

ESE 531: Digital Signal Processing

Lec 10: February 14th, 2017
Practical and Non-integer Sampling, Multi-rate Sampling



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Lecture Outline

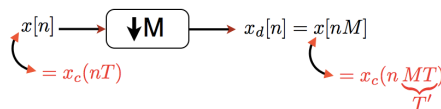
- Downsampling/Upsampling
- Practical Interpolation
- Non-integer Resampling
- Multi-Rate Processing
 - Interchanging Operations
- Polyphase Decomposition
- Multi-Rate Filter Banks

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Downsampling

- Definition: Reducing the sampling rate by an integer number



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Downsampling

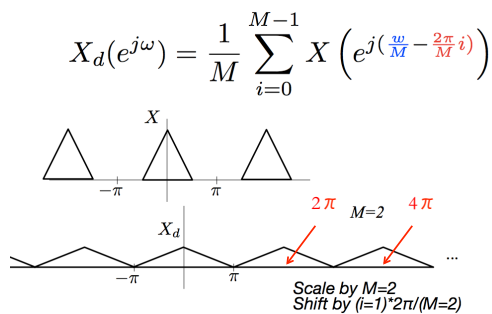
$$\begin{aligned}
 X_d(e^{j\omega}) &= \frac{1}{MT} \sum_k X_c \left(j \left(\frac{\omega}{MT} - \frac{2\pi}{MT} k \right) \right) \\
 &= \frac{1}{M} \sum_{i=0}^{M-1} \underbrace{\frac{1}{T} \sum_{r=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{MT} - \frac{2\pi}{MT} i - \frac{2\pi}{T} r \right) \right)}_{X(e^{j(\frac{\omega}{M}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

stretch by M replicate

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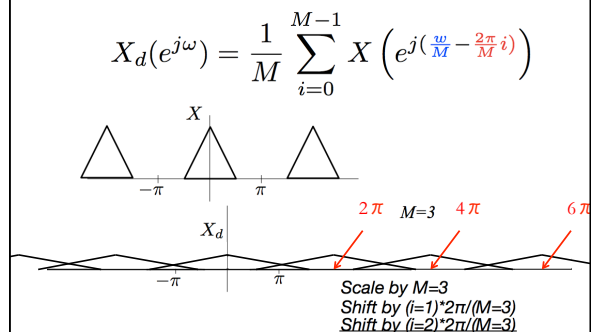
Example



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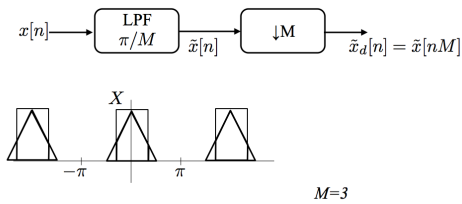
Example



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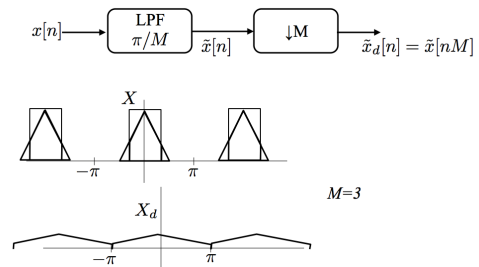
Example



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Example



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Upsampling

- Definition: Increasing the sampling rate by an integer number

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

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

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

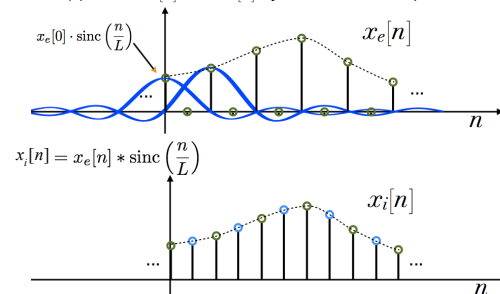
$$(1) \text{ Generate: } x_e[n] = \begin{cases} x[n/L] & n = 0, \pm L, \pm 2L, \dots \\ 0 & \text{otherwise} \end{cases}$$

