

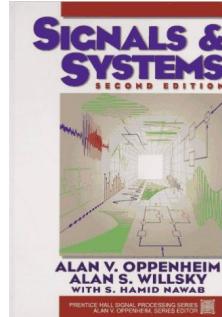
ESE 531: Digital Signal Processing

Lec 10: February 15th, 2018
 Practical and Non-integer Sampling, Multi-rate Sampling



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Signals and Systems Review



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Lecture Outline

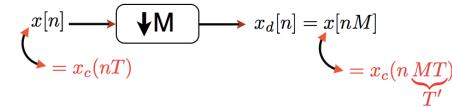
- ❑ Review: Downsampling/Upsampling
- ❑ Non-integer Resampling
- ❑ Multi-Rate Processing
 - Interchanging Operations
- ❑ Polyphase Decomposition
- ❑ Multi-Rate Filter Banks

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Downsampling

- ❑ Definition: Reducing the sampling rate by an integer number



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

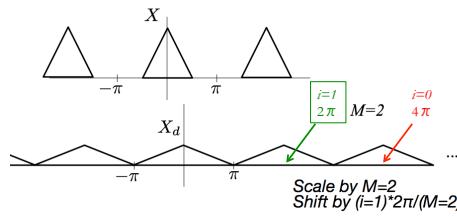
stretch by M replicate

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Example

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

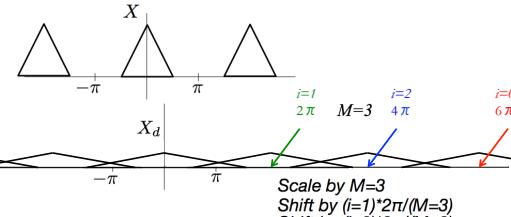


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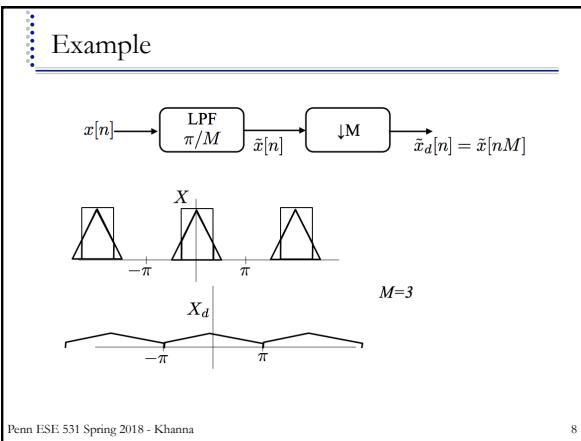
Example

