



CHAPTER 4

THE CONTINUOUS- TIME FOURIER TRANSFORM

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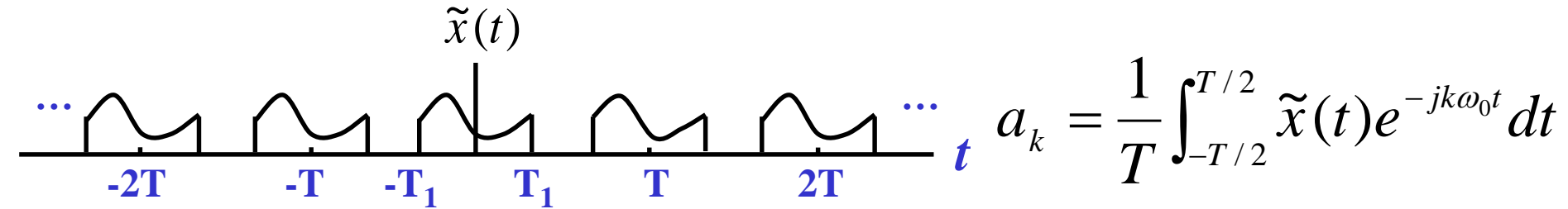
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Sampling

- Representation of continuous-time **aperiodic** signals as linear combination of complex exponentials — Inverse Fourier Transform
- Frequency spectrum of aperiodic signals — Fourier Transform
- Applications of continuous-time Fourier transform

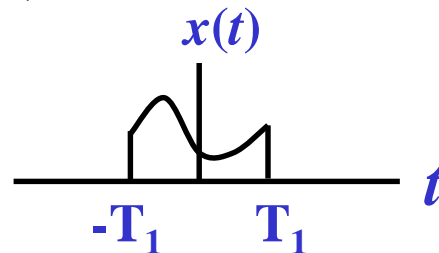
4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

4.1.1 Fourier Transform and Inverse Fourier Transform



$$Ta_k = \int_{-T/2}^{T/2} \tilde{x}(t) e^{-jk\omega_0 t} dt$$

As $T \rightarrow \infty$, $\lim_{T \rightarrow \infty} Ta_k = \int_{-\infty}^{\infty} x(t) e^{-jk\omega_0 t} dt = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$



Use $X(j\omega)$ to denote this integral, then we have:

Frequency spectrum of $x(t)$ \rightarrow $X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$ \leftarrow Fourier transform of $x(t)$

4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

Since $X(j\omega) = \lim_{T \rightarrow \infty} T a_k = \lim_{\omega_0 \rightarrow 0} 2\pi \frac{a_k}{\omega_0},$

$X(j\omega)$ is actually spectrum-density function(频谱密度函数).

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

As $T \rightarrow \infty$, $x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$ ← Inverse Fourier transform

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

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

Fourier Transform
Pair

4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

Comparing the synthesis equations in:

$$CTFS : x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} \quad CTFT : x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$$

$$a_k \Leftrightarrow \frac{X(j\omega) d\omega}{2\pi}$$

This fact means: Aperiodic signals can also be decomposed as linear combination of infinite numbers of complex exponentials, which occur at a continuum of frequencies, but have amplitudes infinitesimally small — approaching zero!

$|X(j\omega)|$ indicate the relative amplitudes of all components, and angle $\angle X(j\omega)$ indicate the phases of all components.

4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

An useful relationship: $a_k = \frac{1}{T} X(j\omega) \Big|_{\omega=k\omega_0}$

Where, $\tilde{x}(t) \xleftrightarrow{FS} a_k$, $x(t) \xleftrightarrow{FT} X(j\omega)$.

This relationship is also valid for aperiodic signals with unlimited duration!

➤ Convergence of Fourier Transforms

Dirichlet conditions:

1. $x(t)$ is absolutely integrable; that is $\int_{-\infty}^{\infty} |x(t)| dt < \infty$
 2. $x(t)$ have a finite number of maxima and minima within any finite interval.
 3. $x(t)$ have a finite number of discontinuities within any finite interval.
- Furthermore, each of these discontinuities must be finite.

If **impulse functions are permitted** in the transform, some signals which are **not absolutely integrable** over an infinite interval, can also be considered to have Fourier transforms. This will be convenient in the discussion of Fourier methods.

4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

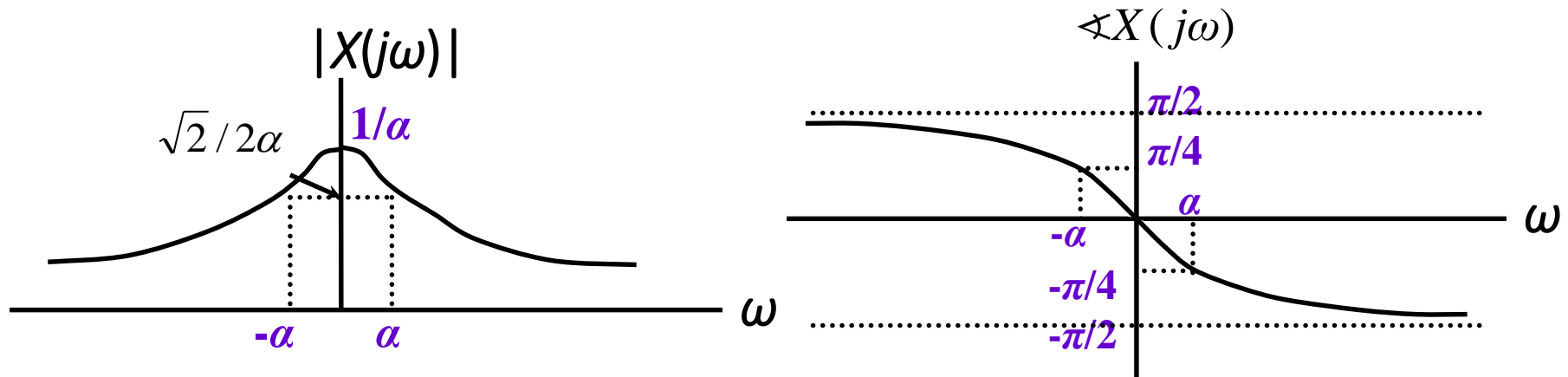
4.1.2 Examples

Example 4.1

Consider the signal $x(t) = e^{-\alpha t} u(t)$ $\alpha > 0$.

$$X(j\omega) = \int_0^{\infty} e^{-\alpha t} e^{-j\omega t} dt = -\frac{1}{\alpha + j\omega} e^{-(\alpha + j\omega)t} \bigg|_0^{\infty} = \frac{1}{\alpha + j\omega}, \quad \alpha > 0$$

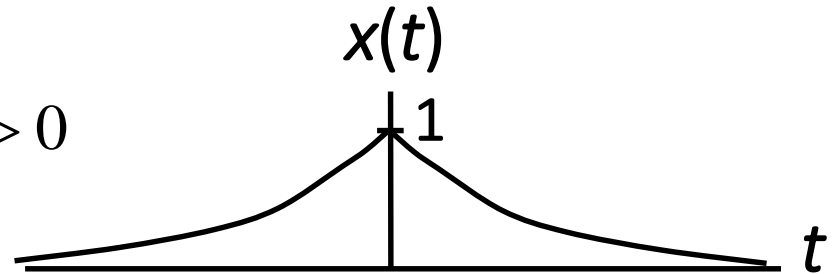
$$|X(j\omega)| = \frac{1}{\sqrt{\alpha^2 + \omega^2}}, \quad \angle X(j\omega) = -\tan^{-1}\left(\frac{\omega}{\alpha}\right)$$



4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

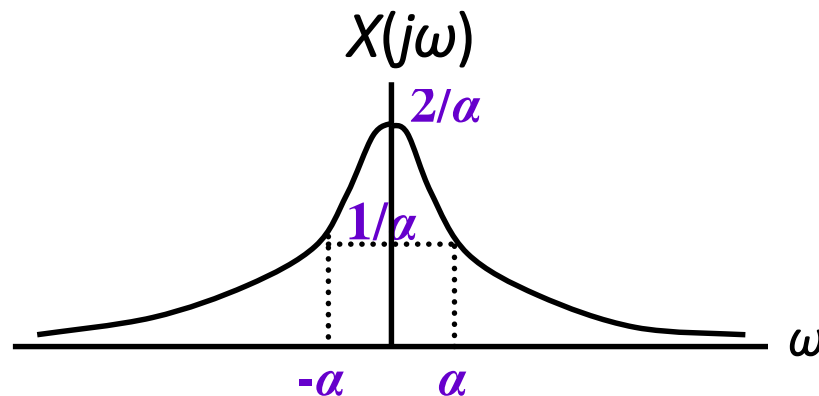
Example 4.2

Consider the signal $x(t) = e^{-\alpha|t|}$, $\alpha > 0$



Sol: The Fourier transform of the signal is

$$\begin{aligned} X(j\omega) &= \int_{-\infty}^{\infty} e^{-\alpha|t|} e^{-j\omega t} dt = \int_{-\infty}^0 e^{\alpha t} e^{-j\omega t} dt + \int_0^{\infty} e^{-\alpha t} e^{-j\omega t} dt \\ &= \frac{1}{\alpha - j\omega} + \frac{1}{\alpha + j\omega} = \frac{2\alpha}{\alpha^2 + \omega^2} \end{aligned}$$



4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

Example 4.3

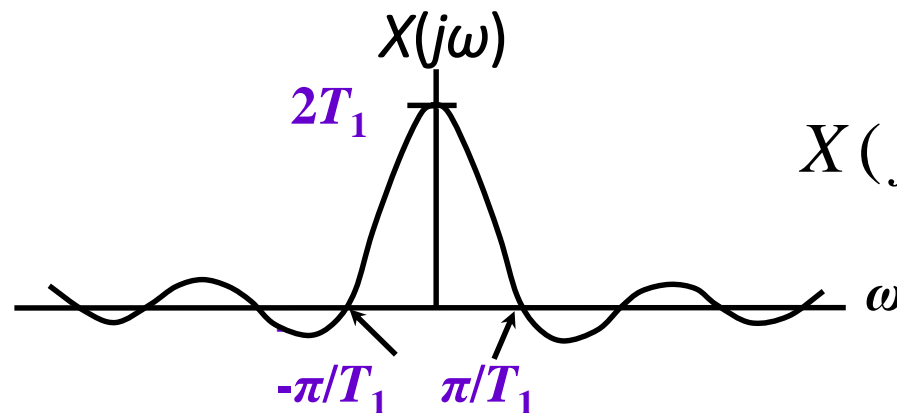
Determine the Fourier transform of the unit impulse $x(t) = \delta(t)$

$$X(j\omega) = \int_{-\infty}^{\infty} \delta(t) e^{-j\omega t} dt = 1$$

That is, the unit impulse has a Fourier transform consisting of equal contribution at all frequencies. This spectrum is referred to as **white-spectrum**.

Example 4.4

Consider the rectangular pulse signal $x(t) = \begin{cases} 1, & |t| < T_1 \\ 0, & |t| > T_1 \end{cases}$



$$X(j\omega) = \int_{-T_1}^{T_1} e^{-j\omega t} dt = \frac{2 \sin \omega T_1}{\omega}$$

4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform

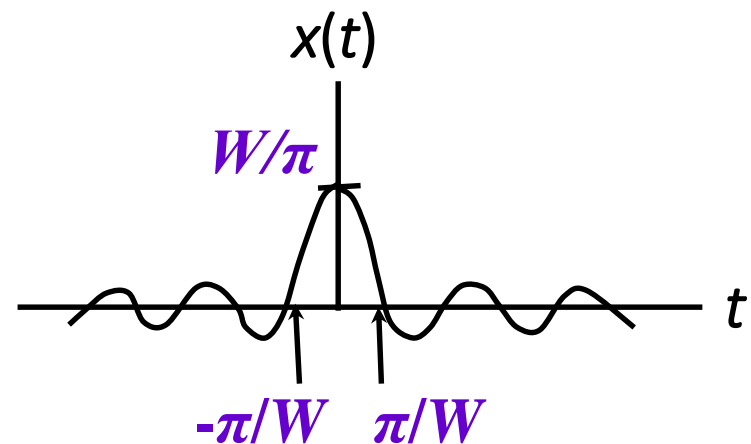
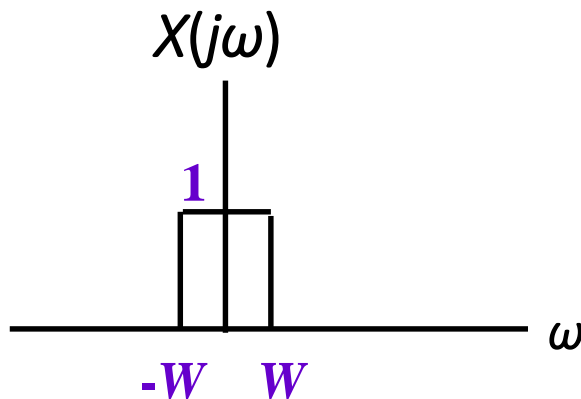
Example 4.5