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Upsampling

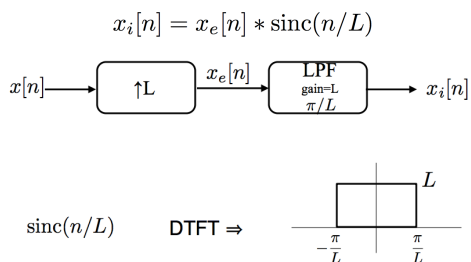
(2) Obtain $x_i[n]$ from $x_e[n]$ by bandlimited interpolation:



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Frequency Domain Interpretation



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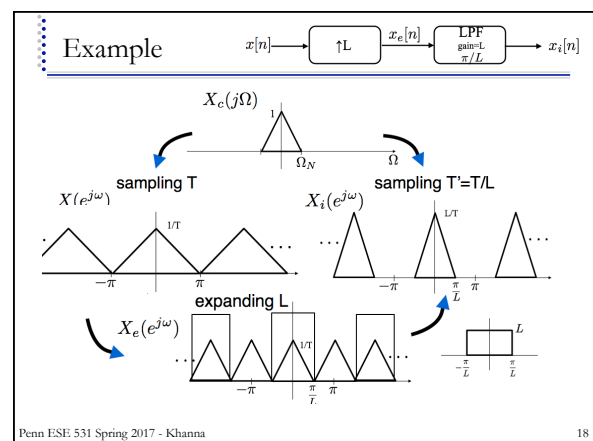
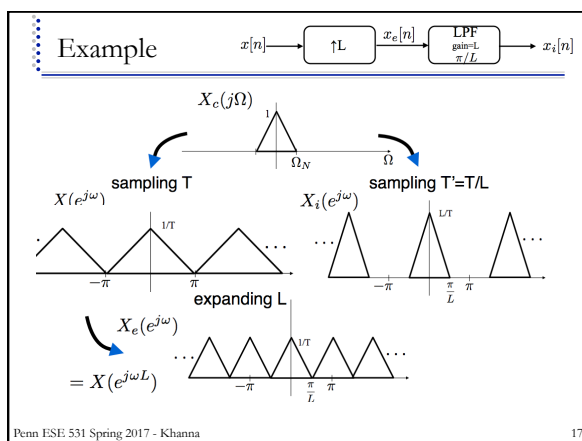
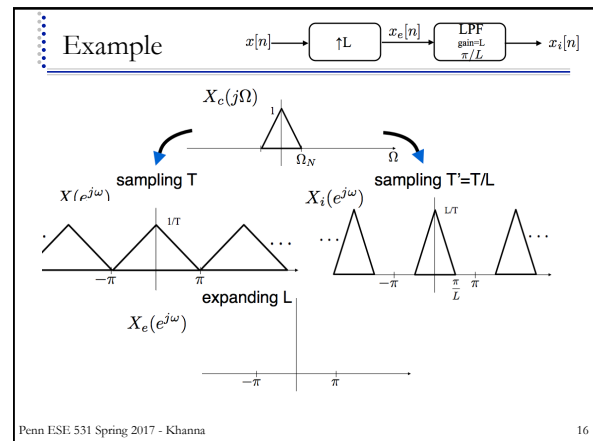
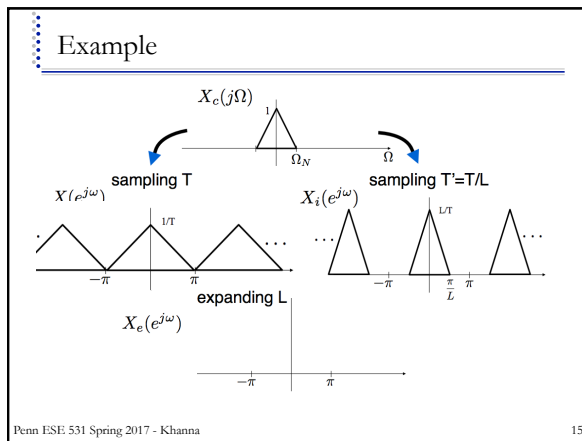
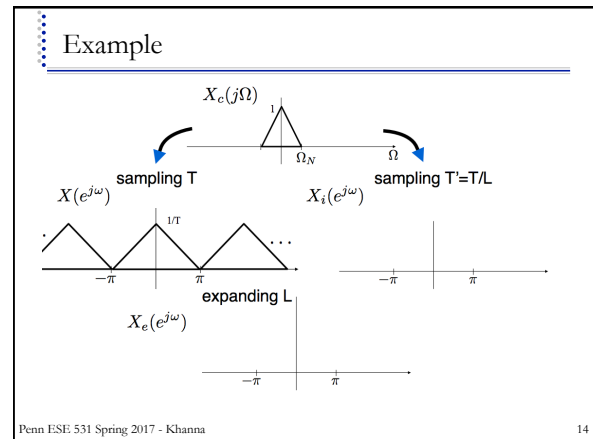
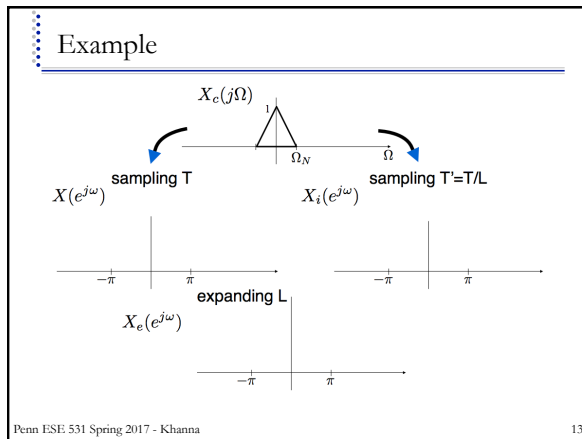
Frequency Domain Interpretation

