Signals & Systems

Spring 2019

https://sites.google.com/site/ntusands/

https://ceiba.ntu.edu.tw/1072EE2011_04

Yu-Chiang Frank Wang 王鈺強, Associate Professor Dept. Electrical Engineering, National Taiwan University

Ch. 4 Continuous-Time Fourier Transform

- Sec. 4.1 Representation of Aperiodic Signals:

 The Continuous-Time Fourier Transform
- Sec. 4.2 The Fourier Transform for Periodic Signals
- Sec. 4.3 Properties of the Continuous-Time Fourier Transform
- Sec. 4.4 The Convolution Property
- Sec. 4.5 The Multiplication Property
- Sec. 4.6 Tables of Fourier Properties and of Basic Fourier Transform Pairs
- Sec. 4.7 Systems Characterized by Linear Constant Coefficient Differential Equations

Sect. 4.3 Properties of CTFT

Integration

$$x(t) \stackrel{\mathcal{F}}{\longleftrightarrow} X(j\omega)$$

$$\Rightarrow \int_{-\infty}^{t} x(\tau)d\tau \stackrel{\mathcal{F}}{\longleftrightarrow} \frac{1}{j\omega} X(j\omega) \left\{ \pi X(j0)\delta(\omega) \right\}^{\gamma}.$$

$$y(t) = \int_{-\infty}^{t} x(\tau)d\tau \implies \frac{dy(t)}{dt} = x(t)$$

$$Y(j\omega) = \frac{1}{j\omega}X(j\omega)$$
 Indeterminate at ω =0 as a result of the differentiation that destroys the dc component of $y(t)$

The value at ω =0 is modified by including an impulse in the transform:

$$Y(j\omega) = \frac{1}{j\omega}X(j\omega) + \pi X(j0)\delta(\omega)$$

$$x(t) \neq y(t) \stackrel{FT}{\longleftarrow} X(j\omega)U(j\omega) = X(j\omega)T \frac{1}{j\omega} + \pi \delta(\omega)$$

$$\begin{array}{c} \begin{array}{c} \begin{array}{c} \begin{array}{c} \\ \\ \end{array} \end{array} \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array} \begin{array}{c} \\ \end{array}$$

$$\frac{d}{d\tau} Sgn(t) \stackrel{FT}{\longleftarrow} j\omega S(j\omega)$$

$$2 S(t) \stackrel{FT}{\longleftarrow} j\omega S(j\omega) - 2 S(t) \stackrel{FT}{\longleftarrow} 1$$

$$S(j\omega) = \frac{2}{j\omega}$$

Sect. 4.3 Convolution Property

$$y(t) = x(t) * h(t) \longleftrightarrow Y(j\omega) = X(j\omega)H(j\omega)$$

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

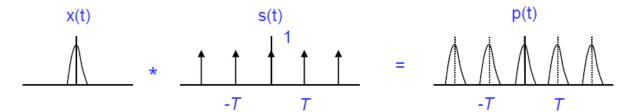
$$\Rightarrow Y(j\omega) = F\{y(t)\} = \int_{-\infty}^{+\infty} \left[\int_{-\infty}^{+\infty} x(\tau)h(t-\tau)d\tau \right] e^{-j\omega t}dt$$

$$= \int_{-\infty}^{+\infty} x(\tau) \left[\int_{-\infty}^{+\infty} h(t-\tau)e^{-j\omega t}dt \right] d\tau$$

$$= \int_{-\infty}^{+\infty} x(\tau) \left[e^{-j\omega \tau} H(j\omega) \right] d\tau$$

$$= H(j\omega) \int_{-\infty}^{+\infty} x(\tau)e^{-j\omega \tau}d\tau = H(j\omega)X(j\omega)$$

Fourier Transform of an Arbitrary Periodic Signal



By convolution,

$$p(t) = x(t) * s(t) = x(t) * \sum_{k=-\infty}^{\infty} \delta(t - kT) = \sum_{k=-\infty}^{\infty} x(t - kT)$$

$$p(t) = x(t) * s(t) = x(t) * \sum_{k=-\infty}^{\infty} \delta(t - kT) = \sum_{k=-\infty}^{\infty} x(t - kT)$$
From Example 3.8 or Example 4.8, we have
$$s(t) = \sum_{k=-\infty}^{\infty} \delta(t - kT) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_0 t} = \frac{1}{T} \sum_{k=-\infty}^{\infty} e^{jk\omega_0 t}$$
That is, we have two expressions for a periodic impulse train.

Express them in terms of FT, we have

$$S(j\omega) = \sum_{k=-\infty}^{\infty} e^{-jk\omega T} = \frac{1}{T} \sum_{k=-\infty}^{\infty} 2\pi \delta(\omega - k\omega_0)$$
wherefore

Therefore,

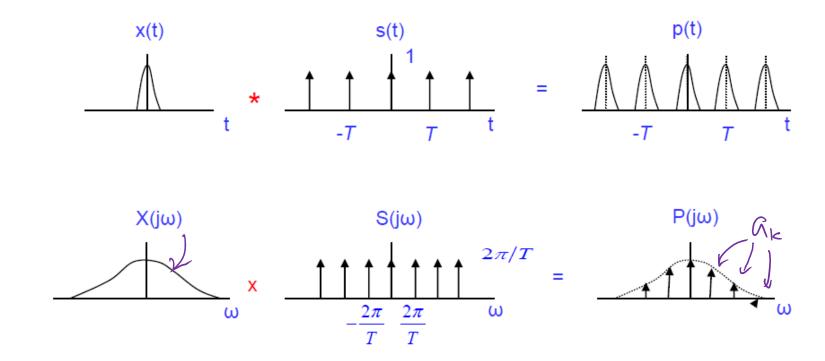
$$P(j\omega) = X(j\omega)S(j\omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} X(j\omega)\delta(\omega - k\omega)$$

se train.

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

$$a_k = \frac{1}{T} \int_{-T/2}^{+T/2} x(t) e^{-jk\omega_0 t} dt$$

- Fourier Transform of an Arbitrary Periodic Signal
- We see that the FT of periodic function consists of impulses in frequency at multiples of the fundamental frequency.
- Thus, CT periodic signals can be represented by a countably infinite number of complex exponentials.



Fourier Transform of an Arbitrary Periodic Signal

• Recall that
$$X(j\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t}dt = \int_{-T/2}^{T/2} x(t)e^{-j\omega t}dt$$

• Therefore,
$$\frac{X(jk\omega_0)}{T} = \frac{1}{T} \int_{-T/2}^{T/2} x(t)e^{-jk\omega_0 t} dt = a_k$$
 (=Fourier series coefficient)

which is *exactly the same equation* as Eqn. (4.10). Therefore, for an arbitrary CT periodic signal, it's FT consists of impulses (located at the harmonic frequencies) whose areas are proportional to the FS coeff.

We can also conclude that the FT of a periodic signal is related to its FS coefficients a_k by

$$P(j\omega) = \sum_{k=-\infty}^{\infty} 2\pi a_k \delta(\omega - k\omega_0),$$

which is exactly the same as Eqn. (4.22) (i.e., $X(j\omega)$).

