

Digital signal processing

Chapter 1. Data sampling and reconstruction

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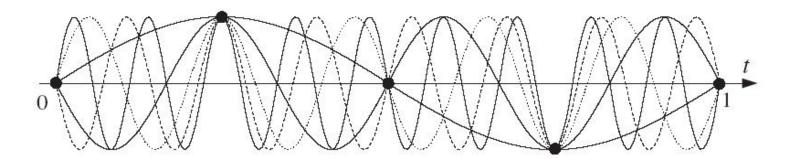
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Study Points

- Sampling Theorem
- Antialiasing
- Reconstruction
- Lab 1

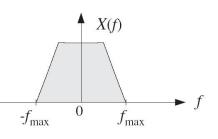
Sampling theorem: Physical meanings

■ Without additional conditions, a signal cannot be uniquely represented by a series of samplings:



Sampling theorem states that for accurate representation of a signal x(t) by its time samples x(nT), two conditions:

1. x(t) must be bandlimited, that is, its frequency spectrum must be limited to contain frequencies up to some maximum frequency, say f_{max} , and no frequencies beyond that.



2. The sampling rate f_s must be chosen to be at least twice the maximum frequency f_{max} , that is, $f_s > 2 f_{\text{max}}$

Sampling theorem: illustration

The analog signal x(t) is periodically measured every T seconds. ideal sampler x(t)analog signal \bigwedge X(f) $-f_{\text{max}}$ Convolution multiply * $x_p(t) = \sum_{n=0}^{+\infty} x(t)\delta(t - nT)$ $X_p(f) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} X(f - kf_s)$ replication original $X(f+2f_s)$ $X(f+f_s)$ X(f) $X(f-f_s)$ _Nyquist_ interval

Sampling theorem: illustration detail

A series of pulses p(t) with a period T are given to sample an arbitrary signal x(t)

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

Multiply p(t) and x(t)

$$x_{p}(t) = x(t) p(t)$$

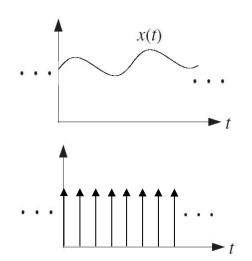
$$x_{p}(t) = \sum_{n=-\infty}^{\infty} x(nT)\delta(t-nT)$$

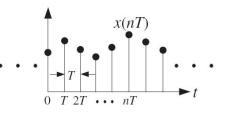
The property of Fourier transform is given as

$$\mathcal{F}[x(t)y(t)] = \frac{1}{2\pi} [X(j\omega) * Y(j\omega)] = X(jf) * Y(jf)$$

The Fourier transform of $x_p(t)$ is given as:

$$X_{p}(j\omega) = \frac{1}{2\pi}X(j\omega)*P(j\omega)$$





Sampling theorem: illustration detail

Define $\omega_s = 2 \pi / T$ as the sampling (angle) frequency, the Fourier transform of p(t) can be given as:

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

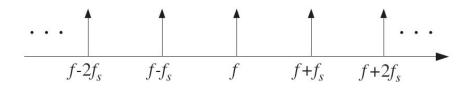
Obtain the Fourier coefficients with inverse transform:

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

Substituting a_k into $P(j\omega)$

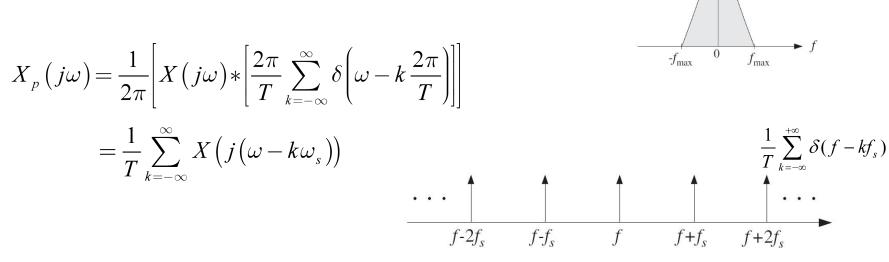
$$P(j\omega) = \sum_{k=-\infty}^{+\infty} a_k 2\pi \delta(\omega - k\omega_s) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\omega - k\frac{2\pi}{T})$$

