient sequence.

#### **Periodic Convolution**

Let x and y be two periodic functions with the same period T. If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} a_k$  and  $y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} b_k$ , then

$$x \circledast y(t) \stackrel{\texttt{CTFS}}{\longleftrightarrow} Ta_k b_k.$$

In other words, periodic convolution of two functions corresponds to the multiplication (up to a scale factor) of their Fourier-series coefficient sequences.

### **Multiplication**

Let x and y be two periodic functions with the same period. If  $x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} a_k$  and  $y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} b_k$ , then

$$x(t)y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} \sum_{n=-\infty}^{\infty} a_n b_{k-n}$$

- As we shall see later, the above summation is the DT convolution of a and b.
- In other words, the multiplication of two periodic functions corresponds to the DT convolution of their corresponding Fourier-series coefficient sequences.

### **Differentiation in Time**

We have

$$\frac{d}{dt}x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} jk\omega_0 X[k].$$

# **Scaling in Time**

We have

$$x(at) \stackrel{\text{CTFS}}{\longleftrightarrow} X[k],$$

but the spacing between the Fourier coefficients changes from  $\omega_0$  to  $a\omega_0$ .

### **Even and Odd Symmetry**

■ For a *T*-periodic function *x* with Fourier series coefficient sequence *c*, the following properties hold:

$$x$$
 is even  $\Leftrightarrow c$  is even; and  $x$  is odd  $\Leftrightarrow c$  is odd.

In other words, the even/odd symmetry properties of x and c always match.

#### **Real-Valued Functions**

A function x is real if and only if its Fourier series coefficient sequence c satisfies

$$c_k = c_{-k}^*$$
 for all  $k$ 

(i.e., c is conjugate symmetric).

Thus, for a real-valued function, the negative-indexed Fourier series coefficients are *redundant*, as they are completely determined by the nonnegative-indexed coefficients.

#### **Real-Valued Functions**

From properties of complex numbers, one can show that  $c_k = c_{-k}^*$  is equivalent to

$$|c_k| = |c_{-k}|$$
 and  $\arg c_k = -\arg c_{-k}$ 

(i.e.,  $|c_k|$  is *even* and  $\arg c_k$  is *odd*).

Note that x being real does not necessarily imply that c is real.

#### Parseval's Relation

A function x and its Fourier series coefficient sequence a satisfy the following relationship:

$$\frac{1}{T} \int_{T} |x(t)|^{2} dt = \sum_{k=-\infty}^{\infty} |a_{k}|^{2}.$$

- The above relationship is simply stating that the amount of energy in x (i.e.,  $\frac{1}{T} \int_T |x(t)|^2 dt$ ) and the amount of energy in the Fourier series coefficient sequence a (i.e.,  $\sum_{k=-\infty}^{\infty} |a_k|^2$ ) are equal.
- In other words, the transformation between a function and its Fourier series coefficient sequence preserves energy.

### **Other Properties**

- For a T-periodic function x with Fourier-series coefficient sequence c, the following properties hold:
  - $\mathbf{1}$   $c_0$  is the average value of x over a single period;
  - $\mathbf{z}$  x is real and even  $\Leftrightarrow$  c is real and even; and
  - $\mathbf{S}$  is real and odd  $\Leftrightarrow c$  is purely imaginary and odd.

# **Table of CTFS Properties**

$$x(t) \stackrel{\text{CTFS}}{\longleftrightarrow} a_k$$
 and  $y(t) \stackrel{\text{CTFS}}{\longleftrightarrow} b_k$ 

Property	Time Domain	Fourier Domain
Linearity	$\alpha x(t) + \beta y(t)$	$\alpha a_k + \beta b_k$
Translation	$x(t-t_0)$	$e^{-jk(2\pi/T)t_0}a_k$
Modulation	$e^{jM(2\pi/T)t}x(t)$	$a_{k-M}$
Reflection	x(-t)	$a_{-k}$
Conjugation	$x^*(t)$	$a_{-k}^*$
Periodic Convolution	$x \circledast y(t)$	$Ta_kb_k$
Multiplication	x(t)y(t)	$\sum_{n=-\infty}^{\infty} a_n b_{k-n}$