Sect. 4.4 Multiplication Property

Multiplication Property

$$r(t) = x(t)y(t) \longleftrightarrow R(j\omega) = \frac{1}{2\pi}X(j\omega)*Y(j\omega)$$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X(j\phi) e^{j\phi t} d\phi, \quad y(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} Y(j\eta) e^{j\eta t} d\eta$$

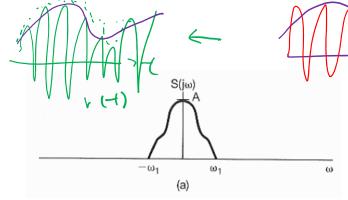
$$r(t) = x(t)y(t) = \frac{1}{(2\pi)^2} \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} X(j\phi) Y(j\eta) e^{j(\phi+\eta)t} d\eta d\phi \quad \leftarrow \eta = \omega - \phi$$

$$= (\frac{1}{2\pi}) \left(\frac{1}{2\pi} \right) \int_{-\infty}^{+\infty} \left[\int_{-\infty}^{+\infty} X(j\phi) Y(j\omega - j\phi) d\phi \right] e^{j\omega t} d\omega$$
inverse Fourier transform

Multiplication in time corresponds to convolution in frequency.

Multiply one signal by another in time is referred to as amplitude modulation.

• Example 4.21 Amplitude Modulation (AM)

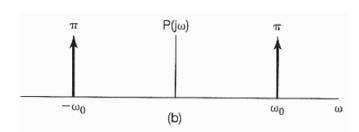


A band-limited signal

$$s(t) \stackrel{\mathcal{F}}{\longleftrightarrow} S(j\omega)$$

$$p(t) \stackrel{\mathcal{F}}{\longleftrightarrow} P(j\omega)$$

$$r(t) = s(t)p(t)$$



$$R(j\omega) = \frac{1}{2\pi} \left[S(j\omega) * P(j\omega) \right]$$

$$A/2 - \omega_0$$

$$(-\omega_0 - \omega_1) (-\omega_0 + \omega_1)$$

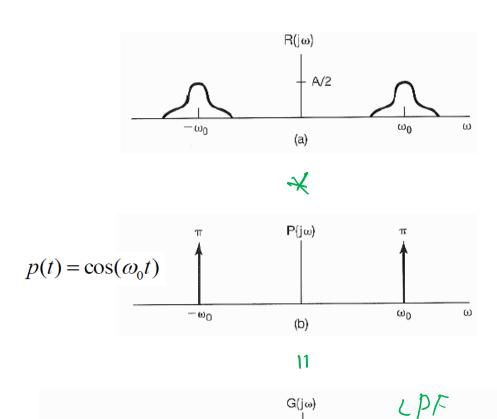
$$(c)$$

$$p(t) = \cos(\omega_0 t) = \frac{1}{2} \left[e^{j\omega_0 t} + e^{-j\omega_0 t} \right]$$

$$P(j\omega) = \pi \delta(\omega - \omega_0) + \pi \delta(\omega + \omega_0)$$

$$\begin{split} R(j\omega) &= \frac{1}{2\pi} \, S(j\omega) * P(j\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S(j\theta) P(j(\omega - \theta)) d\theta \\ &= \frac{1}{2} S(j(\omega - \omega_0)) + \frac{1}{2} S(j(\omega + \omega_0)) \end{split}$$

Example 4.22 Demodulation



(c)

 $-2\omega_0$

$$G(j\omega) = \frac{1}{2\pi} [R(j\omega) * P(j\omega)]$$

Ch. 4 Continuous-Time Fourier Transform

- Sec. 4.1 Representation of Aperiodic Signals:

 The Continuous-Time Fourier Transform
- Sec. 4.2 The Fourier Transform for Periodic Signals
- Sec. 4.3 Properties of the Continuous-Time Fourier Transform
- Sec. 4.4 The Convolution Property
- Sec. 4.5 The Multiplication Property
- Sec. 4.6 Tables of Fourier Properties and of Basic Fourier Transform Pairs*
- Sec. 4.7 Systems Characterized by Linear Constant Coefficient Differential Equations

A Useful Class of CT LTI Systems

$$x(t) \longrightarrow \text{LTI System} \longrightarrow y(t)$$

$$a_N \frac{d^N y(t)}{dt^N} + a_{N-1} \frac{d^{N-1} y(t)}{dt^{N-1}} + \dots + a_1 \frac{dy(t)}{dt} + a_0 y(t)$$

$$= b_M \frac{d^M x(t)}{dt^M} + b_{M-1} \frac{d^{M-1} x(t)}{dt^{M-1}} + \dots + b_1 \frac{dx(t)}{dt} + b_0 x(t)$$

$$\sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k} = \sum_{k=0}^M b_k \frac{d^k x(t)}{dt^k}$$

$$F\left\{\sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k}\right\} = F\left\{\sum_{k=0}^M b_k \frac{d^k x(t)}{dt^k}\right\}$$

A Useful Class of CT LTI Systems (cont'd)

$$\sum_{k=0}^{N} a_k F\left\{\frac{d^k y(t)}{dt^k}\right\} = \sum_{k=0}^{M} b_k F\left\{\frac{d^k x(t)}{dt^k}\right\}$$

$$\sum_{k=0}^{N} a_k (j\omega)^k Y(j\omega) = \sum_{k=0}^{M} b_k (j\omega)^k X(j\omega)$$

$$Y(j\omega) \left[\sum_{k=0}^{N} a_k (j\omega)^k\right] = X(j\omega) \left[\sum_{k=0}^{M} b_k (j\omega)^k\right]$$

$$\Rightarrow H(j\omega) = \frac{Y(j\omega)}{X(j\omega)} = \frac{\sum_{k=0}^{M} b_k (j\omega)^k}{\sum_{k=0}^{N} a_k (j\omega)^k} = \frac{b_M (j\omega)^M + \dots + b_1 (j\omega) + b_0}{a_N (j\omega)^N + \dots + a_1 (j\omega) + a_0}$$

• Example 4.24

$$\frac{dy(t)}{dt} + ay(t) = x(t)$$

$$j\omega Y(j\omega) + aY(j\omega) = X(j\omega)$$

$$(j\omega + a)Y(j\omega) = X(j\omega)$$

$$\Rightarrow H(j\omega) = \frac{Y(j\omega)}{X(j\omega)} = \frac{1}{j\omega + a}$$

$$\Rightarrow h(t) = e^{-at}u(t)$$

Example 4.25

$$\frac{d^2y(t)}{dt^2} + 4\frac{dy(t)}{dt} + 3y(t) = \frac{dx(t)}{dt} + 2x(t)$$

$$\Rightarrow H(j\omega) = \frac{(j\omega) + 2}{(j\omega)^2 + 4(j\omega) + 3} = \frac{(j\omega + 2)}{(j\omega + 1)(j\omega + 3)}$$