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

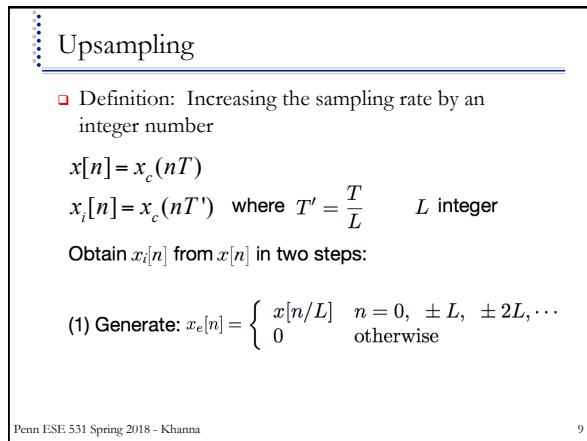


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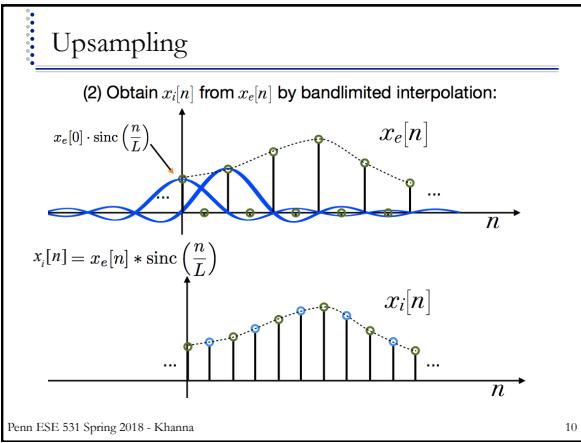
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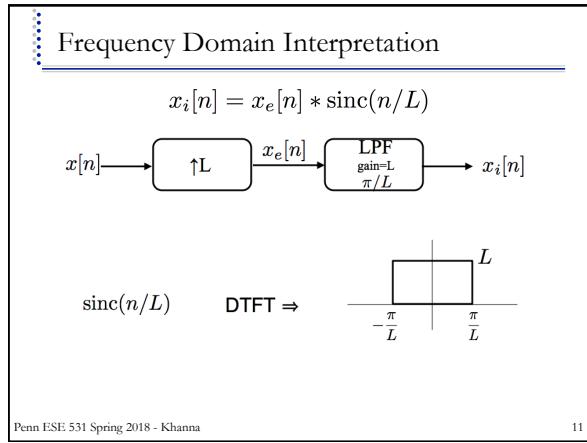
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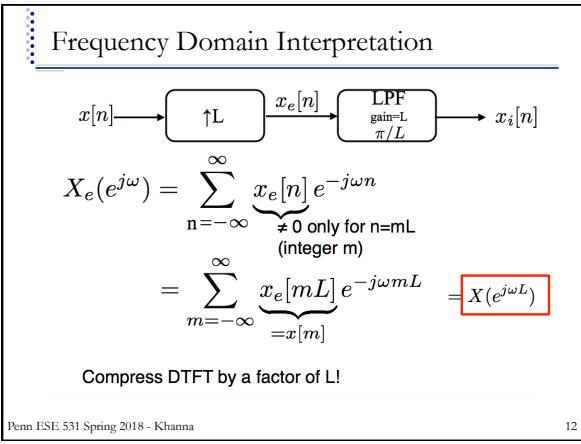
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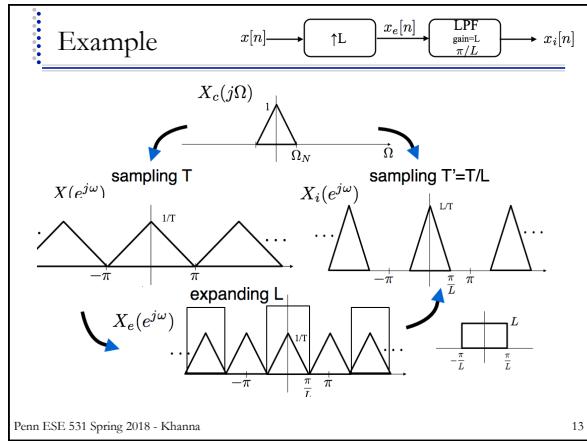
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Non-integer Resampling



Non-integer Resampling

◻ $T' = TM/L$

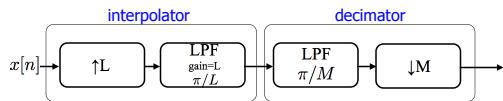
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Non-integer Resampling

◻ $T' = TM/L$

- Upsample by L, then downsample by M



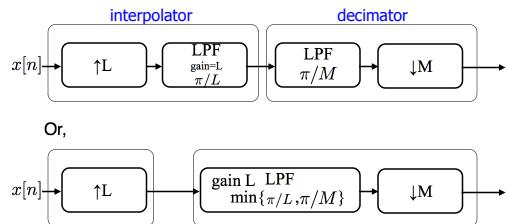
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Non-integer Resampling