$$X_e(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x_e[n] e^{-j\omega n} = \sum_{m=-\infty}^{\infty} \underbrace{x_e[mL]}_{=x[m]} e^{-j\omega mL} = X(e^{j\omega L})$$

Compress DTFT by a factor of L!

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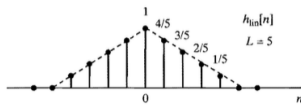
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Practical Interpolation

- Interpolate with simple, practical filters
 - Linear interpolation – samples between original samples fall on a straight line connecting the samples
 - Convolve with triangle instead of sinc

$$h_{lin}[n] = \begin{cases} 1 - |n|/L, & |n| \leq L, \\ 0, & \text{otherwise,} \end{cases}$$

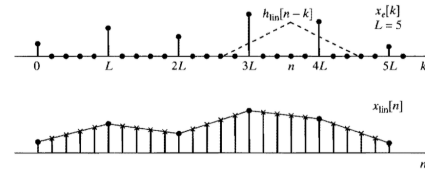


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Practical Interpolation

- Interpolate with simple, practical filters
 - Linear interpolation – samples between original samples fall on a straight line connecting the samples
 - Convolve with triangle instead of sinc

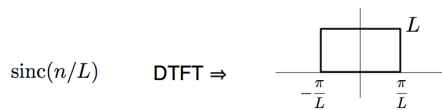
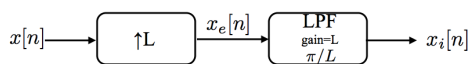


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Frequency Domain Interpretation

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

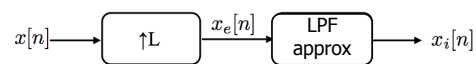


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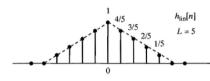
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Linear Interpolation -- Frequency Domain

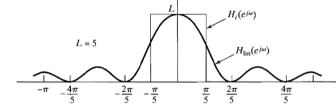
$$x_i[n] = x_e[n] * h_{lin}[n]$$



$$h_{lin}[n] = \begin{cases} 1 - |n|/L, & |n| \leq L, \\ 0, & \text{otherwise,} \end{cases}$$



DTFT $\Rightarrow$

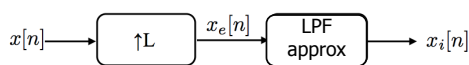


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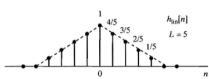
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Linear Interpolation -- Frequency Domain