$$= \frac{1/2}{j\omega + 1} + \frac{1/2}{j\omega + 3}$$

$$\Rightarrow h(t) = \frac{1}{2}e^{-t}u(t) + \frac{1}{2}e^{-3t}u(t)$$

Ch. 4 Continuous-Time Fourier Transform

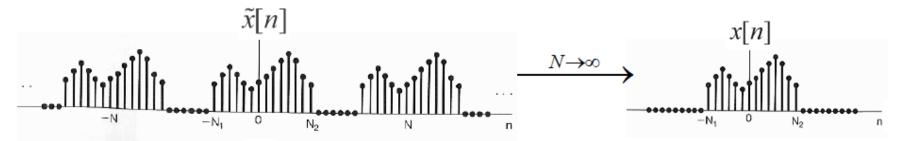
- Sec. 4.1 Representation of Aperiodic Signals:
 The Continuous-Time Fourier Transform
- Sec. 4.2 The Fourier Transform for Periodic Signals
- Sec. 4.3 Properties of the Continuous-Time Fourier Transform
- Sec. 4.4 The Convolution Property
- Sec. 4.5 The Multiplication Property
- Sec. 4.6 Tables of Fourier Properties and of Basic Fourier Transform Pairs
- Sec. 4.7 Systems Characterized by Linear Constant Coefficient Differential Equations

Ch. 5 Discrete-Time Fourier Transform

- Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform
- Sec. 5.2 The Fourier Transform for Periodic Signals
- Sec. 5.3 Properties of the Discrete-Time Fourier Transform
- Sec. 5.4 The Convolution Property
- Sec. 5.5 The Multiplication Property
- Sec. 5.6 Tables of FT Properties and Basic FT Pairs
- Sec. 5.7 Duality
- Sec. 5.8 Systems Characterized by Linear Constant Coefficient Differential Equations

Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform

Develop DT FT for Aperiodic Signals



As $N \to \infty$, $\tilde{x}[n] = x[n]$ for any finite value of n. We will use this relation to derive the DTFT of aperiodic signals. Recall the FS representation of DT signals:

$$\tilde{x}[n] = \sum_{k=\langle N \rangle} a_k e^{jk\omega_0 n}$$

$$a_k = \frac{1}{N} \sum_{n=\langle N \rangle} \tilde{x}[n] e^{-jk\omega_0 n}$$

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

Develop DT FT for Aperiodic Signals



Since $\tilde{x}[n] = x[n]$ within any period $\langle N \rangle$, we have

$$a_{k} = \frac{1}{N} \sum_{n = < N >} \tilde{x}[n] e^{-jk\omega_{0}n} = \frac{1}{N} \sum_{n = -N_{1}}^{N_{2}} x[n] e^{-jk\omega_{0}n} = \frac{1}{N} \sum_{n = -\infty}^{+\infty} x[n] e^{-jk\omega_{0}n}$$
Define
$$X(e^{j\omega}) = \sum_{n = -\infty}^{\infty} x[n] e^{-j\omega n} \qquad () \uparrow \uparrow \uparrow)$$

then we have

$$a_k = \frac{1}{N} X(e^{jk\omega_0}).$$

Substituting this a_k to the synthesis equation yields

$$\widetilde{x[n]} = \sum_{k=< N>} \frac{1}{N} X(e^{jk\omega_0}) e^{jk\omega_0 n}.$$

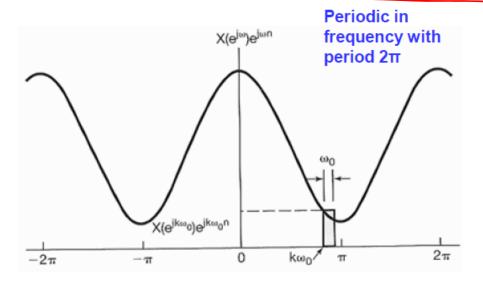
Since $\omega_0 = 2\pi / N$, or equivalently, $1/N = \omega_0 / 2\pi$,

$$\tilde{x}[n] = \frac{1}{2\pi} \sum_{k=< N>} X(e^{jk\omega_0}) e^{jk\omega_0 n} \omega_0.$$

- Develop DT FT for Aperiodic Signals (cont'd)
- Note: 1) Both $X(e^{j\omega})$ and $e^{j\omega n}$ are periodic in ω with period 2π , so is the product $X(e^{j\omega})e^{j\omega n}$.
 - 2) The total interval of integration will become 2π since the summation is carried over N consecutive of intervals of width $\omega_0 = 2\pi / N$.

Therefore, as $N \to \infty$, $\tilde{x}[n] \to x[n]$, $\omega_0 \to 0$, we have

$$\tilde{x}[n] = \frac{1}{2\pi} \sum_{k=\langle N \rangle} X(e^{jk\omega_0}) e^{jk\omega_0 n} \omega_0 \qquad \rightarrow \qquad \left[x[n] = \frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega. \right]$$



Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform

DTFT

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

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

Inverse Fourier transform Synthesis equation

Fourier transform Analysis equation

Recall the CTFT:

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

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

Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform

Periodicity

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

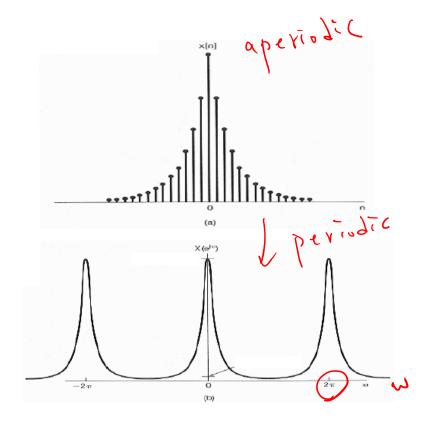
$$X(e^{j(\omega+2\pi)}) = \sum_{n=-\infty}^{+\infty} x[n]e^{-j(\omega+2\pi)n}$$

$$= \sum_{n=-\infty}^{+\infty} x[n]e^{-j\omega n}e^{-j2\pi n}$$

$$= \sum_{n=-\infty}^{+\infty} x[n]e^{-j\omega n}$$

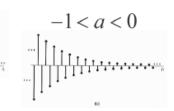
$$\therefore X(e^{j(\omega+2\pi)}) = X(e^{j\omega})$$

 $\Rightarrow X(e^{j\omega})$ is periodic with period 2π

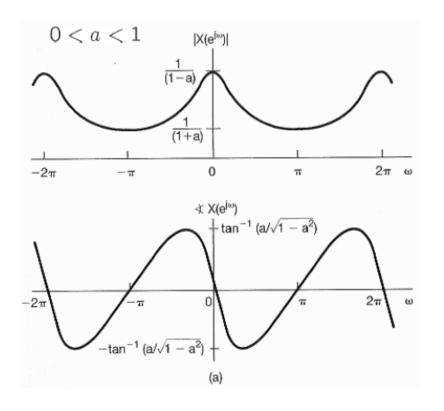


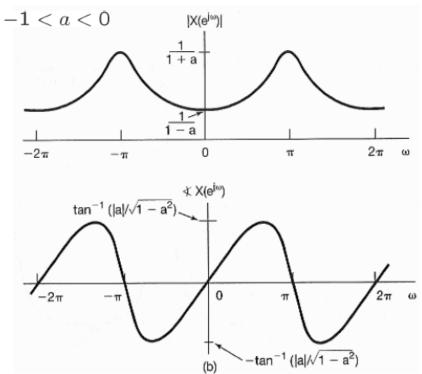
• Example 5.1

$$x[n] = a^n u[n], \quad |a| < 1$$



$$\Rightarrow X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} a^n u[n] e^{-j\omega n} = \sum_{n=0}^{\infty} (ae^{-j\omega})^n = \frac{1}{1 - ae^{-j\omega}}$$