Consider the signal $x(t)$ whose Fourier transform is

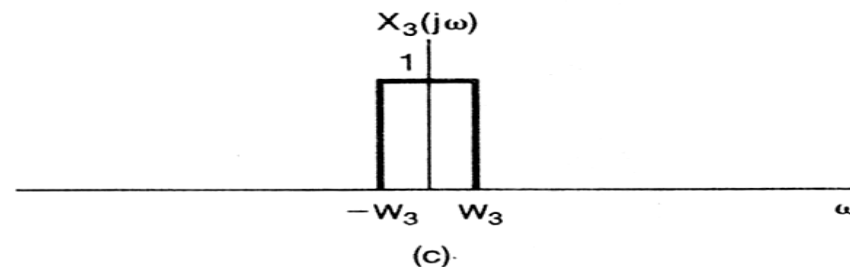
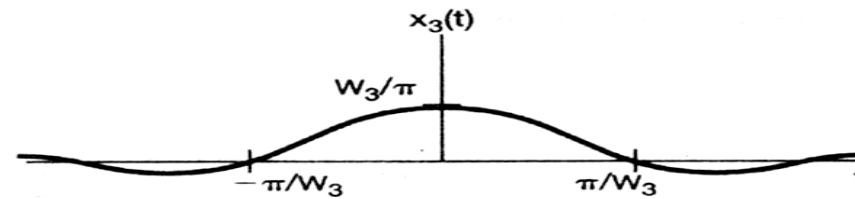
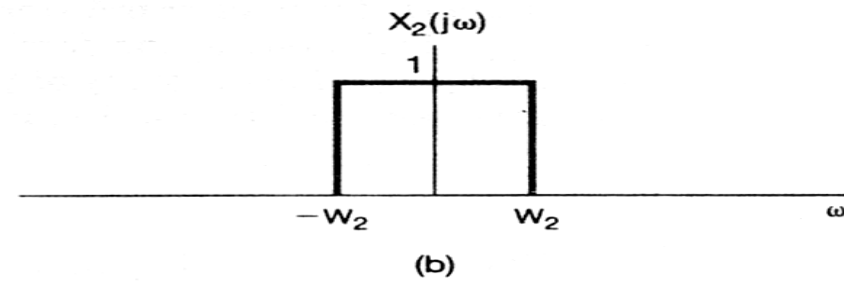
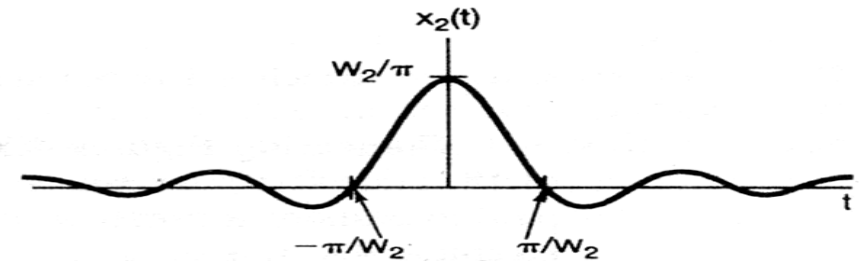
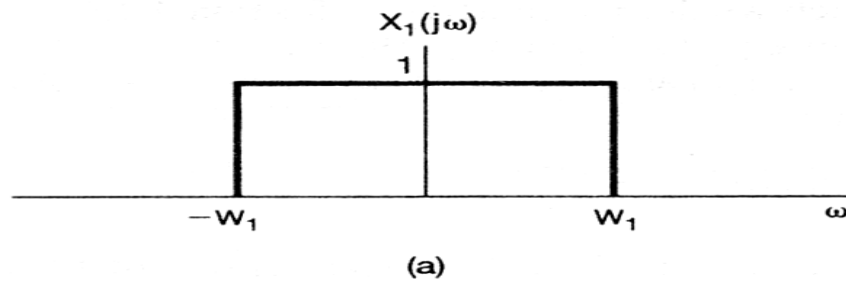
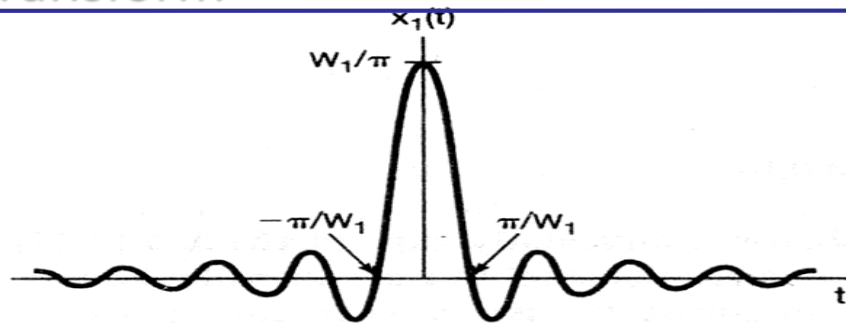
$$X(j\omega) = \begin{cases} 1, & |\omega| < W \\ 0, & |\omega| > W \end{cases}$$

Sol: Using the synthesis equation, we can determine

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



4.1 Representation of Aperiodic Signals: The Continuous-Time Fourier Transform



4.2 The Fourier Transform for Periodic Signals

$$\mathcal{F}\{e^{j\omega_0 t}\} = ?$$

From the analysis equation,

$$X(j\omega) = \int_{-\infty}^{\infty} e^{j\omega_0 t} e^{-j\omega t} dt = \frac{1}{j(\omega_0 - \omega)} e^{j(\omega_0 - \omega)t} \bigg|_{-\infty}^{\infty}$$

Does not converge !

We have obtained $\delta(t) \leftrightarrow 1$, so $\delta(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} 1 \cdot e^{j\omega t} d\omega$

$$2\pi\delta(t) = \int_{-\infty}^{\infty} 1 \cdot e^{j\omega t} d\omega$$

$$2\pi\delta(-\omega) = \int_{-\infty}^{\infty} e^{-ja\omega} da \Rightarrow 1 \xleftrightarrow{FT} 2\pi\delta(-\omega) = 2\pi\delta(\omega)$$

$$2\pi\delta(\omega_0 - \omega) = \int_{-\infty}^{\infty} e^{-jt(\omega - \omega_0)} dt = \int_{-\infty}^{\infty} e^{j\omega_0 t} e^{-j\omega t} dt \Rightarrow \mathcal{F}\{e^{j\omega_0 t}\} = 2\pi\delta(\omega - \omega_0)$$

The Fourier transform of the complex exponential signal $e^{j\omega_0 t}$ is an impulse located at $\omega = \omega_0$ with its area 2π .

4.2 The Fourier Transform for Periodic Signals

For an arbitrary periodic signal $x(t)$, if it can be represented by Fourier series as

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

We can obtain its Fourier transform as

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

The Fourier transform of a periodic signal with Fourier series coefficients $\{a_k\}$ can be interpreted as a train of impulses occurring at the harmonically related frequencies and for which the area of the impulse at the k th harmonic frequency $k\omega_0$ is 2π times the k th Fourier series coefficient a_k .

4.2 The Fourier Transform for Periodic Signals

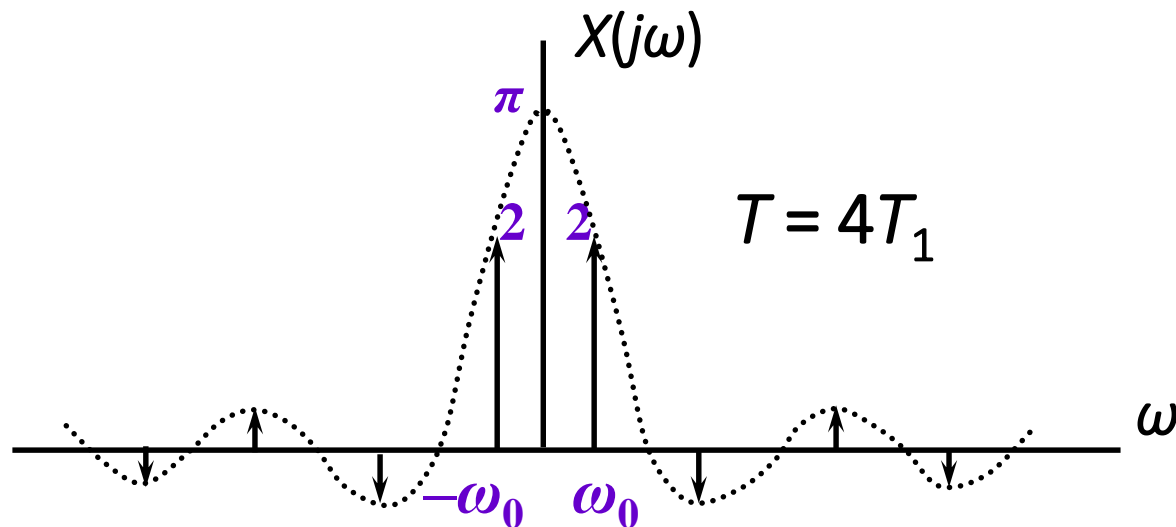
Example 4.6

Consider the periodic square wave with FS coefficients are

$$a_k = \frac{\sin k\omega_0 T_1}{k\pi}$$

Sol: From the formula its Fourier transform is

$$X(j\omega) = \sum_{k=-\infty}^{\infty} 2\pi \cdot \frac{\sin k\omega_0 T_1}{k\pi} \delta(\omega - k\omega_0) = \sum_{k=-\infty}^{\infty} \frac{2 \sin k\omega_0 T_1}{k} \delta(\omega - k\omega_0)$$



4.2 The Fourier Transform for Periodic Signals

Example 4.7

Find the Fourier Transforms of $x_1(t) = \sin \omega_0 t$ and $x_2(t) = \cos \omega_0 t$.

Sol: The Fourier series coefficients for $x_1(t)$ are

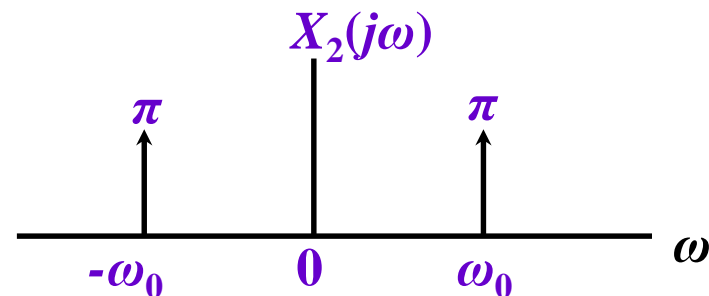
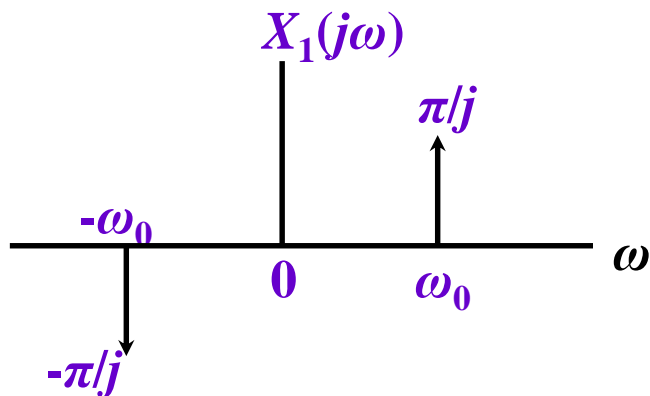
$$a_1 = \frac{1}{2j}, \quad a_{-1} = -\frac{1}{2j}, \quad a_k = 0, \quad k \neq \pm 1$$

The Fourier series coefficients for $x_2(t)$ are

$$a_1 = a_{-1} = \frac{1}{2}, \quad a_k = 0, \quad k \neq \pm 1$$

$$\sin \omega_0 t \longleftrightarrow j\pi [\delta(\omega + \omega_0) - \delta(\omega - \omega_0)]$$

$$\cos \omega_0 t \longleftrightarrow \pi [\delta(\omega + \omega_0) + \delta(\omega - \omega_0)]$$

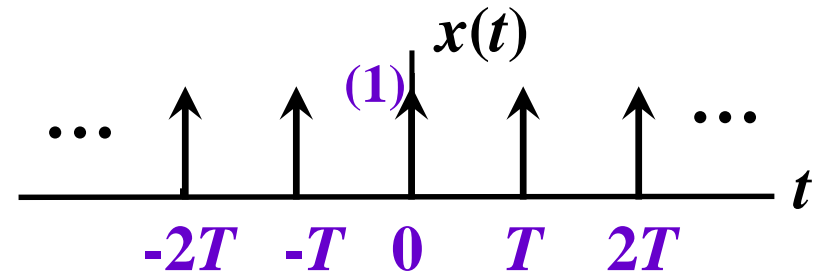


4.2 The Fourier Transform for Periodic Signals

Example 4.8

Consider the impulse train

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

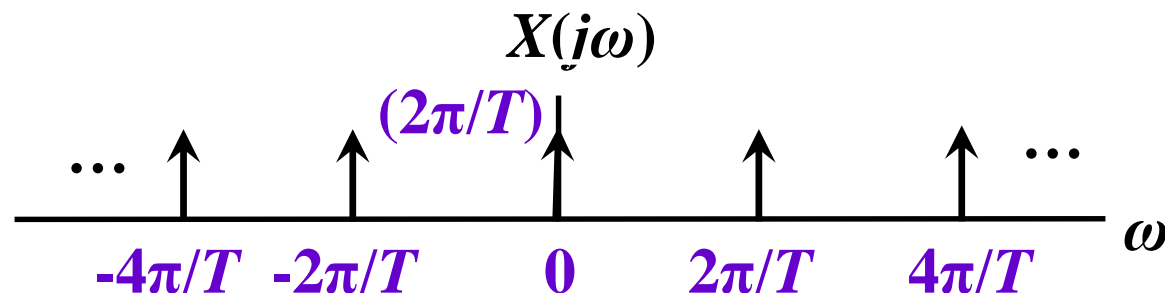


Sol: The Fourier series coefficients for this signal are

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

Thus, its Fourier transform is

$$X(j\omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\omega - \frac{2k\pi}{T}) = \omega_0 \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_0)$$



4.3 Properties of The Continuous-Time Fourier Transform

4.3.1 Linearity

If $x(t) \xleftrightarrow{FT} X(j\omega)$ and $y(t) \xleftrightarrow{FT} Y(j\omega)$

then $ax(t) + by(t) \xleftrightarrow{FT} aX(j\omega) + bY(j\omega)$

4.3.2 Time Shifting

If $x(t) \xleftrightarrow{FT} X(j\omega)$,

$$\mathcal{F}\{e^{-j\omega t_0} X(j\omega)\} = \mathcal{F}\{X(j\omega)\} - \omega t_0$$

then $x(t - t_0) \xleftrightarrow{FT} e^{-j\omega t_0} X(j\omega)$

A signal which is shifted in time does not have the magnitude of its Fourier transform altered. The effect of a time shift on a signal is to introduce into its transform a phase shift, namely, $-\omega t_0$.

4.3 Properties of The Continuous-Time Fourier Transform

Example 4.9

Determine the Fourier transform of $x(t)$ shown in the figure.

Sol: Since

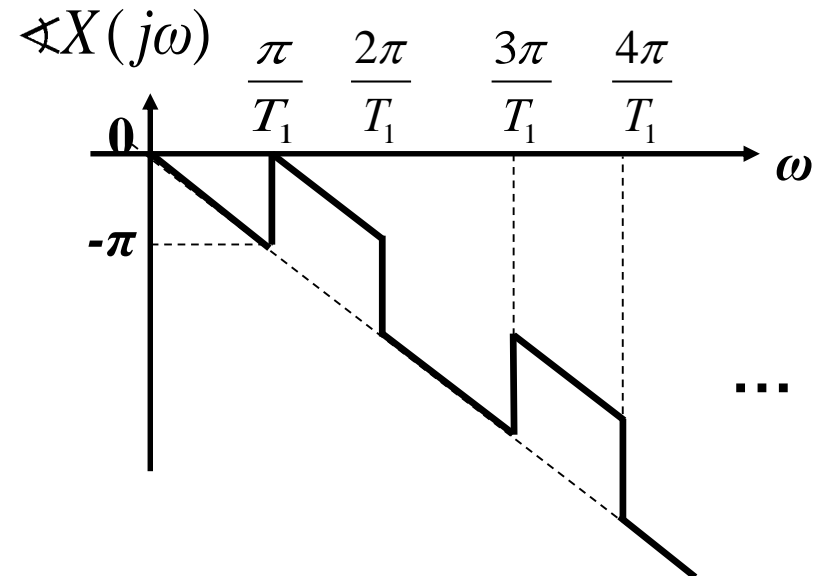
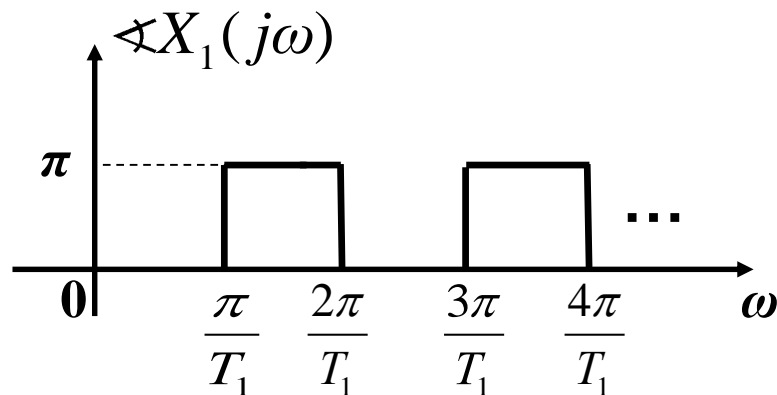
$$x_1(t) = u(t + T_1) - u(t - T_1) \leftrightarrow X_1(j\omega) = \frac{2 \sin \omega T_1}{\omega}$$

$$\text{and } x(t) = Ax_1(t - T_1)$$

From the time shifting property, we have

$$X(j\omega) = AX_1(j\omega)e^{-j\omega T_1} = \frac{2A \sin \omega T_1}{\omega} e^{-j\omega T_1}$$

$$\angle X(j\omega) = \angle X_1(j\omega) - \omega T_1$$



4.3 Properties of The Continuous-Time Fourier Transform

4.3.3 Conjugation and Conjugate Symmetry

If $x(t) \xleftrightarrow{FT} X(j\omega)$, then $x^*(t) \xleftrightarrow{FT} X^*(-j\omega)$