# Table of CTFS Properties 2

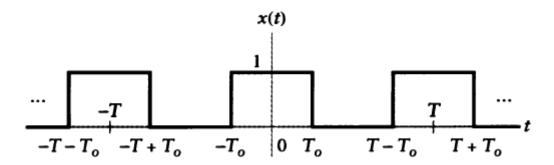
Property	
Parseval's Relation	$\frac{1}{T} \int_T  x(t) ^2 dt = \sum_{k=-\infty}^{\infty}  a_k ^2$
Even Symmetry	$x$ is even $\Leftrightarrow a$ is even
Odd Symmetry	$x$ is odd $\Leftrightarrow a$ is odd
Real / Conjugate Symmetry	$x$ is real $\Leftrightarrow a$ is conjugate symmetric

**Continuous Time Fourier Transform (CTFT)** 

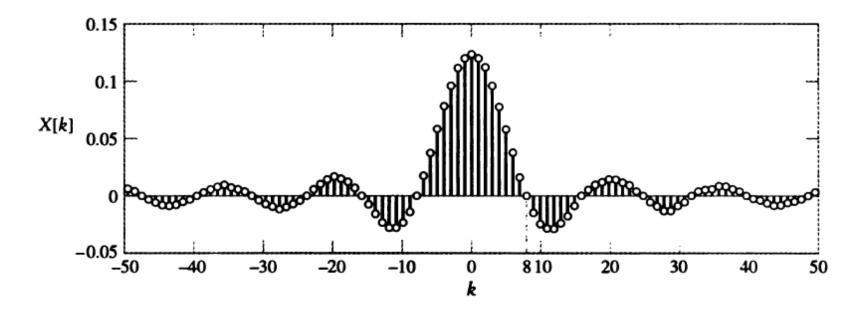
#### **Motivation for CTFT**

- The (CT) Fourier series provide an extremely useful representation for periodic functions.
- Often, however, we need to deal with functions that are not periodic.
- A more general tool than the Fourier series is needed in this case.
- The (CT) Fourier transform can be used to represent both periodic and aperiodic functions.
- Since the Fourier transform is essentially derived from Fourier series through a limiting process, the Fourier transform has many similarities with Fourier series.

Recall the example of periodic rectangular pulses x(t) discussed earlier.

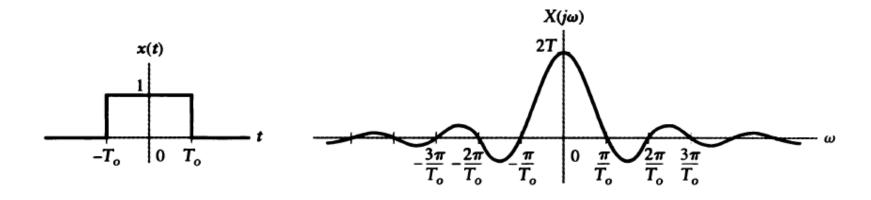


Recall the corresponding CTFS coefficients X[k] obtained earlier.



If we make the period T grow very large, then x(t) becomes a non-periodic signal and the fundamental frequency  $\omega_0 = \frac{2\pi}{T}$  reduces to  $\Delta\omega \approx 0$ .

• Then, it is intuitively clear that in the limit  $T \to \infty$  the discrete-index function X[k] becomes a function  $X(\omega)$  of continuous frequency  $\omega$ .



The FS coefficient X[k] becomes

$$X[k] = \frac{2T_0}{T} \frac{\sin(k2\pi T_0/T)}{k2\pi T_0/T} = \frac{2T_0}{T} \frac{\sin(k\Delta\omega T_0)}{k\Delta\omega T_0} = \frac{1}{2\pi} \frac{2\sin(k\Delta\omega T_0)}{k\Delta\omega} \Delta\omega,$$

and in the limit  $T \to \infty$ , we have  $\Delta \omega \to d\omega$  and  $k\Delta \omega \to \omega$ , and the FS representation of x(t) becomes

$$x(t) = \sum_{k=-\infty}^{\infty} X[k] e^{jk\omega_0 t} \xrightarrow[T \to \infty]{} \frac{1}{2\pi} \int_{-\infty}^{\infty} \underbrace{\frac{2\sin(\omega T_0)}{\omega}}_{X(\omega)} e^{j\omega t} d\omega$$

