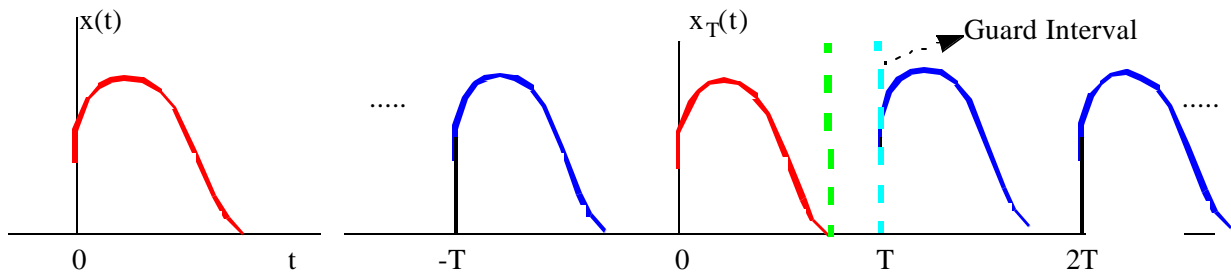


## Chap 5. Continuous-Time Fourier Transform and Applications

### 5.1 Illustrative Definition of Fourier Transform

In this chapter, we will develop the basis for Fourier analysis of non-periodic signals, which is the only group of signals meaningful in engineering and real-life applications. Traditionally, Fourier analysis is presented by giving the definitions as we did for Laplace transforms in Chapter 3. Here, we will approach it as a limiting behavior of Fourier series analysis for periodic signals. To do that, let us consider an aperiodic (non-periodic) signal  $x(t)$  and form its periodic extension  $x_T(t)$  by repeating it every  $T$  seconds, where  $T$  is the period.



It is clear from above that the expansion signal can be written as:

$$x_T(t) = \begin{cases} x(t) & \text{if } 0 \leq t < T \\ x(t) = x(t + nT) & \text{for every integer } n \end{cases} \quad (5.1)$$

The Fourier series representation for this new periodic signal is simply:

$$\text{Synthesis Equation: } x_T(t) = \sum_{k=-\infty}^{\infty} F_k \cdot e^{jk\omega_0 t} \quad (5.2a)$$

$$\text{Analysis Equation: } F_k = \frac{1}{T} \cdot \int_{\langle T \rangle} x_T(t) \cdot e^{-jk\omega_0 t} dt \quad (5.2b)$$

In the limit as  $T \Rightarrow \infty$ , we observe that  $\omega_0 = \frac{2\pi}{T} \Rightarrow d\omega$ , which is an infinitesimally small quantity in frequency-domain and it implies that  $k\omega_0 \Rightarrow \omega$ , a continuous variable and  $\frac{1}{T} \Rightarrow \frac{d\omega}{2\pi} = df$ . Finally, with this limiting behavior, the summation in (5.2a) becomes an integral as follows:

$$F_k = \frac{d\omega}{2\pi} \cdot \int_{-\infty}^{\infty} x_T(t) \cdot e^{-j\omega t} dt \Rightarrow \frac{F_k}{d\omega} = \frac{1}{2\pi} \int_{-\infty}^{\infty} x_T(t) \cdot e^{-j\omega t} dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} x(t) \cdot e^{-j\omega t} dt \quad (5.3)$$

and

$$x(t) = \int_{-\infty}^{\infty} \left( \int_{-\infty}^{\infty} x(t) \cdot e^{-j\omega t} dt \right) \cdot e^{j\omega t} \frac{d\omega}{2\pi} = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) \cdot e^{j\omega t} d\omega \quad (5.4)$$

With this result, we have implicitly defined the Fourier transform relationships for  $x(t)$ .

$$\text{Analysis Equation: } X(\omega) = \int_{-\infty}^{\infty} x(t) \cdot e^{-j\omega t} dt = F\{x(t)\} \quad (5.5a)$$

