

Introduction to Discrete-Time Signals and Systems

## Analog to Digital Conversion (Sampling)

### Lecture #37

The material to be covered in this lecture is as follows:

- Introduction to discrete-time signals and systems
- Analog to Digital Conversion
  - Sampling (Ideal and Non-ideal)
  - Quantization
  - Encoding

After finishing this lecture you should be able to:

- Distinguish discrete from continuous signals
- Perform the steps required for analog to digital conversion
  - Find the proper sampling instants and the associated sampled values
  - Perform quantization and estimate its effects on the signal quality
  - Convert the quantized values into code words
- Sketch the spectrum of the sampled signal using ideal and non-ideal sampling

## Introduction to Discrete-Time Signals and Systems

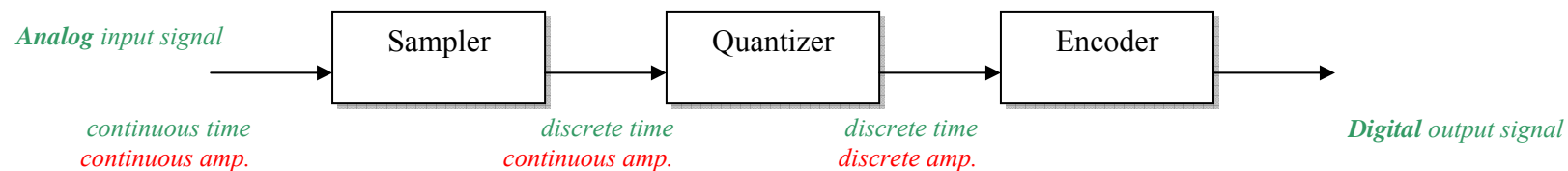
- Signals in life can be analog or digital.
- Nowadays, with the advances in digital systems and personal computers one can do advance processing for digital signals. This includes: compression, encryption, error-control coding....
- There are many other advantages for digital systems.
- To be able to process an analog signal in the same way it has to be converted to a digital form.
- For the conversion to be accomplished there are three main steps
  - Sampling
  - Quantization
  - Encoding
- The analog signal is converted into discrete-time signal by means of sampling
- Discrete-time signals are defined by specifying the value of the signal only at discrete times (sampling instants)

## Analog to Digital Conversion

- Include the animation already prepared (flash) analog digital

# Analog to Digital Conversion

The stages for analog to digital conversion may be summarized in the following figure



The emphasis on the remaining part will be on discrete-time signals which are signals after the sampler. We will assume that the error introduced by the quantizer to be relatively ignorable.

## Example 37.1

Given the signal

$$x(t) = 8[1 + \cos(120\pi t)\cos(100\pi t)]$$

which is sampled at the rate of (50 samples per second). Each sample is quantized to the closest integer between 0 and 15. Each of the integer values is encoded using a 4 bit code word according to the usual binary representation of integers (i.e. 0=0000, 1=0001, ....., 15=1111). Determine the sampled value, the quantized value and the binary code for the first three samples starting at  $t=0$ .

Answer

Sampling frequency,  $f_s = 150$  Hz.

Sampling Interval,  $T_s = \frac{1}{f_s} = \frac{1}{150} = 6.67$  ms.

Sampling instants,  $t = nT_s$ ,  $n = 0, 1, 2, 3, \dots$   
 $= 0, 6.67, 13.33$  ms

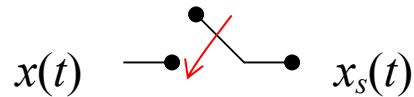
$n$	Time (ms)	Sampled value	Quantized value	Binary Code
0	0	16	15	1111
1	6.67	11.23	11	1011
2	13.33	6.76	7	0111

The quantized values can be found by substituting  $t \leftarrow nT_s$ .