➤ If $x(t)$ is real, then $X(-j\omega) = X^*(j\omega)$

$$\operatorname{Re}\{X(j\omega)\} = \operatorname{Re}\{X(-j\omega)\}, \operatorname{Im}\{X(j\omega)\} = -\operatorname{Im}\{X(-j\omega)\}$$

even $\nearrow |X(j\omega)| = |X(-j\omega)|$, $\nwarrow X(j\omega) = -\nwarrow X(-j\omega)$ *odd*

➤ If $x(t)$ is both real and even, so is $X(j\omega)$, i.e., $X(j\omega) = \operatorname{Re}\{X(j\omega)\}$

➤ If $x(t)$ is real and odd, $X(j\omega)$ is purely imaginary and odd, i.e., $X(j\omega) = j \operatorname{Im}\{X(j\omega)\}$

➤
$$x(t) = x_e(t) + x_o(t)$$

$$x_e(t) \xleftrightarrow{FT} \operatorname{Re}\{X(j\omega)\} \quad x_o(t) \xleftrightarrow{FT} j \operatorname{Im}\{X(j\omega)\}$$

4.3 Properties of The Continuous-Time Fourier Transform

Example 4.10

Consider again the Fourier transform evaluation of $x(t) = e^{-\alpha|t|}$

Sol: $e^{-\alpha|t|} = e^{-\alpha t}u(t) + e^{\alpha t}u(-t) = 2 \left[\frac{e^{-\alpha t}u(t) + e^{\alpha t}u(-t)}{2} \right] = 2\text{Ev} \{ e^{-\alpha t}u(t) \}$

and
$$e^{-\alpha t}u(t) \xleftrightarrow{FT} \frac{1}{\alpha + j\omega}$$

From the symmetry properties of the Fourier transform, we have

$$X(j\omega) = 2 \operatorname{Re} \left\{ \frac{1}{\alpha + j\omega} \right\} = \frac{2\alpha}{\alpha^2 + \omega^2}$$

4.3 Properties of The Continuous-Time Fourier Transform

4.3.4 Differentiation and Integration

If $x(t) \xleftrightarrow{FT} X(j\omega)$, then $\frac{dx(t)}{dt} \xleftrightarrow{FT} j\omega X(j\omega)$

$$\frac{d^n x(t)}{dt^n} \xleftrightarrow{FT} (j\omega)^n X(j\omega)$$

Consequence: Strengthening the high-frequencies in the signal.

$$\int_{-\infty}^t x(\tau) d\tau \xleftrightarrow{FT} \frac{1}{j\omega} X(j\omega) + \pi X(0) \delta(\omega)$$

Consequence: Strengthening the low-frequencies in the signal.

Example 4.11

Determine the Fourier transform of the unit step $x(t) = u(t)$.

Sol : $u(t) = \frac{1}{2} + \frac{1}{2} \text{sgn}(t)$ $\text{sgn}(t) = \lim_{\alpha \rightarrow 0} [e^{-\alpha t} u(t) - e^{\alpha t} u(-t)] \quad (\alpha > 0)$

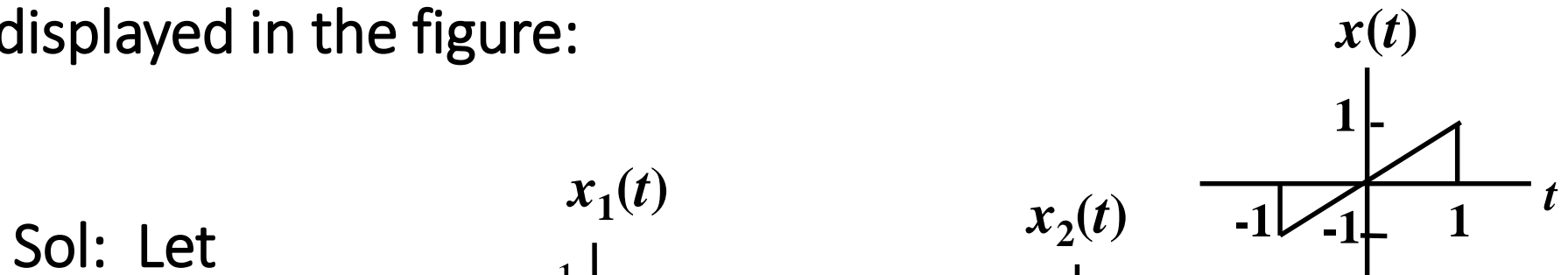
$$\mathcal{F}\{\text{sgn}(t)\} = \lim_{\alpha \rightarrow 0} \left(\frac{1}{\alpha + j\omega} - \frac{1}{\alpha - j\omega} \right) = \frac{2}{j\omega}$$

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

4.3 Properties of The Continuous-Time Fourier Transform

Example 4.12

Calculate the Fourier transform $X(j\omega)$ for the signal $x(t)$ displayed in the figure:



$$g(t) = \frac{dx(t)}{dt} = \text{[rectangular pulse from } -1 \text{ to } 1 \text{ with height } 1] + \text{[two impulses at } t = -1 \text{ and } t = 1 \text{ with magnitude } -1 \text{]}$$

$$X_1(j\omega) = \left(\frac{2 \sin \omega}{\omega} \right), \quad X_2(j\omega) = -e^{j\omega} - e^{-j\omega}$$

$$G(j\omega) = j\omega X(j\omega) = \left(\frac{2 \sin \omega}{\omega} \right) - e^{j\omega} - e^{-j\omega}$$

$$X(j\omega) = \frac{2 \sin \omega}{j\omega^2} - \frac{2 \cos \omega}{j\omega}$$

4.3 Properties of The Continuous-Time Fourier Transform

4.3.5 Time and Frequency Scaling

If $x(t) \xleftrightarrow{FT} X(j\omega)$, then $x(at) \xleftrightarrow{FT} \frac{1}{|a|} X\left(\frac{j\omega}{a}\right)$, a is real

A linear scaling in time by a factor of a corresponds to a linear scaling in frequency by a factor of $1/a$, and vice versa.

Specially, when $a = -1$, we have $x(-t) \xleftrightarrow{FT} X(-j\omega)$

4.3.6 Duality

If $x(t) \xleftrightarrow{FT} X(j\omega)$, then $X(jt) \xleftrightarrow{FT} 2\pi x(-\omega)$

Example 4.13

Using duality to find $G(j\omega)$ of the signal $g(t) = \frac{2}{1+t^2}$.

Sol: From pair $x(t) = e^{-|t|} \xleftrightarrow{FT} X(j\omega) = \frac{2}{1+\omega^2}$

By duality property, we can write

$$g(t) = \frac{2}{1+t^2} \xleftrightarrow{FT} G(j\omega) = 2\pi e^{-|-\omega|} = 2\pi e^{-|\omega|}$$

4.3 Properties of The Continuous-Time Fourier Transform

Duality property shows that for any Fourier transform pair there is a *dual pair* with the time and frequency variables interchanged.

Differentiation in Frequency-domain:

$$-jtx(t) \xleftrightarrow{FT} \frac{dX(j\omega)}{d\omega}$$

Integration in Frequency-domain:

$$-\frac{1}{jt} x(t) + \pi x(0)\delta(t) \xleftrightarrow{FT} \int_{-\infty}^{\omega} X(\eta) d\eta$$

Frequency Shifting:

$$e^{j\omega_0 t} x(t) \xleftrightarrow{FT} X(j(\omega - \omega_0))$$

4.3 Properties of The Continuous-Time Fourier Transform

4.3.7 Parseval's Relation

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

energy-density spectrum
(能量密度谱)

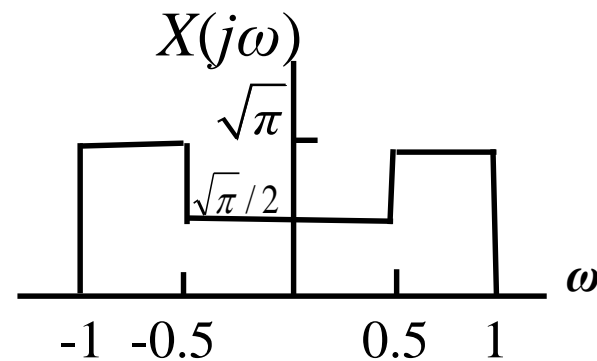
Parseval's relation says that this total energy may be determined either by computing energy per unit time ($|x(t)|^2$) and integrating over all time or by computing the energy per unit frequency ($|X(j\omega)|^2 / 2\pi$) and integrating over all frequencies.

4.3 Properties of The Continuous-Time Fourier Transform

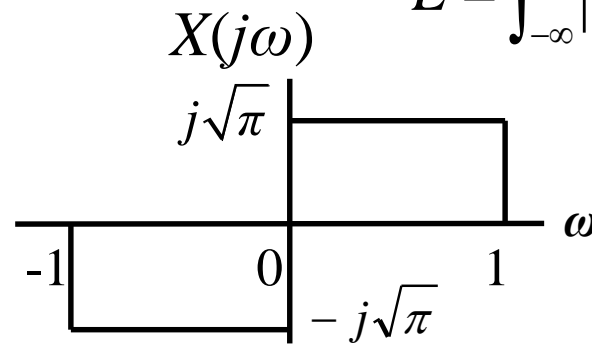
Example 4.14

For each of the Fourier transforms shown in figures (a) and (b), evaluate the following time-domain expressions:

$$E = \int_{-\infty}^{\infty} |x(t)|^2 dt, \quad D = \left. \frac{d}{dt} x(t) \right|_{t=0}$$



(a)



(b)

Sol: $E_a = \frac{5}{8}$ $E_b = 1$ (Parseval's Relation)

$$g(t) = \frac{d}{dt} x(t) \xleftrightarrow{FT} G(j\omega) = j\omega X(j\omega)$$

$$D = g(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} G(j\omega) d\omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} j\omega X(j\omega) d\omega$$

$$D_a = 0 \quad D_b = -\frac{\sqrt{\pi}}{2\pi}$$

4.4 The Convolution Property

Consider the convolution integral: $y(t) = x(t) * h(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau$

The Fourier transform of $y(t)$ is:

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

Interchanging the order of integration and noting that $x(\tau)$ does not depend on t , we have $Y(j\omega) = \int_{-\infty}^{\infty} x(\tau) \left[\int_{-\infty}^{\infty} h(t - \tau)e^{-j\omega t} dt \right] d\tau$


$$e^{-j\omega\tau} H(j\omega)$$

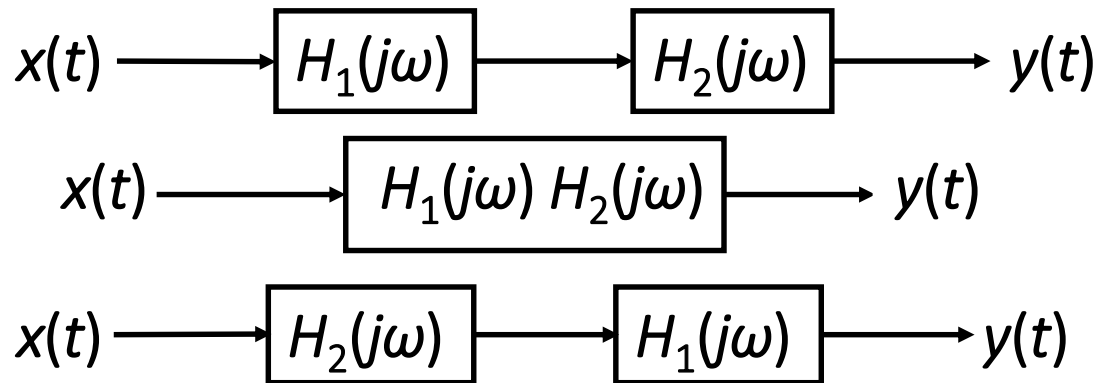
$$Y(j\omega) = \int_{-\infty}^{\infty} x(\tau)e^{-j\omega\tau} H(j\omega)d\tau = H(j\omega) \int_{-\infty}^{\infty} x(\tau)e^{-j\omega\tau} d\tau = H(j\omega)X(j\omega)$$

$$y(t) = h(t) * x(t) \xleftrightarrow{FT} Y(j\omega) = H(j\omega)X(j\omega)$$

The Fourier transform maps the convolution of two signals into the product of their Fourier transforms. $H(j\omega)$, the frequency response, is the Fourier transform of the impulse response $h(t)$. It captures the change in complex amplitude of the Fourier transform of the input at each frequency ω .

4.4 The Convolution Property

➤ The frequency response $H(j\omega)$ also can characterize an LTI system, just as its inverse transform, the unit impulse response $h(t)$.



Three equivalent LTI systems. Here, each LTI system is represented by $H(j\omega)$

➤ $H(j\omega)$ **cannot** be defined for every LTI system.

➤ Since essentially all **physical or practical** signals satisfy the last two conditions in Dirichlet conditions, the condition of absolutely integrable becomes the determining factor which can guarantee the existence of the Fourier transform $H(j\omega)$ of $h(t)$. That is, only **a stable LTI system has a frequency response $H(j\omega)$** .

4.4 The Convolution Property

Example 4.15

Consider an integrator — that is, an LTI system specified by the equation

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

Sol: Since $y(t) = x(t) * u(t) = \int_{-\infty}^{\infty} x(\tau) u(t - \tau) d\tau = \int_{-\infty}^t x(\tau) d\tau$

So the impulse response for this system is the unit step $u(t)$.

The frequency response of the system is $H(j\omega) = \frac{1}{j\omega} + \pi\delta(\omega)$

Using the convolution property, we have

$$Y(j\omega) = H(j\omega)X(j\omega)$$

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

$$= \frac{1}{j\omega} X(j\omega) + \pi X(0) \delta(\omega)$$

4.4 The Convolution Property

Example 4.16

Find the response of an LTI system with $h(t) = e^{-at}u(t)$, $a > 0$ to the input signal $x(t) = e^{-bt}u(t)$, $b > 0$.

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

Expand $Y(j\omega)$ in a *partial-fraction expansion* (部分分式展开法).

