

## ESE 531: Digital Signal Processing

Lec 9: February 9th, 2017  
CT Processing, Downsampling/Upsampling



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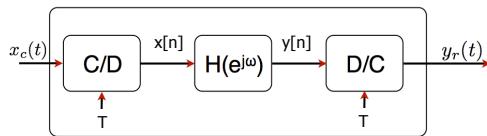
## Lecture Outline

- DT processing of CT signals
  - Impulse Invariance
- CT processing of DT signals (why??)
- Downsampling
- Upsampling

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### Discrete-Time Processing of Continuous Time



- If  $x_c(t)$  is bandlimited by  $\pi/T$ 
  - I.e.  $X_c(j\Omega) = 0$  for  $|\Omega| > \pi/T$
- then,

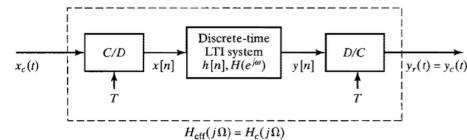
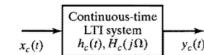
$$\frac{Y_r(j\Omega)}{X_c(j\Omega)} = H_{\text{eff}}(j\Omega) = \begin{cases} H(e^{j\omega}) & |\omega| < \pi/T \\ 0 & \text{else} \end{cases}$$

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### Impulse Invariance

- Want to implement continuous-time system in discrete-time



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### Impulse Invariance

- With  $H_c(j\Omega)$  bandlimited, choose

$$H(e^{j\omega}) = H_c(j\frac{\omega}{T}), \quad |\omega| < \pi$$

- With the further requirement that T be chosen such that

$$H_c(j\Omega) = 0, \quad |\Omega| \geq \pi/T$$

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### Impulse Invariance

- With  $H_c(j\Omega)$  bandlimited, choose

$$H(e^{j\omega}) = H_c(j\frac{\omega}{T}), \quad |\omega| < \pi$$

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$$h[n] = Th_c(nT)$$

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## Impulse Invariance

- Let,

$$h[n] = h_c(nT)$$

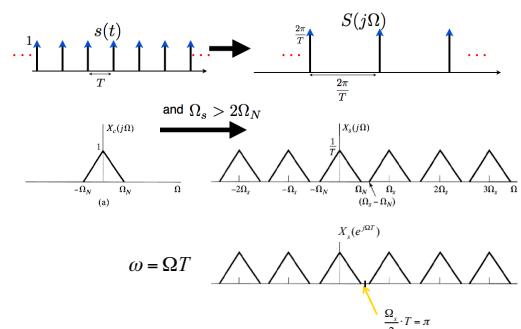
- If sampling at Nyquist Rate then

$$H(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} H_c \left[ j \left( \frac{\omega}{T} - \frac{2\pi k}{T} \right) \right]$$

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## Frequency Domain Analysis



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## Impulse Invariance

- Let,

$$h[n] = h_c(nT) \quad H_c(j\Omega) = 0, \quad |\Omega| \geq \pi/T$$

- If sampling at Nyquist Rate then

$$H(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} H_c \left[ j \left( \frac{\omega}{T} - \frac{2\pi k}{T} \right) \right]$$

$$H(e^{j\omega}) = \frac{1}{T} H_c \left( j \frac{\omega}{T} \right), \quad |\omega| < \pi$$

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## Impulse Invariance

- Let,

$$h[n] = Th_c(nT) \quad H_c(j\Omega) = 0, \quad |\Omega| \geq \pi/T$$

- If sampling at Nyquist Rate then

$$H(e^{j\omega}) = \frac{T}{T} \sum_{k=-\infty}^{\infty} H_c \left[ j \left( \frac{\omega}{T} - \frac{2\pi k}{T} \right) \right]$$

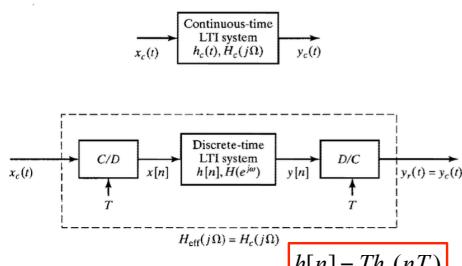
$$H(e^{j\omega}) = T H_c \left( j \frac{\omega}{T} \right), \quad |\omega| < \pi$$

