ECE 6560
Multirate Signal Processing
Decimation and Interpolation

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References


Filter-Decimator Implementation (1)

- Deriving a computational structure

\[
\begin{align*}
  w(n) &= \sum_{p=-\infty}^{\infty} h(p) \cdot x(n - p) \\
  y(m) &= w(m \cdot M) = \sum_{p=-\infty}^{\infty} h(p) \cdot x(m \cdot M - p) \\
  p &= r \cdot M + \rho \\
  y(m) &= w(m \cdot M) = \sum_{r=-\infty}^{M-1} \sum_{\rho=0}^{M-1} h(r \cdot M + \rho) \cdot x(m \cdot M - r \cdot M - \rho) \\
  y(m) &= \sum_{\rho=0}^{M-1} \left\{ \sum_{r=-\infty}^{\infty} h(r \cdot M + \rho) \cdot x((m-r) \cdot M - \rho) \right\}
\end{align*}
\]
Filter-Decimator Implementation (2)

\[ y(m) = \sum_{\rho=0}^{M-1} \left\{ \sum_{r=-\infty}^{\infty} h(r \cdot M + \rho) \cdot x((m-r) \cdot M - \rho) \right\} \]

Implementation:  
(1) Generate polyphase elements
(2) Sum the polyphase elements

\[ y_\rho(m) = \sum_{r=-\infty}^{\infty} h(r \cdot M + \rho) \cdot x((m-r) \cdot M - \rho) \quad \text{for} \quad \rho = 0 : M - 1 \]

\[ y(m) = \sum_{\rho=0}^{M-1} y_\rho(m) \]
Filter-Decimator Implementation (3)

- Using a causal filter of length \(N=\lambda M\)

Implementation:
1. Generate polyphase elements
2. Sum the polyphase elements

\[
y_\rho(m) = \sum_{r=0}^{\lambda-1} h(r \cdot M + \rho) \cdot x((m-r) \cdot M - \rho) \quad \text{for} \quad \rho = 0 : M - 1
\]

\[
y(m) = \sum_{\rho=0}^{M-1} y_\rho(m)
\]

Coefficient and Data Sets

\[
h_\rho(r) = h(r \cdot M + \rho) \quad \text{for} \quad r = 0 : \lambda - 1
\]

\[
x_\rho((m-r) \cdot M) = x((m-r) \cdot M - \rho) \quad \text{for} \quad \rho = 0 : M - 1
\]

\[
y_\rho(m) = \sum_{r=0}^{\lambda-1} h_\rho(r) \cdot x_\rho((m-r) \cdot M) \quad \text{for} \quad \rho = 0 : M - 1
\]
Polyphase Filter-Decimator

\[ y(m) = \sum_{\rho=0}^{M-1} \left\{ \sum_{r=0}^{M-1} h(r \cdot M + \rho) \cdot x((m-r) \cdot M - \rho) \right\} \]
## Matrix Implementation

\[
h_\rho(r) = h(r\cdot M + \rho) \quad \text{for} \quad r = 0 : \lambda - 1
\]
\[
x_\rho((m-r)\cdot M) = x((m-r)\cdot M - \rho) \quad \text{for} \quad \rho = 0 : M - 1
\]
\[
y_\rho(m) = \sum_{r=0}^{\lambda-1} h_\rho(r) \cdot x_\rho((m-r)\cdot M) \quad \text{for} \quad \rho = 0 : M - 1
\]

\[
y(m) = \sum_{\rho=0}^{M-1} y_\rho(m)
\]

<table>
<thead>
<tr>
<th>(y_0(m))</th>
<th>(y_\rho(m))</th>
<th>(y_{M-1}(m))</th>
</tr>
</thead>
<tbody>
<tr>
<td>(x_{m-0})</td>
<td>(x_{m-r\cdot M-0})</td>
<td>(x_{m-(\lambda-1)\cdot M-0})</td>
</tr>
<tr>
<td>(\vdots)</td>
<td>(\ddots)</td>
<td>(\vdots)</td>
</tr>
<tr>
<td>(x_{m-\rho})</td>
<td>(x_{m-r\cdot M-\rho})</td>
<td>(x_{m-(\lambda-1)\cdot M-\rho})</td>
</tr>
<tr>
<td>(\vdots)</td>
<td>(\ddots)</td>
<td>(\vdots)</td>
</tr>
<tr>
<td>(x_{m-(M-1)})</td>
<td>(x_{m-r\cdot M-(M-1)})</td>
<td>(x_{m-(\lambda-1)\cdot M-(1)})</td>
</tr>
</tbody>
</table>

\[
y(m) = \sum_{\rho=0}^{M-1} y_\rho(m)
\]
Example Matlab Code Execution for $y(m)$ given $h$ and $x(m-(\lambda M-1):m-0)$

```matlab
Xmatrix = flipud(fliprl(reshape(x,M,lambda));
Hmatrix= reshape(h,M,lambda);
Yrho = sum(Xmatrix.*Hmatrix,2);
Y = sum(Yrho);
```

Updating Xmatrix

```matlab
Xmatrix(:,2:lambda) = Xmatrix(:,1:lambda-1);
Xmatrix(:,1) = flipud(x(1:M).
```

```matlab
T)
```
Filter and Decimate

- Input at 20 kHz, with an output data rate only at 400 Hz.
  - Compute the 360 tap filter once every 50 input data cycles, but the data locations must be maintained.
  - What if we took the 360 tap filter and partitioned it into 50 sampled filters?
    \[ \phi_i = h(i : 50 : end) \]
  - Each filter would get one new input every 50 clock cycles.
  - These is a polyphase implementation.

