



## **Lecture 2 Outlines**

### **1. Sampling Theorem**

❖ **Impulse Sampling**

❖ **Natural Sampling**

❖ **Differences between Impulse Sampling  
and Natural Sampling**

### **2. Application of Sampling process**



## Sampling Theorem

The link between an analog waveform and its sampled version is provided by what is known as the sampling process.

A band limited signal having no spectral components above ( $f_m$  Hz) can be determined uniquely by values sampled at uniform intervals of  $T_s$  second, where:

$$T_s = \frac{1}{2f_m}$$

Stated another way, the upper limit on  $T_s$  can be expressed in terms of the *sampling rate*, denoted  $f_s = \frac{1}{T_s}$

The restriction, stated in terms of sampling rate, is known as the Nyquist criterion. The statement is

$$f_s \geq 2f_m$$

The sampling rate ( $f_s = 2f_m$ ) also called *Nyquist rate*.

The Nyquist criterion is a theoretically sufficient condition to allow an analog signal to be reconstructed completely from a set of uniformly spaced discrete time samples.



### ❖ Impulse Sampling

Assume an analog waveform  $x(t)$ , as shown in Fig. (a), with a Fourier transform,  $X(f)$ , which is zero outside the interval  $(-f_m < f < f_m)$ , as shown in Fig. (b). The sampling of  $x(t)$  can be viewed as the product of  $x(t)$  with a train of unit impulses functions,  $x_s(t)$ , shown in Fig. (c), and defined as follows:

$$x_s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$$

Let us choose  $T_s = \frac{1}{2f_m}$ , so that Nyquist rate is just satisfied.

Using shifting property of the impulse function the  $x_s(t)$ , shown in Fig. (e), can be given by

$$\begin{aligned} x_s(t) &= x(t)x_s(t) = \sum_{n=-\infty}^{\infty} x(t)\delta(t - nT_s) \\ &= \sum_{n=-\infty}^{\infty} x(nT_s)\delta(t - nT_s) \end{aligned}$$

Using frequency convolution property of Fourier transform, the time product  $x(t)x_s(t)$  transforms to the frequency domain convolution  $X(f) \otimes X_s(f)$ , where  $X_s(f)$  is the Fourier transform of  $x_s(t)$  and given by

$$X_s(f) = \frac{1}{T_s} \sum_{n=-\infty}^{\infty} \delta(f - nf_s)$$



The convolution with an impulse function simply shifts the original function, as follows:

$$X(f) \otimes \delta(f - nf_s) = X(f - nf_s)$$

The Fourier transform of the sampled waveform,  $X_s(f)$ , can be given by:

$$\begin{aligned} X_s(f) &= X(f) \otimes X_\delta(f) = X(f) \otimes \left[ \frac{1}{T_s} \sum_{n=-\infty}^{\infty} \delta(f - nf_s) \right] \\ &= \frac{1}{T_s} \sum_{n=-\infty}^{\infty} X(f - nf_s) \end{aligned}$$

Figure below shows the sampling theorem using the frequency convolution property of the Fourier transform (Impulse sampling).

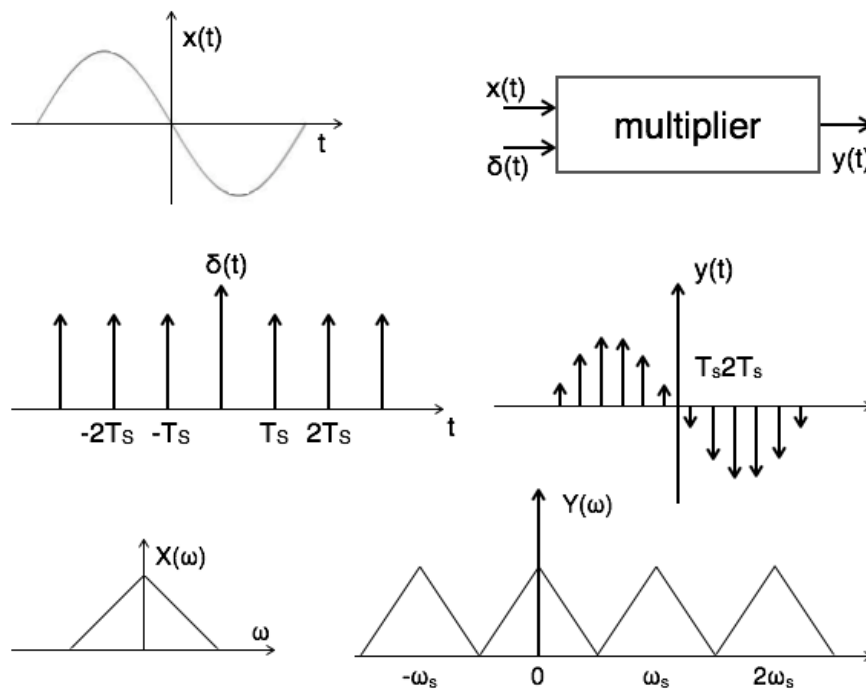




Figure (1): Sampling theorem using impulse method

### ❖ Natural Sampling

In this way the band limited analog signal  $x(t)$ , shown in Fig. (2-a), is multiplied by the pulse train or switching waveform  $x_p(t)$ , shown in Fig. (2-b). Each pulse in  $x_p(t)$  has width  $T$  and amplitude  $1/T$ .

The resulting sampled data sequence,  $x_s(t)$ , is shown in Fig. (2-c) and is expressed as

$$x_s(t) = x(t)x_p(t)$$

Figure below shows the sampling theorem using the shifting property of the Fourier transform (Natural sampling).

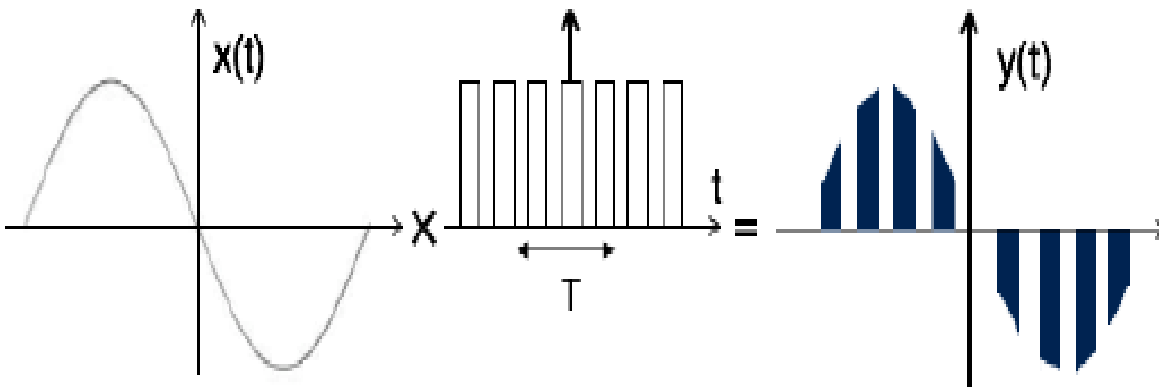


Figure (2): Sampling theorem using the shifting property of the Fourier transform (Natural sampling).

**Note:** The sampling here is termed natural sampling, since the top of each pulse in the  $x_s(t)$  sequence retains the shape of its corresponding analog segment during the pulse interval.