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## Impulse Invariance

- Want to implement continuous-time system in discrete-time

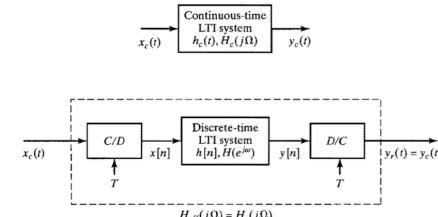


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## Example: DT Lowpass Filter

- We wish to implement a lowpass filter with cutoff frequency  $\Omega_c$  on continuous time signal in discrete time with the following system

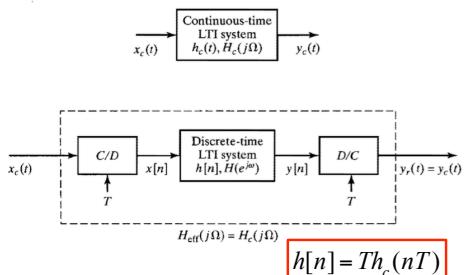


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## Impulse Invariance

- Want to implement continuous-time system in discrete-time



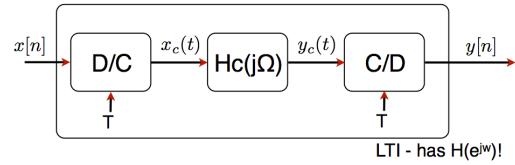
$$h[n] = Th_c(nT)$$

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## Continuous-Time Processing of Discrete-Time

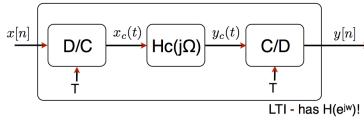
- Useful to interpret DT systems with no simple interpretation in discrete time



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## Continuous-Time Processing of Discrete-Time

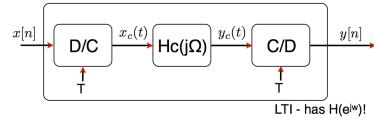


$$X_c(j\Omega) = \begin{cases} TX(e^{j\Omega T}) & |\Omega| < \pi / T \\ 0 & \text{else} \end{cases}$$

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## Continuous-Time Processing of Discrete-Time



$$X_c(j\Omega) = \begin{cases} TX(e^{j\Omega T}) & |\Omega| < \pi / T \\ 0 & \text{else} \end{cases}$$

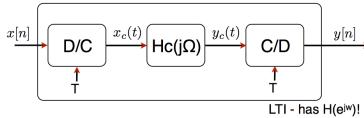
Also bandlimited

$$\rightarrow Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$$

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## Continuous-Time Processing of Discrete-Time



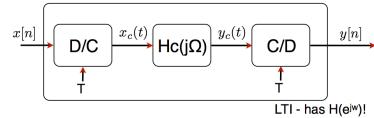
$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$$

$$Y(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} Y_c[j(\Omega - k\Omega_s)] \Big|_{\Omega=\omega/T} = \frac{1}{T} Y_c(j\Omega) \Big|_{\Omega=\omega/T}$$

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## Continuous-Time Processing of Discrete-Time



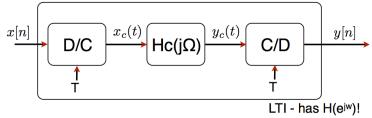
$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$$

$$Y(e^{j\omega}) = \frac{1}{T} Y_c(j\Omega) \Big|_{\Omega=\omega/T}$$

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### Continuous-Time Processing of Discrete-Time



$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega) \quad Y(e^{j\omega}) = \frac{1}{T}Y_c(j\Omega) \Big|_{\Omega=\omega/T}$$

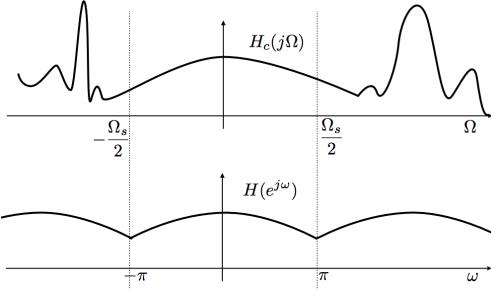
$$\begin{aligned} Y(e^{j\omega}) &= \frac{1}{T}H_c(j\Omega) \Big|_{\Omega=\omega/T} X_c(j\Omega) \Big|_{\Omega=\omega/T} \\ &= \frac{1}{T}H_c(j\Omega) \Big|_{\Omega=\omega/T} (TX(e^{j\omega})) \quad |\omega| < \pi \end{aligned}$$

$$H(e^{j\omega})$$

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### Example



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### Example: Non-integer Delay

