

Chap6 Sampling of continuoustime signals

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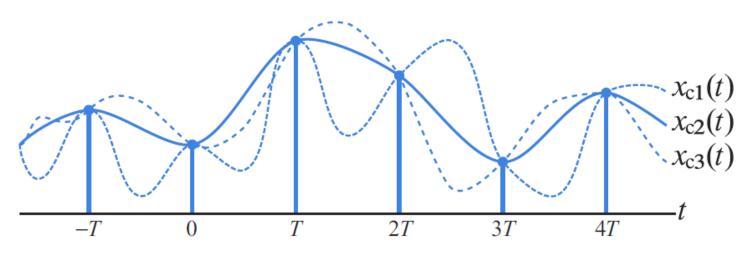


Chap 6 Sampling of continuoustime signals

- 6.1 Ideal periodic sampling of continuous-time signals
- 6.2 Reconstruction of a bandlimited signal from samples
- 6.3 The effect of undersampling: aliasing
- 6.4 Discrete-time processing of continuous-time signals
- 6.5 Practical sampling and reconstruction
- 6.6 Sampling of bandpass signals
- 6.7 Image sampling and reconstruction



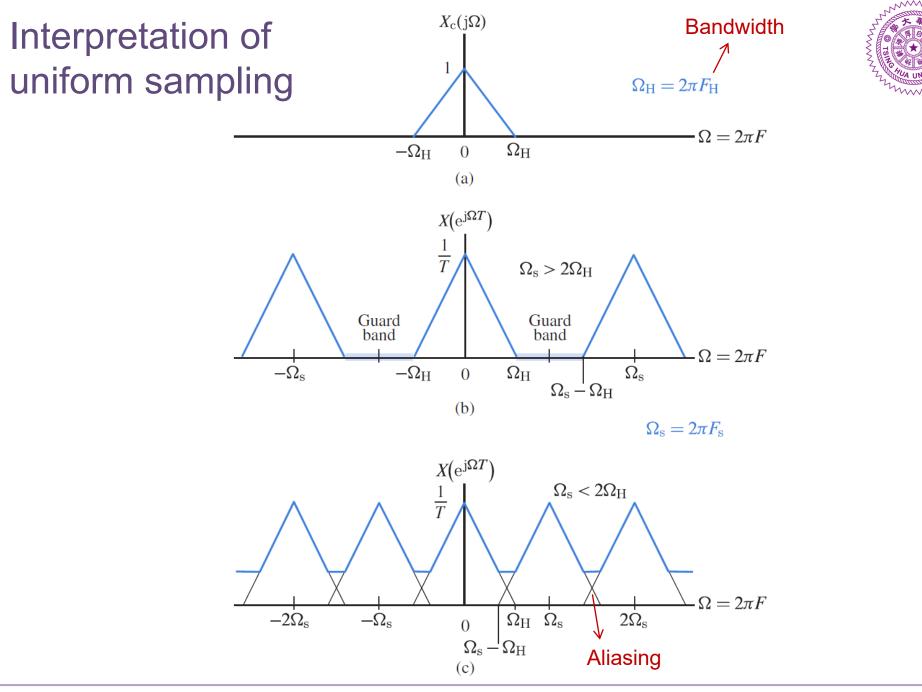
Many continuous-time signals could lead to the same sampled discrete-time signal => Non-invertible?





Frequency-domain relationship $X_{c}(j\Omega) = \int_{-\infty}^{\infty} x_{c}(t)e^{-j\Omega t}dt, \qquad \qquad X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n},$ $x_{c}(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X_{c}(j\Omega) e^{j\Omega t}d\Omega \qquad \qquad x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})e^{j\omega n}d\omega,$ $x[n] = x_{c}(nT) \qquad \qquad \omega = \Omega T = 2\pi FT = 2\pi \frac{F}{F_{s}} = 2\pi f$

$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j\frac{\omega}{T} - j\frac{2\pi}{T}k \right)$$
$$X(e^{j\Omega T}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j\Omega - j\frac{2\pi}{T}k \right)$$





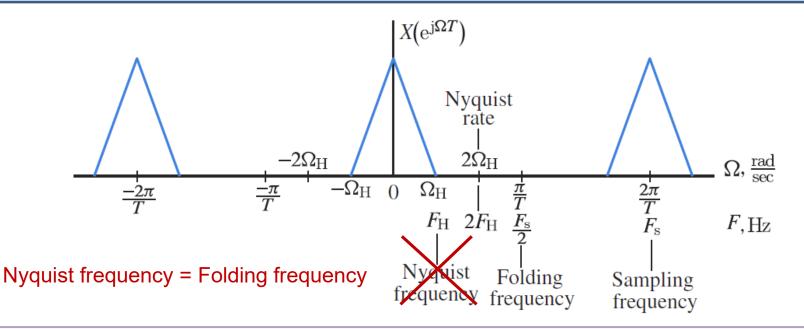
Sampling theorem: Let $x_c(t)$ be a continuous-time bandlimited signal with Fourier transform

$$X_{\rm c}(j\Omega) = 0 \quad \text{for}|\Omega| > \Omega_{\rm H}. \tag{6.18}$$

Then $x_c(t)$ can be uniquely determined by its samples $x[n] = x_c(nT)$, where $n = 0, \pm 1, \pm 2, \ldots$, if the sampling frequency Ω_s satisfies the condition