In general, for any periodic signal x(t) with period T, we have

$$x(t) = \sum_{k=-\infty}^{\infty} X[k] e^{jk\omega_0 t} = \sum_{k=-\infty}^{\infty} \left(\frac{1}{T} \int_{-T/2}^{T/2} x(t) e^{-jk(2\pi/T)t} dt\right) e^{jk\omega_0 t}$$

$$\xrightarrow[T\to\infty]{} \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt\right) e^{j\omega t} d\omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{j\omega t} d\omega,$$

where 
$$X(\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t}dt$$
 is called the **Fourier**

**Transform** of the **non-periodic** signal x(t).

#### Dirichlet Conditions for the Existence of CTFT

The foregoing equation representing x(t) as an integration (weighted sum) of a continuum of frequency components and the equation representing the corresponding Fourier Transform  $X(\omega)$  are valid at all points, except at discontinuities, only if the **Dirichlet** conditions are satisfied:

• x(t) is absolutely integrable, i.e., if

$$\int_{-\infty}^{\infty} |x(t)| dt \le M_x < \infty.$$

• x(t) has a finite number of maxima, minima and discontinuities in any finite interval, and the size of each discontinuity is finite.

#### Dirichlet Conditions for the Existence of CTFT

If the FT exists, then we say that x(t) and  $X(\omega)$  form a **FT pair** and denote this relationship as

$$x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega).$$

Example: Consider the signal  $x(t) = e^{-at}u(t)$ . Does this signal have a Fourier Transform? If yes, find the FT.

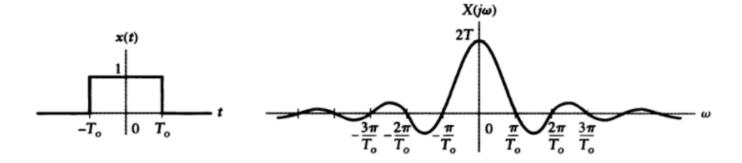
• Solution: FT exists only if a > 0. If FT exists, then we have  $X(\omega) = \frac{1}{a + j\omega}$   $\Rightarrow |X(\omega)| = \left(a^2 + \omega^2\right)^{-1/2} \text{ and } \arg\{X(\omega)\} = -\tan^{-1}(\omega/a).$ 

### CTFT Example

Example: Consider the (single) rectangular pulse again. We have

$$X(\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t}dt = \int_{-T_0}^{T_0} e^{-j\omega t}dt = \begin{cases} \frac{2}{\omega}\sin(\omega T_0) & \omega \neq 0\\ 2T_0 & \omega = 0. \end{cases}$$

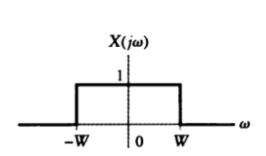
- Write  $X(\omega)$  in terms of sinc(.) function.
- Sketch the magnitude and phase spectra.

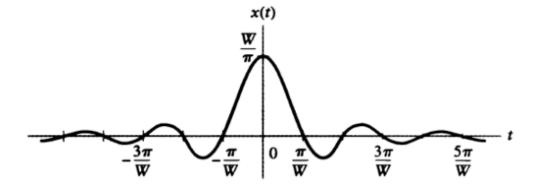


### **Inverse Fourier Transform Example**

**Example:** Consider a rectangular spectrum in the frequency domain shown below. We have

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{j\omega t} d\omega = \int_{-W}^{W} e^{j\omega t} d\omega = \begin{cases} \frac{\sin(Wt)}{\pi t} & t \neq 0 \\ W/\pi & t = 0. \end{cases}$$





### **Fourier Transforms of Some Important Signals**

Problem: Find the Fourier Transform of each of the following signals: (i) A unit impulse function, (ii) A unit DC signal, (iii) A unit step function, (iv)  $x(t) = e^{j\omega_0 t}$ , (v)  $x(t) = \cos(\omega_0 t)$ , (vi)  $x(t) = \sin(\omega_0 t)$ 

**Solution:** (i) 1, (ii)  $2\pi\delta(\omega)$  (requires duality property, to be discussed later), (iii)  $\pi\delta(\omega) + (1/j\omega)$  (requires differentiation property, to be discussed later), (iv)  $2\pi\delta(\omega-\omega_0)$ , (v)  $\pi[\delta(\omega-\omega_0)+\delta(\omega+\omega_0)]$ , (vi)  $j\pi[\delta(\omega-\omega_0)-\delta(\omega+\omega_0)]$ .