• Example 5.2 $x[n] = a^{|n|}, |a| < 1$

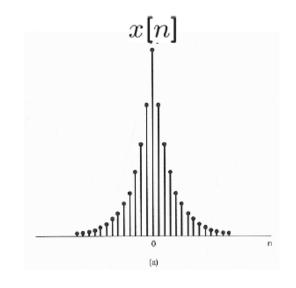
$$X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} a^{|n|} e^{-j\omega n}$$

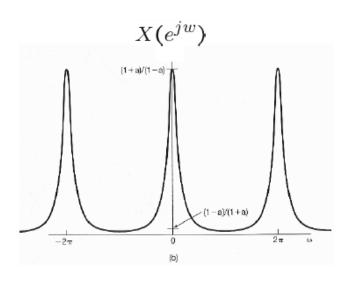
$$= \sum_{n=0}^{+\infty} a^n e^{-j\omega n} + \sum_{n=-\infty}^{-1} a^{-n} e^{-j\omega n}$$

$$= \sum_{n=0}^{+\infty} (ae^{-j\omega})^n + \sum_{m=1}^{+\infty} (ae^{j\omega})^m$$

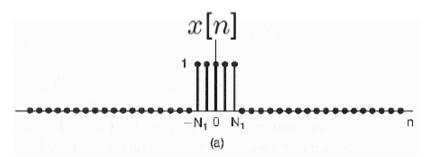
$$= \frac{1}{1 - ae^{-j\omega}} + \frac{ae^{j\omega}}{1 - ae^{j\omega}}$$

$$= \frac{1 - a^2}{1 - 2a\cos\omega + a^2}$$





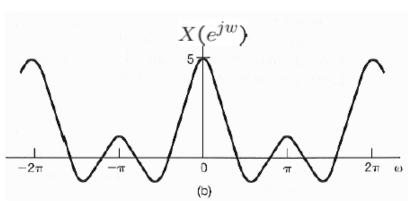
• Example 5.3
$$x[n] = \begin{cases} 1, & |n| \le N_1 \\ 0, & |n| > N_1 \end{cases}$$



$$X(e^{j\omega}) = \sum_{n=-N_1}^{N_1} e^{-j\omega n}$$

$$= e^{-j\omega(-N_1)} + \dots + e^{-j\omega(N_1)}$$

$$= e^{-j\omega(-N_1)} \left(\frac{1 - (e^{-j\omega})^{2N_1 + 1}}{1 - (e^{-j\omega})}\right)$$



$$= e^{j\omega(N_1)} \left(\frac{(e^{-j\omega})^{N_1 + \frac{1}{2}} ((e^{j\omega})^{N_1 + \frac{1}{2}} - (e^{-j\omega})^{N_1 + \frac{1}{2}})}{(e^{-j\omega/2})((e^{j\omega/2}) - (e^{-j\omega/2}))} \right)$$

$$= \frac{\sin(\omega(N_1 + \frac{1}{2}))}{\sin(\omega/2)}$$

Convergence of DTFT

- Sufficient Conditions for the Convergence of FT
- Derivation of FT suggests the same convergence condition as that of FS.

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

If x(t) has finite energy (that is, it is square integrable),

$$\int_{-\infty}^{+\infty} \left| x(t) \right|^2 dt < \infty$$

Then we are guaranteed that $X(j\omega)$ is finite and that

$$\int_{-\infty}^{+\infty} \left| e(t) \right|^2 dt = 0.$$

That is, there is no energy in their difference, even if $\hat{x}(t)$ may differ significantly at individual values of t.

Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform

Convergence of DTFT

The analysis equation

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

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

will converge if x[n] is absolutely summable,

$$\sum_{n=-\infty}^{+\infty} |x[n]| < \infty,$$

or if x[n] has finite energy,

$$\sum_{n=-\infty}^{+\infty} |x[n]|^2 < \infty,$$

very much like its counterpart in the CT case.

Sec. 5.1 Representation of Aperiodic Signals: The Discrete-Time Fourier Transform

Convergence of DTFT

But the synthesis equation

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

has no convergence issue associated with it because the integral is over a finite interval of integration.

Also, in contrast to the CT case, the Gibbs phenomenon does not exist if we approximate x[n] by

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

where $W \le \pi$. The amplitude of the oscillations exhibited in $\hat{x}[n]$ relative to the magnitude of $\hat{x}[0]$ decreases as W is increased.

Revisit of Sect. 3.6 FS Representation of DT Periodic Signals

Partial Sum

$$N = 9, 2N_1 + 1 = 5$$

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

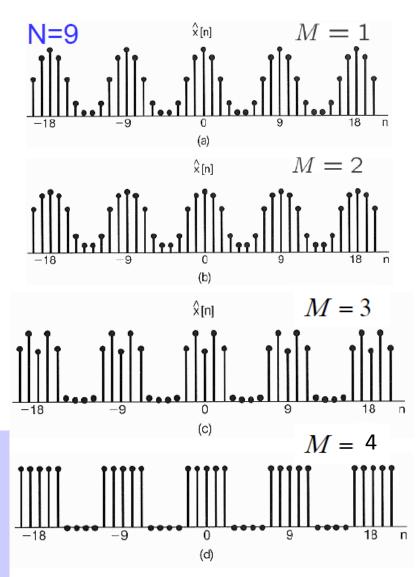
If N is odd

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

If N is even

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

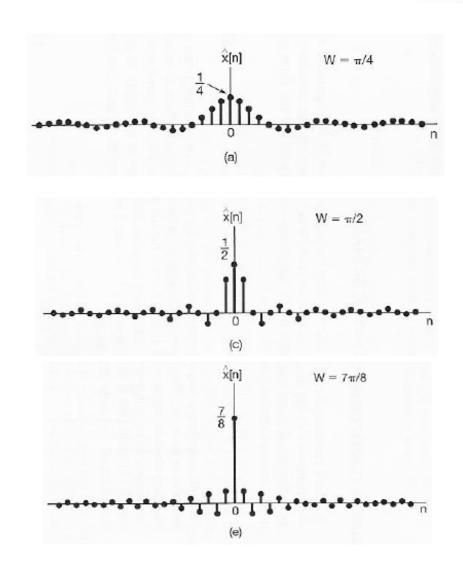
Gibbs phenomenon does not exist for DT signals because DT signals are represented by a finite number of FS coefficients. For the same reason, there is no convergence issue with DTFS.