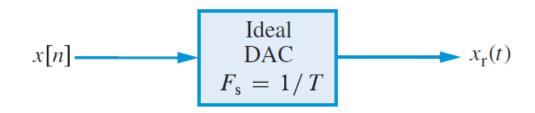
$$\Omega_{\rm s} = \frac{2\pi}{T} \ge 2\Omega_{\rm H}.\tag{6.19}$$

Sampling frequency \geq Nyquist rate

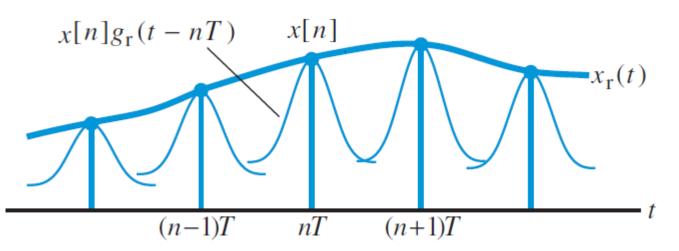


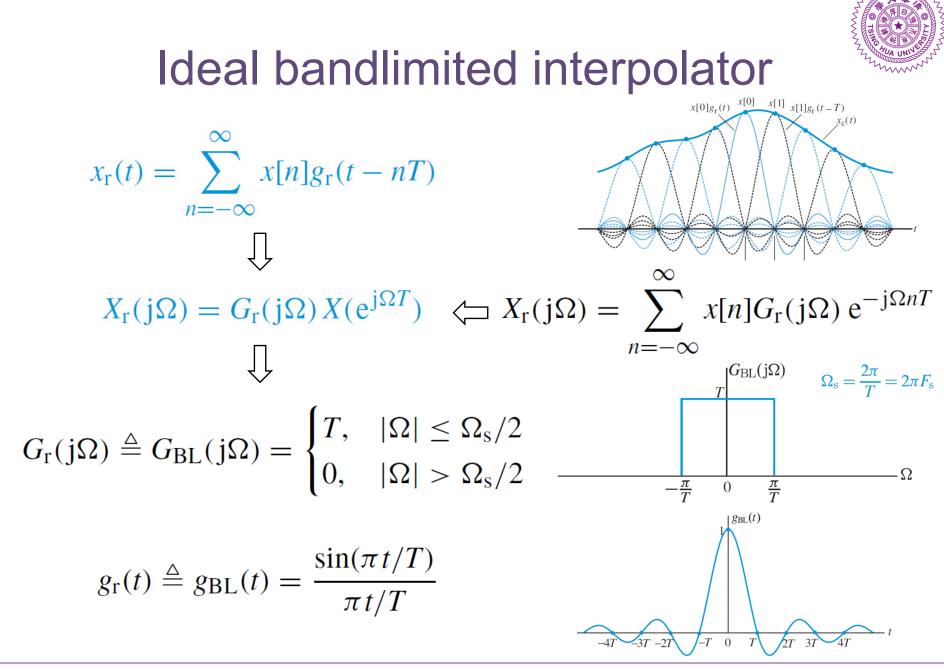
Reconstruction from discrete-time signals





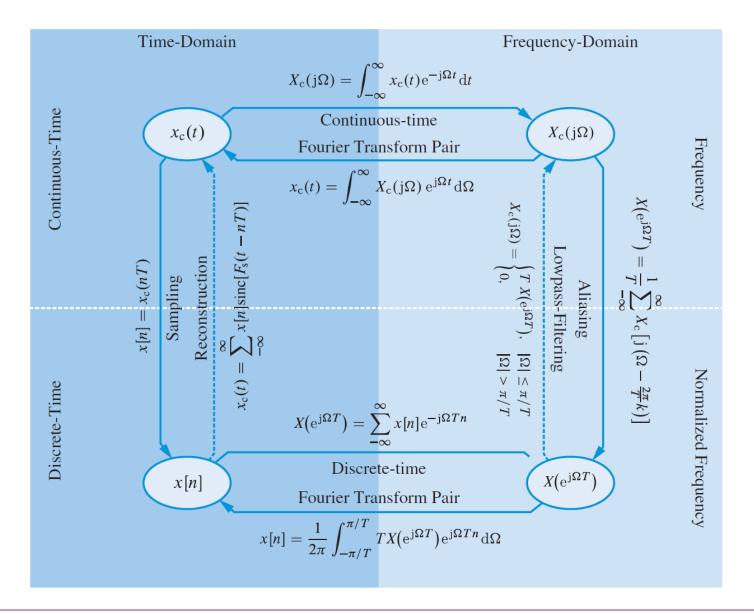
We can fully recover a bandlimited signal based on sampling theorem. => How to do it? Ideal vs. practical?