◻ $T' = TM/L$

- Upsample by L, then downsample by M

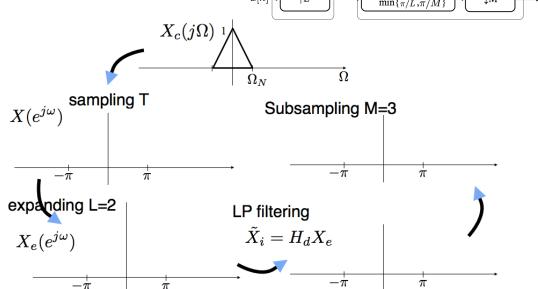


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Example

◻ $T' = 3/2T \rightarrow L=2, M=3$

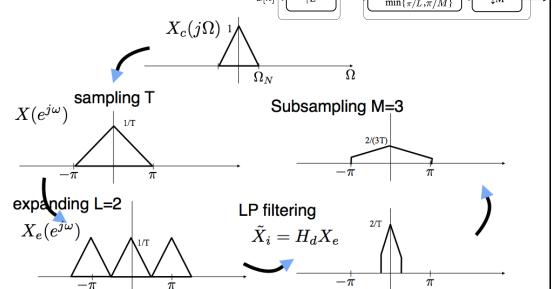


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Example

◻ $T' = 3/2T \rightarrow L=2, M=3$

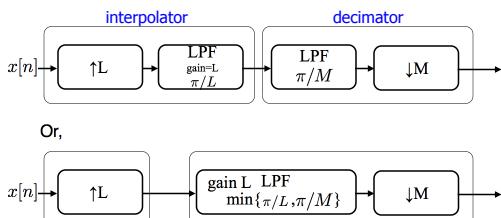


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Non-integer Sampling

- $T' = TM/L$
 - Downsample by M , then upsample by L ?



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Multi-Rate Signal Processing

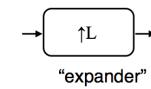
- What if we want to resample by 1.01T?
 - Expand by $L=100$
 - Filter $\pi/101$ (\$\$\$\$\$)
 - Downsample by $M=101$

- Fortunately there are ways around it!
 - Called multi-rate
 - Uses compressors, expanders and filtering

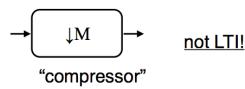
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Interchanging Operations



Upsampling
-expanding in time
-compressing in frequency



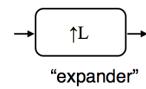
Downsampling
-compressing in time
-expanding in frequency

not LT!!

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Interchanging Operations - Expander

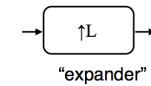


Upsampling
-expanding in time
-compressing in frequency

$$x[n] \rightarrow H(z) \rightarrow \uparrow L \rightarrow y[n] \quad ? \quad x[n] \rightarrow \uparrow L \rightarrow H(z) \rightarrow y[n]$$

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Interchanging Operations - Expander



Upsampling
-expanding in time
-compressing in frequency

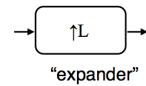
$$x[n] \rightarrow H(z) \rightarrow \uparrow L \rightarrow y[n] \quad ? \quad x[n] \rightarrow \uparrow L \rightarrow H(z) \rightarrow y[n]$$

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

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Interchanging Operations - Expander



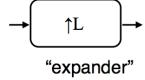
Upsampling
-expanding in time
-compressing in frequency

$$x[n] \rightarrow H(z) \rightarrow \uparrow L \rightarrow y[n] \quad \neq \quad x[n] \rightarrow \uparrow L \rightarrow H(z) \rightarrow y[n]$$

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

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Interchanging Operations - Expander



Upsampling
-expanding in time
-compressing in frequency

$$x[n] \rightarrow H(z) \rightarrow \begin{matrix} \uparrow L \\ \downarrow M \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} \uparrow L \\ X(e^{j\omega L}) \end{matrix} \rightarrow H(z^L) \rightarrow y[n]$$

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Interchanging Operations - Compressor



Downsampling
-compressing in time
-expanding in frequency

$$x[n] \rightarrow \begin{matrix} \downarrow M \\ H(z) \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} H(z^M) \\ \downarrow M \end{matrix} \rightarrow \tilde{y}[n]$$

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Interchanging Operations - Compressor

$$x[n] \rightarrow \begin{matrix} \downarrow M \\ H(z) \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} H(z^M) \\ \downarrow M \end{matrix} \rightarrow \tilde{y}[n]$$