The results of Fourier transform are shown in the figure



Sampling theorem: illustration detail

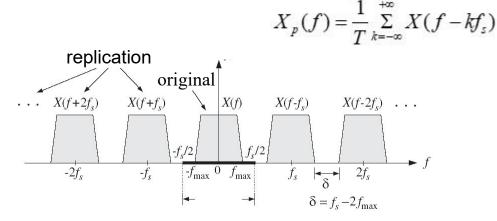
Assume the spectrum of x(t) is as shown in the schematic.



Express the Fourier transform with f_s

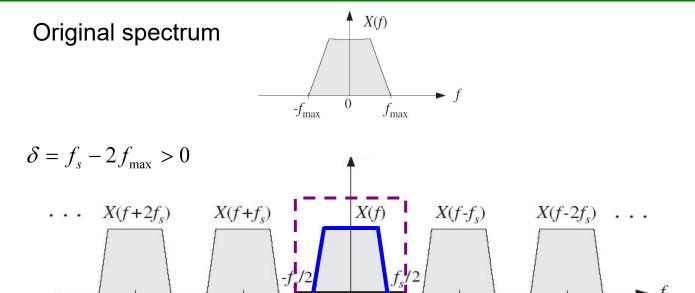
$$X_p(f) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} X(f - kf_s)$$

The result of the convolution is:

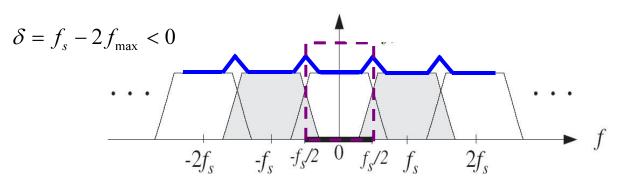


Sampling theorem: Aliasing phenomenon

 $-2f_s$



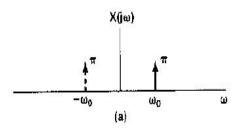
Spectrum of the sampling signal satisfying Sampling theorem



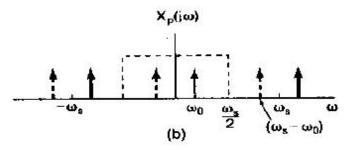
Aliasing phenomenon

Sampling theorem: Aliasing phenomenon

Original signal

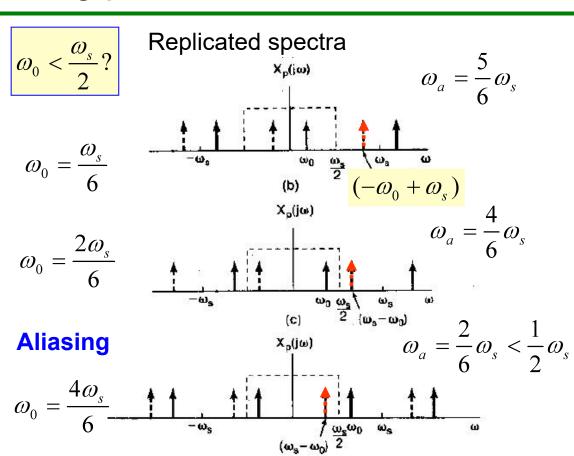


Sampled signal with

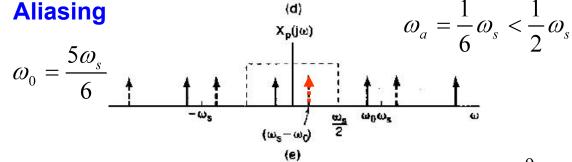


Replicated frequencies

$$\omega_a = (\pm \omega_0 + k\omega_s)$$
 $k \neq 0$



Aliasing



(d)

Sampling theorem: Aliasing phenomenon

The signal

$$x(t) = \sin(\pi t) + 4\sin(3\pi t)\cos(2\pi t)$$

where t is in msec, is sampled at a rate of 3 kHz. Determine the signal $x_a(t)$ aliased with x(t). Then, determine two other signals $x_1(t)$ and $x_2(t)$ that are aliased with the same $x_a(t)$, that is, such that $x_1(nT) = x_2(nT) = x_a(nT)$.