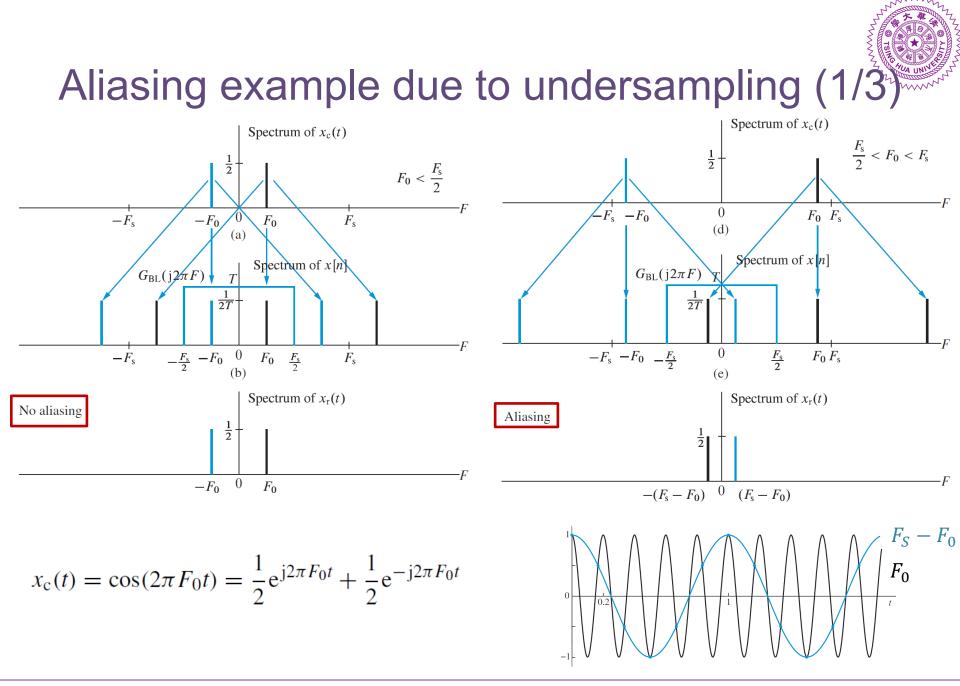




Periodic sampling vs. Bandlimited reconstruction







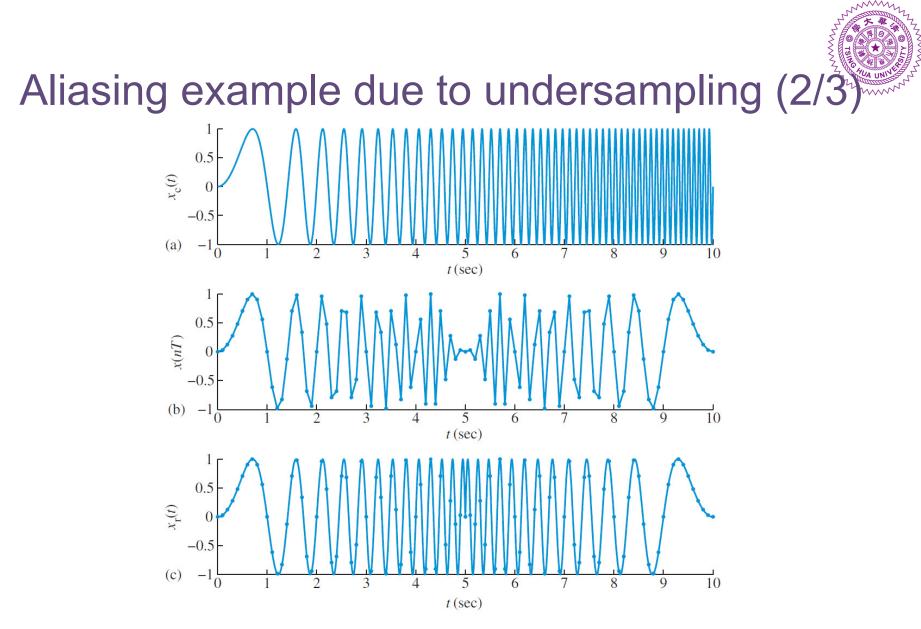


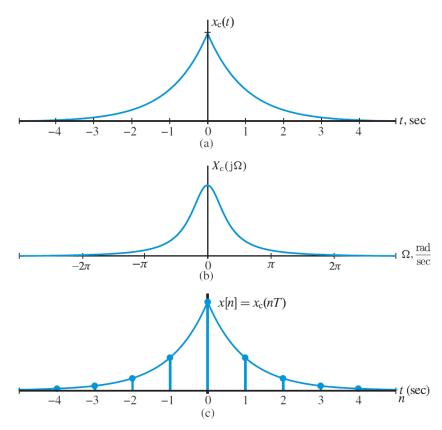
Figure 6.15 Sampling a continuous-time linear FM signal: (a) signal, (b) samples connected by line segments, and (c) output of ideal DAC.



