

ESE 531: Digital Signal Processing

Lec 8: February 12th, 2019
Sampling and Reconstruction



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Lecture Outline

- Review
 - Ideal sampling
 - Frequency response of sampled signal
 - Reconstruction
 - Anti-aliasing filtering
- DT processing of CT signals
 - Impulse Invariance
- CT processing of DT signals (why??)

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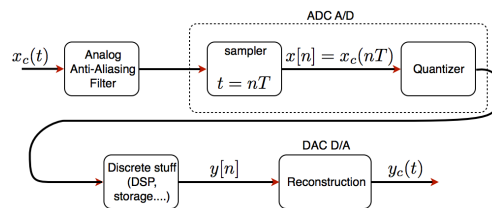
Last Time...

Sampling, Frequency Response of Sampled
Signal, Reconstruction, Anti-aliasing filtering



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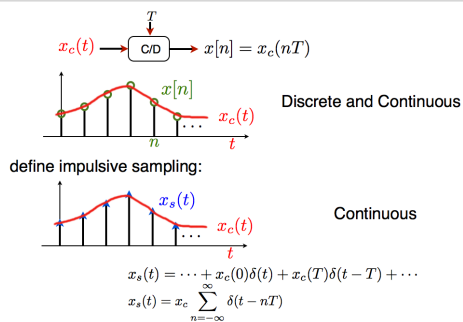
DSP System



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Ideal Sampling Model



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Frequency Domain Analysis

- How is $x[n]$ related to $x_s(t)$ in frequency domain?

$$x[n] = x_c(nT) \quad x_s(t) = x_c \sum_{n=-\infty}^{\infty} \delta(t - nT)$$

$$x_s(t) : \text{C.T.} \quad X_s(j\Omega) = \sum_n x_c(nT) e^{-j\Omega nT}$$

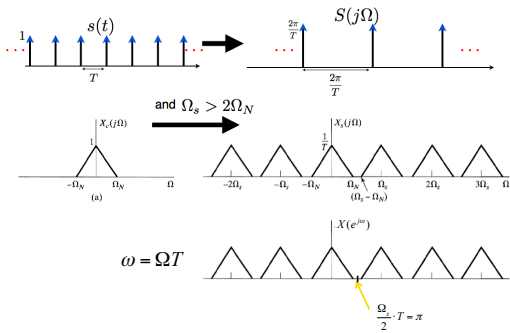
$$x[n] : \text{D.T.} \quad X(e^{j\omega}) = \sum_n x[n] e^{-j\omega n} \quad \omega = \Omega T$$

$$X(e^{j\omega}) = X_s(j\Omega)|_{\Omega=\omega/T} \quad X_s(j\Omega) = X(e^{j\omega})|_{\omega=\Omega T}$$

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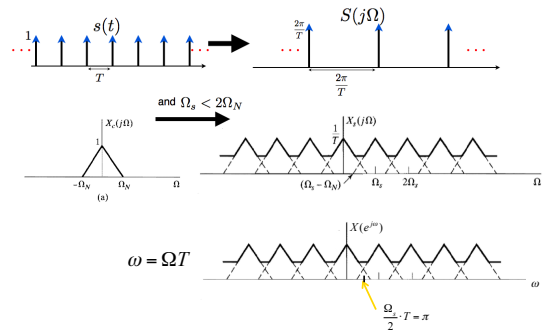
Frequency Domain Analysis



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Frequency Domain Analysis



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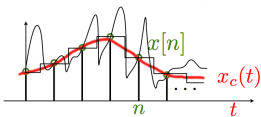
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Reconstruction of Bandlimited Signals

- Nyquist Sampling Theorem: Suppose $x_c(t)$ is bandlimited. I.e.

$$X_c(j\Omega) = 0 \quad \forall \quad |\Omega| \geq \Omega_N$$

- If $\Omega_s \geq 2\Omega_N$, then $x_c(t)$ can be uniquely determined from its samples $x[n] = x_c(nT)$
- Bandlimitedness is the key to uniqueness