- What is the time domain operation when  $\Delta$  is non-integer? I.e  $\Delta=1/2$

$$H(e^{j\omega}) = e^{-j\omega\Delta}$$

$\delta[n] \leftrightarrow 1$   
 $\delta[n-n_d] \leftrightarrow e^{-jn\omega_d}$

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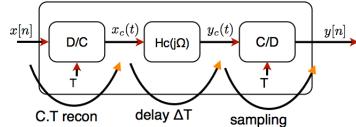
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### Example: Non-integer Delay

- What is the time domain operation when  $\Delta$  is non-integer? I.e  $\Delta=1/2$

$$H(e^{j\omega}) = e^{-j\omega\Delta}$$

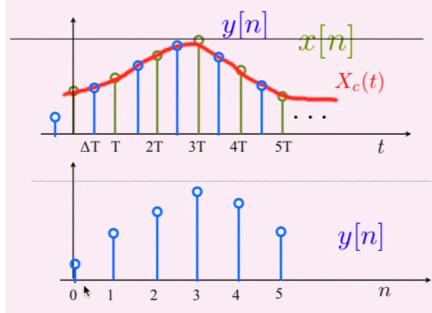
Let:  $H_c(j\Omega) = e^{-j\Omega\Delta T}$  delay of  $\Delta T$  in time



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### Example: Non-integer Delay

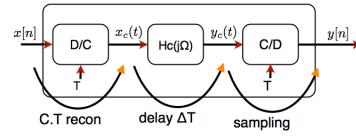


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### Example: Non-integer Delay

- The block diagram is for interpretation/analysis only



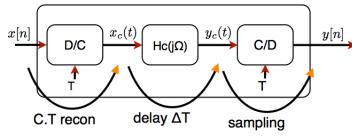
$$y_c(t) = x_c(t - T\Delta)$$

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### Example: Non-integer Delay

- The block diagram is for interpretation/analysis only



$$y_c(t) = x_c(t - T\Delta)$$

$$\begin{aligned} y[n] &= y_c(nT) = x_c(nt - T\Delta) \\ &= \sum_k x[k] \text{sinc}\left(\frac{t - kT - T\Delta}{T}\right) \Big|_{t=nT} \\ &= \sum_k x[k] \text{sinc}(n - k - \Delta) \end{aligned}$$

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### Example: Non-integer Delay

- Delay system has an impulse response of a sinc with a continuous time delay

$$h[n] = \text{sinc}(n - \Delta)$$

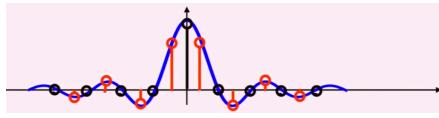
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### Example: Non-integer Delay

- Delay system has an impulse response of a sinc with a continuous time delay

$$h[n] = \text{sinc}(n - \Delta)$$



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### Downsampling

- Similar to C/D conversion
  - Need to worry about aliasing
  - Use anti-aliasing filter to mitigate effects

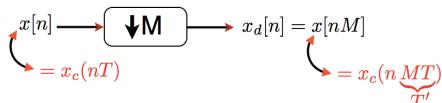
- If your discrete time signal is finely sample almost like a CT signal
  - Downsampling is just like sampling (C/D conversion)

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### Downsampling

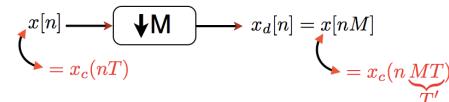
- Definition: Reducing the sampling rate by an integer number



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### Downsampling



The DTFT:

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

$$X_d(e^{j\omega}) = \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right)$$

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## Downsampling

The DTFT:

$$X(e^{j\omega}) = \frac{1}{T} \sum_k X_c \left( j \left( \underbrace{\frac{\omega}{T}}_{\Omega} - \underbrace{\frac{2\pi}{T} k}_{\Omega_s} \right) \right)$$

$$X_d(e^{j\omega}) = \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right)$$