Aliasing example due to undersampling (3/3)

$$x_{c}(t) = e^{-A|t|} \xleftarrow{\text{CTFT}} X_{c}(j\Omega) = \frac{2A}{A^{2} + \Omega^{2}}. \quad A > 0$$
$$x[n] = x_{c}(nT) = e^{-A|n|T} = (e^{-AT})^{|n|} = a^{|n|}, \quad a \triangleq e^{-AT}$$

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n} = \frac{1-a^2}{1-2a\cos(\omega)+a^2}, \quad \omega = \Omega/F_s$$



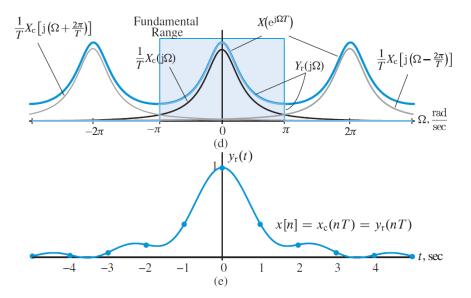
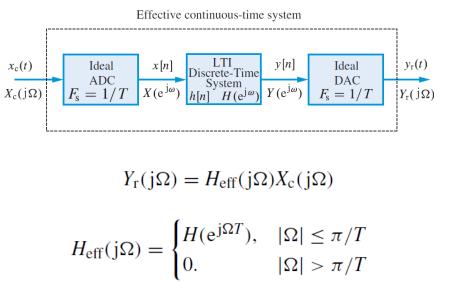


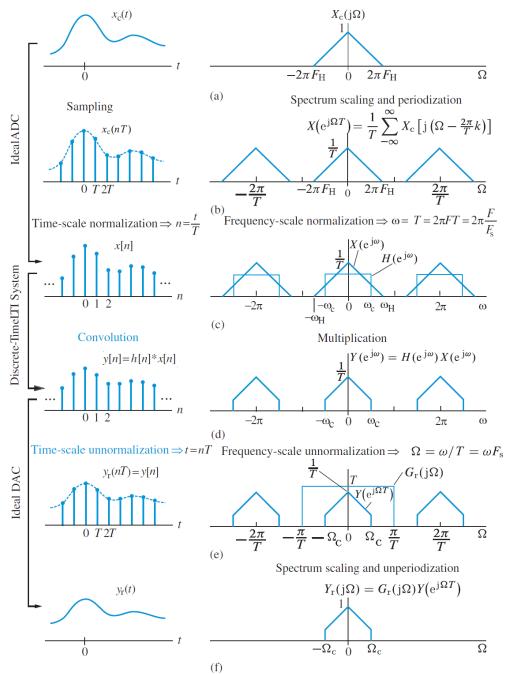
Figure 6.16 Aliasing effects in sampling and reconstruction of a continuous-time nonbandlimited signal: (a) continuous-time signal $x_c(t)$, (b) spectrum of $x_c(t)$, (c) discrete-time signal x[n] sampled at T = 1 s, (d) spectrum of x[n], and (e) bandlimited reconstruction $y_r(t)$. In this case, aliasing distortion is unavoidable.

Discrete-time filtering of continuous-time signals



Constraints:

- Ideal ADC and DAC
- Bandlimited processing





Example of ideal bandlimited differentiator

 $H_{\rm c}(\mathbf{j}\Omega) = \begin{cases} \mathbf{j}\Omega, & |\Omega| \le \Omega_{\rm H} \end{cases}$

Bandlimited differentiator

Discrete filter

$$\begin{bmatrix} 0. & \text{otherwise} \\ & \bigcup & \omega = \Omega T \text{ and } \Omega_{\mathrm{H}} T = \pi \\ H(\mathrm{e}^{\mathrm{j}\omega}) = \frac{1}{T} H_{\mathrm{c}}(\mathrm{j}\omega/T) = \frac{\mathrm{j}\omega}{T^{2}}. \quad |\omega| \leq \pi \\ & \bigcup \\ & \prod \\ h[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(\frac{\mathrm{j}\omega}{T^{2}}\right) \mathrm{e}^{\mathrm{j}\omega n} \mathrm{d}\omega = \begin{cases} 0, & n = 0 \\ \frac{\mathrm{cos}(\pi n)}{nT^{2}}. & n \neq 0 \end{cases}$$



Example of second-order system (1/2)