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

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Interchanging Operations - Compressor

$$x[n] \rightarrow \begin{matrix} \downarrow M \\ H(z) \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} H(z^M) \\ \downarrow M \end{matrix} \rightarrow \tilde{y}[n]$$

$$\begin{aligned} Y(e^{j\omega}) &= H(e^{j\omega}) \left(\frac{1}{M} \sum_{i=0}^{M-1} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \underbrace{H \left(e^{j(\omega - 2\pi i)} \right)}_{H(e^{j\omega})} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \end{aligned}$$

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Interchanging Operations - Compressor

$$x[n] \rightarrow \begin{matrix} \downarrow M \\ H(z) \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} H(z^M) \\ \downarrow M \end{matrix} \rightarrow \tilde{y}[n]$$

$$\begin{aligned} Y(e^{j\omega}) &= H(e^{j\omega}) \left(\frac{1}{M} \sum_{i=0}^{M-1} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \underbrace{H \left(e^{j(\omega - 2\pi i)} \right)}_{H(e^{j\omega})} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} H \left(e^{jM(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \end{aligned}$$

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Interchanging Operations - Compressor

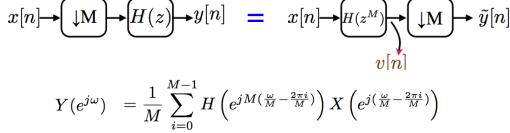
$$x[n] \rightarrow \begin{matrix} \downarrow M \\ H(z) \end{matrix} \rightarrow y[n] = x[n] \rightarrow \begin{matrix} H(z^M) \\ \downarrow M \end{matrix} \rightarrow \tilde{y}[n]$$

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

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Interchanging Operations - Compressor



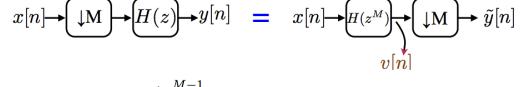
$$Y(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} H\left(e^{jM(\frac{\omega}{M} - \frac{2\pi i}{M})}\right) X\left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})}\right)$$

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

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Interchanging Operations - Compressor



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

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

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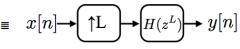
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Interchanging Operations - Summary

Filter and expander



Expander and expanded filter*



Compressor and filter



Expanded filter* and compressor

*Expanded filter = expanded impulse response, compressed freq response

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Multi-Rate Signal Processing

- What if we want to resample by 1.01T?
 - Expand by L=100
 - Filter $\pi/101$ (\$\$\$\$\$)
 - Downsample by M=101

- Fortunately there are ways around it!

- Called multi-rate
- Uses compressors, expanders and filtering

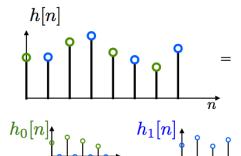
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Polyphase Decomposition

- We can decompose an impulse response (of our filter) to:

$$h[n] = \sum_{k=0}^{M-1} h_k[n-k]$$



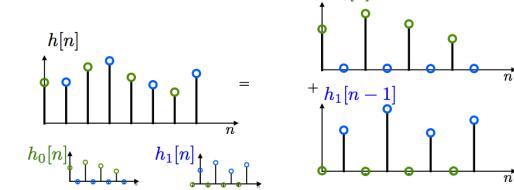
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Polyphase Decomposition

- We can decompose an impulse response (of our filter) to:

$$h[n] = \sum_{k=0}^{M-1} h_k[n-k]$$



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Polyphase Decomposition

$$h_k[n] \rightarrow \boxed{\downarrow M} \rightarrow e_k[n]$$

$$e_k[n] = h_k[nM]$$

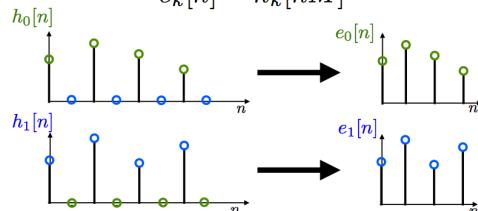
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Polyphase Decomposition