- ❑ Want to relate  $X_d(e^{j\omega})$  to  $X(e^{j\omega})$  not  $X_c(j\Omega)$
- ❑ Separate sum into two sums—fine sum and coarse sum (i.e like counting minutes within hours)

## Downsampling

The DTFT:

$$X(e^{j\omega}) = \frac{1}{T} \sum_k X_c \left( j \left( \underbrace{\frac{\omega}{T}}_{\Omega} - \underbrace{\frac{2\pi}{T} k}_{\Omega_s} \right) \right)$$

$$X_d(e^{j\omega}) = \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right)$$

- ❑  $k = rM + i$ 
  - $i = 0, 1, \dots, M-1$
  - $r = -\infty, \dots, \infty$

## Downsampling

$$\begin{aligned} X_d(e^{j\omega}) &= \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \frac{1}{T} \sum_{r=-\infty}^{\infty} X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} i - \frac{2\pi}{T} r \right) \right) \end{aligned}$$

## Downsampling

$$\begin{aligned} X_d(e^{j\omega}) &= \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \frac{1}{T} \sum_{r=-\infty}^{\infty} X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} i - \frac{2\pi}{T} r \right) \right) \\ X(e^{j\omega}) &= \frac{1}{T} \sum_k X_c \left( j \left( \underbrace{\frac{\omega}{T}}_{\Omega} - \underbrace{\frac{2\pi}{T} k}_{\Omega_s} \right) \right) \quad X(e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)}) \end{aligned}$$

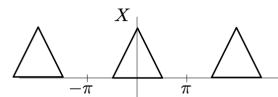
## Downsampling

$$\begin{aligned} X_d(e^{j\omega}) &= \frac{1}{MT} \sum_k X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \frac{1}{T} \sum_{r=-\infty}^{\infty} X_c \left( j \left( \frac{\omega}{MT} - \frac{2\pi}{MT} i - \frac{2\pi}{T} r \right) \right) \\ X(e^{j\omega}) &= \frac{1}{T} \sum_k X_c \left( j \left( \underbrace{\frac{\omega}{T}}_{\Omega} - \underbrace{\frac{2\pi}{T} k}_{\Omega_s} \right) \right) \quad X(e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)}) \\ X_d(e^{j\omega}) &= \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right) \end{aligned}$$

stretch by M      replicate

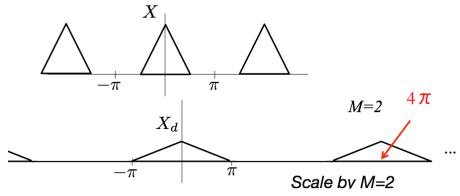
## Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$



### Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$

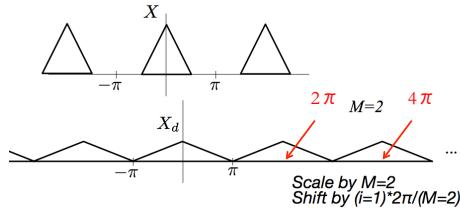


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### Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$

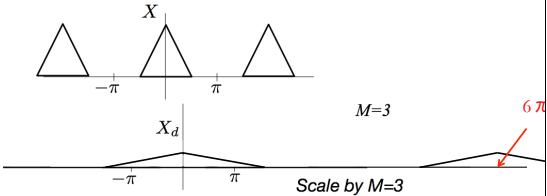


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### Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$

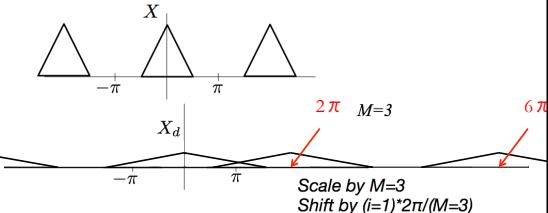


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### Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$

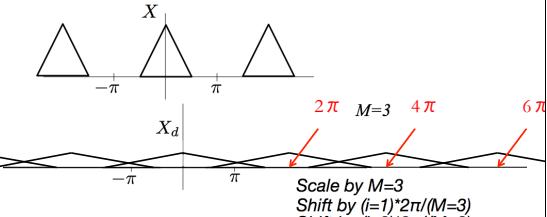


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### Example

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X \left( e^{j(\frac{\omega}{M} - \frac{2\pi}{M} i)} \right)$$