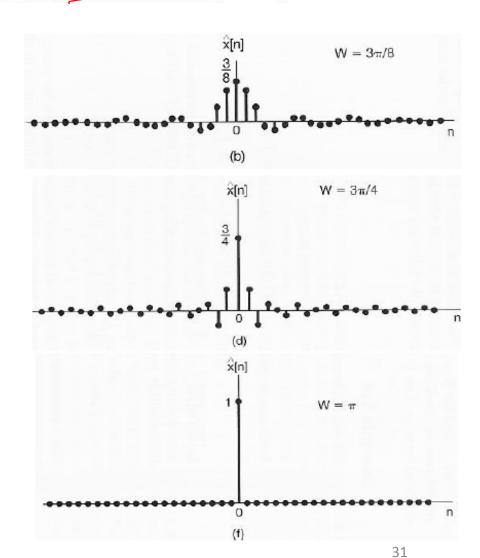


• Convergence of DTFT

$$x[n] = \delta[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega}) = 1$$

$$\hat{x}[n] = \frac{1}{2\pi} \int_{-W}^{W} X(e^{j\omega}) e^{j\omega n} d\omega = \frac{\sin Wn}{\pi n}$$





Sect. 5.2 FT for *Periodic* Signals

Reall:

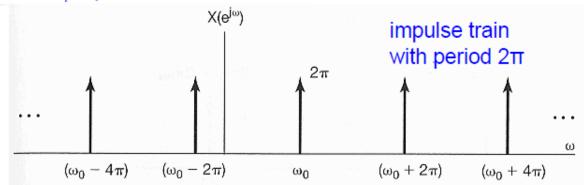
 $e^{j\omega_0 t} \stackrel{F}{\longleftrightarrow} 2\pi \delta(\omega - \omega_0)$

in the CT domain.

FT from FS

$$x[n] = e^{j\omega_0 n}, \quad \omega_0 = \frac{2\pi}{N}$$

$$\Rightarrow X(e^{j\omega}) = \sum_{l=-\infty}^{+\infty} 2\pi \delta(\omega - \omega_0 - 2\pi l)$$



Proof:

$$\frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega = \frac{1}{2\pi} \int_{2\pi} \sum_{l=-\infty}^{+\infty} 2\pi \delta(\omega - \omega_0 - 2\pi l) e^{j\omega n} d\omega$$
$$= e^{j(\omega_0 + 2\pi r)n} = e^{j\omega_0 n}$$

Sect. 5.2 FT for *Periodic* Signals

FT from FS (cont'd)

Thus, for a periodic sequence x[n] with period N and with the FS representation

$$x[n] = \sum_{k=< N>} a_k e^{jk\omega_0 n},$$

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

its FT is related to its Foureir coefficient by

$$X(e^{j\omega}) = \sum_{k=-\infty}^{+\infty} 2\pi a_k \delta(\omega - k\omega_0).$$

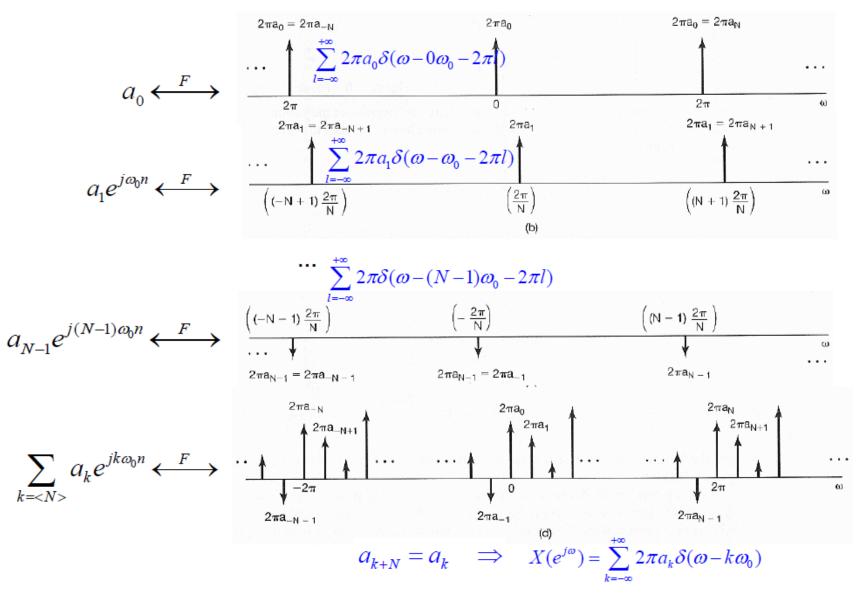
The FT of a periodic signal can be directly constructed from its Fourier coefficients.

We can verify this equation graphically by expressing x[n] as

$$x[n] = a_0 + a_1 e^{j\omega_0 n} + a_2 e^{j2\omega_0 n} + \dots + a_{N-1} e^{j(N-1)\omega_0 n},$$

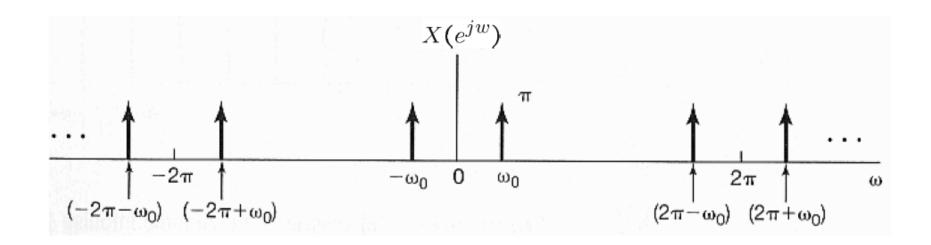
plot the FT of each term, and then superimpose them.

• FT from FS (cont'd)



• Example 5.5

$$x[n] = \cos(\omega_0 n) = \frac{e^{j\omega_0 n} + e^{-j\omega_0 n}}{2}, \qquad \omega_0 = \frac{2\pi}{5}$$
$$X(e^{j\omega}) = \sum_{l=-\infty}^{+\infty} \pi \delta(\omega - \frac{2\pi}{5} - 2\pi l) + \sum_{l=-\infty}^{+\infty} \pi \delta(\omega + \frac{2\pi}{5} - 2\pi l)$$



• Example 5.6 DTFT of Impulse Trains

$$x[n] = \sum_{k=-\infty}^{+\infty} \delta[n - kN]$$

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

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

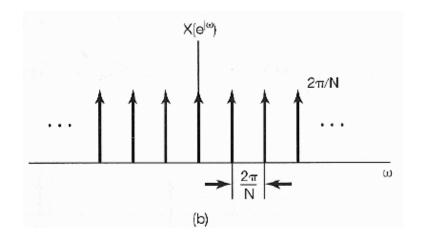
$$= \frac{1}{N}$$

$$\Rightarrow X(e^{j\omega}) = \frac{2\pi}{N} \sum_{k=-\infty}^{+\infty} \delta(\omega - k \frac{2\pi}{N})$$

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

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

$$X(e^{j\omega}) = \sum_{k=-\infty}^{+\infty} 2\pi a_k \delta(\omega - \frac{2\pi k}{N})$$



Sect. 5.3 Properties of DTFT

Recall that...