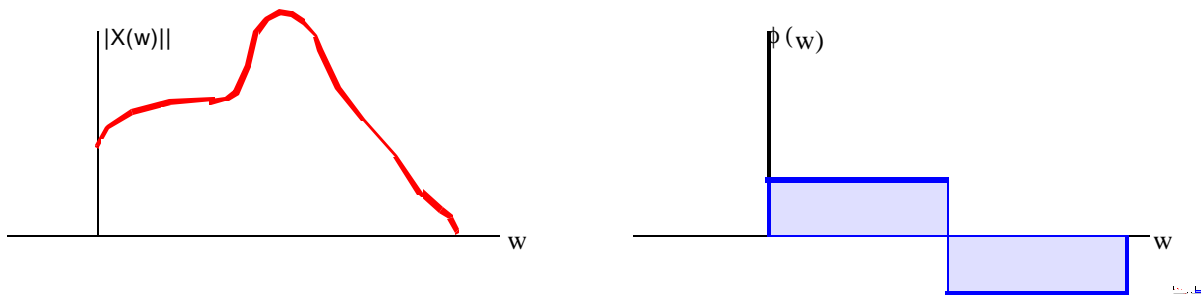
$$\text{Synthesis Equation: } x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) \cdot e^{j\omega t} d\omega = F^{-1}\{X(\omega)\} \quad (5.5b)$$

These two equations are called Fourier transform pair and normally shown by:  $x(t) \Leftrightarrow X(\omega)$ .

In general,  $X(\omega)$  is a complex function of the real-valued frequency variable  $\omega$  and it is written in terms of magnitude and phase terms:

$$X(\omega) = |X(\omega)| \cdot e^{j\phi(\omega)} \quad (5.6)$$

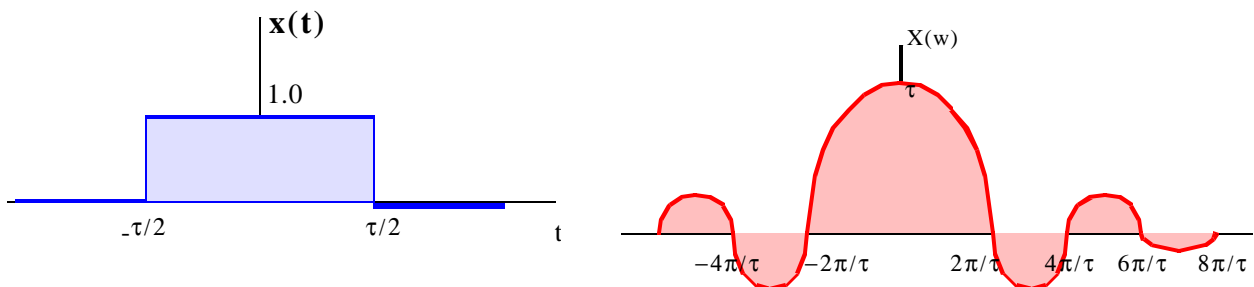
Therefore, we plot one curve for the magnitude and another one for the phase of a given expression.



**Dirichlet Conditions:** Fourier transform exists if:

1.  $x(t)$  is absolutely integrable:  $\int_{-\infty}^{\infty} |x(t)| dt < \infty$
2.  $x(t)$  is a *well-behaving* function. That is, only a finite number of jumps of finite size, minima and maxima occur within any finite interval:  $t_1 < t < t_2$ .

**Example 5.1** Find the Fourier transform of a rectangular pulse (gate function, rectangular time-window).



$$\begin{aligned}
 X(w) &= \int_{-\infty}^{\infty} x(t).e^{-j\omega t} dt = \int_{-t/2}^{t/2} 1.e^{-j\omega t} dt = \left. \frac{e^{-j\omega t}}{-j\omega} \right|_{t=-t/2}^{t/2} = \frac{1}{-j\omega} . [e^{-j\omega t/2} - e^{+j\omega t/2}] \\
 &= \frac{1}{-j\omega} . [-2j.Sin(\frac{\omega t}{2})] = \frac{2}{\omega} . Sin(\frac{\omega t}{2}) = t.Sinc(\frac{\omega t}{2p}) = t.Sa(\frac{\omega t}{2})
 \end{aligned} \tag{5.7}$$

- It is easy to observe that the resulting spectrum is real and symmetrical for all values of frequency  $\omega$ .
- This implies that the phase response is zero:  $\mathbf{f}(\omega) = 0$  for all  $\omega$ .
- The magnitude is decreasing as a function of  $1/\omega$ .
- Approximately, 90% of energy content of this spectrum is under the *main lobe* of the plot:  $-2p/\omega < \omega < 2p/\omega$ . In other words, the energy under all the *tail lobes* is about 10%.