Figure 5.3 50-to-1 Polyphase Partition and Down Sampling of Low-pass Filter
MATLAB Polyphase Filter Decimator

- **Chap5_2.m**
  
  ```matlab
  lambda = length(h1)/polytaps;
  M = polytaps;

  Pfilter1 = reshape(h1,M,lambda);

  xarray = zeros(M,lambda);
  xshift = [zeros(lambda-1,1) eye(lambda-1) ; zeros(1,lambda)];

  for ii = 1:numblocks
    xarray = xarray * xshift;
    Tindex = 1+((ii-1)*M:(ii*M-1))';
    xarray(:,1) = flipud(TestSig(Tindex));
    yvect(:,ii) = sum(((xarray)) .* Pfilter1,2);
  end
  yout1 = sum(yvect);'
  ```
The Z-Transform of w, x and y

\[ w'(n) = \begin{cases} 
  w(n), & n = 0, \pm M, \pm 2M \ldots \\
  0, & \text{otherwise}
\end{cases} \]

\[ w'(n) = w(n) \left[ \frac{1}{M} \sum_{l=0}^{M-1} \exp \left( j2\pi \cdot \frac{l \cdot n}{M} \right) \right], \quad -\infty < n < \infty \]

The discrete Fourier series of an impulse train with period \( M \)

\[ Y(z) = \sum_{m=-\infty}^{\infty} y(m) \cdot z^{-m} \]

\[ Y(z) = \sum_{m=-\infty}^{\infty} w'(mM) \cdot z^{-m} = \sum_{m=-\infty}^{\infty} w'(m) \cdot z^{-m/M} \]

\[ Y(z) = \sum_{m=-\infty}^{\infty} w(m) \left[ \frac{1}{M} \sum_{l=0}^{M-1} \exp \left( j2\pi \cdot \frac{l \cdot m}{M} \right) \right] \cdot z^{-m/M} \]
Studying the Z-Transform (2)

\[
Y(z) = \sum_{m=-\infty}^{\infty} w(m) \left[ \frac{1}{M} \sum_{l=0}^{M-1} \exp\left( j2\pi \cdot \frac{l \cdot m}{M} \right) \right] \cdot z^{-m/M}
\]

\[
Y(z) = \frac{1}{M} \sum_{l=0}^{M-1} \sum_{m=-\infty}^{\infty} w(m) \cdot \exp\left( j2\pi \cdot \frac{l \cdot m}{M} \right) \cdot z^{-m/M}
\]

\[
Y(z) = \frac{1}{M} \sum_{l=0}^{M-1} W \left( \exp\left( - j2\pi \cdot \frac{l}{M} \right) \cdot z^{-1/M} \right)
\]

Since

\[
W(z) = H(z) \cdot X(z)
\]

\[
Y(z) = \frac{1}{M} \sum_{l=0}^{M-1} H \left( \exp\left( - j2\pi \cdot \frac{l}{M} \right) \cdot z^{1/M} \right) \cdot X \left( \exp\left( - j2\pi \cdot \frac{l}{M} \right) \cdot z^{1/M} \right)
\]

Z-Transform in terms of the Frequency

\[ z = \exp(jw') \quad \text{where} \quad w' = 2\pi \cdot f \cdot T' \]

\[
Y(e^{jw'}) = \frac{1}{M} \sum_{l=0}^{M-1} H \left( \exp \left( -\frac{j2\pi \cdot l}{M} \right) \cdot \exp \left( j \frac{w}{M} \right) \right) \cdot X \left( \exp \left( -\frac{j2\pi \cdot l}{M} \right) \cdot \exp \left( j \frac{w}{M} \right) \right)
\]

\[
Y(e^{jw}) = \frac{1}{M} \sum_{l=0}^{M-1} H \left( \exp \left( j \frac{w - 2\pi \cdot l}{M} \right) \right) \cdot X \left( \exp \left( \frac{w - 2\pi \cdot l}{M} \right) \right)
\]

\[
Y(e^{jw}) = \frac{1}{M} \left\{ H(e^{jw}) \cdot X(e^{jw}) + H(e^{(w-2\pi)/M}) \cdot X(e^{(w-2\pi)/M}) + \cdots \right\}
\]

Note: The Nyquist zone are included/aliased based on the filter response. For a low pass filter signal or filter …

\[
Y(e^{jw'}) = \frac{1}{M} H(e^{jw'}) \cdot X(e^{jw'}) = \frac{1}{M} X(e^{jw'})
\]