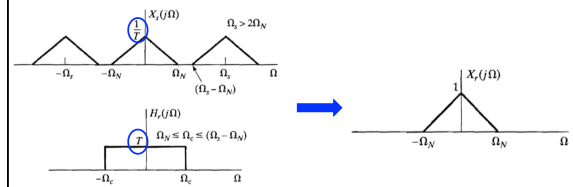
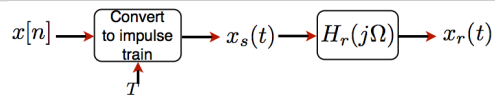


Multiple signals go through the samples, but only one is bandlimited within our sampling band

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Reconstruction in Frequency Domain

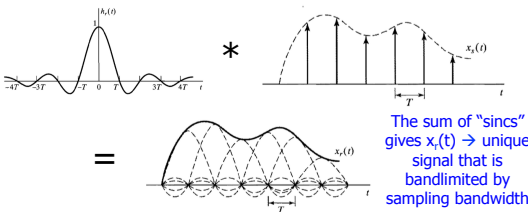


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Reconstruction in Time Domain

$$\begin{aligned} x_r(t) &= x_s(t) * h_r(t) = \left(\sum_n x[n] \delta(t - nT) \right) * h_r(t) \\ &= \sum_n x[n] h_r(t - nT) \end{aligned}$$

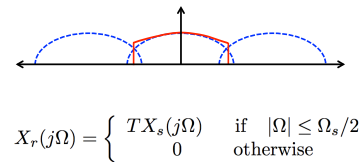


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Aliasing

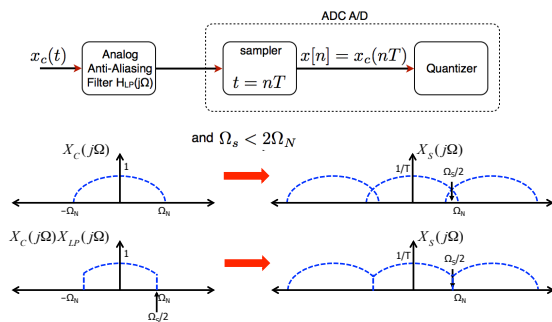
- If $\Omega_N > \Omega_s/2$, $x_r(t)$ is an aliased version of $x_c(t)$



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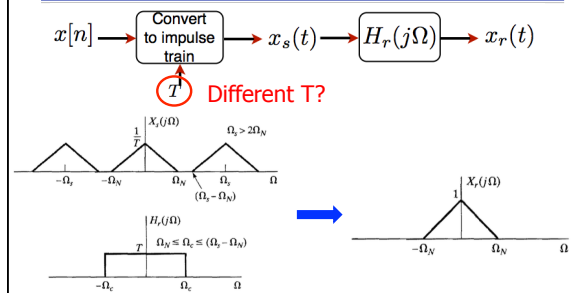
Anti-Aliasing Filter



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Reconstruction in Frequency Domain

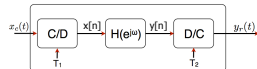


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Signal Processing

- Use theory of sampling (C/D) and reconstruction (D/C) to implement signal processing
- Two cases:
 - Discrete-time processing of continuous-time signals

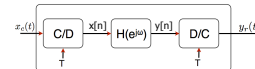


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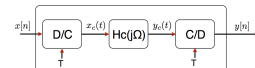
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Signal Processing

- Use theory of sampling (C/D) and reconstruction (D/C) to implement signal processing
- Two cases:
 - Discrete-time processing of continuous-time signals



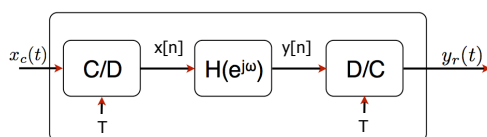
- Continuous-time processing of discrete-time signals



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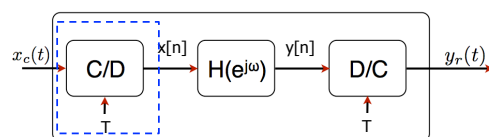
Discrete-Time Processing of Continuous Time