Continuous system

$$\frac{\mathrm{d}^2 y_{\mathrm{c}}(t)}{\mathrm{d}t^2} + 2\zeta \,\Omega_{\mathrm{n}} \frac{\mathrm{d}y_{\mathrm{c}}(t)}{\mathrm{d}t} + \Omega_{\mathrm{n}}^2 y_{\mathrm{c}}(t) = \Omega_{\mathrm{n}}^2 x_{\mathrm{c}}(t)$$

$$H_{\rm c}(s) = \frac{Y_{\rm c}(s)}{X_{\rm c}(s)} = \frac{\Omega_{\rm n}^2}{s^2 + 2\zeta \Omega_{\rm n} s + \Omega_{\rm n}^2} \qquad \text{Non-bandlimited}$$

$$h_{\rm c}(t) = \frac{\Omega_{\rm n}}{\sqrt{1-\zeta^2}} e^{-\zeta \Omega_{\rm n} t} \sin\left[\left(\Omega_{\rm n}\sqrt{1-\zeta^2}\right)t\right] u(t)$$

$$\square \text{ Discrete sampling}$$

(impulse-invariance transformation)

$$h[n] = h_{\rm c}(nT) = \frac{\Omega_{\rm n}}{\sqrt{1-\zeta^2}} e^{-\zeta \Omega_{\rm n} nT} \sin\left[\left(\Omega_{\rm n}\sqrt{1-\zeta^2}\right)nT\right] u(n)$$

Discrete filter

$$= \frac{\Omega_{\rm n}}{\sqrt{1-\zeta^2}} \left({\rm e}^{-\zeta \,\Omega_{\rm n} T} \right)^n \sin \left[\left(\Omega_{\rm n} T \sqrt{1-\zeta^2} \right) n \right] u(n).$$



Example of second-order system (2/2)

Z transform

$$H(z) = \frac{\Omega_{\rm n}}{\sqrt{1-\zeta^2}} \frac{e^{-\zeta \Omega_{\rm n} T} \sin\left(\Omega_{\rm n} T \sqrt{1-\zeta^2}\right) z^{-1}}{1 - 2e^{-\zeta \Omega_{\rm n} T} \cos\left(\Omega_{\rm n} T \sqrt{1-\zeta^2}\right) z^{-1} + e^{-2\zeta \Omega_{\rm n} T} z^{-2}}$$

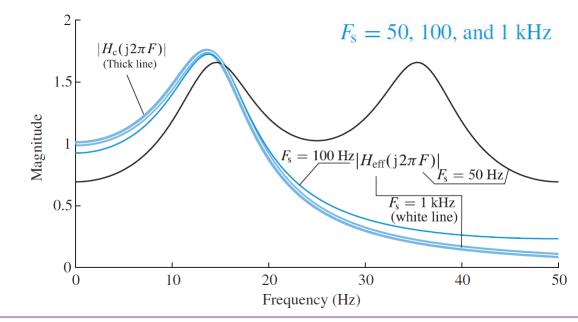
Difference equation

 $\zeta = 0.3, \, \Omega_{\rm n} = 30\pi$

uation
$$y[n] = \frac{\Omega_n}{\sqrt{1-\zeta^2}} e^{-\zeta\Omega_n T} \sin\left(\Omega_n T \sqrt{1-\zeta^2}\right) x[n-1]$$

 $+ 2e^{-\zeta\Omega_n T} \cos\left(\Omega_n T \sqrt{1-\zeta^2}\right) y[n-1] - e^{-2\zeta\Omega_n T} y[n-2]$

Example

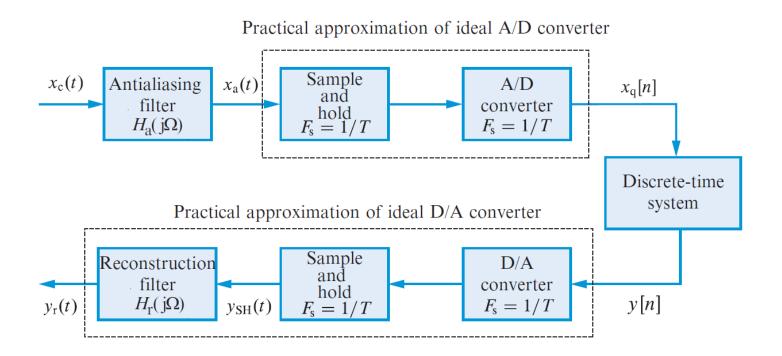




Practical discrete-time processing of continuous signals

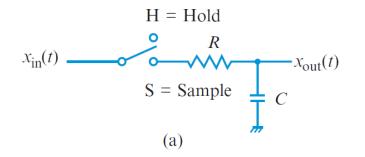
Main differences from ideal processing:

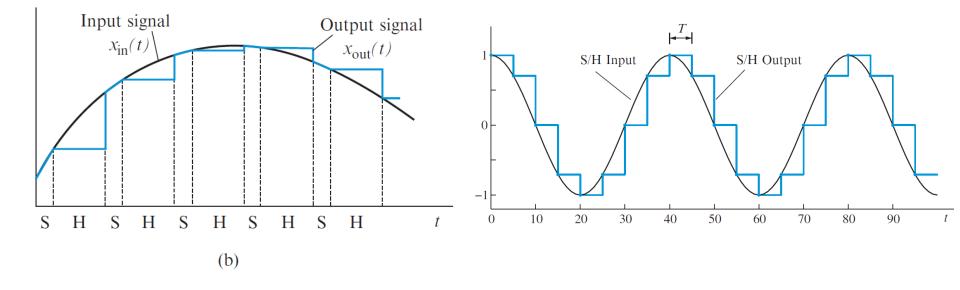
- Practical signals are time-limited, thus not bandlimited => Need analog antialiasing pre-filtering
- Impulse sampling is not practical for ADC
- Discrete signal values are quantized
- Ideal interpolator (sinc) is not practical for DAC
- => Use sample-and-hold circuits instead
- => Need to consider quantization noise
- => Use S/H reconstruction