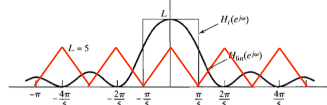
$$x_i[n] = x_e[n] * h_{lin}[n]$$



$$h_{lin}[n] = \begin{cases} 1 - |n|/L, & |n| \leq L, \\ 0, & \text{otherwise,} \end{cases}$$



DTFT $\Rightarrow$

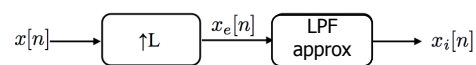


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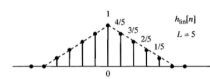
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Linear Interpolation -- Frequency Domain

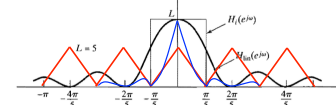
$$x_i[n] = x_e[n] * h_{lin}[n]$$



$$h_{lin}[n] = \begin{cases} 1 - |n|/L, & |n| \leq L, \\ 0, & \text{otherwise,} \end{cases}$$



DTFT $\Rightarrow$



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Non-integer Sampling

□ $T' = TM/L$

- Upsample by L, then downsample by M



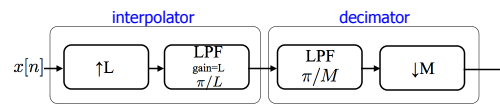
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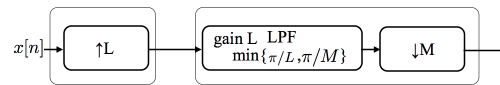
Non-integer Sampling

□ $T' = TM/L$

- Upsample by L, then downsample by M



Or,

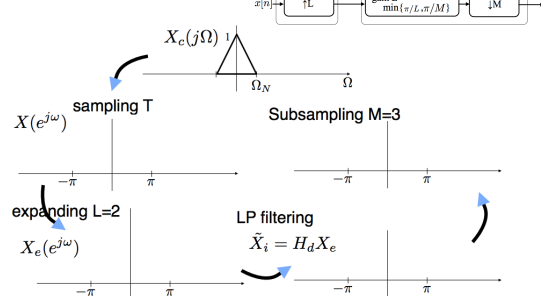


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Example

□ $T' = 3/2T \rightarrow L=2, M=3$

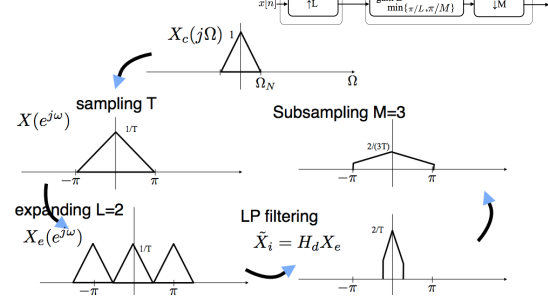


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Example

□ $T' = 3/2T \rightarrow L=2, M=3$



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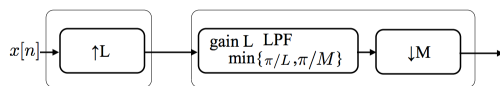
Non-integer Sampling

□ $T' = TM/L$

- Downsample by M, then upsample by L?



Or,



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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 $M=101$

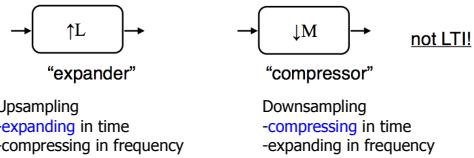
- Fortunately there are ways around it!