**Example 5.2** Find the Fourier transform of a delta function. We will solve this problem using the Sifting Theorem definition of a delta function:

$$F\{V_0 \mathbf{d}(t)\} = V_0 \int_{-\infty}^{\infty} \mathbf{d}(t).e^{-j\omega t} dt = V_0 \Rightarrow F\{\mathbf{d}(t)\} = 1 \Rightarrow \mathbf{d}(t) \Leftrightarrow 1 \tag{5.9}$$

Similarly, we can use the inversion formula to have:

$$\mathbf{d}(t) = \frac{1}{2p} \int_{-\infty}^{\infty} 1.e^{j\omega t} d\omega \tag{5.10}$$

From this final result we can deduce that the Fourier transform of a constant is a delta function in frequency-domain:

$$F\{V_0\} = 2p.V_0.\mathbf{d}(\omega) \tag{5.11}$$

**Example 5.3** Find the Fourier transform of a complex exponential harmonic function:

$$F\{e^{j\omega_0 t}\} = \int_{-\infty}^{\infty} e^{j\omega_0 t}.e^{-j\omega t} dt = \int_{-\infty}^{\infty} e^{-j(\omega-\omega_0)t} dt = 2p\mathbf{d}(\omega-\omega_0) \quad e^{j\omega_0 t} \Leftrightarrow 2p\mathbf{d}(\omega-\omega_0) \tag{5.12}$$

On the other hand, the Fourier transform of a one-sided decaying exponential function is:

$$\begin{aligned}
 x(t) &= e^{-at}u(t) \quad \text{and} \quad a > 0 \\
 X(\omega) &= \int_{-\infty}^{\infty} e^{-at}.u(t).e^{-j\omega t} dt = \int_0^{\infty} 1.e^{-(a+j\omega)t} dt = \frac{1}{a+j\omega}
 \end{aligned} \tag{5.13}$$

**Example 5.4** Find the Fourier transform of a periodic signal  $x(t)$  with a period  $T_0 = \frac{2p}{\omega_0}$ . As we know from the previous chapter that the periodic functions have Fourier series representation:

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

Let us take Fourier transform of this equation term-by-term:

$$X(\omega) = F\{x(t)\} = \sum_{k=-\infty}^{\infty} F_k F\{e^{jk\omega_0 t}\} = \sum_{k=-\infty}^{\infty} 2p \cdot F_k \mathbf{d}(\omega - k\omega_0) = \sum_{k=-\infty}^{\infty} 2p \cdot F_k \mathbf{d}\left(\omega - k \frac{2p}{T_0}\right) \quad (5.14)$$

Therefore, the Fourier transform is a sequence of impulse functions regardless of the actual shape of the signal  $x(t)$ . In other words, the Fourier transform of all periodic functions is a family of impulses. What makes them different for various  $x(t)$  shapes are the values of the coefficients  $\{F_k\}$ .

**Example 5.5** Using the results from the previous example, we are asked to find the Fourier transform of an impulse train.

$$x(t) = \sum_{k=-\infty}^{\infty} \mathbf{d}(t - kT_0)$$

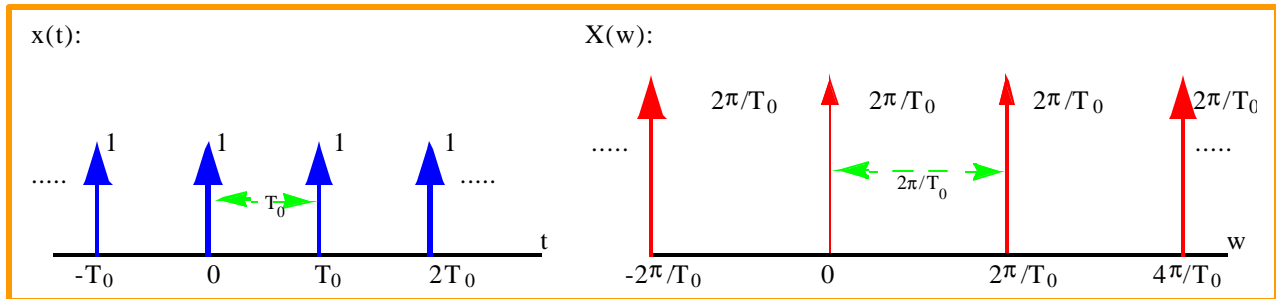
$$X(\omega) = \sum_{k=-\infty}^{\infty} 2p \cdot F_k \mathbf{d}\left(\omega - k\omega_0\right) = \sum_{k=-\infty}^{\infty} 2p \cdot F_k \mathbf{d}\left(\omega - k \frac{2p}{T_0}\right) \quad (5.15)$$

$$F_k = \frac{1}{T_0} \cdot \int_{\langle T_0 \rangle} x(t) \cdot e^{-jk\omega_0 t} dt = \frac{1}{T_0} \cdot \int_{\langle T_0 \rangle} \mathbf{d}(t - kT_0) \cdot e^{-jk\omega_0 t} dt = \frac{1}{T_0}$$