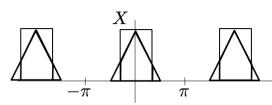


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### Example

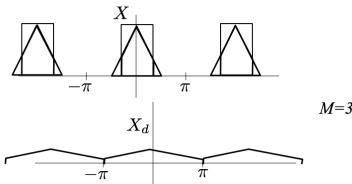
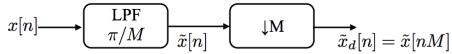
$$x[n] \rightarrow \text{LPF } \frac{\pi}{M} \tilde{x}[n] \rightarrow \downarrow M \rightarrow \tilde{x}_d[n] = \tilde{x}[nM]$$



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### Example



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### Upsampling

- Much like D/C converter
- Upsample by A LOT  $\rightarrow$  almost continuous

#### Intuition:

- Recall our D/C model:  $x[n] \rightarrow x_s(t) \rightarrow x_c(t)$
- Approximate " $x_s(t)$ " by placing zeros between samples
- Convolve with a sinc to obtain " $x_c(t)$ "

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### Upsampling

- Definition: Increasing the sampling rate by an integer number

$$x[n] = x_c(nT)$$

$$x_i[n] = x_c(nT') \text{ where } T' = \frac{T}{L} \quad L \text{ integer}$$

Obtain  $x_i[n]$  from  $x[n]$  in two steps:

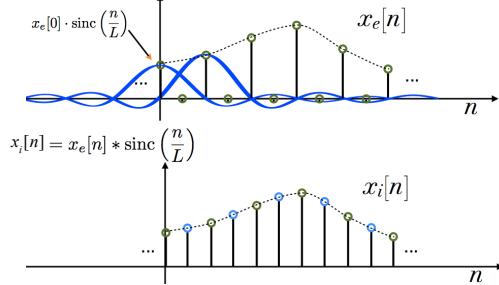
$$(1) \text{ Generate: } x_e[n] = \begin{cases} x[n/L] & n = 0, \pm L, \pm 2L, \dots \\ 0 & \text{otherwise} \end{cases}$$

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### Upsampling

(2) Obtain  $x_i[n]$  from  $x_e[n]$  by bandlimited interpolation:



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### Upsampling

$$x_i[n] = x_e[n] * \text{sinc}(n/L)$$

$$x_e[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n - kL]$$

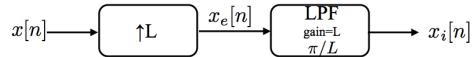
$$x_i[n] = \sum_{k=-\infty}^{\infty} x[k] \text{sinc}\left(\frac{n - kL}{L}\right)$$

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### Frequency Domain Interpretation

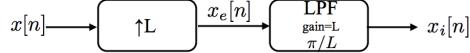
$$x_i[n] = x_e[n] * \text{sinc}(n/L)$$



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### Frequency Domain Interpretation

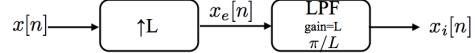


$$X_e(e^{j\omega}) = \sum_{n=-\infty}^{\infty} \underbrace{x_e[n]}_{\neq 0 \text{ only for } n=mL \text{ (integer } m)} e^{-j\omega n}$$

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### Frequency Domain Interpretation



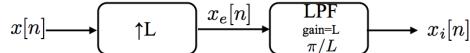
$$\begin{aligned} X_e(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} \underbrace{x_e[n]}_{\neq 0 \text{ only for } n=mL \text{ (integer } m)} e^{-j\omega n} \\ &= \sum_{m=-\infty}^{\infty} \underbrace{x_e[mL]}_{=x[m]} e^{-j\omega mL} \end{aligned}$$

Compress DTFT by a factor of L!

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### Frequency Domain Interpretation



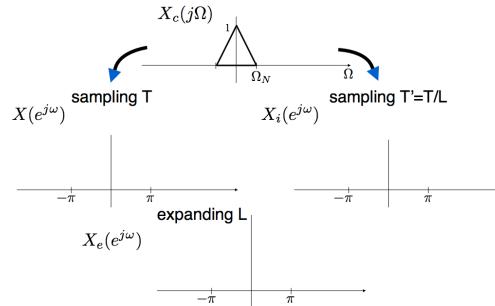
$$\begin{aligned} X_e(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} \underbrace{x_e[n]}_{\neq 0 \text{ only for } n=mL \text{ (integer } m)} e^{-j\omega n} \\ &= \sum_{m=-\infty}^{\infty} \underbrace{x_e[mL]}_{=x[m]} e^{-j\omega mL} = X(e^{j\omega L}) \end{aligned}$$

Compress DTFT by a factor of L!