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

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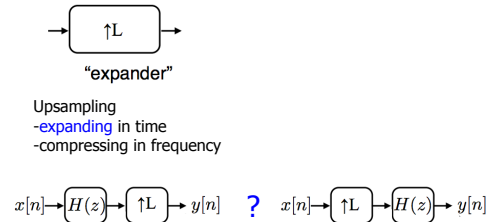
Interchanging Operations



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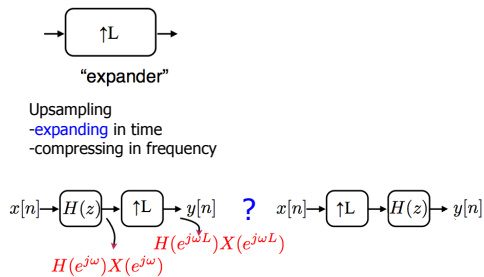
Interchanging Operations - Expander



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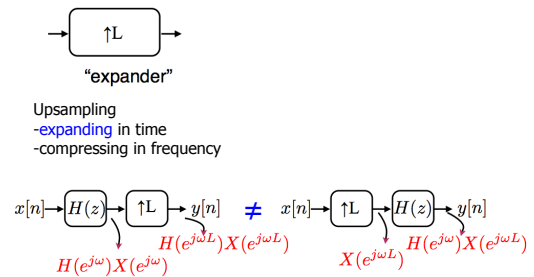
Interchanging Operations - Expander



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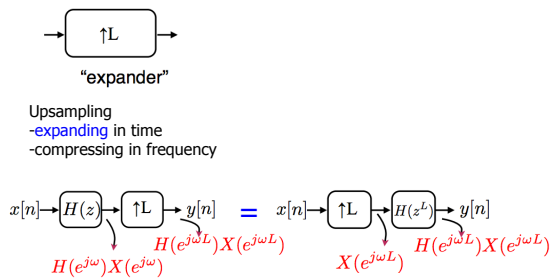
Interchanging Operations - Expander



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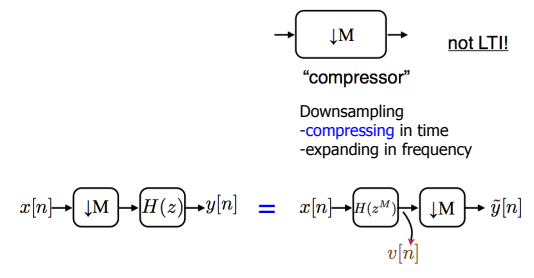
Interchanging Operations - Expander



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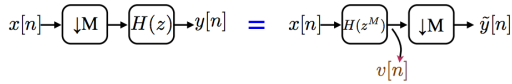
Interchanging Operations - Compressor



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Interchanging Operations - Compressor

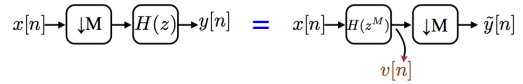


$$Y(e^{j\omega}) = H(e^{j\omega}) \left(\frac{1}{M} \sum_{i=0}^{M-1} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \right)$$

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Interchanging Operations - Compressor

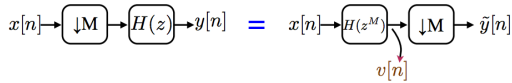


$$\begin{aligned} Y(e^{j\omega}) &= H(e^{j\omega}) \left(\frac{1}{M} \sum_{i=0}^{M-1} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \underbrace{H \left(e^{j(\omega - 2\pi i)} \right)}_{H(e^{j\omega})} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})}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

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Interchanging Operations - Compressor

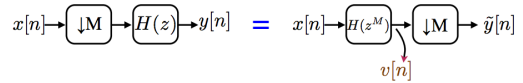


$$\begin{aligned} Y(e^{j\omega}) &= H(e^{j\omega}) \left(\frac{1}{M} \sum_{i=0}^{M-1} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})} \right) \right) \\ &= \frac{1}{M} \sum_{i=0}^{M-1} \underbrace{H \left(e^{j(\omega - 2\pi i)} \right)}_{H(e^{j\omega})} X \left(e^{j(\frac{\omega}{M} - \frac{2\pi i}{M})}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

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Interchanging Operations - Compressor



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

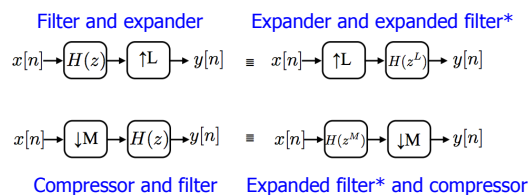
After compressing

$$V(e^{j\omega}) = H(e^{j\omega M}) X(e^{j\omega})$$

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Interchanging Operations - Summary