Let us substitute these coefficient values in to the result above to obtain:

$$X(\omega) = \sum_{k=-\infty}^{\infty} 2p \cdot \frac{1}{T_0} \mathbf{d}\left(\omega - k \frac{2p}{T_0}\right) = \frac{2p}{T_0} \cdot \sum_{k=-\infty}^{\infty} \mathbf{d}\left(\omega - k \frac{2p}{T_0}\right) \quad (5.16)$$

Thus, we conclude that the Fourier transform of an impulse train is another impulse train in the frequency-domain with different strengths in the coefficient set.



## 5.2 Properties of Fourier Transforms

1. **Linearity:** The Fourier transform is a linear transform.

$$a \cdot f(t) + b g(t) \Leftrightarrow a \cdot F(\omega) + b G(\omega) \quad (5.17)$$

2. **Symmetry:** If  $x(t)$  is a real signal then  $X(-\omega) = X^*(\omega)$  or in polar form:

$$X(\omega) = |X(\omega)| e^{j\mathbf{f}(\omega)} \quad \text{and} \quad X^*(\omega) = X(-\omega) = |X(\omega)| e^{-j\mathbf{f}(\omega)} \quad (5.18)$$

Therefore, we have an even-symmetry of the amplitude spectrum and an odd-symmetry for the phase spectrum.

**Table 5.1 Fourier Transform Pairs for Selected Signals**

1. 1	$2\pi \delta(\omega)$
2. $u(t)$	$\pi \delta(\omega) + \frac{1}{j\omega}$
3. $\delta(t)$	1
4. $\delta(t - t_0)$	$\exp[-j\omega t_0]$
5. $\text{rect}(t/\tau)$	$\tau \text{sinc} \frac{\omega\tau}{2\pi} = \frac{2 \sin \omega\tau/2}{\omega}$
6. $\frac{\omega_B}{\pi} \text{sinc} \frac{\omega_B t}{\pi} = \frac{\sin \omega_B t}{\pi t}$	$\text{rect}(\omega/2\omega_B)$
7. $\text{sgn } t$	$\frac{2}{j\omega}$
8. $\exp[j\omega_0 t]$	$2\pi \delta(\omega - \omega_0)$
9. $\sum_{n=-\infty}^{\infty} a_n \exp[jn\omega_0 t]$	$2\pi \sum_{n=-\infty}^{\infty} a_n \delta(\omega - n\omega_0)$
10. $\cos \omega_0 t$	$\pi [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$
11. $\sin \omega_0 t$	$\frac{\pi}{j} [\delta(\omega - \omega_0) - \delta(\omega + \omega_0)]$
12. $(\cos \omega_0 t)u(t)$	$\frac{\pi}{2} [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)] + \frac{j\omega}{\omega_0^2 - \omega^2}$
13. $(\sin \omega_0 t)u(t)$	$\frac{\pi}{2j} [\delta(\omega - \omega_0) - \delta(\omega + \omega_0)] + \frac{\omega_0}{\omega_0^2 - \omega^2}$
14. $\cos \omega_0 t \text{rect}(t/\tau)$	$\tau \text{sinc} \frac{(\omega - \omega_0)\tau}{2\pi}$
15. $\exp[-at]u(t), \text{Re}\{a\} > 0$	$\frac{1}{a + j\omega}$
16. $t \exp[-at]u(t), \text{Re}\{a\} > 0$	$\left(\frac{1}{a + j\omega}\right)^2$
17. $\frac{t^{n-1}}{(n-1)!} \exp[-at]u(t), \text{Re}\{a\} > 0$	$\frac{1}{(a + j\omega)^n}$
18. $\exp[-a t ], a > 0$	$\frac{2a}{a^2 + \omega^2}$
19. $ t  \exp[-a t ], \text{Re}\{a\} > 0$	$\frac{4aj\omega}{a^2 + \omega^2}$
20. $\frac{1}{a^2 + t^2}, \text{Re}\{a\} > 0$	$\frac{\pi}{a} \exp[-a \omega ]$
21. $\frac{t}{a^2 + t^2}, \text{Re}\{a\} > 0$	$\frac{-j\pi\omega \exp[-a \omega ]}{2a}$
22. $\exp[-at^2], a > 0$	$\sqrt{\frac{\pi}{a}} \exp\left[-\frac{\omega^2}{4a}\right]$
23. $\Delta(t/\tau)$	$\tau \text{sinc}^2 \frac{\omega\tau}{2\pi}$
24. $\sum_{n=-\infty}^{\infty} \delta(t - nT)$	$\frac{2\pi}{T} \sum_{n=-\infty}^{\infty} \delta\left(\omega - \frac{2n\pi}{T}\right)$