Synthesis equation

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

Analysis equation

$$\nabla \tau \tau \tau \qquad X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x[n]e^{-j\omega n}$$

$$X(e^{j\omega}) = F\{x[n]\}$$

$$x[n] = F^{-1}\{X(e^{j\omega})\}$$

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega})$$

$$\frac{1}{1 - ae^{j\omega}} = F\{a^n u[n]\}, \quad |a| < 1$$

$$a^n u[n] = F^{-1}\{\frac{1}{1 - ae^{j\omega}}\}$$

$$a^n u[n] \xleftarrow{F} \frac{1}{1 - ae^{j\omega}}$$

Periodicity of DT Fourier Transform:

$$X(e^{j(\omega+2\pi)}) = X(e^{j\omega})$$

Linearity:

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega})$$

$$y[n] \stackrel{F}{\longleftrightarrow} Y(e^{j\omega})$$

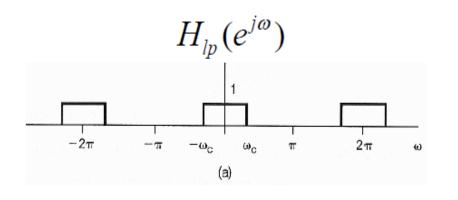
$$\Rightarrow ax[n] + by[n] \stackrel{F}{\longleftrightarrow} aX(e^{j\omega}) + bY(e^{j\omega})$$

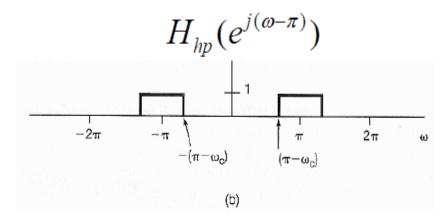
Time & Frequency Shifting:

$$x[n-n_0] \stackrel{F}{\longleftrightarrow} e^{-j\omega n_0} X(e^{j\omega})$$

$$e^{j\omega_0 n} x[n] \stackrel{F}{\longleftrightarrow} X(e^{j(\omega-\omega_0)}) \longrightarrow AM$$

• Example 5.7 Relationship between LPF & HPF





$$H_{hp}(e^{j\omega}) = H_{lp}(e^{j(\omega-\pi)})$$

$$\Rightarrow h_{hp}[n] = e^{j\pi n} h_{lp}[n]$$

$$= (-1)^n h_{lp}[n]$$

$$= h_{hp}[n]$$

$$= h_{hp}[n]$$

$$= h_{hp}[n]$$

Conjugation & Conjugate Symmetry

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega}) \implies x^*[n] \stackrel{F}{\longleftrightarrow} X^*(e^{-j\omega})$$

If
$$x[n]$$
 is real, then $x[n] = x^*[n]$ and $X(e^{-j\omega}) = X^*(e^{j\omega})$.

That is, $X(e^{j\omega})$ is conjugate symmetric and

$$\underline{Ev}\{x[n]\} \longleftrightarrow \underline{Re}\{X(e^{j\omega})\}$$

$$Od\left\{x[n]\right\} \stackrel{F}{\longleftrightarrow} jIm\left\{X(e^{j\omega})\right\}$$

Let
$$X(e^{j\omega}) = Re\{X(e^{j\omega})\} + jIm\{X(e^{j\omega})\}$$

$$\Rightarrow Re\{X(e^{j\omega})\} = Re\{X(e^{-j\omega})\}$$

$$\Rightarrow Im\{X(e^{j\omega})\} = -Im\{X(e^{-j\omega})\}$$

Real part is an even function Imaginary part is an odd function

Let
$$X(e^{j\omega}) = |X(e^{j\omega})| e^{\angle X(e^{j\omega})}$$

$$\Rightarrow$$
 $X(e^{j\omega})$ even, $\angle X(e^{j\omega})$ odd \longrightarrow Magnitude: an even function Phase: an odd function

Conjugation & Conjugate Symmetry

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega}) \implies x^*[n] \stackrel{F}{\longleftrightarrow} X^*(e^{-j\omega})$$

- If $x[n] = x^*[n]$ and x[-n] = x[n]
 - $\Rightarrow X(e^{-j\omega}) = X^*(e^{j\omega}) \text{ and } X(e^{-j\omega}) = X(e^{j\omega})$
 - $\Rightarrow X(e^{j\omega}) = X^*(e^{j\omega})$
 - \Rightarrow If x[n] is real and even, then $X(e^{j\omega})$ is real and even.
- If x[n] is real and odd, then $X(e^{j\omega})$ is pure imaginary and odd.

Differencing & Accumulation

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega})$$

$$x[n] - x[n-1] \stackrel{F}{\longleftrightarrow} (1 - e^{-j\omega}) X(e^{j\omega})$$

Differentiation in Frequency

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega}) \Rightarrow nx[n] \stackrel{F}{\longleftrightarrow} j \frac{d}{d\omega} X(e^{j\omega})$$

Proof:

$$\frac{d}{d\omega}X(e^{j\omega}) = \frac{d}{d\omega}\sum_{n=-\infty}^{+\infty}x[n]e^{-j\omega n}$$
$$= \sum_{n=-\infty}^{+\infty}(-jn)x[n]e^{-j\omega n} = (-j)\sum_{n=-\infty}^{+\infty}(nx[n])e^{-j\omega n}$$

Time Reversal

$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega}) \Rightarrow x[-n] \stackrel{F}{\longleftrightarrow} X(e^{-j\omega})$$

Proof:

$$X(e^{j\omega}) = \sum_{n = -\infty}^{+\infty} x[n]e^{-j\omega n}, \quad X(e^{j(-\omega)}) = \sum_{n = -\infty}^{+\infty} x[n]e^{-j(-\omega)n}$$

Time Expansion

$$x[n] \Rightarrow x[an] = ?$$

If a is an integer and a>1, x[an] is a time-compressed version of x[n]. For example, x[2n] is the even samples of x[n].

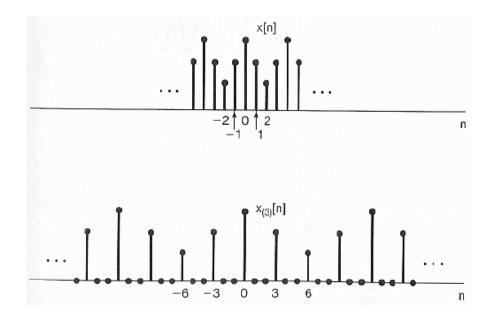
However, if a is not an integer, the value of x[an] is unknown because discrete-time signals are defined over integer intervals. Consequently, we cannot slow down the signal by making a < 1.

We resort to an alternative method (on next page).

Time Expansion

Define

$$x_{(k)}[n] = \begin{cases} x[n/k], & \text{if n is a multiple of } k \\ 0, & \text{otherwise.} \end{cases}$$



 $x_{(k)}[n]$ is obtained by placing k-1 zeros between successive samples of the original signal.