*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

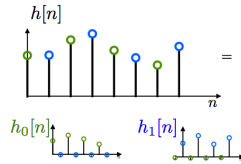
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Polyphase Decomposition

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

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



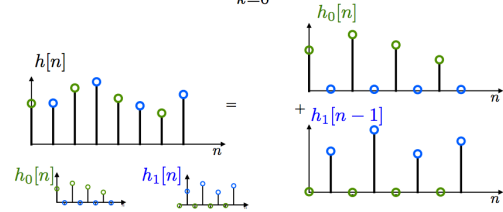
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Polyphase Decomposition

- 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

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

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

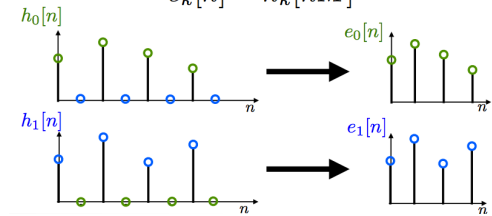
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Polyphase Decomposition

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

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



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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,

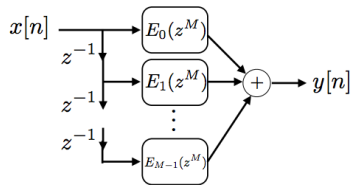
$$H(z) = \sum_{k=0}^{M-1} E_k(z^M) z^{-k}$$

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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

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

Problem:

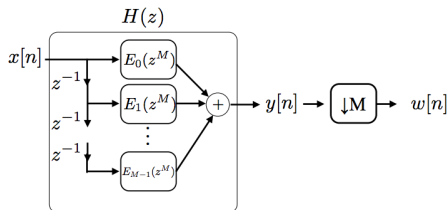
- Compute all $y[n]$ and then throw away -- wasted computation!
- For FIR length $N \rightarrow N$ mults/unit time

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Polyphase Implementation of Decimation

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

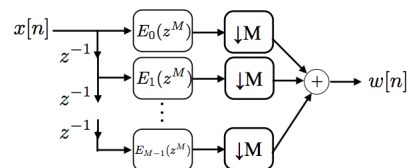


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Polyphase Implementation of Decimation

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



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Interchanging Operations - Summary

$$\begin{aligned} \text{Filter and expander} \quad & x[n] \rightarrow H(z) \rightarrow \uparrow L \rightarrow y[n] \quad \equiv \quad x[n] \rightarrow \uparrow L \rightarrow H(z^L) \rightarrow y[n] \\ \text{Compressor and filter} \quad & x[n] \rightarrow \downarrow M \rightarrow H(z) \rightarrow y[n] \quad \equiv \quad x[n] \rightarrow H(z^M) \rightarrow \downarrow M \rightarrow y[n] \end{aligned}$$

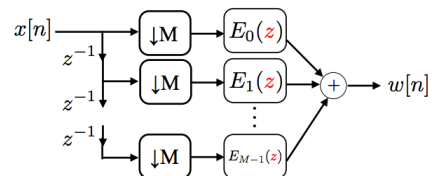
Expander and expanded filter Expanded filter and compressor

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Polyphase Implementation of Decimation

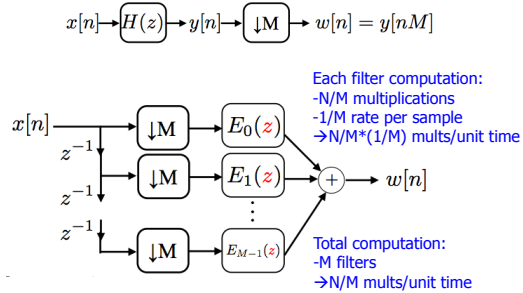
$$x[n] \rightarrow H(z) \rightarrow y[n] \rightarrow \downarrow M \rightarrow w[n] = y[nM]$$



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Polyphase Implementation of Decimation



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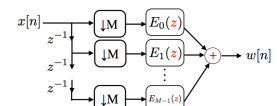
Multi-Rate Signal Processing

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!