# **CTFT Properties**

### Linearity

■ If  $x_1(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_1(\omega)$  and  $x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_2(\omega)$ , then

$$a_1x_1(t) + a_2x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} a_1X_1(\omega) + a_2X_2(\omega),$$

where  $a_1$  and  $a_2$  are arbitrary complex constants.

This is known as the linearity property of the Fourier transform.

### Time Shifting (Translation)

■ If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$x(t-t_0) \stackrel{\text{CTFT}}{\longleftrightarrow} e^{-j\omega t_0} X(\omega),$$

where  $t_0$  is an arbitrary real constant.

■ This is known as the **translation** (or **time-domain shifting**) **property** of the Fourier transform.

### Frequency Shifting (Modulation)

■ If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$e^{j\omega_0 t}x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega - \omega_0),$$

where  $\omega_0$  is an arbitrary real constant.

■ This is known as the modulation (or frequency-domain shifting) property of the Fourier transform.

### Time and Frequency Scaling

■ If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$x(at) \stackrel{\text{CTFT}}{\longleftrightarrow} \frac{1}{|a|} X\left(\frac{\omega}{a}\right),$$

where *a* is an arbitrary nonzero real constant.

■ This is known as the dilation (or time/frequency-scaling) property of the Fourier transform.

### Conjugation

■ If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$x^*(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X^*(-\omega).$$

■ This is known as the **conjugation property** of the Fourier transform.

### **Duality**

- This is known as the duality property of the Fourier transform.
- This property follows from the high degree of symmetry in the forward and inverse Fourier transform equations, which are respectively given by

$$X(\lambda) = \int_{-\infty}^{\infty} x(\theta) e^{-j\theta\lambda} d\theta \quad \text{and} \quad x(\lambda) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\theta) e^{j\theta\lambda} d\theta.$$

That is, the forward and inverse Fourier transform equations are identical except for a *factor of*  $2\pi$  and *different sign* in the parameter for the exponential function.

#### Convolution

■ If  $x_1(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_1(\omega)$  and  $x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_2(\omega)$ , then

$$x_1 * x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_1(\omega) X_2(\omega).$$

- This is known as the convolution (or time-domain convolution) property of the Fourier transform.
- In other words, a convolution in the time domain becomes a multiplication in the frequency domain.
- This suggests that the Fourier transform can be used to avoid having to deal with convolution operations.

### **Multiplication**

If  $x_1(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_1(\omega)$  and  $x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X_2(\omega)$ , then

$$x_1(t)x_2(t) \stackrel{\text{CTFT}}{\longleftrightarrow} \frac{1}{2\pi}X_1 * X_2(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X_1(\theta)X_2(\omega - \theta)d\theta.$$

- This is known as the (time-domain) multiplication (or frequency-domain convolution) property of the Fourier transform.
- In other words, multiplication in the time domain becomes convolution in the frequency domain (up to a scale factor of  $2\pi$ ).
- Do not forget the factor of  $\frac{1}{2\pi}$  in the above formula!
- This property of the Fourier transform is often tedious to apply (in the forward direction) as it turns a multiplication into a convolution.

#### Differentiation in Time Domain

■ If  $x(t) \stackrel{\mathsf{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$\frac{dx(t)}{dt} \stackrel{\text{CTFT}}{\longleftrightarrow} j\omega X(\omega).$$

- This is known as the (time-domain) differentiation property of the Fourier transform.
- Differentiation in the time domain becomes multiplication by  $j\omega$  in the frequency domain.
- Of course, by repeated application of the above property, we have that  $\left(\frac{d}{dt}\right)^n x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} (j\omega)^n X(\omega)$ .
- The above suggests that the Fourier transform might be a useful tool when working with differential (or integro-differential) equations.

### **Differentiation in Frequency Domain**

■ If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$tx(t) \stackrel{\text{CTFT}}{\longleftrightarrow} j \frac{d}{d\omega} X(\omega).$$

This is known as the <u>frequency-domain differentiation property</u> of the Fourier transform.