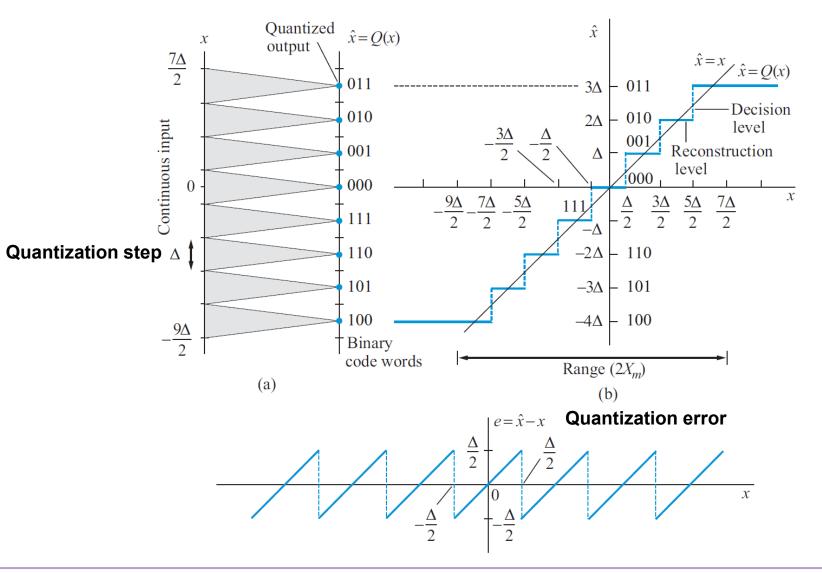
Sample-and-hold circuits for ADC







Quantization of ADC converter





Quantization Noise

Quantization error power (assume error is uniformly distributed)

$$P_{\rm Q} = \frac{1}{2\tau} \int_{-\tau}^{\tau} e_{\rm c}^2(t) dt = \frac{\Delta^2}{12}$$

Signal power

(assume sinusoidal signals)

$$P_{\rm S} = \frac{1}{T_{\rm p}} \int_0^{T_{\rm p}} X_{\rm m}^2 \sin^2\left(\frac{2\pi}{T_{\rm p}}\right) dt = \frac{X_{\rm m}^2}{2}$$
$$x_{\rm c}(t) = X_{\rm m} \sin(\frac{2\pi}{T_{\rm p}}t)$$

Signal-to-quantization-noise-ratio

$$SQNR \triangleq \frac{P_S}{P_Q} = \frac{3}{2} \times 2^{2B}$$

 $SQNR(dB) = 10 \log_{10} SQNR = 6.02B + 1.76$

(one additional bit adds 6dB)



S/H reconstruction

 ∞

S/H amplifier

S/H amplifier

$$x_{SH}(t) = \sum_{n=-\infty} x_q[n]g_{SH}(t - nT)$$

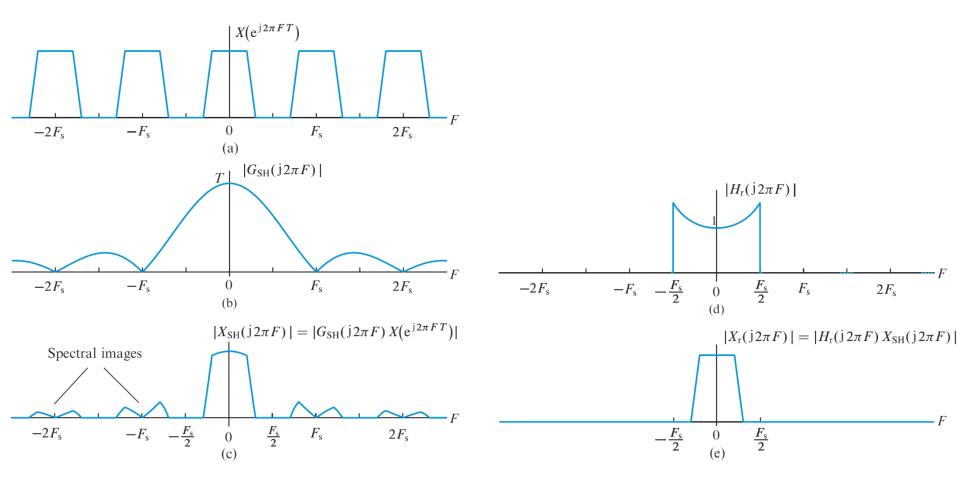
$$g_{SH}(t) = \begin{cases} 1, & 0 \le t \le T \\ 0, & \text{otherwise} \end{cases} \xrightarrow{\text{CTFT}} G_{SH}(j\Omega) = \frac{2\sin(\Omega T/2)}{\Omega} e^{-j\Omega T/2}$$

$$\xrightarrow{\text{Sample and hold}} \xrightarrow{T} \xrightarrow{\text{Ideal bandlimited}} \xrightarrow{\text{interpolator}} G_{BL}(j\Omega)$$