**Example 5.6** Using the properties of real signals we can turn Fourier transform into a sine or a cosine transform. In particular, discrete version of the cosine transform has found a neat application area in image compression based on JPEG and MPEG techniques. These compression techniques and their derivatives attempt to exploit the symmetry properties of imagery in the frequency-domain.

$$\begin{aligned}
X(w) &= \int_{-\infty}^{\infty} x(t).e^{-j\omega t} dt = \int_{-\infty}^{\infty} x(t).[Cos(\omega t) - jSin(\omega t)]dt \\
&= \int_{-\infty}^{\infty} x(t)Cos(\omega t)dt - j \int_{-\infty}^{\infty} x(t)Sin(\omega t)dt \\
&= \int_{-\infty}^{\infty} x(t)Cos(\omega t)dt + 0 \quad (\text{Sin is odd fn.}) \\
&= 2 \int_0^{\infty} x(t).Cos(\omega t)dt
\end{aligned} \tag{5.19}$$

As we see from this last result that the Fourier transform turns into a Fourier Cosine transform for real signals.

3. **Time-Shifting (Delays):** Given:  $x(t) \Leftrightarrow X(w)$  then

$$x(t - T) \Leftrightarrow X(w).e^{-j\omega T} = |X(w)|.e^{j(\omega T - \phi)} \tag{5.20}$$

**Notes:**

- Amplitude spectrum:  $|X(w)|$  is unaffected by a delay (time-shift) of T units.
- However, the phase spectrum is linearly shifted by  $-\omega T$  radians from the phase value  $\phi$  of the original spectrum  $X(w)$ . This linear phase-shift property makes the time-delay issue very useful in many applications.

4. **Time-Scaling (Resolution Change):** Given the Fourier pair:  $x(t) \Leftrightarrow X(w)$  then a time-scale change will correspond to:

$$x(at) \Leftrightarrow \frac{1}{|a|}.X\left(\frac{\omega}{a}\right) \tag{5.21}$$

**Notes:**

- For values of  $a$  larger than "1" this works as a compressor. That is, it counts slower than one time-unit at a time.
- On the other hand, for  $a$  smaller than "1" it works as an expander to increase the resolution in time-domain. The resulting behavior in frequency-domain is just the opposite.

**Example 5.7** Using the above property find the Fourier transform of  $x(t) = a.\text{rect}\left(\frac{at}{T}\right)$

Recalling  $F\left\{\text{rect}\left(\frac{t}{T}\right)\right\} = T.\text{Sinc}\left(\frac{\omega T}{2\pi}\right)$  and the resolution change we get:

$$F\left\{a.\text{rect}\left(\frac{\omega T}{2a\pi}\right)\right\} = T.\text{sin}c\left(\frac{\omega T}{2a\pi}\right) \tag{5.22}$$

5. **Derivatives and Integrals:-Scaling:** Given the Fourier pair:  $x(t) \Leftrightarrow X(w)$  then we have:

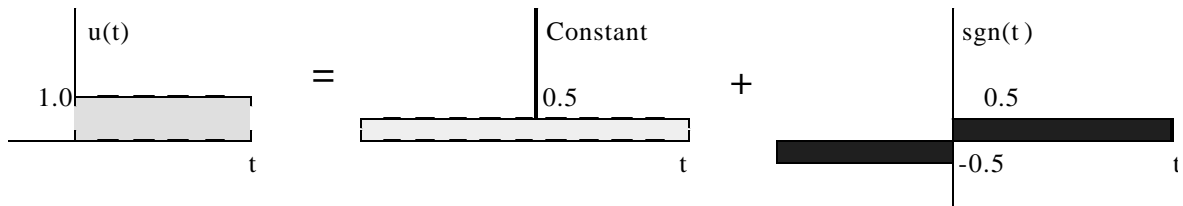
$$\text{a. } \frac{d}{dt}x(t) \Leftrightarrow jw.X(w) \quad (5.23a)$$