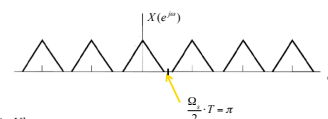
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Discrete-Time Processing of Continuous Time



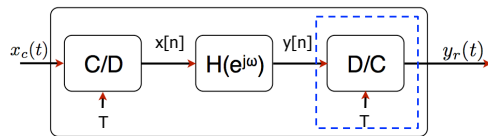
$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left[j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right]$$



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Discrete-Time Processing of Continuous Time



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

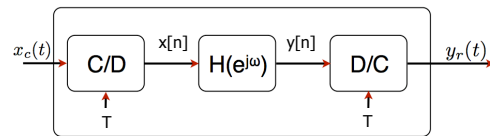
Sum of scaled shifted sincs

$$y_r(t) = \sum_{n=-\infty}^{\infty} y[n] \frac{\sin[\pi(t-nT)/T]}{\pi(t-nT)/T}$$

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Discrete-Time Processing of Continuous Time



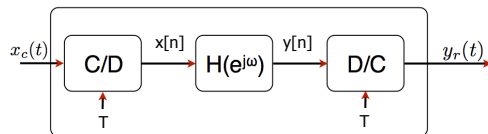
$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left[j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right] \quad y_r(t) = \sum_{n=-\infty}^{\infty} y[n] \frac{\sin[\pi(t-nT)/T]}{\pi(t-nT)/T}$$

- If $h[n]/H(e^{j\omega})$ is LTI
 - Is the whole system from $x_c(t) \rightarrow y_r(t)$ LTI?

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Discrete-Time Processing of Continuous Time



$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left[j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right] \quad y_r(t) = \sum_{n=-\infty}^{\infty} y[n] \frac{\sin[\pi(t-nT)/T]}{\pi(t-nT)/T}$$

- If $x_c(t)$ is bandlimited by $\Omega_s/T = \pi/T$, then,

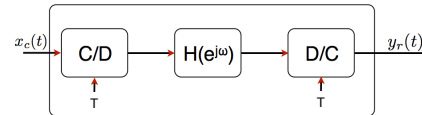
$$\frac{Y_r(j\Omega)}{X_c(j\Omega)} = H_{eff}(j\Omega) = \begin{cases} H(e^{j\omega})|_{\omega=\Omega T} & |\Omega| < \Omega_s/2 \\ 0 & \text{else} \end{cases}$$

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Example 1

- Consider the following system



- Where

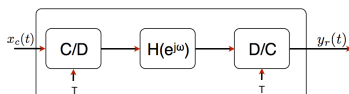
$$H(e^{j\omega}) = \begin{cases} 1 & |\omega| < \omega_c \\ 0 & \omega_c < |\omega| \leq \pi \end{cases}$$
- What is the effective frequency response of the system? What happens to a signal bandlimited by Ω_N ?

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Example 2

- DT implementation of an ideal CT bandlimited differentiator



- The ideal CT differentiator is defined by

$$y_c(t) = \frac{d}{dt}[x_c(t)]$$

- With corresponding

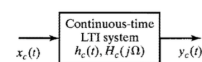
$$H_c(j\Omega) = j\Omega$$

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Impulse Invariance

- Want to implement continuous-time system...

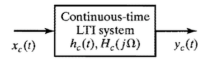


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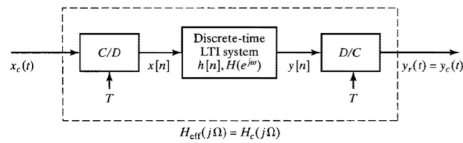
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Impulse Invariance

- Want to implement continuous-time system...