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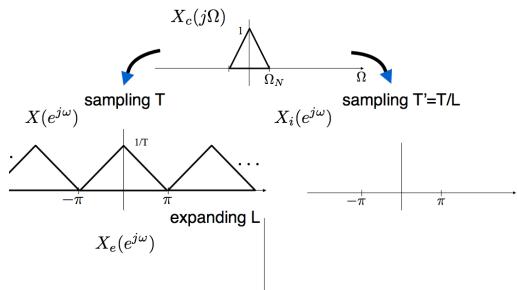
### Example



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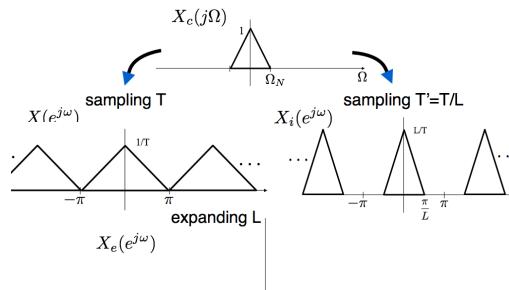
### Example



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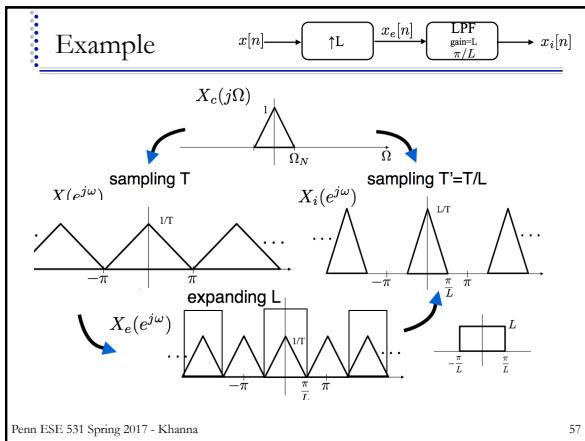
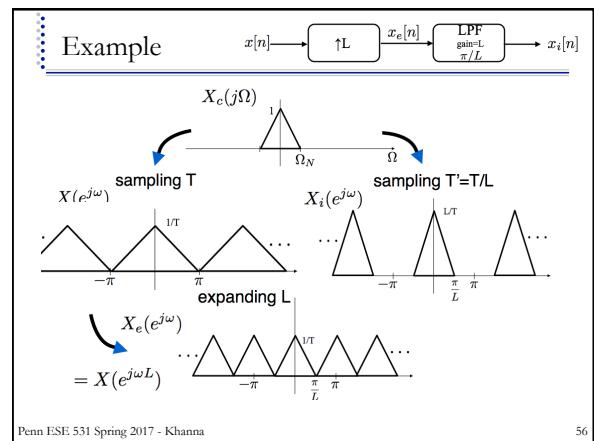
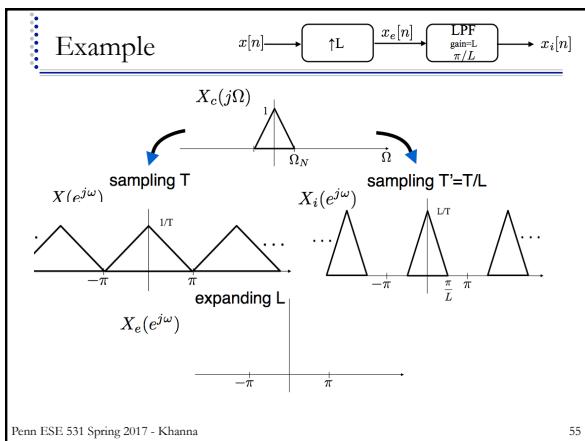
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### Example



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- Big Ideas**
- DT processing of CT signals
    - Impulse Invariance to design DT systems for CT signals
  - CT processing of DT signals
    - Allows for interpretation of DT systems
  - Downsampling
    - Like a C/D converter
  - Upsampling
    - Like a D/C converter
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- Admin**
- HW 3 due Friday
  - HW 4 posted after class
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