$$\text{b. } \frac{d^n}{dt^n}x(t) \Leftrightarrow (jw)^n.X(w) \quad (5.23b)$$

$$\text{c. } \int_{-\infty}^t x(t)dt \Leftrightarrow \mathbf{p}.X(0).\mathbf{d}(w) + \frac{1}{jw}.X(w) \quad (5.23c)$$

$$\text{d. } \text{If there is no D.C. term, i.e., } X(0) = 0 \text{ then } \int_{-\infty}^t x(t)dt \Leftrightarrow \frac{1}{jw}.X(w) \quad (5.23d)$$

**Example 5.8** Using the above property find the Fourier transform of a unit-step function:  $u(t)$ .



As we see from the above figure, the unit-step function can be written as:

$$u(t) = \frac{1}{2} + \frac{1}{2}.Sgn(t) \quad (5.24)$$

But we also know that:

$$\frac{d}{dt}\left\{\frac{1}{2}.Sgn(t)\right\} = \mathbf{d}(t) \quad \text{equivalently: } (jw).F\left\{\frac{1}{2}.Sgn(t)\right\} = 1 \quad (5.25)$$

which results in:

$$F\left\{\frac{1}{2}.Sgn(t)\right\} = \frac{1}{jw} \quad \text{and} \quad F\left\{\frac{1}{2}\right\} = \frac{1}{2}.2\mathbf{p}.\mathbf{d}(w) \quad (5.26)$$

$$F\{u(t)\} = \mathbf{p}.\mathbf{d}(w) + \frac{1}{jw} \quad (5.27)$$

6. **Energy in Aperiodic Signals (Parseval's Theorem):** Recalling energy in time-domain:

$$\begin{aligned} E &= \int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} x(t).x^*(t)dt = \int_{-\infty}^{\infty} x(t).\left[\frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} X^*(w).e^{-jw} dw\right]dt \\ &= \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} X^*(w).\left[\int_{-\infty}^{\infty} x(t).e^{-jw} dt\right]dw = \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} X^*(w).X(w)dw = \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} |X(w)|^2 dw \end{aligned}$$

which implies that the energy is conserved during Fourier transform and it can be summarized by:

$$E = \int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} x(t) \cdot x^*(t) dt = \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} X^*(w) \cdot X(w) dw = \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} |X(w)|^2 dw \quad (5.28)$$

Similarly, the energy content of a signal within a finite band of frequency:  $w_1 \leq w \leq w_2$  will be:

$$\Delta E_x = \frac{1}{2\mathbf{p}} \int_{w_1}^{w_2} |X(w)|^2 dw \quad (5.29)$$

**Example 5.9** Compute the portion of the energy for an exponentially decaying signal in the frequency interval:  $-4 \leq w \leq 4$ . We know the Fourier transform for this signal both from a previous example and the Fourier Tables 5.1 that:

$$x(t) = e^{-t} \cdot u(t) \Rightarrow X(w) = \frac{1}{1+jw} \Rightarrow |X(w)|^2 = \frac{1}{1+w^2}$$

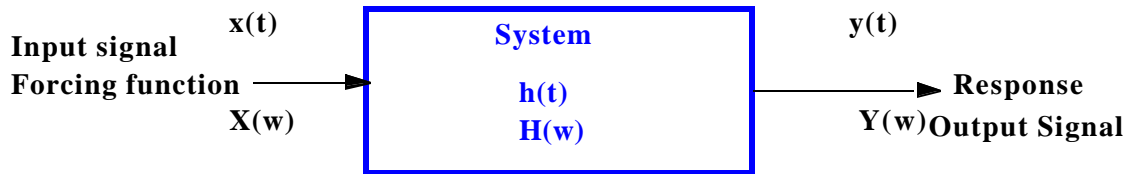
When we substitute this into (5.28) and (5.29) we get:

$$E = \frac{1}{2\mathbf{p}} \int_{-\infty}^{\infty} \frac{1}{1+w^2} dw = \frac{1}{2} \Rightarrow \text{From integration tables.}$$

$$\Delta E_x = \frac{1}{2\mathbf{p}} \int_{-4}^4 |X(w)|^2 dw = \frac{1}{\mathbf{p}} \int_0^4 \frac{1}{1+w^2} dw = \frac{1}{\mathbf{p}} \cdot \arctan(w) \Big|_0^4 \cong 0.422$$

Therefore, the portion of the energy in this finite band is 84.4%.

