



E.g., uniform sampling: x(n) = x<sub>a</sub>(nT) where T is the sampling period

#### **Discrete-Time Signals and Systems**

#### **Reference:**

Sections 6.1, 6.2, 6.4, 6.5 of

John G. Proakis and Dimitris G. Manolakis, *Digital Signal Processing: Principles, Algorithms, and Applications*, 4th edition, 2007.

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Chapter 6: Sampling and Reconstruction of Signals 6.1 Ideal Sampling and Reconstruction of Cts-Time Signals

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#### Sampling Process

- ► To effectively reconstruct an analog signal from its samples, the sampling frequency  $F_s = \frac{1}{T}$  must be selected to be "large enough".
- Sampling in the time-domain:

$$x(n) = x_a(nT), \quad -\infty < n < \infty$$



## Chapter 6: Sampling and Reconstruction of Signals 6.1 Ideal Sampling and Reconstruction of Cts-Time Signals Sampling Process

- ► Time-Domain Sampling: frequency-domain perspective
  - The sampling frequency  $F_s = \frac{1}{T}$  must be selected to be large enough such that the sampling processing does not cause any loss of spectral information (i.e., no aliasing).
- Recall CTFT and DTFT for aperiodic  $x_a(t)$  and x(n):

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Sampling Process

$$LHS = \int_{-\frac{1}{2}}^{\frac{1}{2}} X(f) e^{j2\pi fn} df \quad \text{let } f = \frac{F}{F_s}$$
$$= \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} X\left(\frac{F}{F_s}\right) e^{j2\pi \frac{F}{F_s}n} \frac{dF}{F_s}$$
$$= \frac{1}{F_s} \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} X\left(\frac{F}{F_s}\right) e^{j2\pi \frac{F}{F_s}n} dF$$

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Chapter 6: Sampling and Reconstruction of Signals Sampling Process  $x(n) \equiv x_a(nT) = x_a\left(\frac{n}{F_s}\right)$   $\int_{-\frac{1}{2}}^{\frac{1}{2}} X(f)e^{j2\pi fn}df = \int_{-\infty}^{\infty} X_a(F)e^{j2\pi F\frac{n}{F_s}}dF$   $K(F) = \frac{\int_{-\infty}^{\infty} X_a(F)e^{j2\pi F\frac{n}{F_s}}dF}{KHS}$ 

 We can use the following relationship that comes about from the "reinterpretation stage".

$$t = nT$$
 or  $\frac{1}{F} = \frac{T}{f}$   
 $f = T \cdot F = \frac{F}{F_s} \Longrightarrow f = \frac{F}{F_s}$ 

Chapter 6: Sampling and Reconstruction of Signals 6.1 Ideal Sampling and Reconstruction of Cts-Time Signals  
Sampling Process  

$$RHS = \int_{-\infty}^{\infty} X_a(F) e^{j2\pi \frac{F}{F_s}n} dF$$

$$= \sum_{k=-\infty}^{\infty} \int_{-\frac{F_s}{2}-kF_s}^{+\frac{F_s}{2}-kF_s} X_a(F) e^{j2\pi \frac{F}{F_s}n} dF \quad \text{let } F' = F + kF_s$$

$$= \sum_{k=-\infty}^{\infty} \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} X_a(F' - kF_s) \underbrace{e^{j2\pi \frac{F'}{F_s}n}}_{=e^{j2\pi \frac{F'}{F_s}n}} dF'$$

$$= \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} \left[ \sum_{k=-\infty}^{\infty} X_a(F - kF_s) \right] e^{j2\pi \frac{F_s}{F_s}n} dF$$

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## Sampling Process

Since,

$$LHS = RHS$$

$$\frac{1}{F_s} \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} X\left(\frac{F}{F_s}\right) e^{j2\pi \frac{F}{F_s}n} dF = \int_{-\frac{F_s}{2}}^{\frac{F_s}{2}} \left[\sum_{k=-\infty}^{\infty} X_a (F - kF_s)\right] e^{j2\pi \frac{F}{F_s}n} dF$$

By inspection,

$$X\left(\frac{F}{F_s}\right) = F_s \sum_{k=-\infty}^{\infty} X_a \left(F - kF_s\right)$$

or letting  $f = \frac{F}{F_{e}}$ ,

$$X(f) = F_s \sum_{k=-\infty}^{\infty} X_a (f \cdot F_s - kF_s) = F_s \sum_{k=-\infty}^{\infty} X_a ((f-k)F_s)$$

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Chapter 6: Sampling and Reconstruction of Signals 6.1 Ideal Sampling and Reconstruction of Cts-Time Signals

#### Sampling Theorem

A bandlimited continuous-time signal, with highest frequency (bandwidth) B Hz, can be uniquely recovered from its samples provided that the sampling rate  $F_s \ge 2B$  samples per second.