- ...in discrete-time



Impulse Invariance

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

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

- With the further requirement that T be chosen such that

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

Impulse Invariance

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

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- With the further requirement that T be chosen such that

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

Impulse Invariance

- Let,

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

Impulse Invariance

- Let,

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

- If sampling at Nyquist Rate then

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

Impulse Invariance

- Let,

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

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

- If sampling at Nyquist Rate then

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

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

Impulse Invariance

- Let,

$$h[n] \stackrel{?}{=} h_c(nT)$$

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

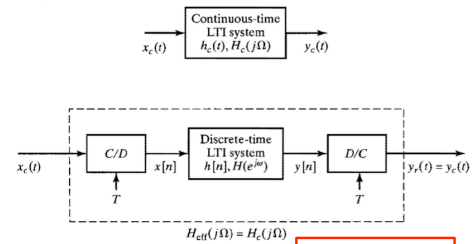
- If sampling at Nyquist Rate then

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

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

Impulse Invariance

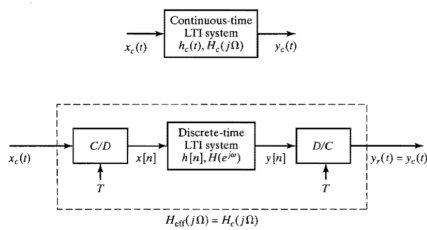
- Want to implement continuous-time system in discrete-time



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

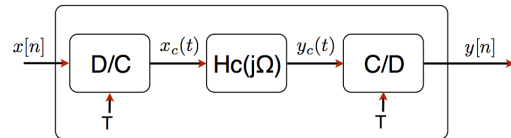
Example 3: DT Lowpass Filter

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



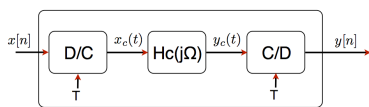
Continuous-Time Processing of Discrete-Time

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



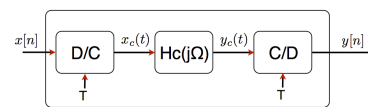
Is the effective $H(e^{j\omega})$ LTI?

Continuous-Time Processing of Discrete-Time



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

Continuous-Time Processing of Discrete-Time

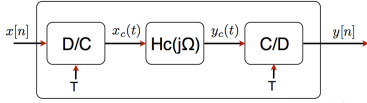


$$Y_c(j\Omega) = \begin{cases} TX(e^{j\Omega T}) & |\Omega| < \pi/T \\ 0 & \text{else} \end{cases}$$

Also bandlimited

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

Continuous-Time Processing of Discrete-Time



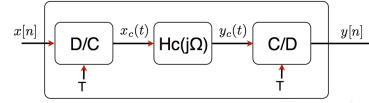
$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$$

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

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Continuous-Time Processing of Discrete-Time



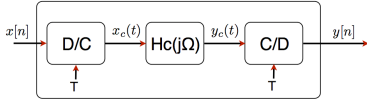
$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega)$$

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Continuous-Time Processing of Discrete-Time

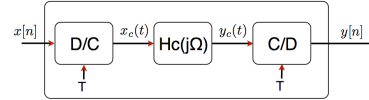


$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega) \quad Y(e^{j\omega}) = \frac{1}{T} Y_c(j\Omega) \Big|_{\Omega=\omega/T}$$

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Continuous-Time Processing of Discrete-Time



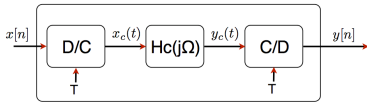
$$Y_c(j\Omega) = H_c(j\Omega)X_c(j\Omega) \quad Y(e^{j\omega}) = \frac{1}{T} Y_c(j\Omega) \Big|_{\Omega=\omega/T}$$

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

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Continuous-Time Processing of Discrete-Time



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

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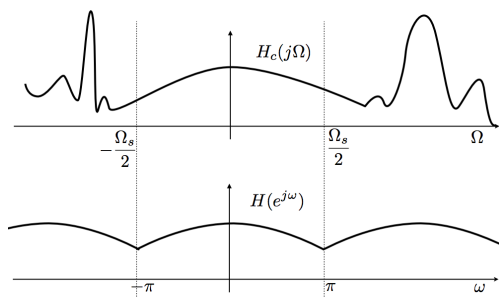
$$= \frac{1}{T} H_c(j\Omega) \Big|_{\Omega=\omega/T} (TX(e^{j\omega})) \quad |\omega| < \pi$$