Solution: To determine the frequency content of x(t), we must express it as a sum of sinusoids. Using the trigonometric identity $2 \sin a \cos b = \sin(a+b) + \sin(a-b)$, we find:

$$x(t) = \sin(\pi t) + 2[\sin(3\pi t + 2\pi t) + \sin(3\pi t - 2\pi t)] = 3\sin(\pi t) + 2\sin(5\pi t)$$

$$f_a = f \mod(f_s)$$

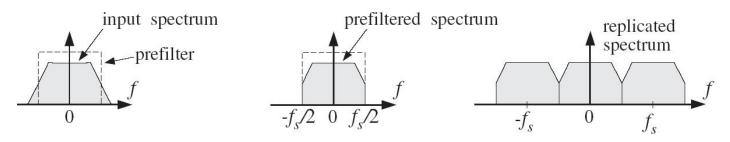
Thus, the frequencies present in x(t) are $f_1 = 0.5$ kHz and $f_2 = 2.5$ kHz. The first already lies in the Nyquist interval [-1.5, 1, 5] kHz so that $f_{1a} = f_1$. The second lies outside and can be reduced mod f_s to give $f_{2a} = f_2 \mod(f_s) = 2.5 \mod(3) = 2.5 - 3 = -0.5$. Thus, the given signal will "appear" as:

$$x_{a}(t) = 3\sin(2\pi f_{1a}t) + 2\sin(2\pi f_{2a}t)$$

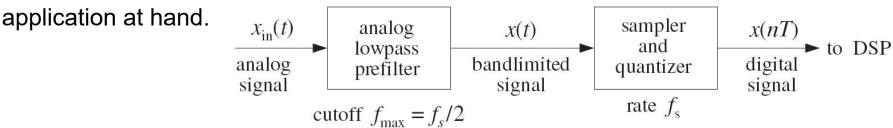
$$= 3\sin(\pi t) + 2\sin(-\pi t) = 3\sin(\pi t) - 2\sin(\pi t)$$

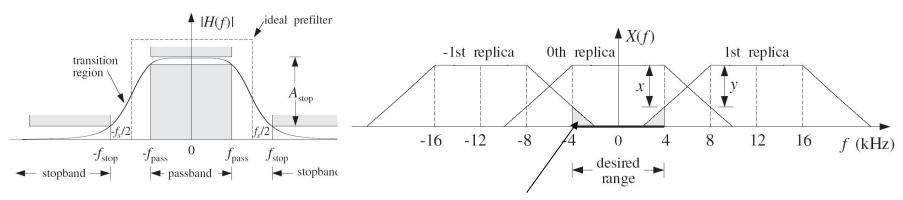
$$= \sin(\pi t)$$

Sampling theorem: Antialiasing prefilter



It should be emphasized that the rate f_s must be chosen to be high enough so that, after the prefiltering operation, the surviving signal spectrum within the Nyquist interval $[-f_s/2, f_s/2]$ contains all the significant frequency components for the