When $b \neq a$, let $Y(j\omega) = \frac{A}{(a + j\omega)} + \frac{B}{(b + j\omega)} = \frac{1}{b - a} \left[\frac{1}{a + j\omega} - \frac{1}{b + j\omega} \right]$

$$A = (a + j\omega)Y(j\omega) \Big|_{j\omega = -a} = \frac{1}{b + j\omega} \Big|_{j\omega = -a} = \frac{1}{b - a}$$

$$B = (b + j\omega)Y(j\omega) \Big|_{j\omega = -b} = \frac{1}{a + j\omega} \Big|_{j\omega = -b} = \frac{-1}{b - a}$$

$$y(t) = \frac{1}{b - a} \left[e^{-at}u(t) - e^{-bt}u(t) \right] \quad b \neq a$$

4.4 The Convolution Property

$$\text{When } b = a, \quad Y(j\omega) = \frac{1}{(a + j\omega)^2} = j \frac{d}{d\omega} \left[\frac{1}{(a + j\omega)} \right]$$

From the differentiation in the frequency-domain property,

$$e^{-at}u(t) \xleftrightarrow{FT} \frac{1}{a + j\omega}$$

$$te^{-at}u(t) \xleftrightarrow{FT} j \frac{d}{d\omega} \left[\frac{1}{a + j\omega} \right] = \frac{1}{(a + j\omega)^2}$$

Consequently,

$$y(t) = te^{-at}u(t) \quad b = a$$

4.4 The Convolution Property

Example 4.17

Determine the response of an ideal low-pass filter to an input signal $x(t)$ that has the form of a *sinc* function, $x(t) = \frac{\sin \omega_i t}{\pi t}$

Sol: The impulse response of the ideal low-pass filter is of a similar form: $h(t) = \frac{\sin \omega_c t}{\pi t}$

$$X(j\omega) = \begin{cases} 1 & |\omega| \leq \omega_i \\ 0 & \text{elsewhere} \end{cases}, \quad H(j\omega) = \begin{cases} 1 & |\omega| \leq \omega_c \\ 0 & \text{elsewhere} \end{cases}$$

Therefore, $Y(j\omega) = \begin{cases} 1 & |\omega| \leq \omega_0 \\ 0 & \text{elsewhere} \end{cases}$ where $\omega_0 = \min(\omega_i, \omega_c)$.

Finally, the inverse Fourier transform of $Y(j\omega)$ is given by

$$y(t) = \begin{cases} \frac{\sin \omega_c t}{\pi t} = h(t) & \text{if } \omega_c \leq \omega_i \\ \frac{\sin \omega_i t}{\pi t} = x(t) & \text{if } \omega_i \leq \omega_c \end{cases}$$

4.5 The Multiplication Property

$$r(t) = s(t)p(t) \xleftrightarrow{FT} R(j\omega) = \frac{1}{2\pi} [S(j\omega) * P(j\omega)]$$

amplitude modulation property (幅度调制定理)

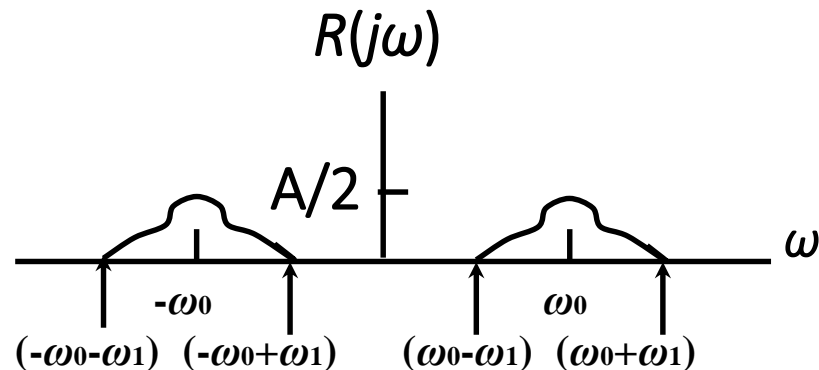
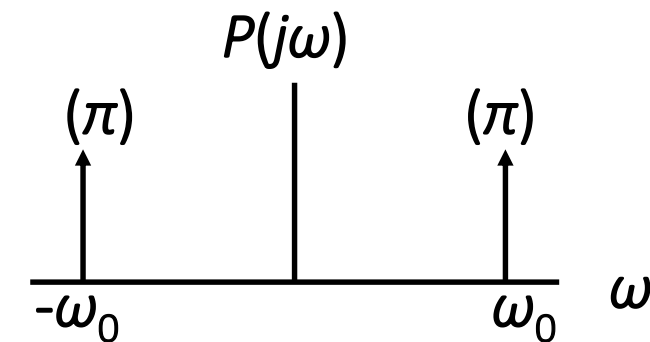
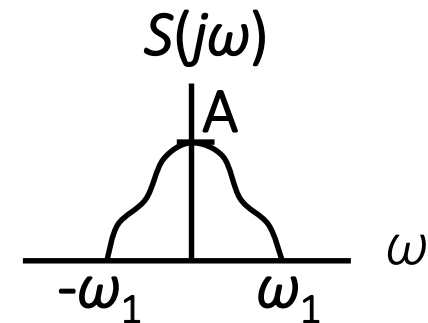
Example 4.18

Let $s(t)$ be a signal whose spectrum $S(j\omega)$ is depicted in the following Figure. Also, consider the signal $p(t) = \cos \omega_0 t$,

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

From the multiplication property:

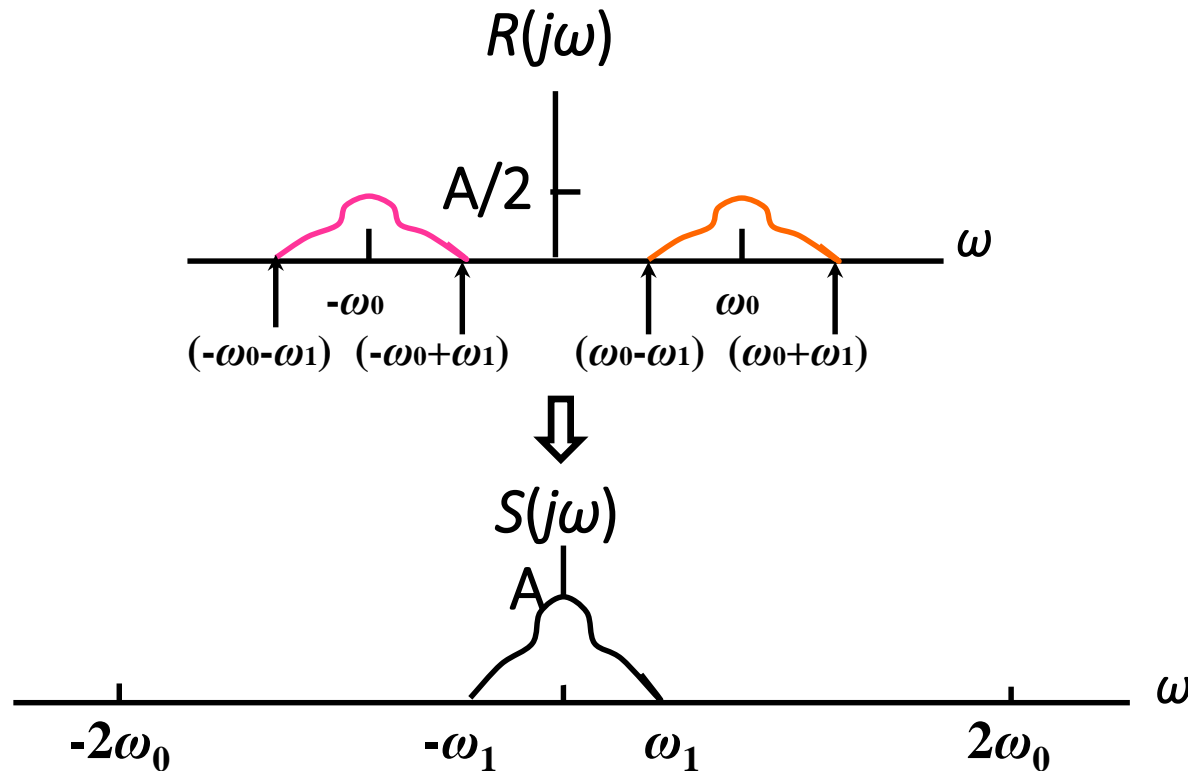
$$R(j\omega) = \frac{1}{2\pi} S(j\omega) * P(j\omega) = \frac{1}{2} S(j(\omega - \omega_0)) + \frac{1}{2} S(j(\omega + \omega_0))$$



4.5 The Multiplication Property

Example 4.19

Let us consider $r(t)$ as obtained in Example 4.18, and let $g(t) = r(t)p(t)$, we will show how to recover the modulated signal $s(t)$.



4.6 Systems Characterized By Linear Constant Coefficient Differential Equations

For a *stable* LTI system which is described by a linear constant-coefficient differential equation of the form :

$$\sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k} = \sum_{k=0}^M b_k \frac{d^k x(t)}{dt^k}$$
$$\mathcal{F} \left\{ \sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k} \right\} = \mathcal{F} \left\{ \sum_{k=0}^M b_k \frac{d^k x(t)}{dt^k} \right\} \quad \sum_{k=0}^N a_k \mathcal{F} \left\{ \frac{d^k y(t)}{dt^k} \right\} = \sum_{k=0}^M b_k \mathcal{F} \left\{ \frac{d^k x(t)}{dt^k} \right\}$$
$$\sum_{k=0}^N a_k (j\omega)^k Y(j\omega) = \sum_{k=0}^M b_k (j\omega)^k X(j\omega) \quad (\text{differentiation})$$

Equivalently,

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

➤ $H(j\omega)$ is a ratio of polynomials in $(j\omega)$, so it is a rational function.

4.6 Systems Characterized By Linear Constant Coefficient Differential Equations

Example 4.20

Consider a stable LTI system that is characterized by the differential equation $\frac{d^2 y(t)}{dt^2} + 4\frac{dy(t)}{dt} + 3y(t) = \frac{dx(t)}{dt} + 2x(t)$, determine its impulse response.

Sol: The frequency response is

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

To determine the corresponding impulse response, we use the method of partial-fraction expansion:

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

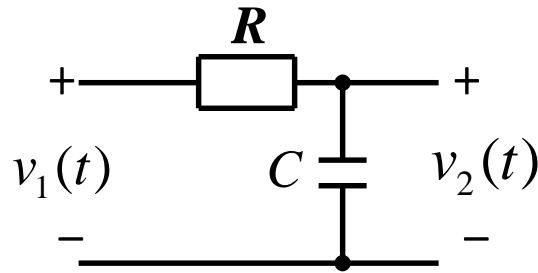
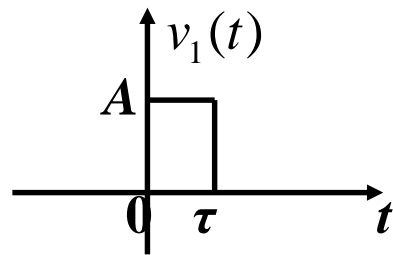
Thus, the impulse response is

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

4.6 Systems Characterized By Linear Constant Coefficient Differential Equations

Example 4.21

A RC circuit is as follows, $v_1(t)$ is the input, determine $v_2(t)$.



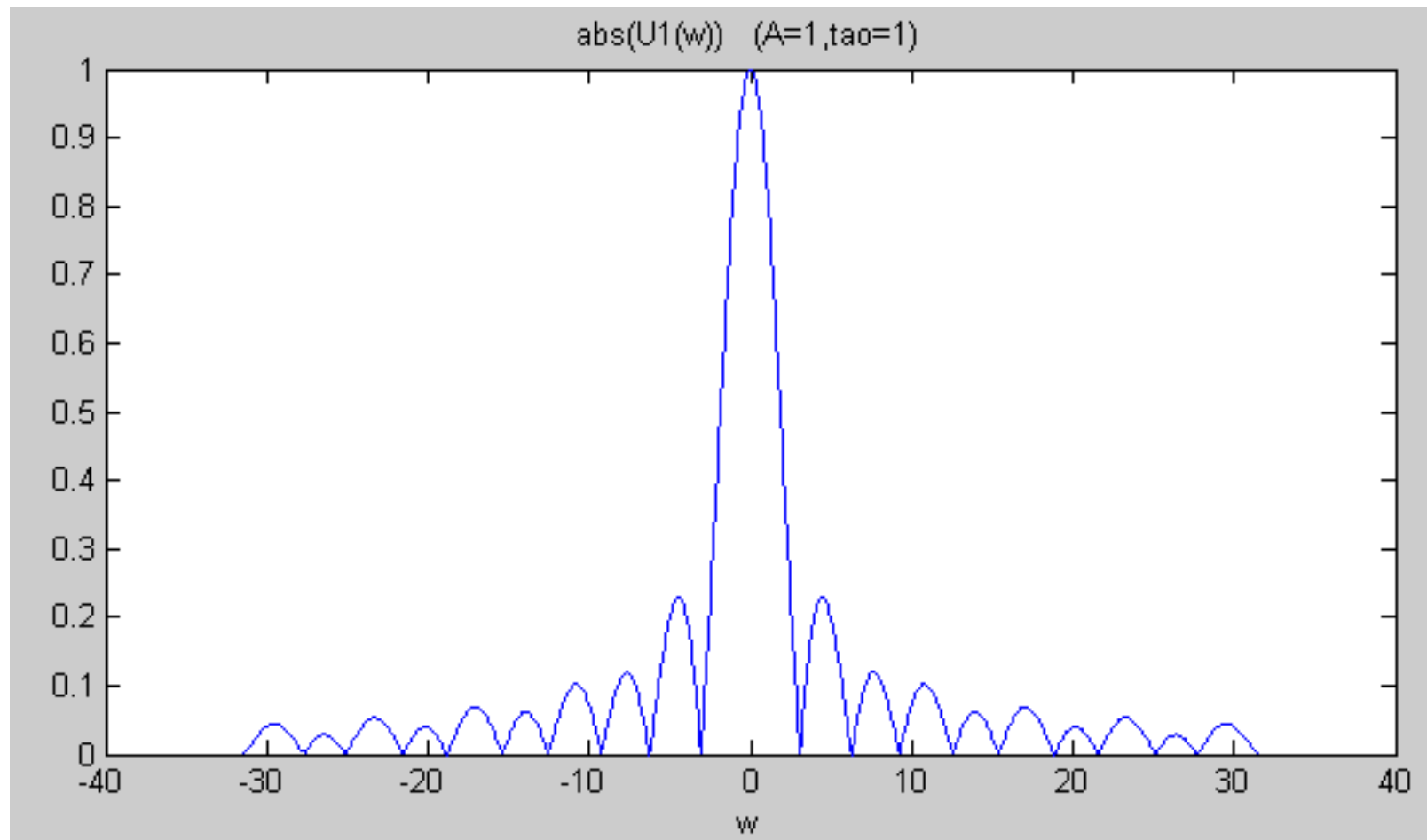
Sol: 1) Compute $V_1(j\omega)$

$$v_1(t) = A[u(t) - u(t - \tau)]$$

$$V_1(j\omega) = A\left[\pi\delta(\omega) + \frac{1}{j\omega} - \pi\delta(\omega)e^{-j\omega\tau} - \frac{e^{-j\omega\tau}}{j\omega}\right]$$

$$= \frac{A}{j\omega}(1 - e^{-j\omega\tau}) = \frac{2A}{\omega} \sin \frac{\omega\tau}{2} e^{-j\frac{\omega\tau}{2}} = A\tau \text{Sa}\left(\frac{\omega\tau}{2}\right) e^{-j\frac{\omega\tau}{2}}$$

4.6 Systems Characterized By Linear Constant Coefficient Differential Equations



4.6 Systems Characterized By Linear Constant Coefficient Differential Equations

2) Compute $H(j\omega)$

$$H(j\omega) = \frac{V_2(j\omega)}{V_1(j\omega)} = \frac{\frac{1}{j\omega C}}{R + \frac{1}{j\omega C}} = \frac{1}{1 + j\omega RC} = \frac{\frac{1}{RC}}{\frac{1}{RC} + j\omega}$$