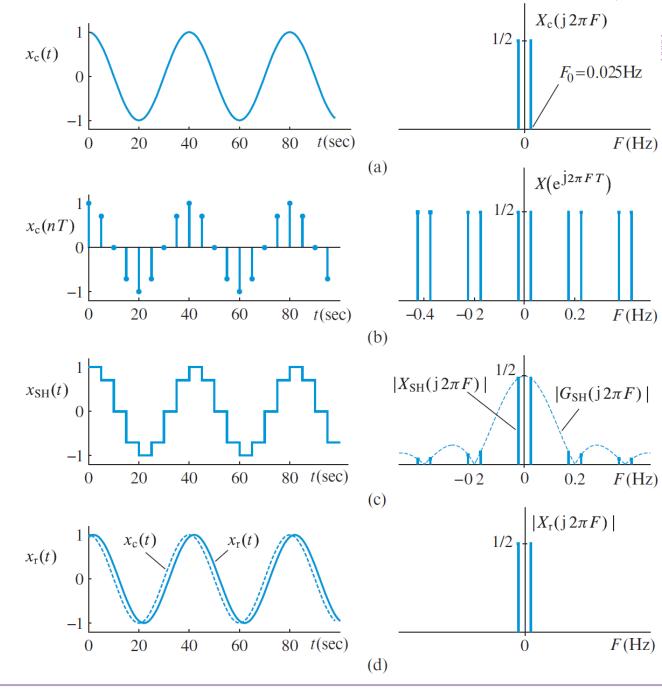
$$\xrightarrow{\frac{2\pi}{T}} -\frac{\pi}{T} \xrightarrow{0} \xrightarrow{\pi} \frac{2\pi}{T} -\frac{2\pi}{\Omega}$$
Reconstruction filter
(anti-imaging and equalization)
$$H_r(j\Omega) = \begin{cases} \frac{\Omega T/2}{\sin(\Omega T/2)} e^{j\Omega T/2}, & |\Omega| < \pi/T \\ 0, & \text{otherwise} \end{cases}$$



Frequency domain of S/H reconstruction

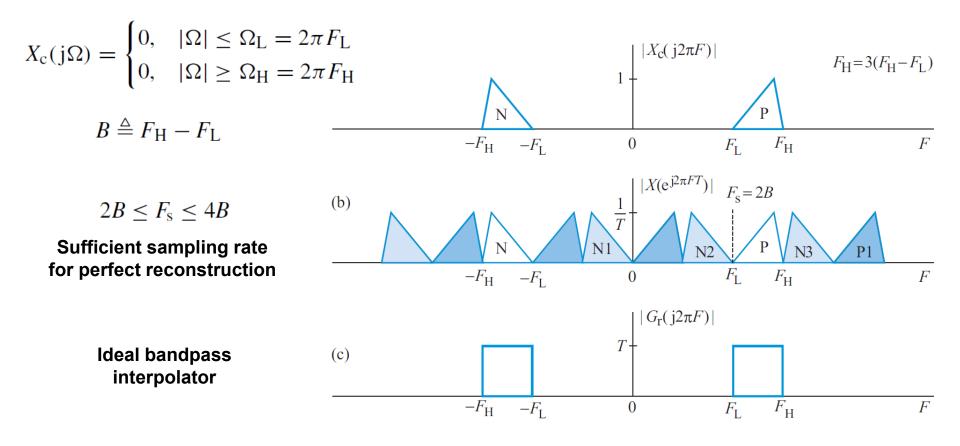


Example of S/H reconstruction





Sampling of bandpass signals





2-D transform for image processing

2D CTFT

$$S_{c}(F_{x}, F_{y}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} s_{c}(x, y) e^{-j2\pi (xF_{x} + yF_{y})} dxdy,$$