$$h_k[n] \rightarrow \boxed{\downarrow M} \rightarrow e_k[n]$$

$$e_k[n] = h_k[nM]$$



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Polyphase Decomposition

$$e_k[n] \rightarrow \boxed{\uparrow M} \rightarrow h_k[n]$$

recall upsampling \Rightarrow scaling

$$H_k(z) = E_k(z^M)$$

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Polyphase Decomposition

$$e_k[n] \rightarrow \boxed{\uparrow M} \rightarrow h_k[n]$$

recall upsampling \Rightarrow scaling

$$H_k(z) = E_k(z^M)$$

Also, recall:

$$h[n] = \sum_{k=0}^{M-1} h_k[n-k]$$

So,

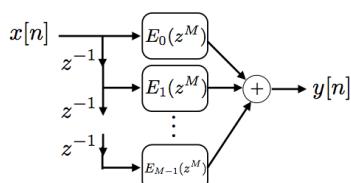
$$H(z) = \sum_{k=0}^{M-1} E_k(z^M)z^{-k}$$

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Polyphase Decomposition

$$H(z) = \sum_{k=0}^{M-1} E_k(z^M)z^{-k}$$



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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \boxed{\downarrow M} \rightarrow w[n] = y[nM]$$

□ Problem:

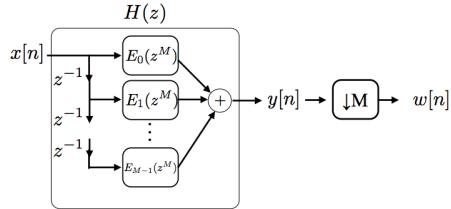
- Compute all $y[n]$ and then throw away -- wasted computation!
- For FIR length $N \rightarrow N$ mults/unit time

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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$

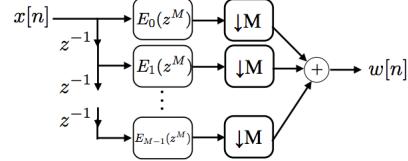


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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$



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Interchanging Operations - Summary

Filter and expander

$$x[n] \rightarrow H(z) \rightarrow \uparrow L \rightarrow y[n] \equiv x[n] \rightarrow \uparrow L \rightarrow H(z^L) \rightarrow y[n]$$

Compressor and filter

$$x[n] \rightarrow \downarrow M \rightarrow H(z) \rightarrow y[n] \equiv x[n] \rightarrow H(z^M) \rightarrow \downarrow M \rightarrow y[n]$$

Expander and expanded filter

$$x[n] \rightarrow \uparrow L \rightarrow H(z^L) \rightarrow y[n]$$

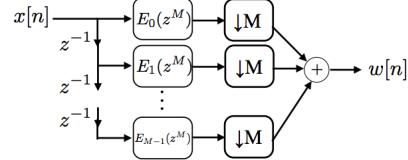
Expanded filter and compressor

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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$

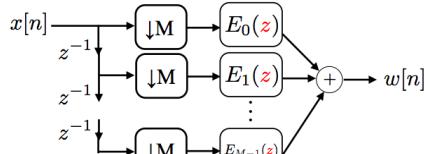


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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$

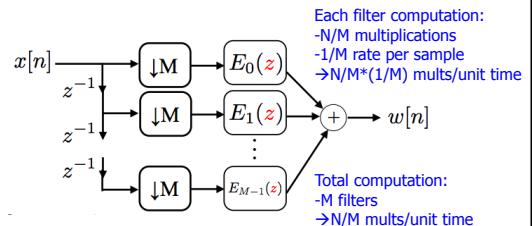


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Polyphase Implementation of Decimation

$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$



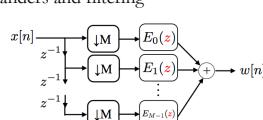
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Multi-Rate Signal Processing

- What if we want to resample by $1.01T$?
 - Expand by $L=100$
 - Filter $\pi/101$ (\$\$\$\$\$)
 - Downsample by $M=101$