Let $\frac{1}{RC} = \alpha = \frac{1}{\tau_0}$

$$H(j\omega) = \frac{\alpha}{\alpha + j\omega} = \frac{\alpha}{\sqrt{\alpha^2 + \omega^2}} e^{-j \arctan \frac{\omega}{\alpha}}$$

4.6 Systems Characterized By Linear Constant Coefficient Differential Equations

3) Compute $V_2(j\omega)$

$$V_2(j\omega) = V_1(j\omega)H(j\omega)$$

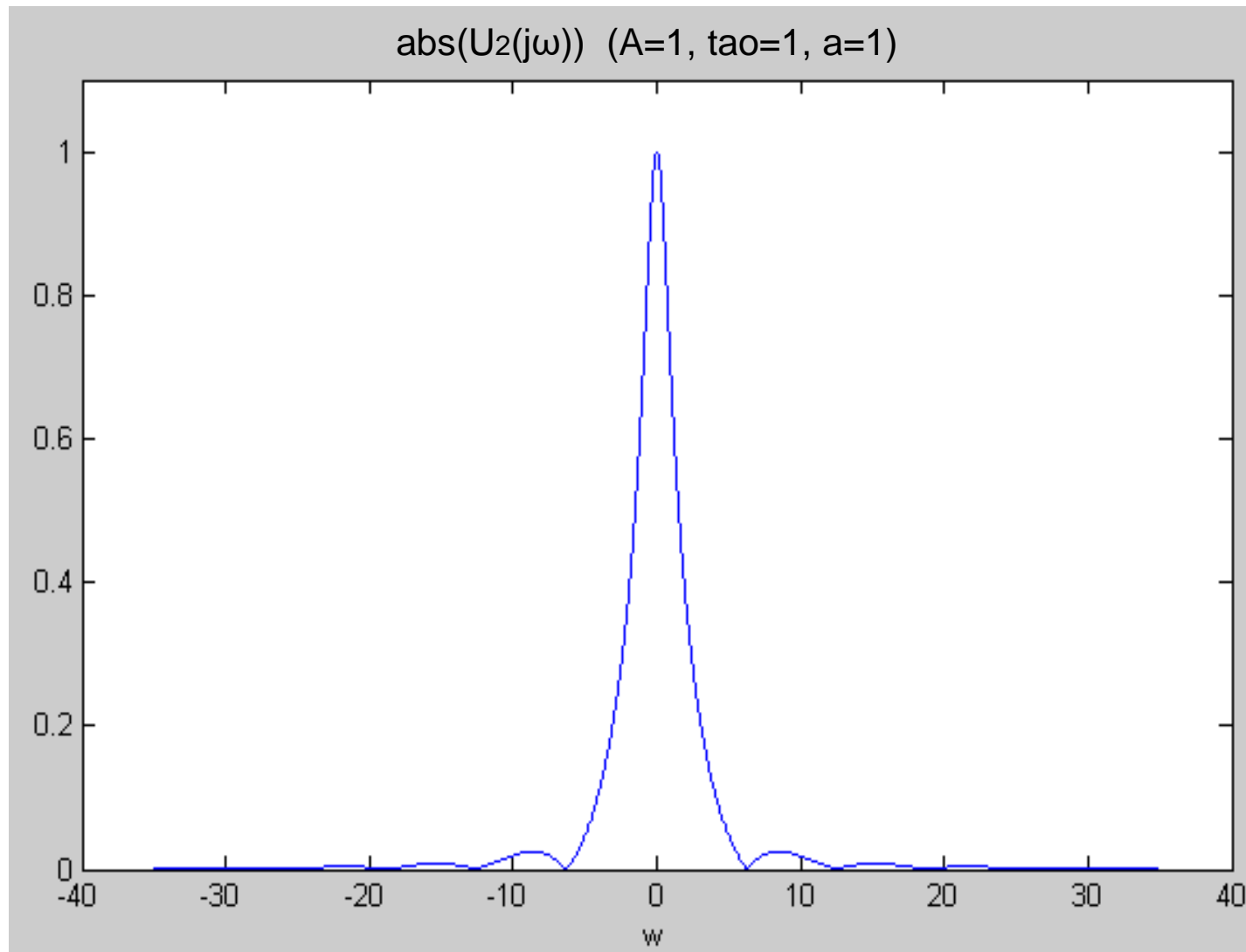
$$= A \tau \text{Sa}\left(\frac{\omega\tau}{2}\right) e^{-j\frac{\omega\tau}{2}} \cdot \frac{\alpha}{\sqrt{\alpha^2 + \omega^2}} e^{-j \arctan \frac{\omega}{\alpha}}$$

$$= \frac{\alpha A \tau}{\sqrt{\alpha^2 + \omega^2}} \text{Sa}\left(\frac{\omega\tau}{2}\right) e^{-j\left(\frac{\omega\tau}{2} + \arctan \frac{\omega}{\alpha}\right)}$$

$$|V_2(j\omega)| = \frac{\alpha A \tau}{\sqrt{\alpha^2 + \omega^2}} \left| \text{Sa}\left(\frac{\omega\tau}{2}\right) \right| = \frac{2\alpha A \left| \sin\left(\frac{\omega\tau}{2}\right) \right|}{\omega \sqrt{\alpha^2 + \omega^2}}$$

$$\varphi_2(\omega) = \begin{cases} -\left(\frac{\omega\tau}{2} + \arctan \frac{\omega}{\alpha}\right), \sin\left(\frac{\omega\tau}{2}\right) > 0 \\ \pm \pi - \left(\frac{\omega\tau}{2} + \arctan \frac{\omega}{\alpha}\right), \sin\left(\frac{\omega\tau}{2}\right) < 0 \end{cases}$$

4.6 Systems Characterized By Linear Constant Coefficient Differential Equations



4.6 Systems Characterized by Equations

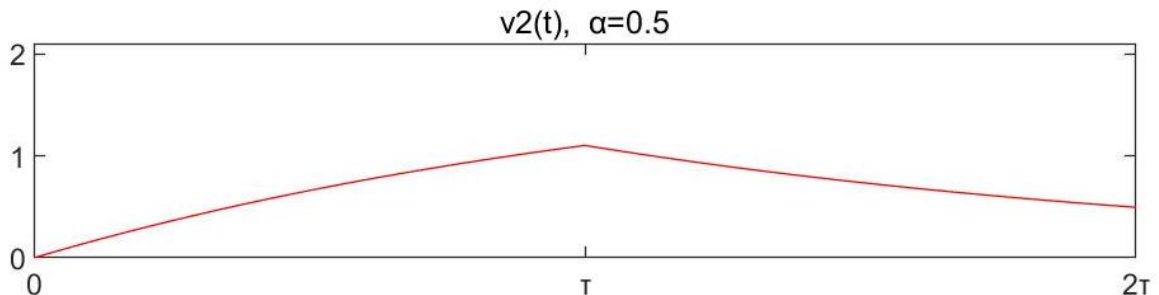
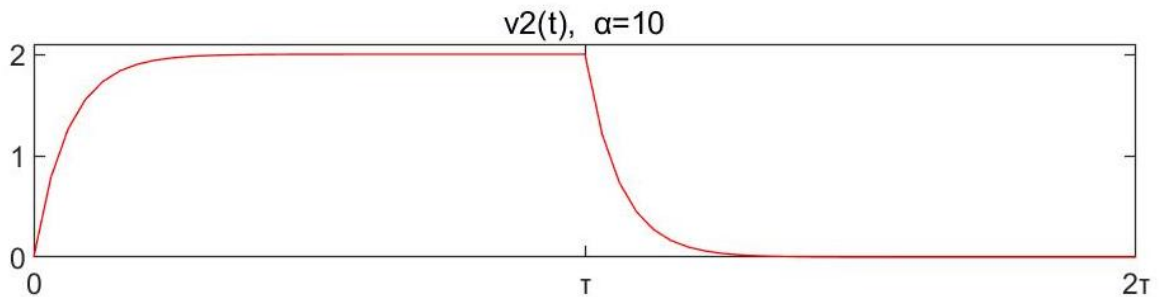
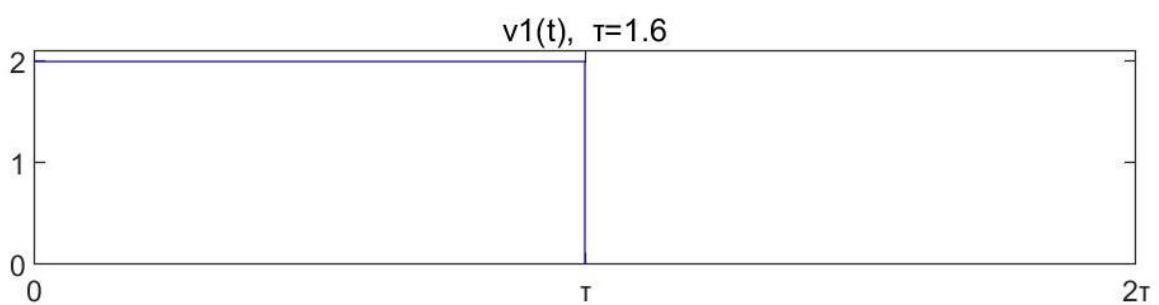
4) Compute $v_2(t) = \mathcal{F}$

$$V_2(j\omega) = A(1 - e^{-j\omega\tau})$$

$$= \frac{A(1 - e^{-j\omega\tau})}{j\omega}$$

$$\therefore v_2(t) = A[u(t) - u(t - \tau)]$$

$$= A(1 - e^{-\alpha t})$$



Rising and falling characteristics in time domain:

$v_1(t) : t=0, \tau$ Change very fast (jump) — **Abundant high frequencies**
 $v_2(t) : t=0, \tau$ Change slowly, need a period of time to rise or fall
 — **High frequencies are attenuated**

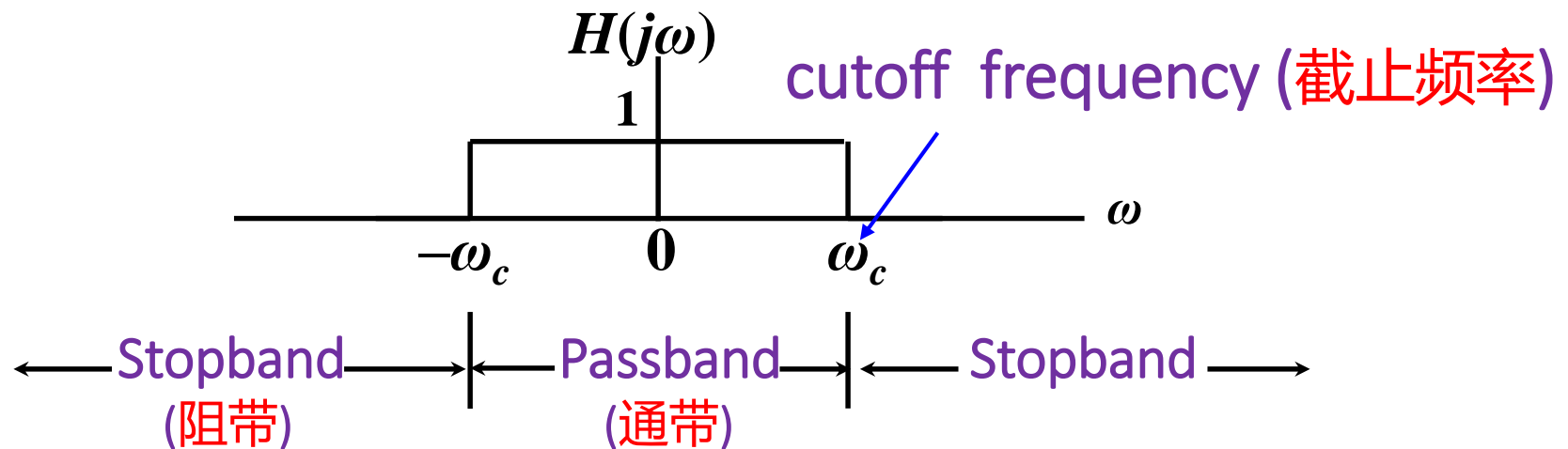
4.7 Frequency-Selective Filters

4.7.1 Introduction to Ideal Frequency-Selective Filters

- *Filtering*: a process in which the relative amplitudes of the frequency components in a signal are changed or some frequency components are eliminated entirely.
- *Frequency-selective filters* (频选滤波器): systems that are designed to pass some frequencies essentially undistorted and significantly attenuate or eliminate others.
- Types of frequency-selective filters
 - ✓ low-pass filter (低通滤波器)
 - ✓ high-pass filter (高通滤波器)
 - ✓ band-pass filter (带通滤波器)
 - ✓ band-stop filter (带阻滤波器)

4.7 Frequency-Selective Filters

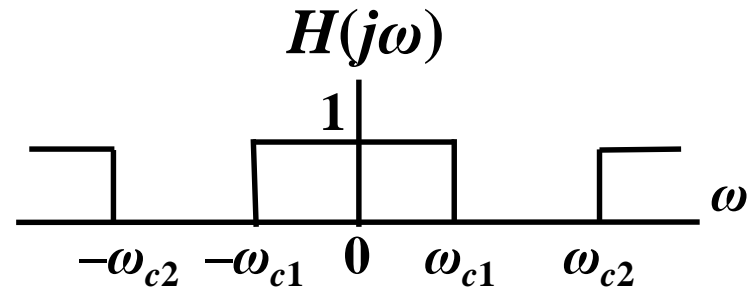
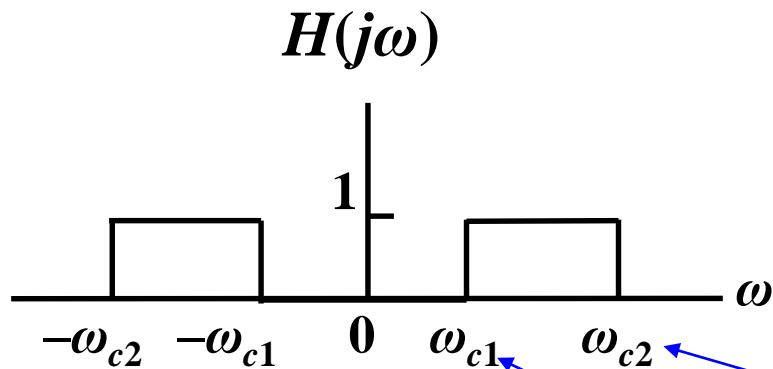
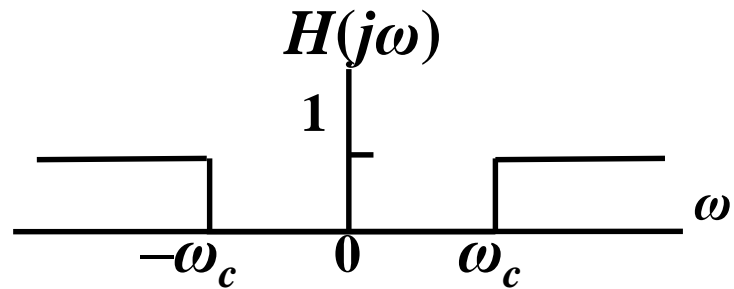
The frequency responses of a *zero-phase ideal* low-pass filter, a *zero-phase ideal* high-pass filter, a *zero-phase ideal* band-pass filter and a *zero-phase ideal* band-stop filter are illustrated in the following figures, respectively:



$$H(j\omega) = \begin{cases} 1, & |\omega| \leq \omega_c \\ 0, & |\omega| > \omega_c \end{cases}$$