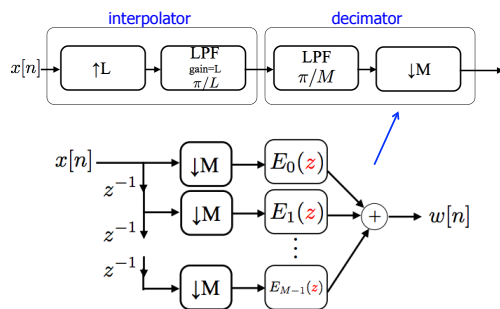
- Called multi-rate
- Uses compressors, expanders and filtering



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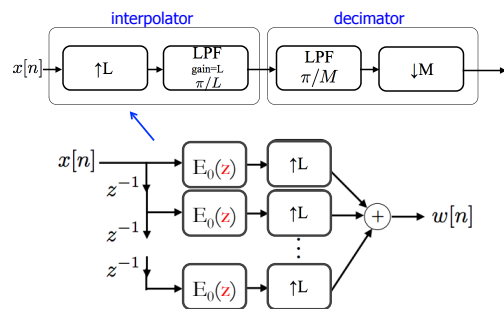
Polyphase Implementation of Decimator



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Polyphase Implementation of Interpolation



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Multi-Rate Filter Banks

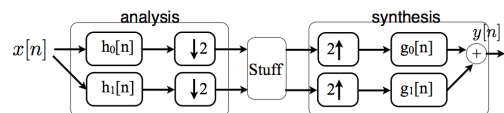
- Use filter banks to operate on a signal differently in different frequency bands
 - To save computation, reduce the rate after filtering

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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

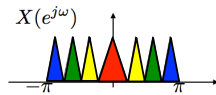
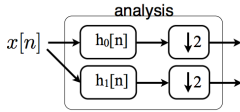


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

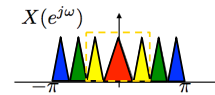
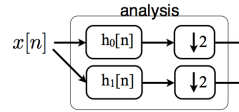


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

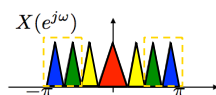
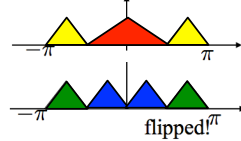
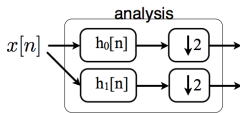


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

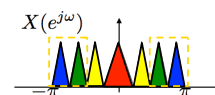
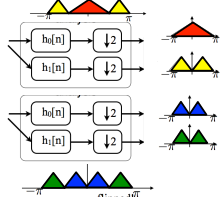
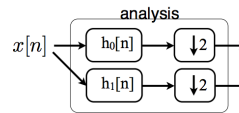


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass



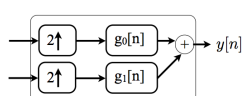
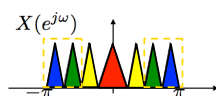
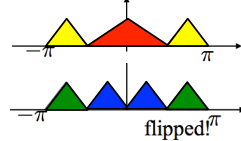
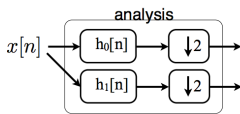
Have to be careful with order!

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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass

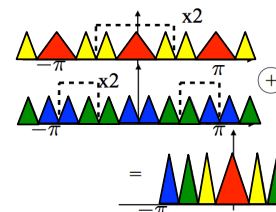
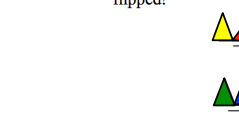
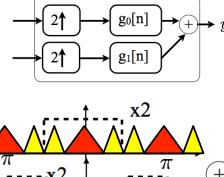
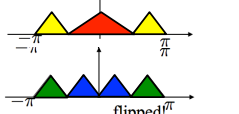


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Multi-Rate Filter Banks

- Assume h_0, h_1 are ideal low/high pass



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- h_0, h_1 are NOT ideal low/high pass



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- ❑ h_0, h_1 are NOT ideal low/high pass



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

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Admin

- HW 4 due Friday
 - Typo in code in MATLAB problem, corrected handout
 - See Piazza for more information