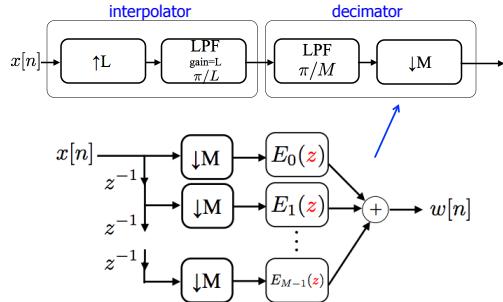
- Fortunately there are ways around it!
 - Called multi-rate
 - Uses compressors, expanders and filtering



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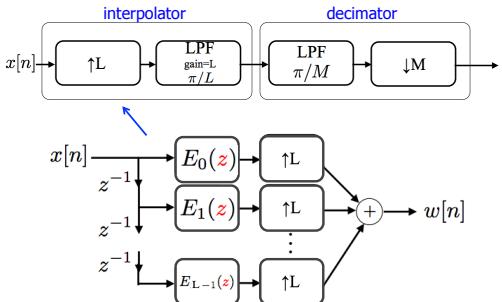
Polyphase Implementation of Decimator



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Polyphase Implementation of Interpolation



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Multi-Rate Filter Banks

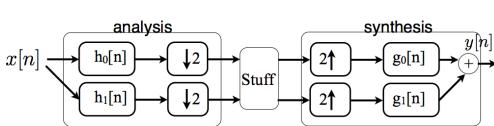
- Use filter banks to operate on a signal differently in different frequency bands
 - To save computation, reduce the rate after filtering

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Multi-Rate Filter Banks

- Use filter banks to operate on a signal differently in different frequency bands
 - To save computation, reduce the rate after filtering
- $h_0[n]$ is low-pass, $h_1[n]$ is high-pass
 - Often $h_1[n] = e^{j\pi n} h_0[n]$ ← shift freq resp by π