4.7 Frequency-Selective Filters

$$H(j\omega) = \begin{cases} 1, & |\omega| \geq \omega_c \\ 0, & |\omega| < \omega_c \end{cases}$$



$$H(j\omega) = \begin{cases} 1, & \omega_{c2} \geq |\omega| \geq \omega_{c1} \\ 0, & |\omega| < \omega_{c1} \text{ or } |\omega| > \omega_{c2} \end{cases}$$

upper cutoff frequency
(上截止频率)
lower cutoff frequency
(下截止频率)

4.7 Frequency-Selective Filters

- The impulse response of the ideal low-pass filter is:

$$h(t) = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} e^{j\omega t} d\omega = \frac{\sin \omega_c t}{\pi t}$$

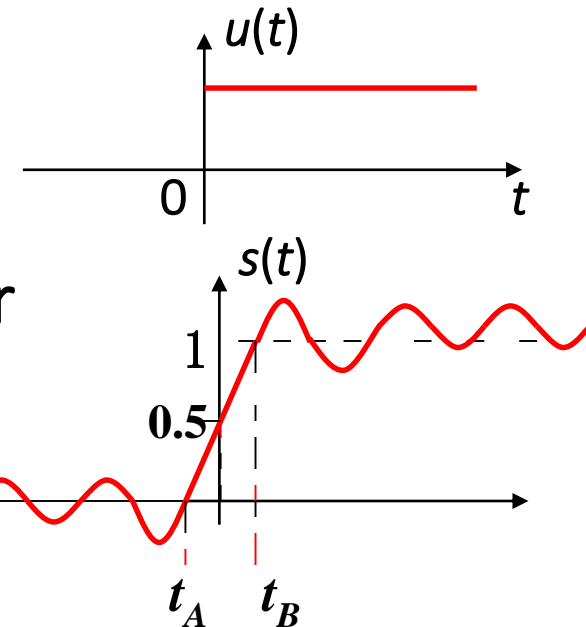
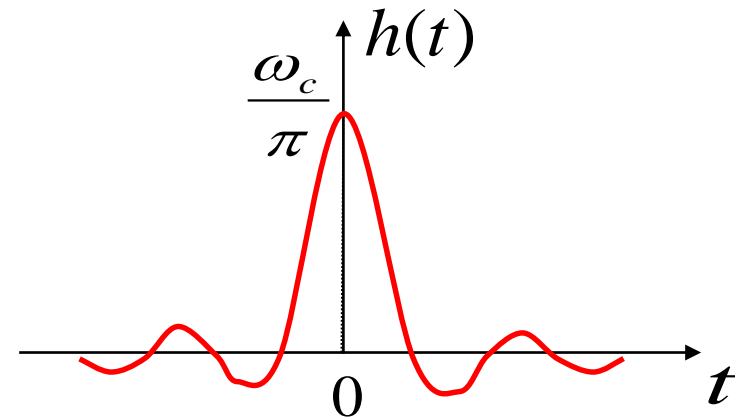
- The step response is:

$$s(t) = \frac{1}{2} + \frac{1}{\pi} Si(\omega_c t)$$

Here, $Si(\omega_c t) = \int_0^{\omega_c t} \frac{\sin x}{x} dx$

Different from the input signal — $u(t)$, output signal $s(t)$ needs a period of time to rise from 0 to 1, because frequencies higher than ω_c in $u(t)$ are rejected. And the period

$$t_r = t_B - t_A = \frac{3.84}{\omega_c}$$



4.7 Frequency-Selective Filters

- The frequency response of the ideal high-pass filter can be represented in terms of the frequency response of the low-pass filter as:

$$H_h(j\omega) = \begin{cases} 1, & |\omega| \geq \omega_c \\ 0, & |\omega| < \omega_c \end{cases} = 1 - \begin{cases} 1, & |\omega| \leq \omega_c \\ 0, & |\omega| > \omega_c \end{cases} = 1 - H_l(j\omega)$$

Thus, the impulse response of the ideal high-pass filter is:

$$h(t) = \delta(t) - \frac{\sin \omega_c t}{\pi t}$$

Clearly $h(t)$ is not causal, so the ideal high-pass filter **cannot be implemented**.

4.7 Frequency-Selective Filters

4.7.2 Realizable Systems and Binding Characteristic of $H(j\omega)$

- A physically realizable system must have its impulse response $h(t)$ satisfy $h(t) = h(t) \cdot u(t)$ (**sufficient and necessary condition**) – Time Domain condition
- Imposing causality on the system restricts the $H(j\omega)$ in significant ways. The **Paley-Wiener Criterion** says if $|H(j\omega)|^2$ is integrable, i.e., $\int_{-\infty}^{\infty} |H(j\omega)|^2 d\omega < \infty$, it can be proved that the **necessary condition** for a realizable $|H(j\omega)|$ (causal system) is:

$$\int_{-\infty}^{\infty} \frac{|\ln |H(j\omega)||}{1 + \omega^2} d\omega < \infty \quad (\text{I})$$

- ✓ To satisfy above condition, $H(j\omega)$ cannot be zero in any band of frequencies (although it can be zero at a finite number of frequencies).

4.7 Frequency-Selective Filters

✓ The decay rate of $|H(j\omega)|$ cannot be greater than that of exponentials.

Consider the causality of system with $|H(j\omega)| = e^{-|\omega|}$,

$$\begin{aligned}\lim_{B \rightarrow \infty} \int_{-B}^B \frac{|\ln |H(j\omega)||}{1 + \omega^2} d\omega &= \lim_{B \rightarrow \infty} \int_{-B}^B \frac{|\ln e^{-|\omega|}|}{1 + \omega^2} d\omega = \lim_{B \rightarrow \infty} \int_{-B}^B \frac{|\omega|}{1 + \omega^2} d\omega \\ &= 2 \times \lim_{B \rightarrow \infty} \int_0^B \frac{\omega}{1 + \omega^2} d\omega = \lim_{B \rightarrow \infty} \ln(1 + \omega^2) \Big|_0^B \rightarrow \infty\end{aligned}$$

This shows that exponential magnitude response $|H(j\omega)| = e^{-|\omega|}$ does not satisfy the equation (I). Consequently, a system with a magnitude response function which decays faster than exponential must be non-causal so that cannot be implemented.

□ It can be proved that a magnitude response function $|H(j\omega)|$ that is composed of rational polynomials can satisfy equation (I).

4.7 Frequency-Selective Filters

➤ Because of the causality restriction ($h(t)=h(t)\cdot u(t)$) there is some kind of mutually binding character between the real and imaginary parts or the magnitude and phase of $H(j\omega)$.

Specifically, if let $H(j\omega) = H_R(j\omega) + jH_I(j\omega)$

then

$$H_R(j\omega) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{H_I(j\eta)}{\omega - \eta} d\eta \quad (\text{II})$$

$$H_I(j\omega) = -\frac{1}{\pi} \int_{-\infty}^{\infty} \frac{H_R(j\eta)}{\omega - \eta} d\eta \quad (\text{III})$$

Equations (II) and (III) are referred to as *Hilbert Transform pair*.

Conclusion: The real and imaginary parts of the transform of a real, causal impulse response $h(t)$ can be determined from one another using the *Hilbert Transform*. (Problems 4.47 and 4.48)
So do the phase and logarithm of the magnitude of $H(j\omega)$.

4.7 Frequency-Selective Filters

4.7.3 A Real Simple RC Low-pass Filter with rational $H(j\omega)$

Input: source voltage $v_s(t)$;

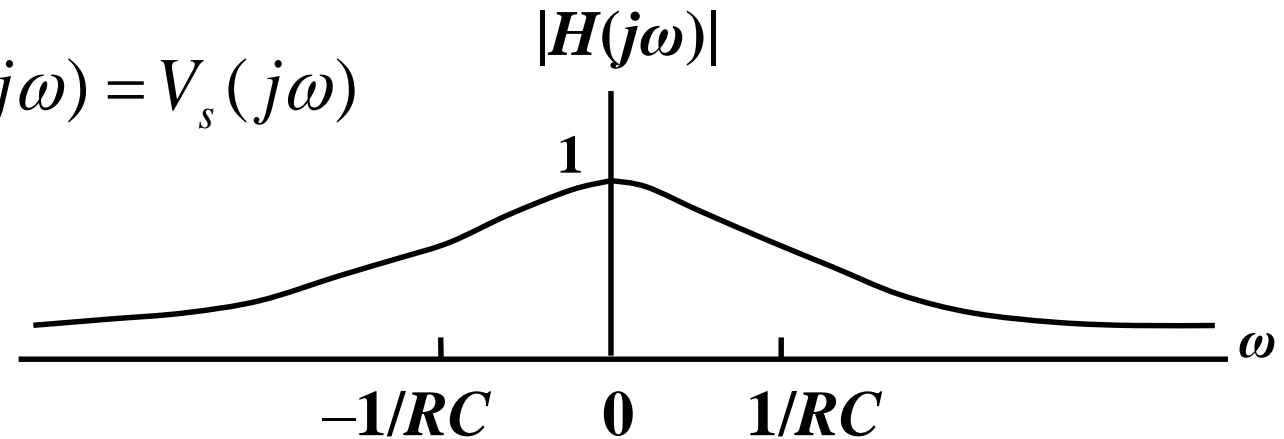
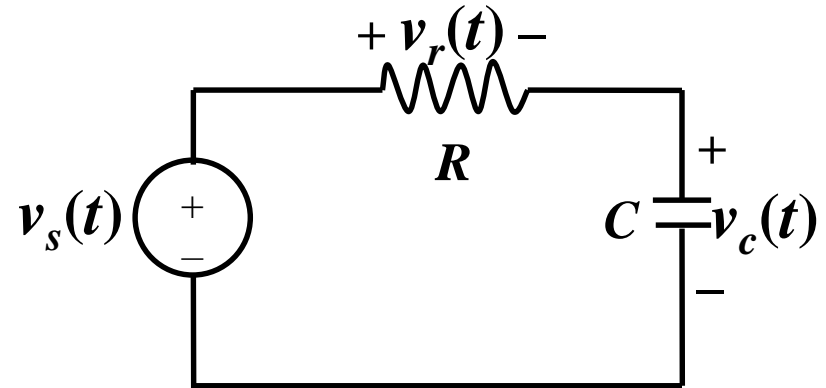
Output: capacitor voltage $v_c(t)$

$$RC \frac{dv_c(t)}{dt} + v_c(t) = v_s(t)$$

$$RCj\omega V_c(j\omega) + V_c(j\omega) = V_s(j\omega)$$

$$H_{lp}(j\omega) = \frac{V_c(j\omega)}{V_s(j\omega)}$$

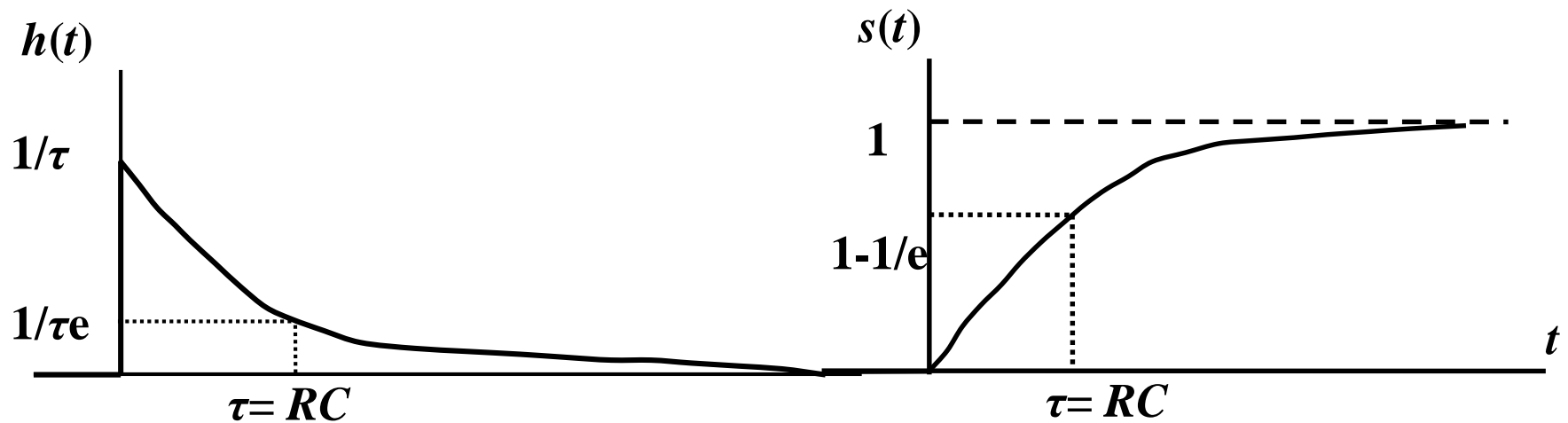
$$= \frac{1}{1 + RCj\omega} = \frac{1/RC}{1/RC + j\omega}$$



4.7 Frequency-Selective Filters

$$h(t) = \frac{1}{RC} e^{-t/RC} u(t)$$

$$s(t) = \left[1 - e^{-t/RC}\right] u(t)$$



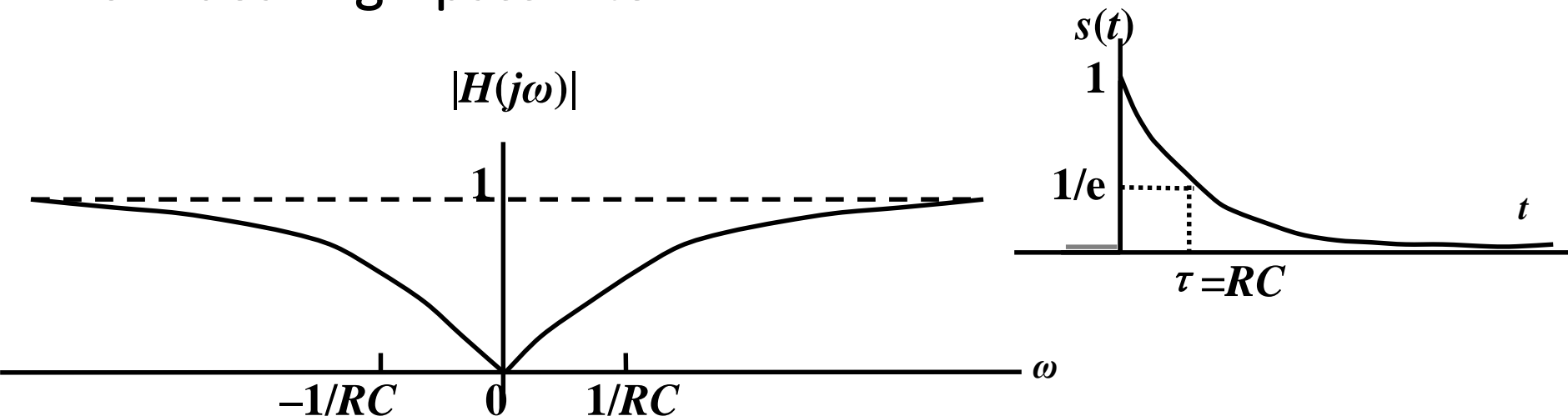
4.7 Frequency-Selective Filters

4.7.4 A Real Simple RC High-pass Filter with rational $H(j\omega)$

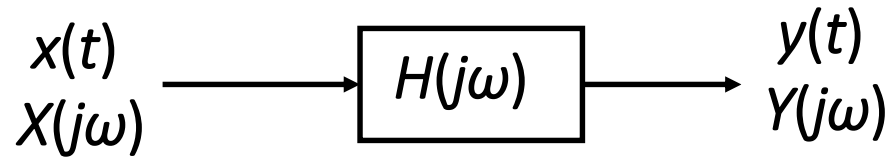
Input: source voltage $v_s(t)$; Output: resistor voltage $v_r(t)$

$$RC \frac{dv_r(t)}{dt} + v_r(t) = RC \frac{dv_s(t)}{dt}$$
$$H_{hp}(j\omega) = \frac{j\omega RC}{1 + j\omega RC} = 1 - \frac{1/RC}{1/RC + j\omega}$$
$$s(t) = e^{-t/RC} u(t)$$

Non-ideal high-pass filter



4.8 Transmission without Distortion



If the waveform of $y(t)$ is different from that of $x(t)$, distortion occurs.