## 7. Convolution Operation:



$$y(t) = x(t) * h(t) = h(t) * x(t) = \int_{-\infty}^{\infty} x(\mathbf{t}) h(t - \mathbf{t}) d\mathbf{t} \quad (5.30a)$$

$$Y(w) = X(w) \cdot H(w) = |Y(w)| e^{j\mathcal{F}(w)} \quad (5.30b)$$

$$|Y(w)| = |X(w)| \cdot |H(w)| \quad \text{and} \quad \angle Y(w) = \angle X(w) + \angle H(w) \quad (5.30c)$$

### Notes:

- Output magnitude spectrum is the **product** of magnitude spectrum of the input and the frequency response of the system.
- The phase spectrum is the **sum** of the phase spectra of the input and the system.

**Example 5.10** Compute the unit-step response to a system with:  $h(t) = e^{-at} \cdot u(t)$ .

$$Y(w) = X(w) \cdot H(w) = F\{x(t)\} \cdot F\{h(t)\}$$



$$Y(w) = [p \cdot d(w) + \frac{1}{jw}]. \frac{1}{a + jw} = \frac{p \cdot d(w)}{a + jw} + \frac{1}{jw} \frac{1}{a + jw} \quad \text{directly from Fourier Tables}$$

Let us note that  $d(w) \neq 0$  only when  $w = 0$ . Therefore, the first term in the last expression is simply:  $1^{st} \text{ Term} = \frac{p}{a} d(w)$ . Using this and partial fraction expansion results in:

$$Y(w) = \frac{p}{a} d(w) + \frac{A}{jw} + \frac{B}{a + jw}$$

$$Bjw + A(a + jw) = 1 \Rightarrow A = \frac{1}{a} \quad \text{and} \quad B = -\frac{1}{a}$$

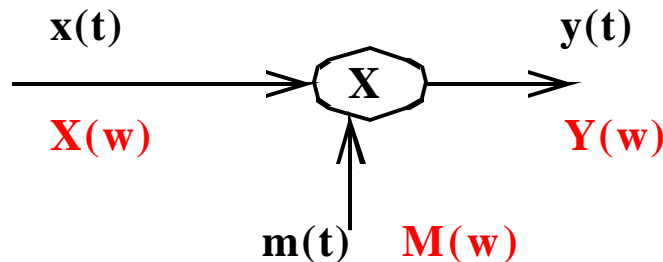
Finally, the output in the frequency-domain will be:

$$Y(w) = \frac{p}{a} d(w) + \frac{1/a}{jw} - \frac{1/a}{a + jw}$$

and the inverse Fourier transform yields the answer:

$$y(t) = \frac{1}{a} \cdot \{1 - e^{-at}\} u(t)$$

#### 8. Modulation Theorem:

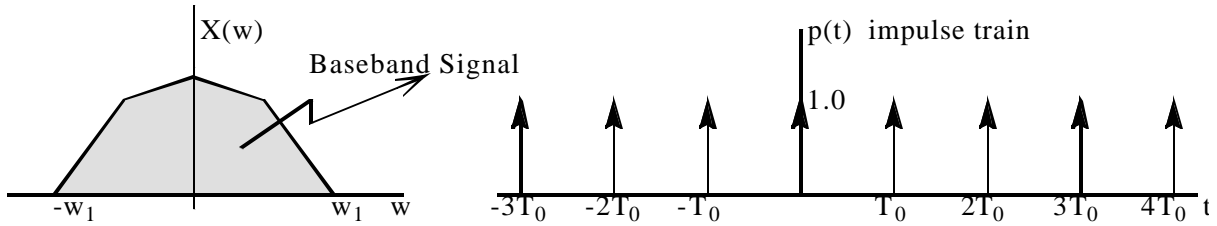


$$y(t) = x(t) \cdot m(t) \quad \Leftrightarrow \quad Y(w) = \frac{1}{2p} [X(w) * M(w)] \quad (5.31)$$

$$Y(w) = \frac{1}{2p} \int_{-\infty}^{\infty} X(y) \cdot H(y - w) dy \quad (5.32)$$

In other words, to replace a time-domain multiplication, we must perform a frequency-domain convolution. As expected, we will attempt to perform time-domain multiplication as much as we can to tackle modulation tasks.