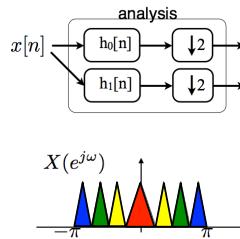


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

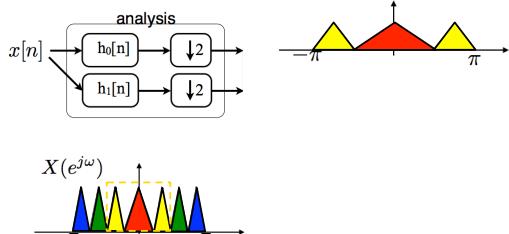


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

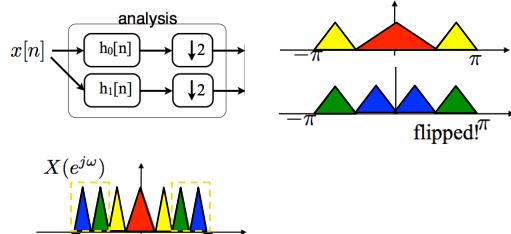


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

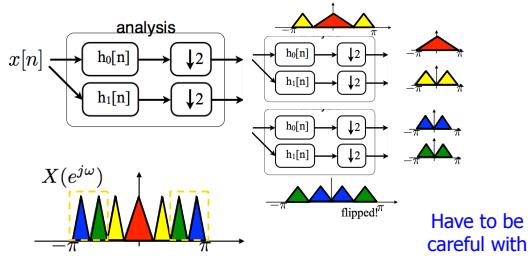


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

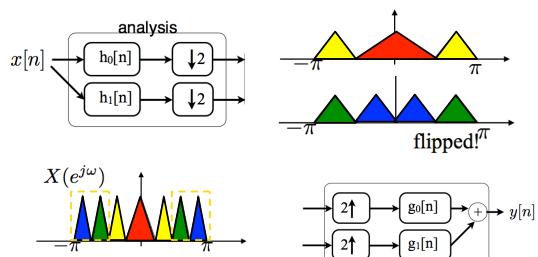


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

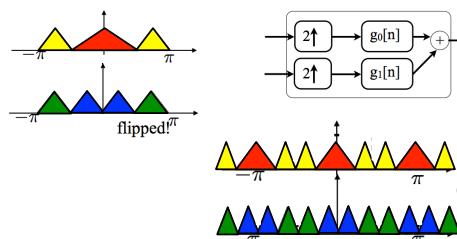


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

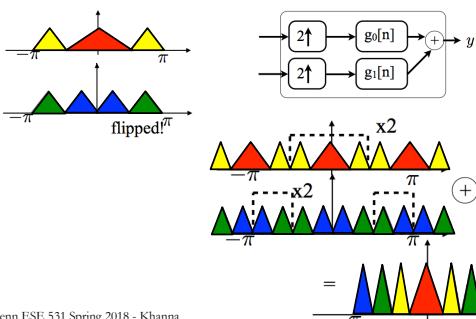


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass with $\omega_c = \pi/2$

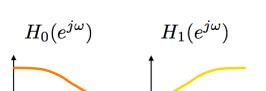
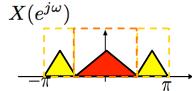
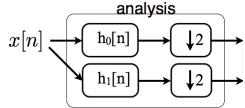


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Multi-Rate Filter Banks

□ h_0, h_1 are NOT ideal low/high pass

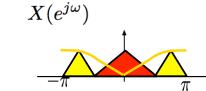
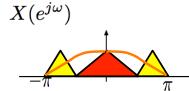
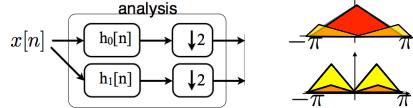


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Non Ideal Filters

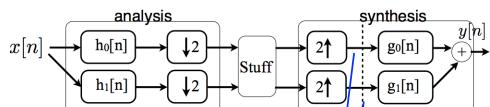
□ h_0, h_1 are NOT ideal low/high pass



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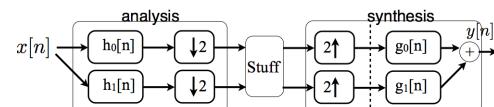
Non Ideal Filters



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Perfect Reconstruction non-Ideal Filters



$$Y(e^{j\omega}) = \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j\omega}) + G_1(e^{j\omega})H_1(e^{j\omega})] X(e^{j\omega}) + \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j(\omega-\pi)}) + G_1(e^{j\omega})H_1(e^{j(\omega-\pi)})] X(e^{j(\omega-\pi)})$$

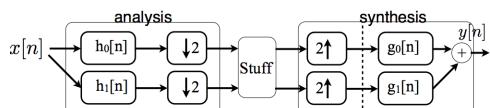
↑ need to cancel!

aliasing

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Quadrature Mirror Filters



Quadrature mirror filters

$$\begin{aligned} H_1(e^{j\omega}) &= H_0(e^{j(\omega-\pi)}) \\ G_0(e^{j\omega}) &= 2H_0(e^{j\omega}) \\ G_1(e^{j\omega}) &= -2H_1(e^{j\omega}) \end{aligned}$$

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Perfect Reconstruction non-Ideal Filters

$$Y(e^{j\omega}) = \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j\omega}) + G_1(e^{j\omega})H_1(e^{j\omega})] X(e^{j\omega}) + \frac{1}{2} [G_0(e^{j\omega})H_0(e^{j(\omega-\pi)}) + G_1(e^{j\omega})H_1(e^{j(\omega-\pi)})] X(e^{j(\omega-\pi)})$$

↑ need to cancel!

aliasing

$$\begin{aligned} H_1(e^{j\omega}) &= H_0(e^{j(\omega-\pi)}) \\ G_0(e^{j\omega}) &= 2H_0(e^{j\omega}) \\ G_1(e^{j\omega}) &= -2H_1(e^{j\omega}) \end{aligned}$$

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Big Ideas

- ❑ Downsampling/Upsampling

- ❑ Practical Interpolation

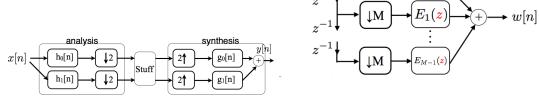
- ❑ Non-integer Resampling

- ❑ Multi-Rate Processing

- Interchanging Operations

- ❑ Polyphase Decomposition

- ❑ Multi-Rate Filter Banks



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- ❑ HW 4 due Sunday

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