Practical antialiasing prefilter

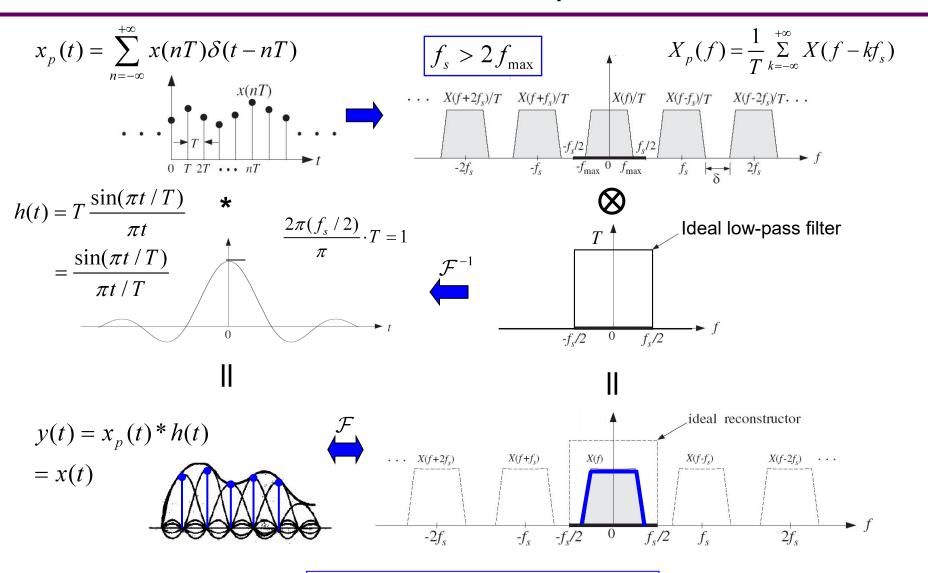
Remainders due to practical prefilter

Sampling theorem: Hardware limits

The sampling theorem provides a lower bound on the allowed values of f_s . The hardware used in the application imposes an upper bound.

In any case, there is a total processing or computation time for acquire sampling, data process and reconstruction, say totally T_{proc} seconds required for each sample. Thus, $T \ge T_{proc}$ or, $f_{proc} = 1/T_{proc}$, $f_s \le f_{proc}$,

$$2f_{max} \le f_s \le f_{proc}$$



Ideal low-pass filter is uncausal

The sampled signal is given as:

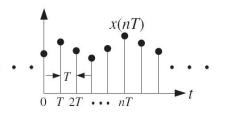
$$X_{p}(t) = \sum_{n=-\infty}^{\infty} x(nT)\delta(t-nT)$$

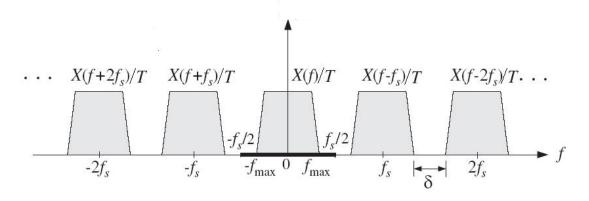
$$X_{p}(f) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} X(f-kf_{s})$$

$$X_{p}(f) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} X(f-kf_{s})$$

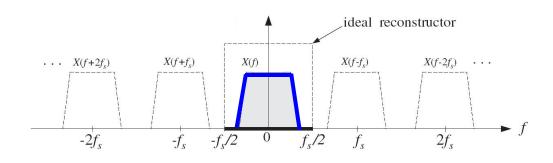
The spectrum of the sampled signal:

$$X_p(f) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} X(f - kf_s)$$





In order to restore the signal, a window with the range $(-f_s/2, f_s/2)$ is given:



The spectrum of an ideal rectangular window in frequency domain is:

$$H(f) = H(j\omega) = \begin{cases} T &, |\omega| < \omega_c \\ 0 &, |\omega| > \omega_c \end{cases}$$

The spectrum of Y(t) can be calculate by :

The property of Fourier transform is given as:

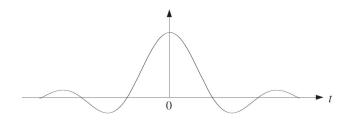
$$\mathcal{F}\left[x(t)*y(t)\right] = X(j\omega)Y(j\omega)$$

y(t) can be calculated by:

$$y(t) = x_p(t) * h(t)$$

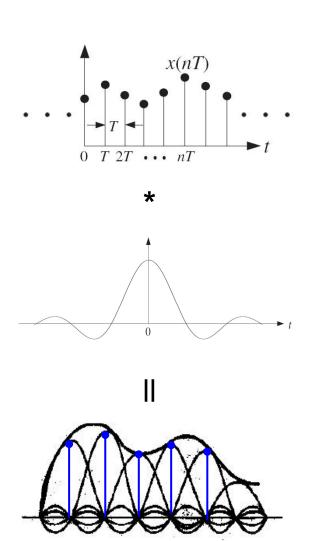
The Inverse Fourier transform of H(f) is:

$$h(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} H(j\omega) e^{j\omega t} d\omega = \frac{T}{2\pi} \int_{-\frac{\omega_s}{2}}^{+\frac{\omega_s}{2}} e^{j\omega t} d\omega = \frac{\sin(\pi t/T)}{\pi t/T}$$