➤ Linear systems can only introduce *linear distortion*, in which there isn't new frequencies generated.

➤ We can see from $Y(j\omega) = |H(j\omega)| |X(j\omega)| e^{j(\arg H(j\omega) + \arg X(j\omega))}$ that linear distortions include *magnitude distortion* and *phase distortion*.

In the case of $y(t) = Kx(t - t_0)$, $x(t)$ is transmitted without distortion.

Applying Fourier transform we have $Y(j\omega) = KX(j\omega)e^{-j\omega t_0}$

$$H(j\omega) = Ke^{-j\omega t_0} \quad \text{or} \quad \begin{cases} |H(j\omega)| = K \\ \angle H(j\omega) = -\omega t_0 \end{cases}$$

4.8 Transmission without Distortion

Suppose $x(t) = E_1 \sin \omega_0 t + E_2 \sin 2\omega_0 t$

Then

$$\begin{aligned} y(t) &= KE_1 \sin(\omega_0 t + \varphi_1) + KE_2 \sin(2\omega_0 t + \varphi_2) \\ &= KE_1 \sin \omega_0 \left(t + \frac{\varphi_1}{\omega_0}\right) + KE_2 \sin 2\omega_0 \left(t + \frac{\varphi_2}{2\omega_0}\right) \end{aligned}$$

To guarantee undistorted,

$$\frac{\varphi_1}{\omega_0} = \frac{\varphi_2}{2\omega_0} = \text{const} = -t_0$$

Consider non-periodic inputs case, we can write without loss of generality that

$$\varphi(\omega) = -\omega t_0 = \angle H(j\omega)$$

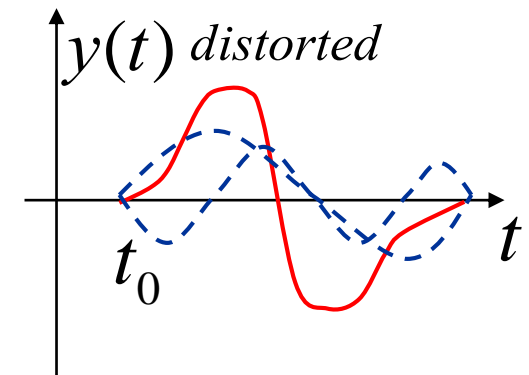
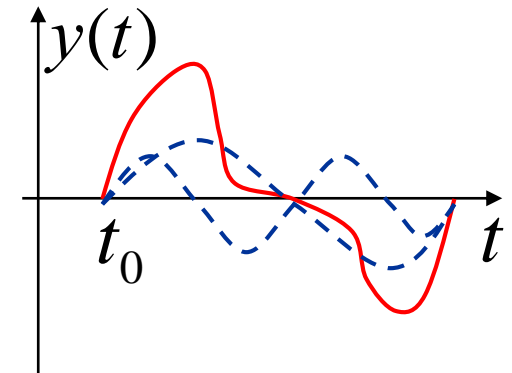
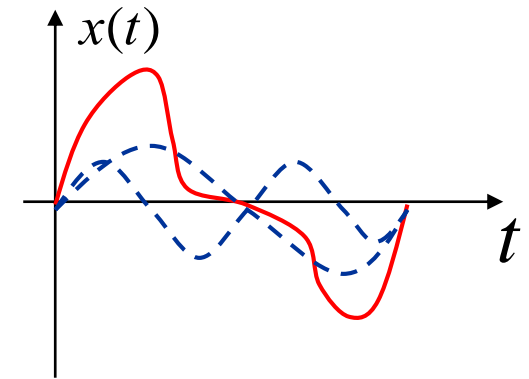
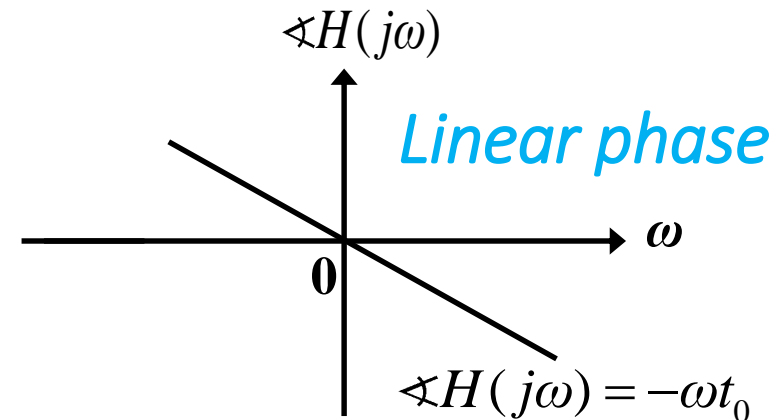
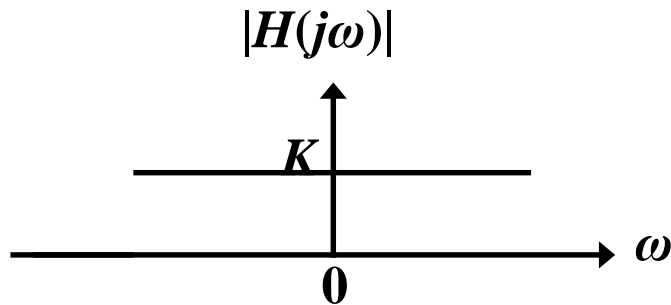


Illustration of phase distortion

4.8 Transmission without Distortion

➤ Conclusions: a linear system that can transmit signals applying to it as input without distortion must have a constant magnitude response and a phase response directly proportional to frequencies.



Linear phase shifts lead to very simple and easily understood change in a signal.

Group Delay: $\tau(\omega) = -\frac{d\{\angle H(j\omega)\}}{d\omega}$

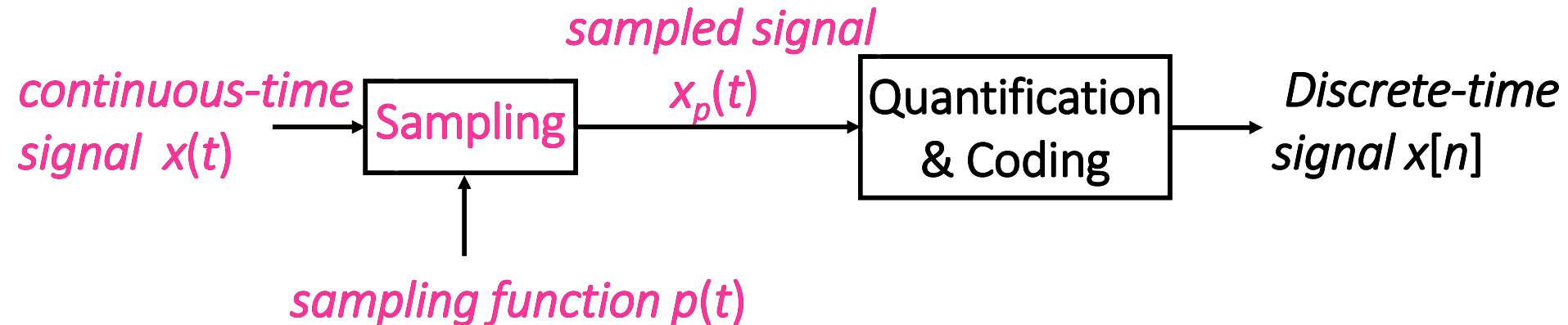
$$h(t) = K\delta(t - t_0)$$

Systems with this impulse response is ideal!

More about this refer to Section 6.2.

4.9 Sampling

Question: Why do we need to study *sampling*?

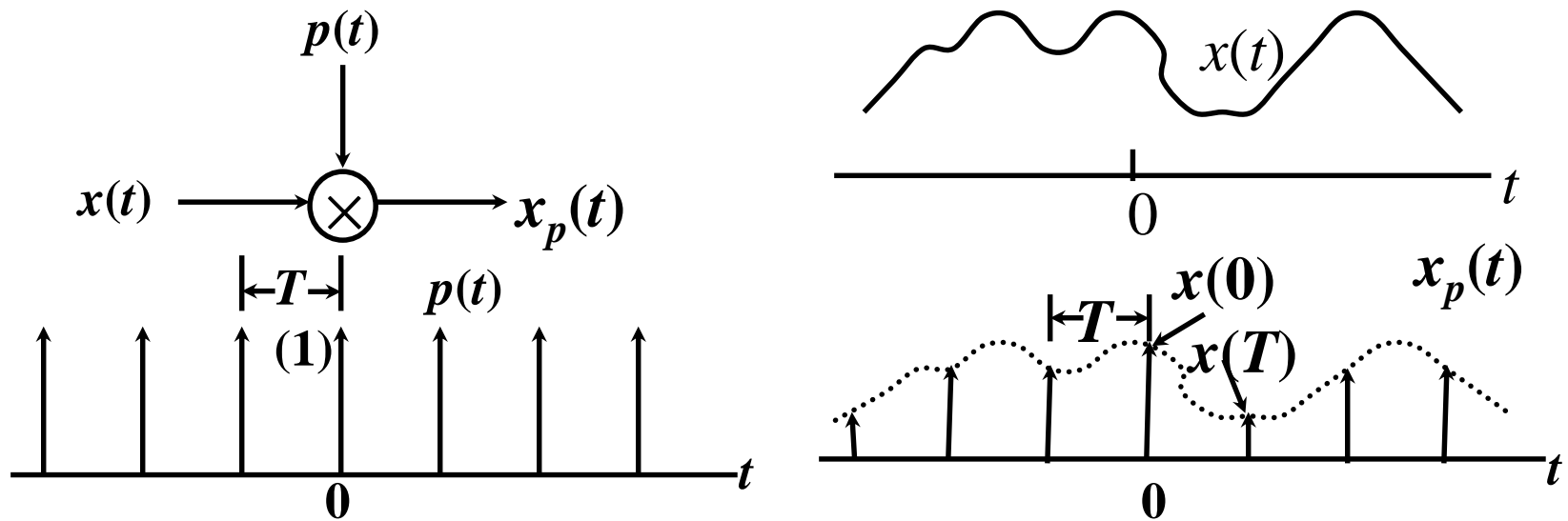


To make sure that $x[n]$ keep most information in $x(t)$, or $x(t)$ can recovered from $x[n]$, two questions need to be answered:

1. What is the Fourier transform of $x_p(t)$? How is it related to that of $x(t)$?
2. How to recover $x(t)$ from $x_p(t)$?

4.9 Sampling

4.9.1 Impulse-train sampling (冲激串采样)



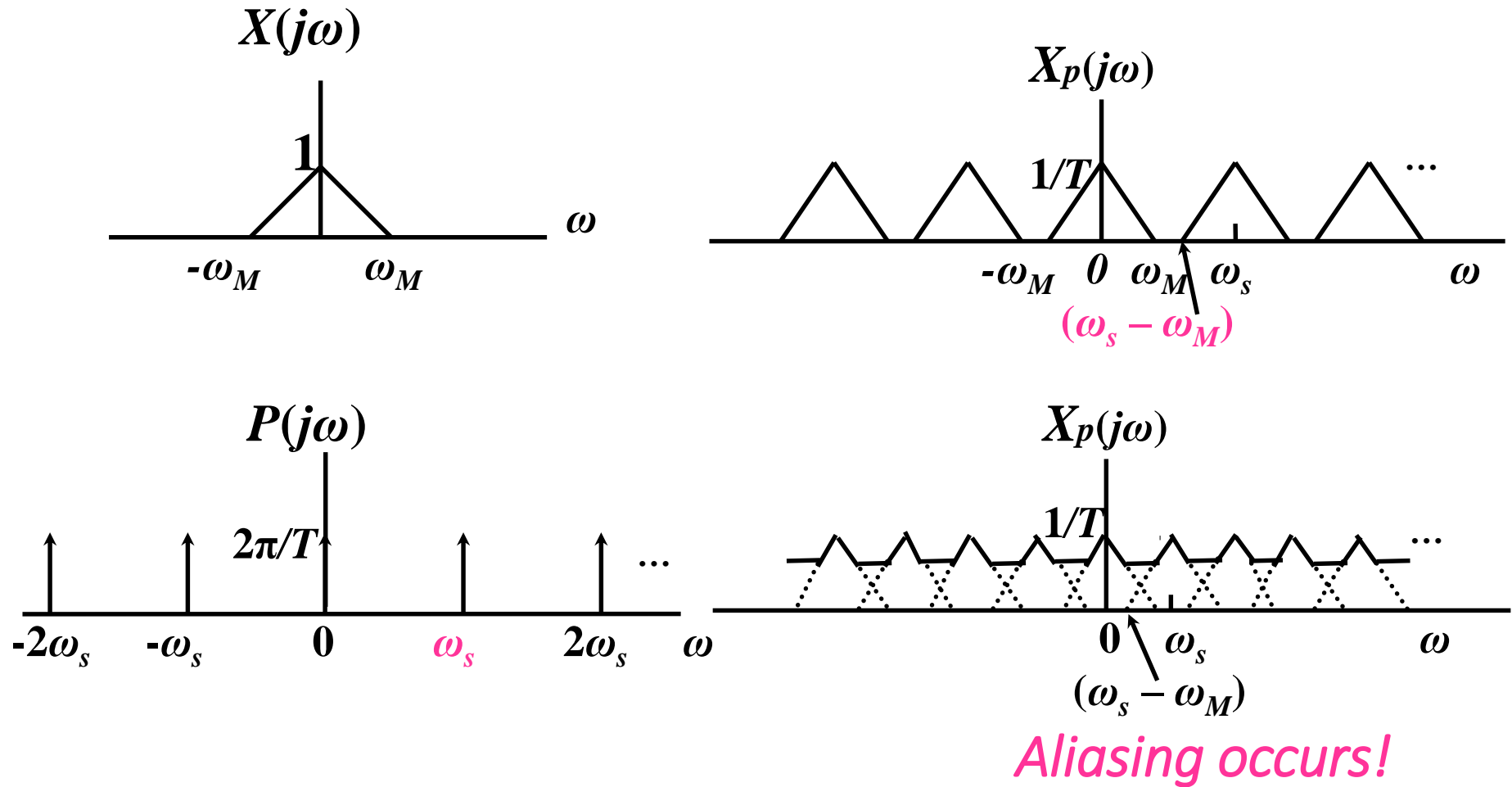
Mechanism of Impulse-train sampling

$$x_p(t) = x(t) \cdot \sum_{n=-\infty}^{\infty} \delta(t - nT) \quad X_p(j\omega) = \frac{1}{2\pi} [X(j\omega) * P(j\omega)] \quad P(j\omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s)$$

$$X_p(j\omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

Answer for Q1: $X_p(j\omega)$ is a periodic function of ω consisting of a superposition of shifted replicas of $X(j\omega)$, scaled by $\frac{1}{T}$.