The quantized values are found by first rounding to the closest integer between 0 and 15 and then represent the answer in binary form. The table summarizes the results. [\(Animate the table\)](#)

# Sampling

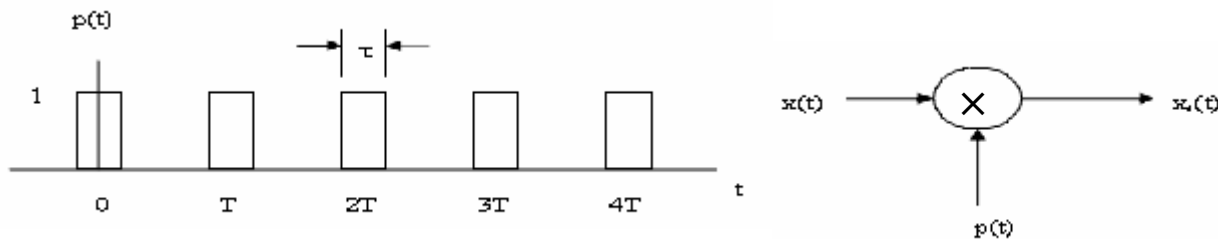
- The sampled signal,  $x_s(t)$  can be generated by applying a switch to the input signal  $x(t)$  as shown in the figure:



The switch closes at the sampling instances.

- Ideally, the switch when it is closed it will pass the input signal to the output and when it is opened nothing will pass to the output.
- Mathematically, this is like multiplying the input signal by another periodic signal,  $p(t)$  which can take only two values 0 or 1.

The signal  $p(t)$  is represented in the figure





where  $T = \frac{1}{f_s}$ , and  $\tau$  is the sampling duration which is theoretically zero.

$$x_s(t) = x(t)p(t) \quad (1)$$

- Since  $p(t)$  is periodic it can be represented by its exponential Fourier series

$$p(t) = \sum_{n=-\infty}^{\infty} C_n e^{j2\pi f_s t} \quad (2)$$

$$\text{where } C_n = \frac{1}{T} \int_{-T/2}^{T/2} p(t) e^{-j2\pi f_s t} dt \quad (3)$$

$f_s$  is the sampling frequency or the frequency of the periodic signal of  $p(t)$

$$f_s = \frac{1}{T} \text{ hertz}$$

by substituting (2) into (1)

$$x_s(t) = \sum_{n=-\infty}^{\infty} C_n x(t) e^{j2\pi f_s t} \quad (4)$$

Now, by substituting (3) into (4) with interchanging the order of summation and integration, the result can be put in the following form

$$x_s(t) = x(t)p(t) = \sum_{n=-\infty}^{\infty} C_n x(t) e^{+jn2\pi f_s t}$$

## Spectrum of Sampled Signal

We can define the Fourier transform of  $x_s(t)$  as,

$$X_s(f) = \int_{-\infty}^{\infty} x_s(t) e^{-j2\pi ft} dt = \int_{-\infty}^{\infty} \sum_{n=-\infty}^{\infty} C_n x(t) e^{+jn2\pi f_s t} e^{-j2\pi ft} dt$$

with interchanging summation & integration

$$X_s(f) = \sum_{n=-\infty}^{\infty} C_n \int_{-\infty}^{\infty} x(t) e^{-j2\pi(f-nf_s)t} dt$$

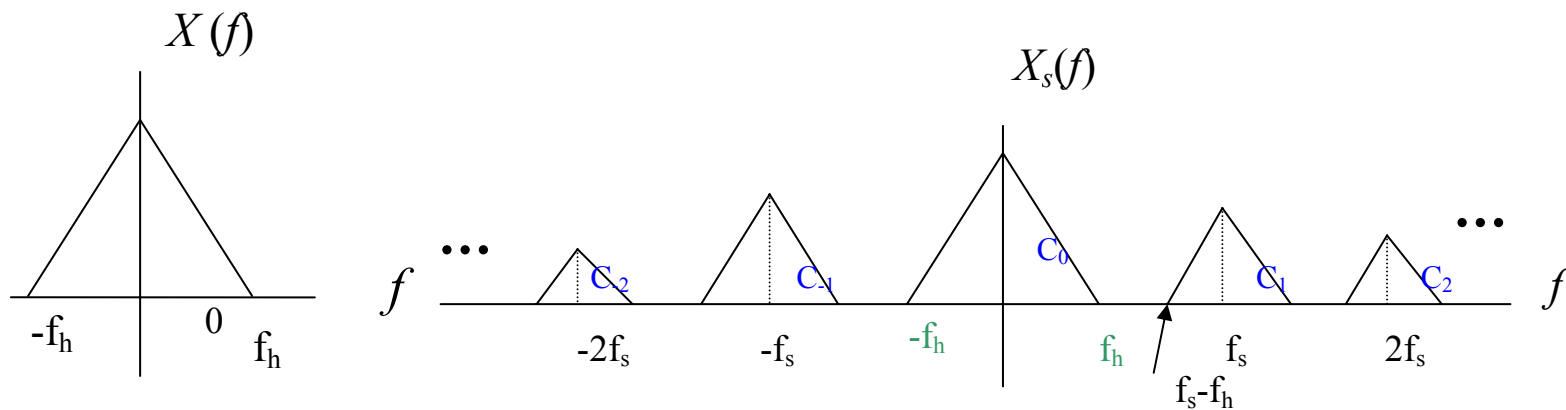
Hence, the Fourier transform of the sampled signal,  $x_s(t)$  is,