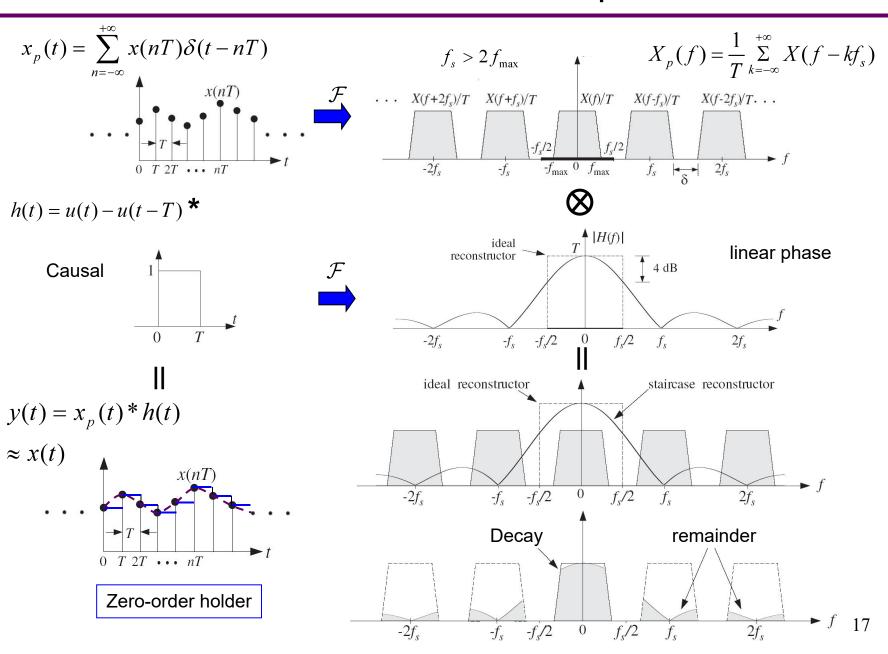
Then the original signal *y* (*t*) can be calculate by:

$$y(t) = x_p(t) * h(t)$$

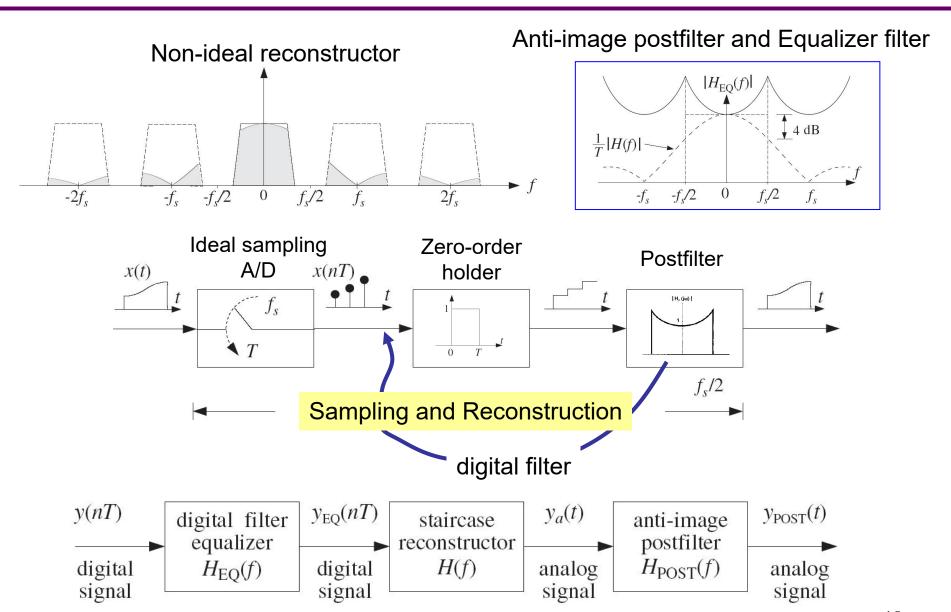


From the Figure given left:

Ideal low-pass filter is non causal

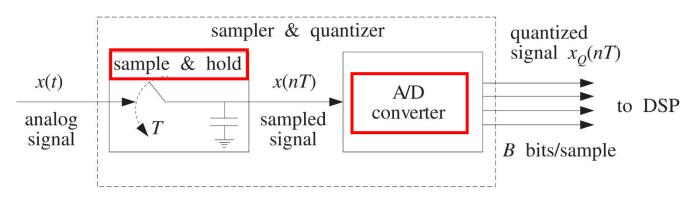


Sampling and Reconstruction without distortion



Quantization Process

Analog to Digital

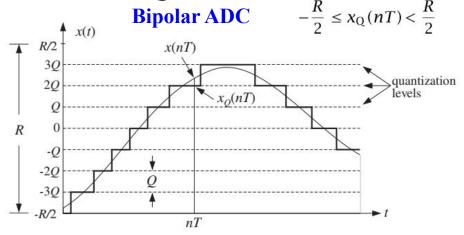


Quantization width or Quantizer resolution Q

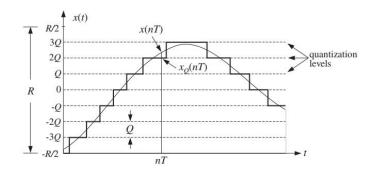
is defined as

$$Q = \frac{R}{2^B}$$

where *R* is the full-scale range



Quantization Process



Quantization error (rounding)

$$e(nT) = x_{Q}(nT) - x(nT)$$
$$e = x_{Q} - x$$

$$-\frac{Q}{2} \le e < \frac{Q}{2}$$

$$\overline{e} = \frac{1}{Q} \int_{-Q/2}^{Q/2} e \, de = 0$$

$$\overline{e^2} = \frac{1}{Q} \int_{-Q/2}^{Q/2} e^2 de = \frac{Q^2}{12}$$

root-mean-square (rms)

$$e_{\rm rms} = \sqrt{\overline{e^2}} = \frac{Q}{\sqrt{12}}$$

Quantization noise

$$p(e) = \begin{cases} \frac{1}{Q} & \text{if } -\frac{Q}{2} \le e \le \frac{Q}{2} \\ 0 & \text{otherwise} \end{cases}$$

$$\int_{-Q/2}^{Q/2} p(e) de = 1$$

$$E[e] = \int_{-Q/2}^{Q/2} ep(e) de$$
 and $E[e^2] = \int_{-Q/2}^{Q/2} e^2 p(e) de$

Signal to Noise Ratio (SNR)

Dynamic range (6 dB per bit rule)

$$20\log_{10}\left(\frac{R}{Q}\right) = 20\log_{10}\left(2^{B}\right) = B \cdot 20\log_{10}2,$$

$$SNR = 20\log_{10}(\frac{R}{Q}) = 6B \text{ dB}$$

ADC Device Price

Device	Bits	Frequency	Price
ADC1175-50CIMTX/NOPB	8bit	50M	19.92
ADC081S021CIMF/NOPB	8bit	200k	4.72
TLC1549IP	10bit	38k	28.9
MCP3021A5T-E/OT	10bit	22.3k	4.42
MCP3208	12bit	100k	20.01
TLC2543IDWR	12bit	66k	37.23
MCP3302-CI/ST	13bit	100k	27.4
AD9240ASZRL	14bit	10M	81.54
AD7616BSTZ	16bit	1M	94.96
ADS1115IDGSR	16bit	860k	23.58
PCAP01AD	17bit	500k	45.94
AD7608BSTZ	18bit	200k	251.46
MCP3422A0-E/MS	18bit	37.5k	19.19
AD7789BRMZ	19bit	16.6k	22.89
AD7781BRUZ	20bit	16.7k	96.08

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