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

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Example



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Example: Non-integer Delay

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

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

$$\delta[n] \leftrightarrow 1$$

$$\delta[n-n_d] \leftrightarrow e^{-j\omega n_d}$$

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Example: Non-integer Delay

- What is the time domain operation when Δ 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 ΔT in continuous time

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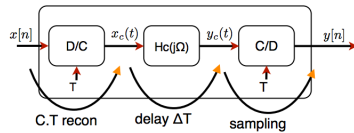
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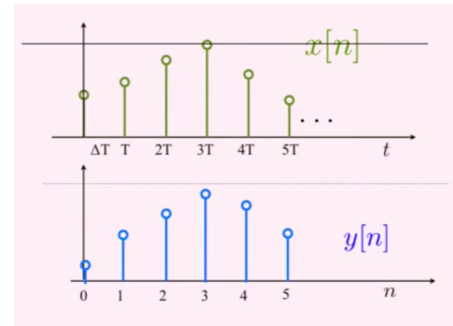
Let: $H_c(j\Omega) = e^{-j\Omega\Delta T}$ delay of ΔT in continuous time



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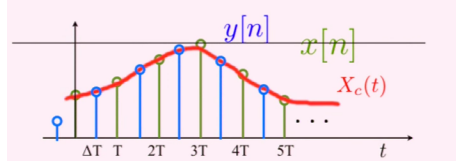
Example: Non-integer Delay



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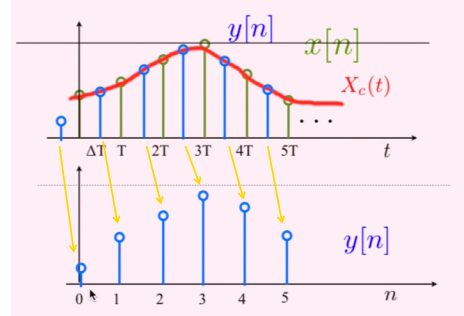
Example: Non-integer Delay



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Example: Non-integer Delay

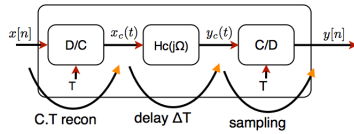


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Example: Non-integer Delay

- The block diagram is for interpretation/analysis only



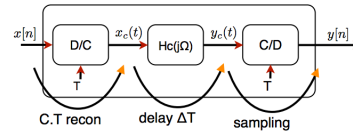
$$y_c(t) = x_c(t - T\Delta)$$

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Example: Non-integer Delay

- The block diagram is for interpretation/analysis only



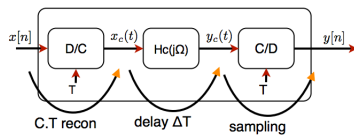
$$y_c(t) = x_c(t - T\Delta) \quad y[n] = y_c(nT) = x_c(nT - T\Delta)$$

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Example: Non-integer Delay

- The block diagram is for interpretation/analysis only



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

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Example: Non-integer Delay

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

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

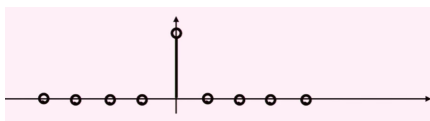
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- My delay system has an impulse response of a sinc with a continuous time delay

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



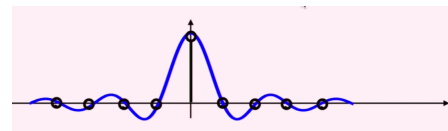
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Example: Non-integer Delay

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

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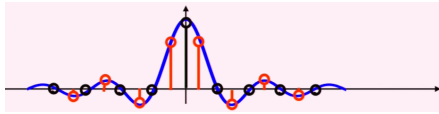
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Big Ideas

- Sampling and reconstruction
 - Rely on bandlimitedness for unique reconstruction
- CT processing of DT
 - Effectively LTI if no aliasing
- DT processing of CT
 - Always LTI
 - Useful for interpretation
- Changing the sampling rates next time
 - Upsampling, downsampling

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- HW 3 due Sunday

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