$$X_s(f) = \sum_{n=-\infty}^{\infty} C_n X(f - nf_s) \quad \text{where} \quad X(f - nf_s) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi(f-nf_s)t} dt$$

$$X_s(f) = \sum_{n=-\infty}^{\infty} C_n X(f - nf_s)$$

## Spectrum of the sampled signal

The spectrum of the sampled continuous times-signal  $x(t)$  is composed of the spectrum of  $x(t)$  plus the spectrum of  $x(t)$  translated to each harmonic of the sampling frequency.



**Note that:**  $X(f)=0$  for  $|f| \geq f_h$  and  $f_s \geq 2f_h$

- From the spectrum of the sampled signal we can clearly see that the original continuous signal can be completely reconstructed by using a low pass filter. Note that constant scaling factor  $C_0$  can be easily accounted for using an amplifier with gain equal to  $1/C_0$
- Now we are ready to state the sampling theorem.

## Sampling Theorem

A bandlimited signal  $x(t)$ , having no frequency components above  $f_h$  Hertz is completely specified by samples that are taken at a uniform rate greater than  $2 f_h$  Hertz. (the time between samples is no more than  $1/(2f_h)$  seconds).

$2f_h$  is known as Nyquist rate.

*Please see if we can do some thing like the visit the website by John Hopkins University or at least provide link*

<http://www.jhu.edu/~signals/sampling/index.html>

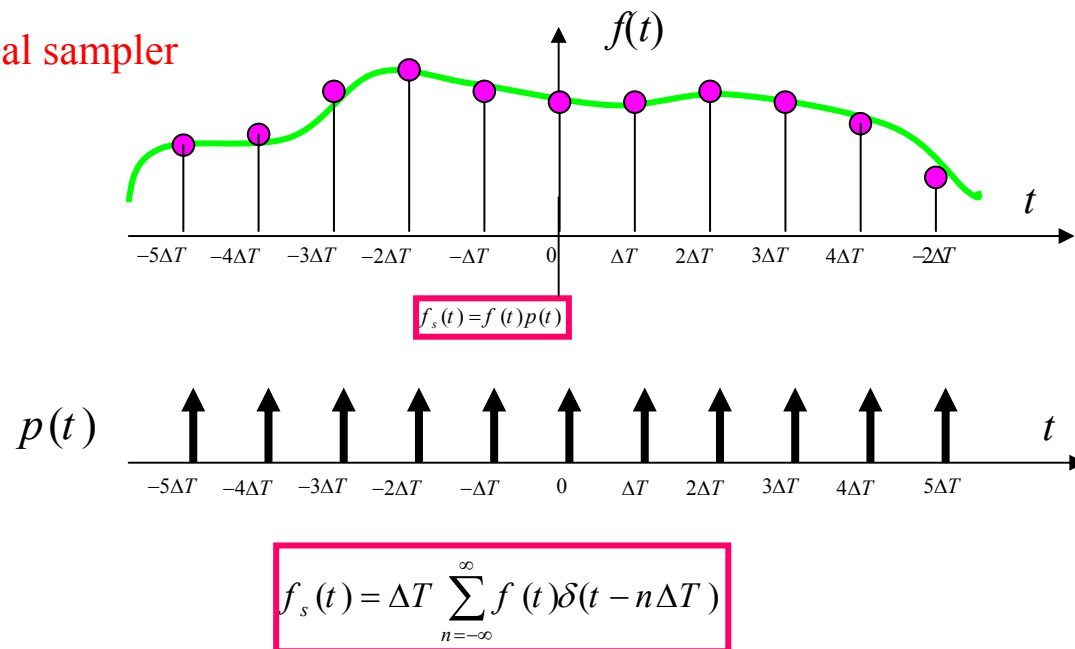
## Ideal Sampling: Impulse-Train Sampling Model