### Integration in Time Domain

If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

$$\int_{-\infty}^{t} x(\tau) d\tau \stackrel{\text{CTFT}}{\longleftrightarrow} \frac{1}{j\omega} X(\omega) + \pi X(0) \delta(\omega).$$

- This is known as the (time-domain) integration property of the Fourier transform.
- Whereas differentiation in the time domain corresponds to *multiplication* by  $j\omega$  in the frequency domain, integration in the time domain is associated with *division* by  $j\omega$  in the frequency domain.
- Since integration in the time domain becomes division by  $j\omega$  in the frequency domain, integration can be easier to handle in the frequency domain.
- The above property suggests that the Fourier transform might be a useful tool when working with integral (or integro-differential) equations.

### **Even and Odd Symmetry**

■ For a function *x* with Fourier transform *X*, the following assertions hold:

x is even  $\Leftrightarrow X$  is even; and x is odd  $\Leftrightarrow X$  is odd.

In other words, the forward and inverse Fourier transforms preserve even/odd symmetry.

#### **Real-Valued Functions**

 $\blacksquare$  A function x is real if and only if its Fourier transform X satisfies

$$X(\omega) = X^*(-\omega)$$
 for all  $\omega$ 

(i.e., X is *conjugate symmetric*).

- Thus, for a real-valued function, the portion of the graph of a Fourier transform for negative values of frequency ω is redundant, as it is completely determined by symmetry.
- From properties of complex numbers, one can show that  $X(\omega) = X^*(-\omega)$  is equivalent to

$$|X(\omega)| = |X(-\omega)|$$
 and  $\arg X(\omega) = -\arg X(-\omega)$ 

(i.e.,  $|X(\omega)|$  is *even* and  $\arg X(\omega)$  is *odd*).

Note that x being real does not necessarily imply that X is real.

#### Parseval's Relation

- Recall that the energy of a function x is given by  $\int_{-\infty}^{\infty} |x(t)|^2 dt$ .
- If  $x(t) \stackrel{\text{CTFT}}{\longleftrightarrow} X(\omega)$ , then

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

(i.e., the energy of x and energy of X are equal up to a factor of  $2\pi$ ).

- This relationship is known as Parseval's relation.
- Since energy is often a quantity of great significance in engineering applications, it is extremely helpful to know that the Fourier transform preserves energy (up to a scale factor).

**Other Properties** 

**Time-Bandwidth Product** 

## **Table of CTFT Properties**

Property	Time Domain	Frequency Domain
Linearity	$a_1x_1(t) + a_2x_2(t)$	$a_1X_1(\omega) + a_2X_2(\omega)$
Time-Domain Shifting	$x(t-t_0)$	$e^{-j\omega t_0}X(\omega)$
Frequency-Domain Shifting	$e^{j\omega_0 t}x(t)$	$X(\omega - \omega_0)$
Time/Frequency-Domain Scaling	x(at)	$\frac{1}{ a }X\left(\frac{\omega}{a}\right)$
Conjugation	$x^*(t)$	$X^*(-\omega)$
Duality	X(t)	$2\pi x(-\omega)$
Time-Domain Convolution	$x_1 * x_2(t)$	$X_1(\omega)X_2(\omega)$
Time-Domain Multiplication	$x_1(t)x_2(t)$	$\frac{1}{2\pi}X_1 * X_2(\omega)$
Time-Domain Differentiation	$\frac{d}{dt}x(t)$	$j\omega X(\omega)$
Frequency-Domain Differentiation	tx(t)	$j\frac{d}{d\omega}X(\omega)$
Time-Domain Integration	$\int_{-\infty}^{t} x(\tau) d\tau$	$\frac{1}{j\omega}X(\omega) + \pi X(0)\delta(\omega)$

### **Table of CTFT Properties**

**Property** 

Parseval's Relation  $\int_{-\infty}^{\infty} |x(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X(\omega)|^2 d\omega$ 

Even Symmetry x is even  $\Leftrightarrow X$  is even

Odd Symmetry x is odd  $\Leftrightarrow X$  is odd

Real / Conjugate Symmetry x is real  $\Leftrightarrow X$  is conjugate symmetric

### **References:**

[1] Simon Haykin and Barry Van Veen, Signals and Systems, Second Edition, John Wiley and Sons, 2003.

[2] Lecture Notes by Michael D. Adams.

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[3] Lecture Notes by Richard Baraniuk.

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