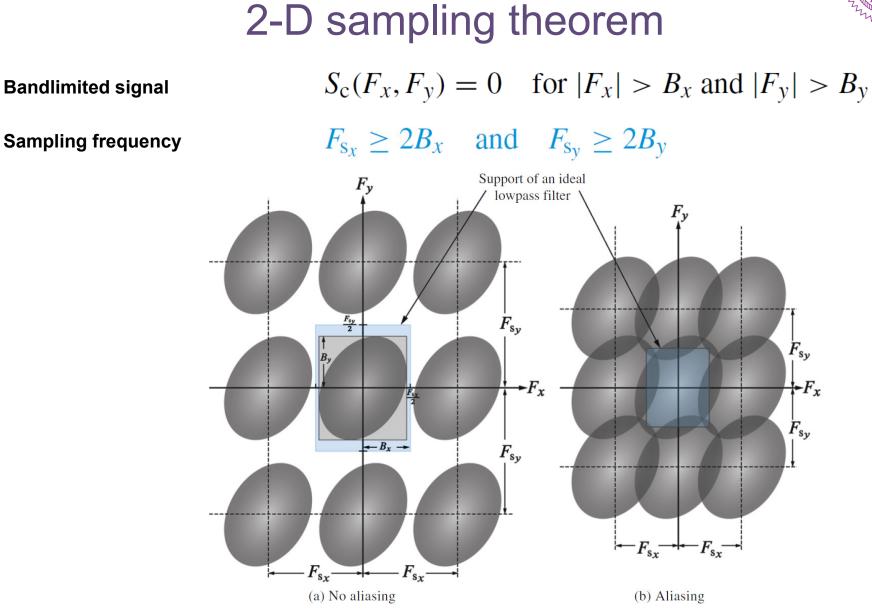
$$s_{c}(x,y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} S_{c}(F_{x},F_{y}) e^{j2\pi(xF_{x}+yF_{y})} dF_{x} dF_{y},$$

2D sampling

$$s[m,n] \triangleq s_{\rm c}(m\Delta x, n\Delta y)$$

2D DTFT $\tilde{S}(F_x, F_y) \triangleq \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} s[m, n] e^{j2\pi (m\Delta x F_x + n\Delta y F_y)}$ $= \frac{1}{\Delta x \Delta y} \sum_{k=-\infty}^{\infty} \sum_{\ell=-\infty}^{\infty} S_c(F_x - kF_{s_x}, F_y - \ell F_{s_y})$







Moiré pattern due to aliasing

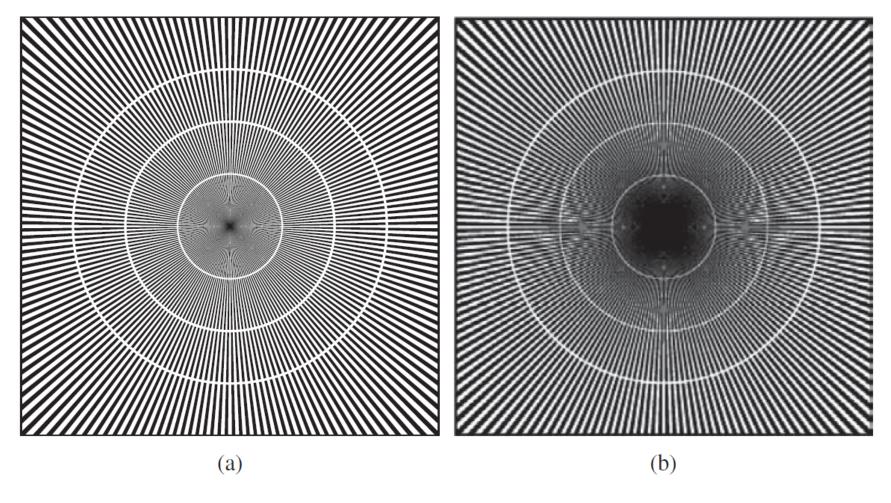


Figure 6.35 Moiré pattern due to aliasing: (a) original pattern, (b) 72 dpi pattern.



Tradeoff between resolution and aliasing

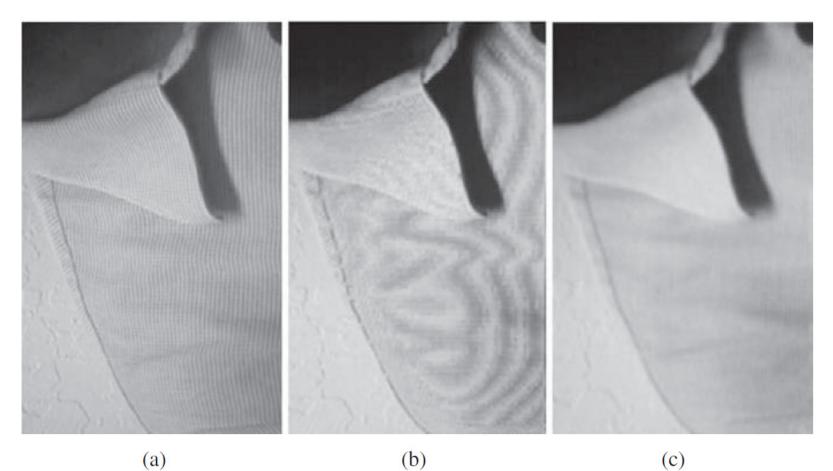


Figure 6.36 Aliasing in resampled images (digital aliasing): (a) original image, (b) resampled without pre-filtering, and (c) resampled with pre-filtering.