Time Expansion

$$X_{(k)}(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x_{(k)}[n]e^{-j\omega n}$$

$$= \sum_{r=-\infty}^{+\infty} x_{(k)}[rk]e^{-j\omega rk} \qquad x_{(k)}[rk] = x[r]$$

$$= \sum_{r=-\infty}^{+\infty} x[r]e^{-jk\omega r}$$

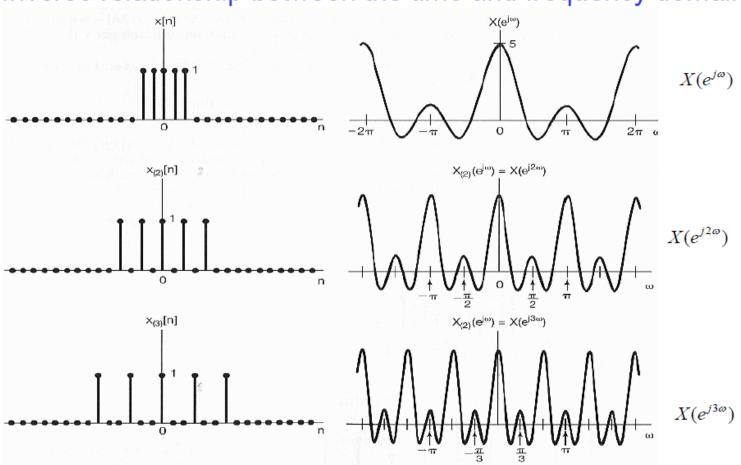
$$= X(e^{jk\omega})$$

$$x_{(k)}[n] \stackrel{F}{\longleftrightarrow} X(e^{jk\omega})$$

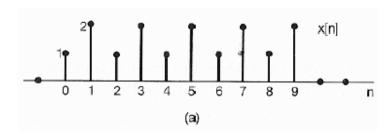
As a signal is spread out and slowed down in time, its FT is compressed.

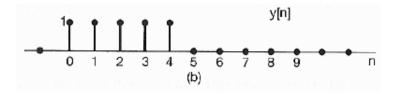
Time Expansion

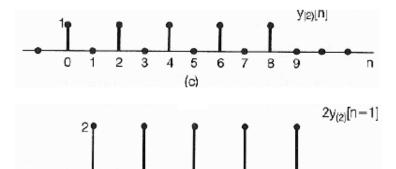
Inverse relationship between the time and frequency domains



• Example 5.9







$$x[n] = y_{(2)}[n] + 2y_{(2)}[n-1]$$

$$Y(e^{j\omega}) = e^{-j2\omega} \frac{\sin(5\omega/2)}{\sin(\omega/2)}$$

$$y[n]$$

$$y[n]$$

$$y(2)[n] = \begin{cases} 2y[n/2], & \text{if } n \text{ is even} \\ 0, & \text{if } n \text{ is odd} \end{cases}$$

$$y_{(2)}[n] \stackrel{F}{\longleftrightarrow} e^{-j4\omega} \frac{\sin(5\omega)}{\sin(\omega)}$$

$$2y_{(2)}[n-1] \stackrel{F}{\longleftrightarrow} 2e^{-j\omega}e^{-j4\omega} \frac{\sin(5\omega)}{\sin(\omega)}$$

$$X(e^{j\omega}) = (1+2e^{-j\omega}) \cdot e^{-j4\omega} \cdot \frac{\sin(5\omega)}{\sin(\omega)}$$

$$X(e^{j\omega}) = (1 + 2e^{-j\omega}) \cdot e^{-j4\omega} \cdot \frac{\sin(5\omega)}{\sin(\omega)}$$

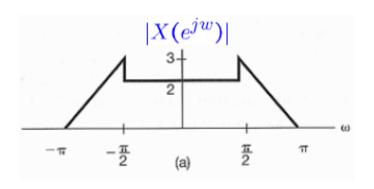
Parseval's relation

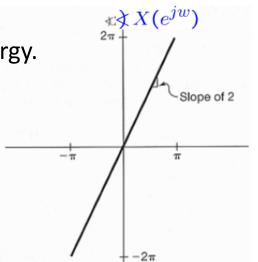
$$x[n] \stackrel{F}{\longleftrightarrow} X(e^{j\omega})$$

$$\sum_{n=-\infty}^{+\infty} |x[n]|^2 = \frac{1}{2\pi} \int_{2\pi} |X(e^{j\omega})|^2 d\omega$$
Total energy

Energy density spectrum

• Example 5.10 Determine if x[n] is periodic/real/even/finite energy.

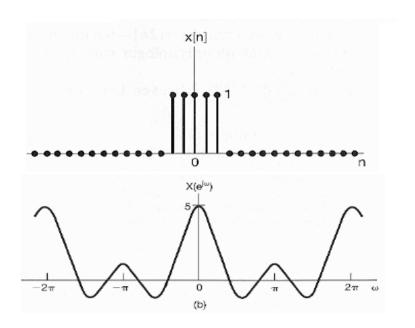


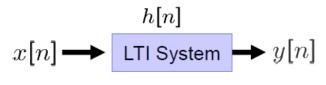


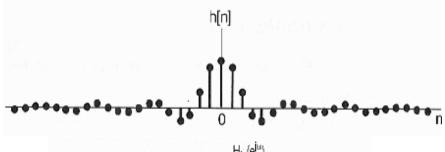
$$X(e^{j\omega}) \neq \text{ impulse train} \qquad \Rightarrow x[n] \text{ is NOT periodic}$$
Even magnitude odd phase $\Rightarrow x[n] \text{ is real}$
 $X(e^{j\omega}) \text{ is not real} \qquad \Rightarrow x[n] \text{ is NOT even}$
 $X(e^{j\omega}) \text{ has finite energy} \qquad \Rightarrow x[n] \text{ is finite}$

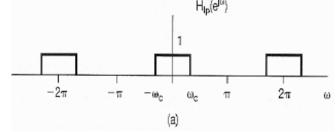
Sect. 5.4 & 5.5 Convolution vs. Multiplication Property

Convolution Property









$$y[n] = x[n] * h[n] \longleftrightarrow Y(e^{j\omega}) = X(e^{j\omega})H(e^{j\omega})$$

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

Example 5.11 Time shifting property

$$h[n]$$

$$x[n] \longrightarrow \text{LTI System} \longrightarrow y[n]$$

$$h[n] = \delta[n - n_0]$$

$$\Rightarrow H(e^{j\omega}) = \sum_{n = -\infty}^{+\infty} \delta[n - n_0] e^{-j\omega n} = e^{-j\omega n_0}$$

$$\Rightarrow Y(e^{j\omega}) = H(e^{j\omega}) X(e^{j\omega})$$

$$= e^{-j\omega n_0} X(e^{j\omega})$$

$$\Rightarrow y[n] = x[n - n_0]$$

• Example 5.13 Determine y[n]

$$x[n] \longrightarrow \text{Filter} \longrightarrow y[n]$$

$$x[n] = a^{n}u[n], \quad |a| < 1 \qquad \Rightarrow H(e^{j\omega}) = \frac{1}{1 - ae^{-j\omega}}$$

$$x[n] = b^{n}u[n], \quad |b| < 1 \qquad \Rightarrow X(e^{j\omega}) = \frac{1}{1 - be^{-j\omega}}$$

$$\Rightarrow Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega})$$

$$= \frac{1}{1 - ae^{-j\omega}} \frac{1}{1 - be^{-j\omega}}$$

• Example 5.13 (cont'd)

if
$$a \neq b$$

$$Y(e^{j\omega}) = \left[\left(\frac{a}{a - b} \right) \frac{1}{1 - ae^{-j\omega}} + \left(\frac{-b}{a - b} \right) \frac{1}{1 - be^{-j\omega}} \right]$$

$$\Rightarrow y[n] = \left(\frac{a}{a - b} \right) a^n u[n] - \left(\frac{b}{a - b} \right) b^n u[n]$$
if $a = b$
$$Y(e^{j\omega}) = \left(\frac{1}{1 - ae^{-j\omega}} \right)^2 = \frac{j}{a} e^{j\omega} \frac{d}{d\omega} \left(\frac{1}{1 - ae^{-j\omega}} \right)$$
since
$$a^n u[n] \stackrel{F}{\longleftrightarrow} \frac{1}{1 - ae^{-j\omega}}$$

$$na^n u[n] \stackrel{F}{\longleftrightarrow} j \frac{d}{d\omega} \left(\frac{1}{1 - ae^{-j\omega}} \right)$$