4.9 Sampling



Effect in the frequency domain of sampling in the time domain

4.9 Sampling

4.9.2 Sampling Theorem (采样定理)

Let $x(t)$ be a band-limited signal with $X(j\omega)=0$ for $|\omega|>\omega_M$. Then $x(t)$ is uniquely determined by its samples $x(nT)$, $n=0, \pm 1, \pm 2, \dots$, if

$$\omega_s > 2\omega_M$$

where

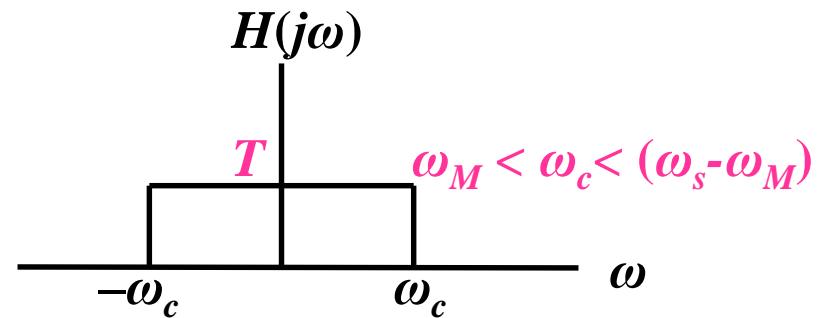
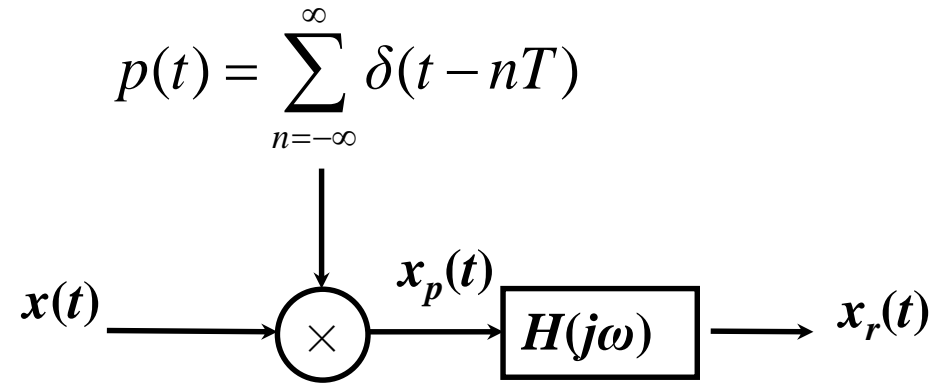
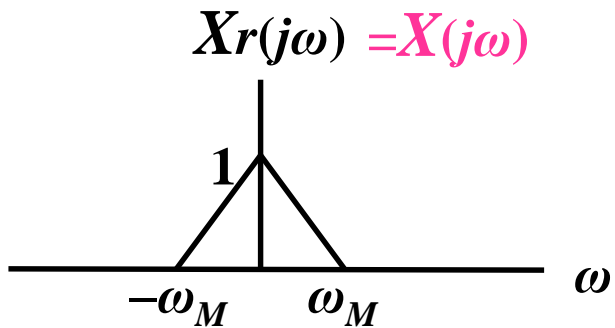
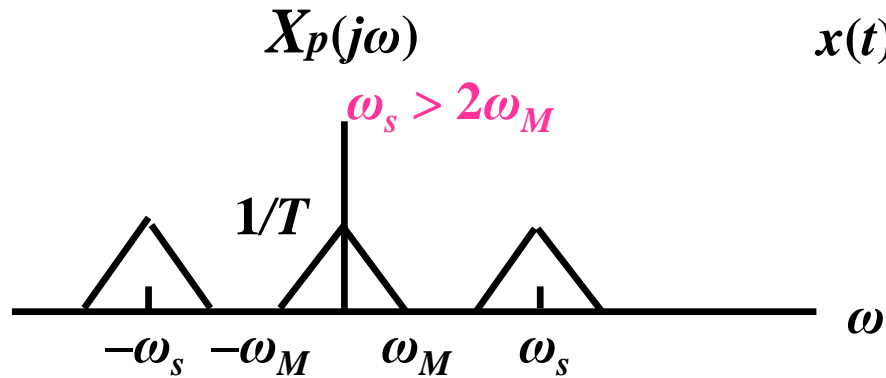
$$\omega_s = \frac{2\pi}{T}$$

Nyquist frequency
(奈奎斯特频率)



Given these samples, we can reconstruct $x(t)$ by generating a periodic impulse train in which successive impulses have amplitudes that are successive sample values. This impulse train is then processed through an ideal low-pass filter with **gain** T and **cutoff frequency** greater than ω_M and less than $\omega_s - \omega_M$. The resulting output signal will exactly equal $x(t)$.

4.9 Sampling



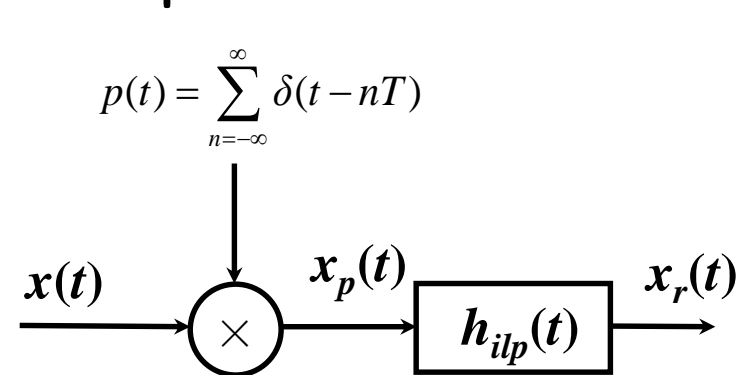
Exact recovery of a continuous-time signal from its samples using an ideal low-pass filter

4.9 Sampling

4.9.3 Reconstruction of A Signal From Its Samples Using Interpolation

➤ *Interpolation* (插值): a procedure in which the fitting (拟合) of a continuous-time signal to a set of sample values. It is commonly used for reconstruction of a function from samples.

➤ Ways to interpolate: zero-order hold interpolation、linear interpolation、band-limited interpolation ...



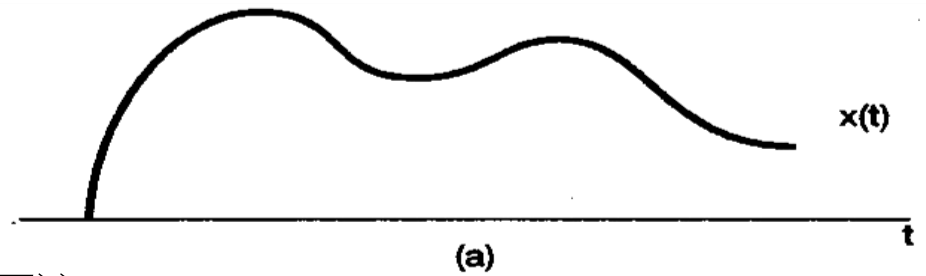
$$x_r(t) = x_p(t) * h(t) = \sum_{n=-\infty}^{\infty} x(nT)h(t - nT)$$

$$h(t) = \frac{\omega_c T \sin(\omega_c t)}{\pi \omega_c t}$$

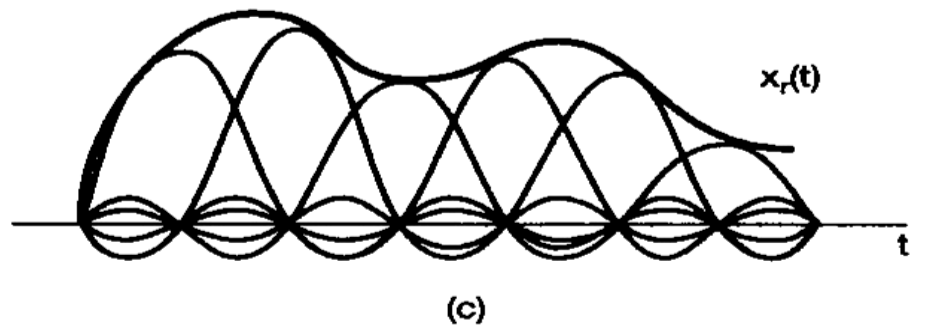
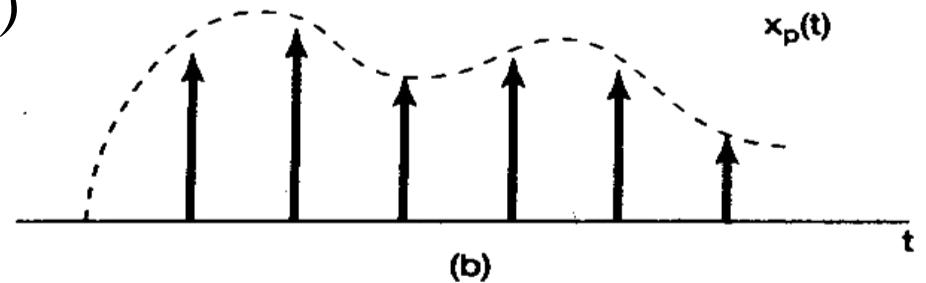
$$x_r(t) = \sum_{n=-\infty}^{\infty} x(nT) \frac{\omega_c T}{\pi} \frac{\sin(\omega_c (t - nT))}{\omega_c (t - nT)}$$

↑
Interpolation formula

4.9 Sampling



$$x_r(t) = \sum_{n=-\infty}^{\infty} x(nT) \frac{\omega_c T}{\pi} \frac{\sin(\omega_c(t - nT))}{\omega_c(t - nT)}$$



Ideal band-limited interpolation using the *sinc* function with $\omega_c = \frac{\omega_s}{2}$

4.9 Sampling

4.9.4 The Effect of Under-Sampling (欠采样): Aliasing(混叠)

$$x(t) = \cos \omega_0 t,$$

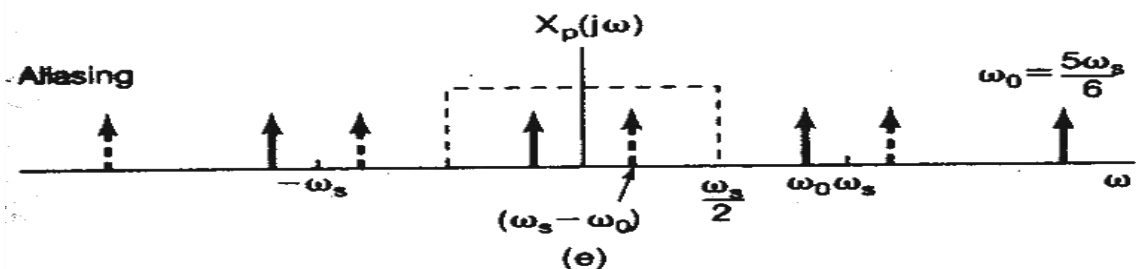
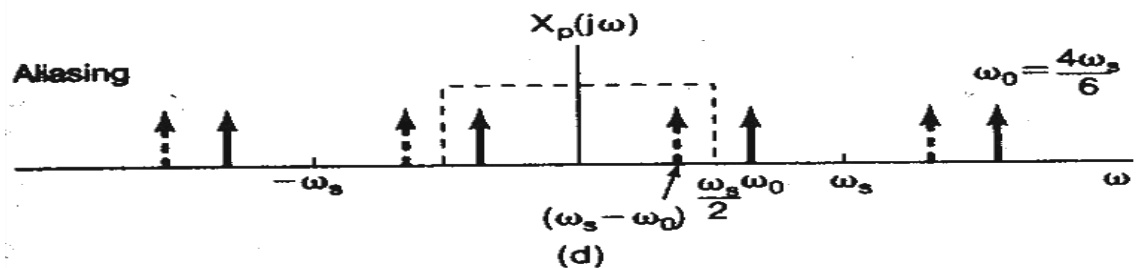
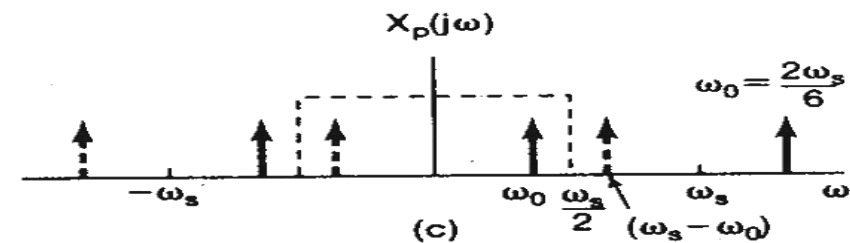
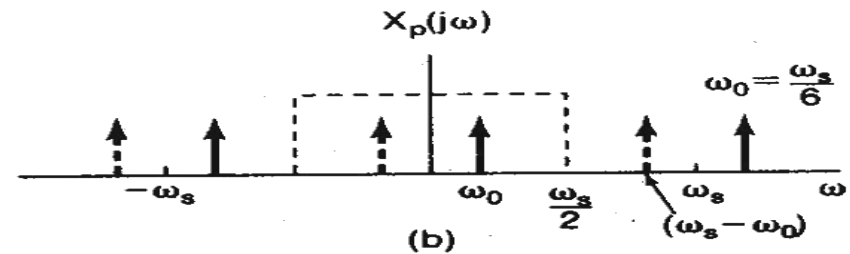
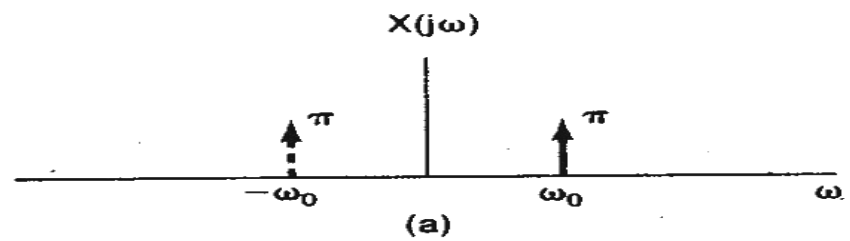
$$x(t) = \cos 1000t$$

is sampled with

$$\omega_s = 1500 \text{ rad/s}$$

$$\text{and } \omega_s = 1200 \text{ rad/s}$$

respectively.



4.10 SUMMARY

- The Fourier transform for both non-periodic and periodic continuous-time signals;
- The properties of the Fourier transform (relationships between characteristics of a continuous-time signal in time and frequency domain);
- Fourier analysis (Frequency domain analysis) for continuous-time LTI systems including both characteristics of systems and responses to some input signals;
- Frequency response and the way to obtain it;
- Continuous-time signals' sampling and their reconstruction.

Homework

4.21 (b)(c) (d) (h) 4.22 (a) (b) (c) (d)

4.24 4.28 (a) 、 (i)(iv)(vi)(viii) in (b)

4.32 (b) (c) 4.34

4.35 7.22 7.23