**Example 5.11** Given a signal with a base-band spectrum (frequency content of the signal in its natural habitat), modulate it with an impulse train of period:  $T_0 = 2p / w_0$ .



The output of the modulator (mixer) is simply:

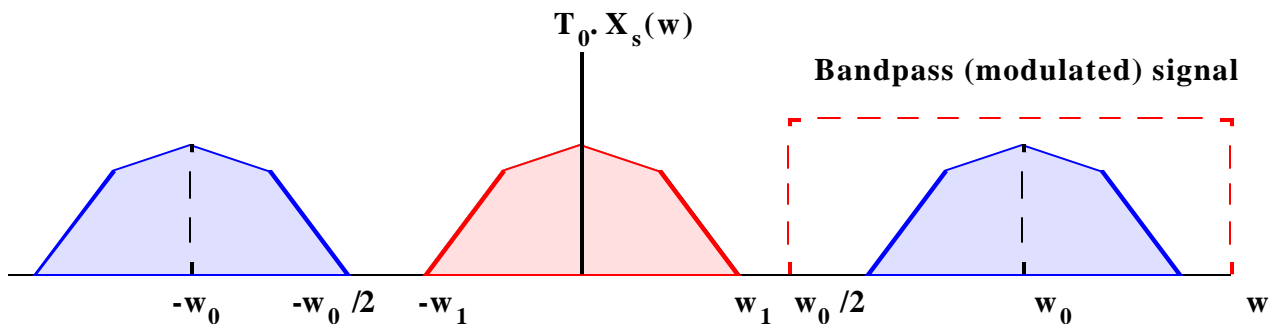
$$x_S(t) = x(t) \cdot p(t) = x(t) \cdot \sum_{n=-\infty}^{\infty} \mathbf{d}(t - nT_0) = \sum_{n=-\infty}^{\infty} x(nT_0) \cdot \mathbf{d}(t - nT_0) \quad (5.33)$$

Here, the impulse train acts like a sampling function, i.e., isolating the signal at a given point. Equivalently, in the frequency-domain:

$$P(w) = F\{p(t)\} = F\left\{\sum_{n=-\infty}^{\infty} \mathbf{d}(t - nT_0)\right\} = \frac{2p}{T_0} \cdot \sum_{k=-\infty}^{\infty} \mathbf{d}\left(w - \frac{2p}{T_0}k\right) \quad (5.34)$$

$$X_S(w) = \frac{1}{2p} X(w) * P(w) = \frac{1}{2p} \cdot \frac{2p}{T_0} \cdot \sum_{k=-\infty}^{\infty} \left[X(w) * \mathbf{d}\left(w - \frac{2p}{T_0}k\right)\right] = \frac{1}{T_0} \cdot \sum_{k=-\infty}^{\infty} X\left(w - \frac{2p}{T_0}k\right) \quad (5.35)$$

It is clear from this result that the output spectrum is a periodic family of the input spectra shifted to the neighborhood of  $\pm kw_0 = \pm k \frac{2p}{T_0}$ .



**Table 5.2 Properties of Continuous-Time Fourier Pairs**

1. Linearity	$\sum_{n=1}^N \alpha_n x_n(t)$	$\sum_{n=1}^N \alpha_n X_n(\omega)$
2. Complex conjugation	$x^*(t)$	$X^*(-\omega)$
3. Time shift	$x(t - t_0)$	$X(\omega) \exp[-j\omega t_0]$
4. Frequency shift	$x(t) \exp[j\omega_0 t]$	$X(\omega - \omega_0)$
5. Time scaling	$x(at)$	$1/ a  X(\omega/a)$
6. Differentiation	$d^n x(t)/dt^n$	$(j\omega)^n X(\omega)$
7. Integration	$\int_{-\infty}^t x(\tau) d\tau$	$\frac{X(\omega)}{j(\omega)} + \pi X(0) \delta(\omega)$
8. Parseval's relation	$\int_{-\infty}^{\infty}  x(t) ^2 dt$	$\frac{1}{2\pi} \int_{-\infty}^{\infty}  X(\omega) ^2 d\omega$
9. Convolution	$x(t) * h(t)$	$X(\omega) H(\omega)$
10. Duality	$X(t)$	$2\pi x(-\omega)$
11. Multiplication by $t$	$(-jt)^n x(t)$	$\frac{d^n X(\omega)}{d\omega^n}$
12. Modulation	$x(t)m(t)$	$\frac{1}{2\pi} X(\omega) * M(\omega)$