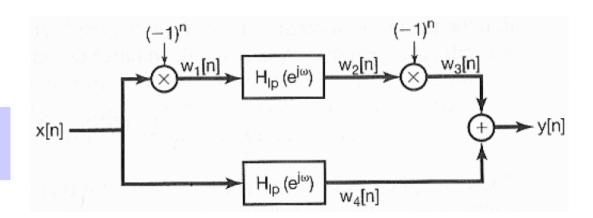
$$(n+1)a^{n+1}u[n+1] \stackrel{F}{\longleftrightarrow} je^{j\omega} \frac{d}{d\omega} \left(\frac{1}{1 - ae^{-j\omega}} \right)$$

$$\Rightarrow y[n] = (n+1)a^n u[n+1] = (n+1)a^n u[n]$$

• Example 5.14

$$H_{lp}(e^{j\omega})$$
: Low-pass filter with $\omega_c = \pi/4$

$$(-1)^n = e^{j\pi n}$$



$$w_{1}[n] = e^{j\pi n} x[n] = (-1)^{n} x[n]$$

$$\Rightarrow W_{1}(e^{j\omega}) = X(e^{j(\omega-\pi)})$$

$$W_{2}(e^{j\omega}) = H_{lp}(e^{j\omega}) X(e^{j(\omega-\pi)})$$

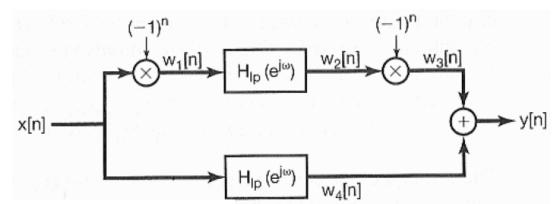
$$w_{3}[n] = e^{j\pi n} w_{2}[n] = (-1)^{n} w_{2}[n]$$

$$\Rightarrow W_{3}(e^{j\omega}) = W_{2}(e^{j(\omega-\pi)}) = H_{lp}(e^{j(\omega-\pi)}) X(e^{j(\omega-2\pi)})$$

$$= H_{lp}(e^{j(\omega-\pi)}) X(e^{j\omega})$$

• Example 5.14

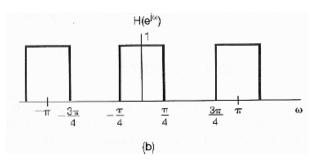
 $H_{lp}(e^{j\omega})$: Low-pass filter with $\omega_c = \pi/4$



$$W_4(e^{j\omega}) = H_{lp}(e^{j\omega})X(e^{j\omega})$$

$$\begin{split} Y(e^{j\omega}) &= W_{3}(e^{j\omega}) + W_{4}(e^{j\omega}) \\ &= H_{lp}(e^{j(\omega - \pi)}) X(e^{j\omega}) + H_{lp}(e^{j\omega}) X(e^{j\omega}) \\ &= [H_{lp}(e^{j(\omega - \pi)}) + H_{lp}(e^{j\omega})] X(e^{j\omega}) \end{split}$$

$$H(e^{j\omega}) = H_{lp}(e^{j(\omega-\pi)}) + H_{lp}(e^{j\omega})$$
 high-pass filter low-pass filter bandstop filter



Sect. 5.4 & 5.5 Convolution vs. Multiplication Property

Multiplication Property

$$r[n] = s[n]p[n] \stackrel{F}{\longleftrightarrow} R(e^{j\omega}) = \frac{1}{2\pi} \int_{2\pi} S(e^{j\theta}) P(e^{j(\omega-\theta)}) d\theta$$

$$R(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} r[n]e^{-j\omega n} = \sum_{n=-\infty}^{+\infty} s[n]p[n]e^{-j\omega n}$$

$$= \sum_{n=-\infty}^{+\infty} s[n] \{ \frac{1}{2\pi} \int_{2\pi} P(e^{j\theta}) e^{j\theta n} d\theta \} e^{-j\omega n}$$

$$= \frac{1}{2\pi} \int_{2\pi} P(e^{j\theta}) [\sum_{n=-\infty}^{+\infty} s[n]e^{-j(\omega-\theta)n}] d\theta$$

$$= \frac{1}{2\pi} \int_{2\pi} P(e^{j\theta}) S(e^{j(\omega-\theta)}) d\theta$$

$$= \frac{1}{2\pi} \int_{2\pi} P(e^{j(\omega-\theta)}) S(e^{j\theta}) d\theta$$

Periodic convolution

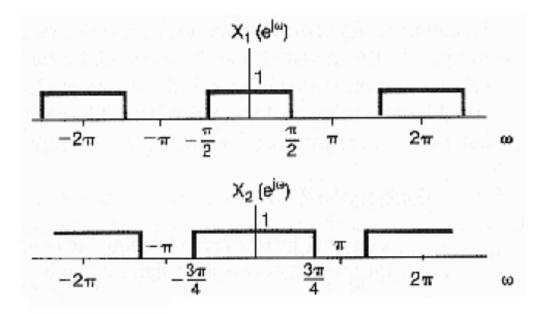
• Example 5.15

Converting periodic convolution into ordinary convolution

$$x[n] = x_1[n]x_2[n]$$

$$x_1[n] = \frac{\sin(\frac{\pi}{2}n)}{\pi n}$$

$$x_2[n] = \frac{\sin(\frac{3\pi}{4}n)}{\pi n}$$



$$X(e^{j\omega}) = \frac{1}{2\pi} \int_{2\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta$$

We can convert this equation to an ordinary convolution. Define

$$\widehat{X}_{1}(e^{j\theta}) = \begin{cases} X_{1}(e^{j\theta}), & \text{for } -\pi < \theta \leq \pi \\ 0, & \text{otherwise.} \end{cases}$$

• Example 5.15 Converting periodic convolution into ordinary convolution

$$\begin{split} X(e^{j\omega}) &= \frac{1}{2\pi} \int_{-\pi}^{+\pi} \hat{X}_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \\ &= \frac{1}{2\pi} \int_{-\infty}^{+\infty} \hat{X}_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \end{split}$$