Perfect reconstruction is possible via the ideal interpolation formula:

$$x_{a}(t) = \sum_{n=-\infty}^{\infty} \underbrace{x_{a}(nT)}_{=\times(n)} \frac{\sin(\frac{\pi}{T}(t-nT))}{\frac{\pi}{T}(t-nT)}$$

See Figure 6.1.2 of text

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Chapter 6: Sampling and Reconstruction of Signals 6.1 Ideal Sampling and Reconstruction of Cts-Time Signals

#### Aliasing

- Sampling and reconstruction of nonbandlimited signals results in aliasing.
- The degree of aliasing/quality of the reconstruction depends on the sampling rate in relation to the decay of the analog signal spectrum.
- ► Example:

$$x_a(t)=e^{-A|t|} \stackrel{\mathcal{F}}{\longleftrightarrow} X_a(F)=rac{2A}{A^2+(2\pi F)^2}, \ A>0$$

See Figure 6.1.8 of text



ideal sampling and interpolation assumed:



#### Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

#### Prefilter



- ensures that bandwidth is limited to avoid aliasing or reduce subsequent computational requirements
- rejects additive noise in higher frequency ranges

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- Q: Is there a discrete-time system such that the overall system above is equivalent to a continuous-time LTI system?
- A: Yes if  $x_a(t)$  is bandlimited and  $F_s > 2B$ .

#### Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

#### Discrete-Time System Design

• Consider a desired continuous-time LTI system:

$$y_{a}(t) = h_{a}(t) * x_{a}(t) = \int_{-\infty}^{\infty} h_{a}(\tau) x_{a}(t-\tau) dt$$
$$Y_{a}(F) = H_{a}(F) X_{a}(F)$$

• If  $x_a(t)$  is bandlimited and  $F_s > 2B$  (no overlap), then

$$X(F) = \frac{1}{T} X_a(F) \quad \text{for } |F| \le \frac{F_s}{2}$$

See Figure 6.2.2 of text

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Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

Discrete-Time System Design

► The desired equivalent cts-time system is:

 $Y_a(F) = H_a(F)X_a(F)$ 

• The actual overall response assuming a dst-time filter H(F) is:

 $Y_{a}(F) = \begin{cases} H(F)X_{a}(F) & |F| \leq \frac{F_{s}}{2} \\ 0 & \text{otherwise} \end{cases}$ 

Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

### Discrete-Time System Design

Recall, from ideal interpolation and assuming dst-time system H(F):

$$Y_{a}(F) = \begin{cases} TY(F) & |F| \leq \frac{F_{s}}{2} \\ 0 & \text{otherwise} \end{cases}$$
$$= \begin{cases} TH(F)X(F) & |F| \leq \frac{F_{s}}{2} \\ 0 & \text{otherwise} \end{cases}$$
$$= \begin{cases} TH(F)\frac{1}{T}X_{a}(F) & |F| \leq \frac{F_{s}}{2} \\ 0 & \text{otherwise} \end{cases}$$
$$= \begin{cases} H(F)X_{a}(F) & |F| \leq \frac{F_{s}}{2} \\ 0 & \text{otherwise} \end{cases}$$
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Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

## Discrete-Time System Design

Therefore,

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$$H_a(F) = \begin{cases} H(F) & |F| \le \frac{F_s}{2} \\ 0 & |F| > \frac{F_s}{2} \end{cases}$$

Naturally,

$$H(F) = H_a(F) \text{ for } |F| \le \frac{F_s}{2}$$
$$H(F) = \sum_{k=-\infty}^{\infty} H_a(F - kF_s)$$
$$h(n) = T h_a(nT)$$



Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

#### Example: Ideal bandlimited differentiator

Choosing  $F_s = 2F_c$ , we define the ideal discrete-time differentiator as:

$$H(F) = H_a(F) = j2\pi F, \quad |F| \le \frac{F_s}{2}$$

and

$$H(F) = \sum_{k=-\infty}^{\infty} H_a(F - kF_s)$$

See Figure 6.2.5 of text

Chapter 6: Sampling and Reconstruction of Signals 6.2 Dst-Time Processing of Cts-Time Signals

### Example: Ideal bandlimited differentiator

Goal: to design a discrete-time filter such that the overall system (including A/D and D/A conversion) is defined by

$$y_a(t) = rac{dx_a(t)}{dt}$$

The frequency response of the overall system is given by:

$$H_{a}(F) = \frac{Y_{a}(F)}{X_{a}(F)} = j2\pi F$$
  
= 
$$\begin{cases} j2\pi F & |F| \le F_{c} \\ 0 & |F| > F_{c} \end{cases}$$
 for bandlimited signals

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Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals

#### **Bandlimited Signal**

► A continuous-time bandpass signal with bandwidth B and center frequency F<sub>c</sub> has its frequency content in two frequency bands defined by 0 < F<sub>L</sub> < |F| < F<sub>H</sub>





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Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals

## Arbitrary Band Positioning

• For  $F_s > B$ , aliasing is due to overlap of "positive" spectral band with "negative" or vice versa.



Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals Arbitrary Band Positioning  $(k-1)F_s \leq 2F_L \Longrightarrow F_s \leq \frac{2F_L}{k-1}$  $2F_H \leq kF_s \Longrightarrow F_s \geq \frac{2F_H}{k}$  $\frac{2F_H}{k} \le F_s \le \frac{2F_L}{k-1}$ • If  $F_s$  obeys the above for integer k > 1, aliasing can be avoided. • The k = 1 case corresponds to the Nyquist sampling criterion. • i.e.,  $F_s \geq 2F_H$ • The case k > 1 corresponds to sampling below Nyquist.

• i.e.,  $F_s < 2F_I/(k-1) < 2F_H$  for all k > 1.

Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals

#### Arbitrary Band Positioning

• The maximum value of k shows the number of bands that we can fit in the range from 0 and  $F_{H}$ .

$$(k-1)F_{s} \leq 2F_{L} \Longrightarrow \qquad (k-1)F_{s} \leq 2F_{H} - 2B$$

$$2F_{H} \leq kF_{s} \Longrightarrow \qquad \frac{1}{F_{s}} \leq \frac{k}{2F_{H}}$$
mult. both sides  $\Longrightarrow \qquad k-1 \leq k-\frac{kB}{F_{H}}$ 

$$k \leq \frac{F_{H}}{B}$$

$$k_{max} = \left\lfloor \frac{F_{H}}{B} \right\rfloor$$
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#### Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals

Arbitrary Band Positioning

 Therefore to avoid aliasing, the range of acceptable uniform sampling rates is given by

$$\begin{array}{ll} \frac{2F_{H}}{k} \leq F_{s} & \leq & \frac{2F_{L}}{k-1} \\ & \text{where } k \in \mathbb{Z}^{+} \text{ and } 1 \leq k \leq \left\lfloor \frac{F_{H}}{B} \right. \end{array}$$

See Figure 6.4.3 of text

Chapter 6: Sampling and Reconstruction of Signals 6.4 Sampling & Reconstruction of Cts-Time Bandlimited Signals

# Arbitrary Band Positioning: Minimum Sampling Rate

Recall to avoid aliasing,

$$\frac{2F_H}{k} \le F_s \le \frac{2F_L}{k-1}$$

 Therefore, the minimum sampling rate to avoid aliasing is given by

$$F_{s,\min} = \frac{2F_H}{k_{\max}}$$

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#### Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals

### Sampling of Dst-Time Signals

- Consider  $x_d(n) = x(nD)$  for all n
  - x(n) can be interpreted as the samples of a continuous-time signal x<sub>a</sub>(t) with rate F<sub>s</sub> = <sup>1</sup>/<sub>T</sub>:

$$X(F) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_a(F - kF_s)$$

•  $x_d(n)$  can be interpreted as the samples of  $x_a(t)$  with sampling rate  $\frac{F_s}{D} = \frac{1}{DT}$ ,

$$X_d(F) = \frac{1}{DT} \sum_{k=-\infty}^{\infty} X_a(F - k \frac{F_s}{D})$$

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Sampling of Dst-Time Signals

- In cts-time sampling  $x(n) = x_a(nT)$ ,
  - ► the aperiodic spectrum X<sub>a</sub>(F) is repeated an infinite number of times to create a periodic spectrum covering the infinite frequency range.
- In dst-time sampling  $x_d(n) = x(nD)$ ,
  - the periodic spectrum X(F) is repeated D times covering one period of the periodic frequency domain.

Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals

## Sampling of Dst-Time Signals

Therefore,

$$X_d(F) = \frac{1}{D} \sum_{k=0}^{D-1} X\left(F - k \frac{F_s}{D}\right)$$

► Assuming X<sub>a</sub>(F) = 0, |F| > B, to avoid aliasing for the dst-time sampling, we need:

$$\frac{F_s}{D} \ge 2B \quad \text{or} \quad B \le \frac{F_s}{2D}$$
$$f_{max} \triangleq \frac{B}{F_s} \le \frac{1}{2D} = \frac{f_s}{2} \quad \text{or} \quad \omega_{max} = 2\pi f_{max} \le \frac{\pi}{D} = \frac{\omega_s}{2}$$

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#### Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals

#### Ideal Interpolation

▶ Recall, the general interpolation formula is:

$$x_a(t) = \sum_{n=-\infty}^{\infty} x(n) \frac{\sin(\frac{\pi}{T}(t-nT))}{\frac{\pi}{T}(t-nT)}$$

► For dst-time sampling, it is given by:

$$x_a(t) = \sum_{m=-\infty}^{\infty} x_d(m) \frac{\sin(\frac{\pi}{DT}(t - mDT))}{\frac{\pi}{DT}(t - mDT)}$$

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See Figure 6.5.1 of text

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#### Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals

## Ideal Interpolation

Using

$$x_a(t) = \sum_{m=-\infty}^{\infty} x_d(m) \frac{\sin(\frac{\pi}{DT}(t - mDT))}{\frac{\pi}{DT}(t - mDT)}$$

• and the fact that  $x(n) = x_a(nT)$ 

$$\begin{aligned} x(n) &= x_a(nT) = \sum_{m=-\infty}^{\infty} x_d(m) \frac{\sin(\frac{\pi}{DT}(nT - mDT))}{\frac{\pi}{DT}(nT - mDT)} \\ &= \sum_{m=-\infty}^{\infty} x_d(m) \frac{\sin(\frac{\pi}{D}(n - mD))}{\frac{\pi}{D}(n - mD)} \\ &= \sum_{m=-\infty}^{\infty} x_d(m) g_{BL}(n - mD) \end{aligned}$$
  
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Ideal Interpolation

► where

$$g_{BL}(n) = D \frac{\sin\left(\frac{\pi}{D}n\right)}{\pi n} \stackrel{\mathcal{F}}{\longleftrightarrow} G_{BL}(\omega) = \begin{cases} D & |\omega| \leq \frac{\pi}{D} \\ 0 & \frac{\pi}{D} < |\omega| \leq \pi \end{cases}$$

See Figure 6.5.1 of text

Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals	
Ideal Interpolation	
► Therefore,	
$x(n) = \sum_{m=-\infty}^{\infty} x_d(m)g_{BL}(n-mD)$	
$g_{BL}(n) = D \frac{\sin\left(\frac{\pi}{D}n\right)}{\pi n}$	
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Linear Interpolation  
See Figure 6.5.2 of text.  

$$x_{lin}(t) = x(m-1) + \frac{x(m) - x(m-1)}{DT}(t - (m-1)DT)$$
for  $(m-1)DT \le t \le mDT$ 





Chapter 6: Sampling and Reconstruction of Signals 6.5 Sampling of Dst-Time Signals Ideal Interpolation where  $g_{lin}(n) = \left\{ egin{array}{cc} 1 - rac{|n|}{D} & |n| \leq D \ 0 & |n| > D \end{array} egin{array}{cc} \mathcal{F} \ \mathcal{G}_{lin}(\omega) = rac{1}{D} \left[ rac{\sin(\omega rac{D}{2})}{\sin(rac{\omega}{2})} 
ight]^2 \end{array} 
ight.$ See Figure 6.5.3 of text 46 / 46 Dr. Deepa Kundur (University of Toronto) Efficient Computation of the DFT: FFT Algorithms