Consider  $p(t)$  is composed of an infinite train of impulse functions of period  $T$ . Thus,

$$p(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

which is the sampling function illustrated in the figure below:

Ideal sampler



$$f_s(t) = \Delta T \sum_{n=-\infty}^{\infty} f(t) \delta(t - n\Delta T)$$

## Continue.. Impulse-Train Sampling Model

The values of  $C_n$ , yields

$$C_n = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} \delta(t) e^{-jn2\pi f_s t} dt$$

Evaluated at  $t=0$  (sifting property), Thus

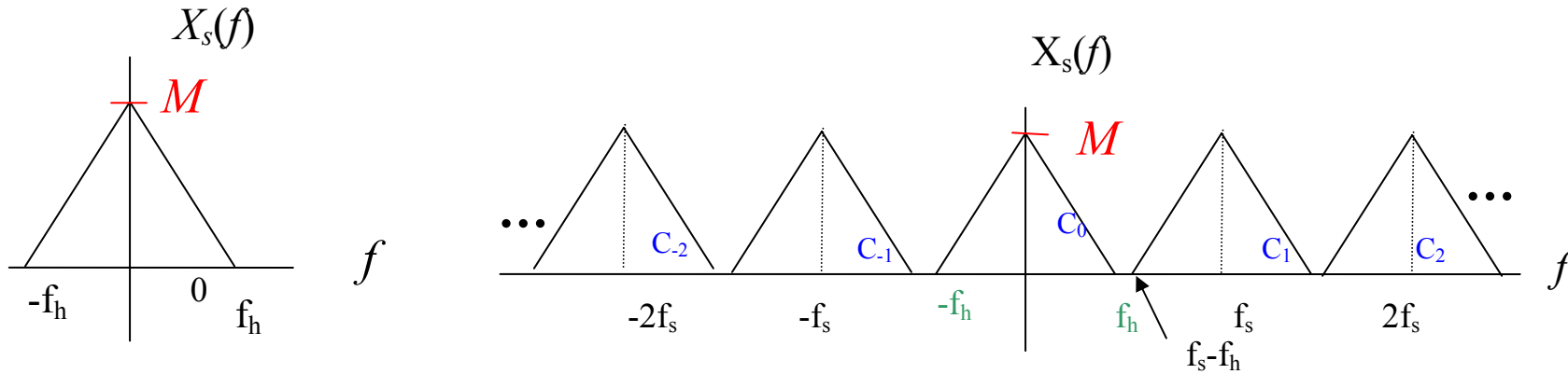
$$C_n = \frac{1}{T} = f_s$$

$$C_n = f_s \quad \text{for all } n$$

Hence, the spectrum of  $x(t)$  yields,

$$X_s(f) = f_s \sum_{-\infty}^{\infty} X(f - nf_s)$$

## Ideal Sampling: Impulse-Train Sampling Model



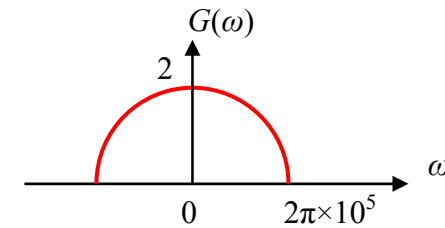
$$X_s(f) = f_s \sum_{n=-\infty}^{\infty} X(f - n f_s)$$

More Examples will be given in the coming lecture when we consider signal reconstruction

$x(t)$ 

## Self Test:

The figure below shows Fourier spectrum of a signal  $g(t)$



1. Determine the Nyquist interval and the sampling rate for  $g(t)$
2. Sketch the spectrum of the sampled signal, if  $g(t)$  is sampled (using uniformly spaced impulses) at **1.5\* Nyquist rate**.

## Animate solution

1. Nyquist Interval = 5 micro seconds

Nyquist rate = 200kHz

2. 1.5\*Nyquist rate=300 kHz