Bandpass Filter

- What would happen if a bandpass filters were used instead of a low pass filter?

\[ Y(e^{jw}) = \frac{1}{M} \left\{ H(e^{jw}) \cdot X(e^{jw}) + H(e^{j(w-2\pi/M)}) \cdot X(e^{j(w-2\pi/M)}) + \ldots \right\} \]

- It should work the same way ....
Nyquist Zones of a Polyphase Filter

• Chap6_1.m and Chap6_2.m
  – Magnitude and phase plots of non-decimated polyphase filter elements.
  – When a complex filter/complex signal mixing is performed, the phase plots are only equal in the Nyquist band of the perfectly centered complex mixing signal.
    • All others should provide phase sums that cancel signal in the Nyquist band.
Crochiere and Rabiner
Filter Decimation Summarized (1)

\[ w(n) = \sum_{k=-\infty}^{\infty} h(k) \cdot x(n-k) \]

\[ y(m) = w(mM) \]

\[ y(m) = \sum_{k=-\infty}^{\infty} h(k) \cdot x(mM-k) = \sum_{k=-\infty}^{\infty} h(mM-k) \cdot x(k) \]

Assume a causal FIR filter of length \( \lambda M \):

\[ h(k) = \begin{cases} 
0, & \text{for } k < 0 \\
h(k), & \text{for } 0 \leq k \leq \lambda M - 1 \\
0, & \text{for } \lambda M - 1 < k 
\end{cases} \]
Filter Decimation Summarized (2)

\[ y(m) = \sum_{k=0}^{\lambda M-1} h(k) \cdot x(mM - k) \]

Let \( k = rM + \rho \)

\[ y(m) = \sum_{r=0}^{\lambda-1} \sum_{\rho=0}^{M-1} h(rM + \rho) \cdot x(mM - rM - \rho) \]

\[ y(m) = \sum_{\rho=0}^{M-1} \sum_{r=0}^{\lambda-1} h(rM + \rho) \cdot x((m-r)M - \rho) \]

\[ y(m) = \sum_{\rho=0}^{M-1} \left\{ \sum_{r=0}^{\lambda-1} h_\rho (r) \cdot x_\rho (m-r) \right\} \]

\[ h_\rho (r) = h(rM + \rho) \]

\[ x_\rho (m-r) = x((m-r)M - \rho) \]

The summation of \( M \lambda \)-tap filters.

The computation is performed once every \( M \) input samples.

Vector Interpretation

\[ y(m) = \sum_{k=0}^{\lambda M-1} h(k) \cdot x(mM - k) \]

\[ h_{\rho}(r) = h(rM + \rho) \]

\[ y(m) = \sum_{\rho=0}^{M-1} \sum_{r=0}^{\lambda-1} h_{\rho}(r) \cdot x_{\rho}(m-r) \]

\[ x_{\rho}(m-r) = x((m-r)M - \rho) \]

x has \( \lambda \) columns of M rows, with columns numbered left-to-right in time as \( r \) increases and bottom-to-top as \( \rho \) increases.

h has \( \lambda \) columns of M rows, with columns numbered left-to-right as \( r \) increases and top-to-bottom as \( \rho \) increases. (x & h must convolve)

To perform point-wise multiplication, if x is stored left-to-right and top-to-bottom, flip\( \text{lr} \) h and then flip\( \text{ud} \) and it will line up correctly for point-wise multiplication and summation. (This can be done to x or h to convolve).

This appears as a “matrix” convolution of the sample vectors.
Polyphase Implementation (1)

Delay Line Based

Block M delays

ρ coefficients applied sequentially
Polyphase Implementation (2)

Commutator Base

\[ x(n) \to h_0(m) \to y(m) \]

\[ h_1(m) \]

\[ h_2(m) \]

\[ h_{M-1}(m) \]

Sum
What about Mixing Prior to Filter Decimation

\[ y_k(m) = \sum_{n=-\infty}^{\infty} h(n) \cdot x(mM - n) \cdot \exp\left(-j2\pi \cdot \frac{(mM-n) \cdot k}{N}\right) \]

\[ y_k(m) = \sum_{n=-\infty}^{\infty} h(n) \cdot x(mM - n) \cdot \exp\left(-j2\pi \cdot m \cdot k \cdot \frac{M}{N}\right) \cdot \exp\left(j2\pi \cdot \frac{n \cdot k}{N}\right) \]

Let \( M = N \)

\[ y_k(m) = \sum_{n=-\infty}^{\infty} \left\{ h(n) \cdot \exp\left(j2\pi \cdot \frac{n \cdot k}{N}\right) \right\} \cdot x(mM - n) \]

Note equivalence to a complex filter.
Mixing Continued

Let \( n = rM + \rho \)

\[
y_k(m) = \sum_{r=0}^{\lambda-1} \sum_{\rho=0}^{M-1} h(rM + \rho) \cdot x(mM - rM + \rho) \cdot \exp\left( j2\pi \cdot \frac{(rM + \rho) \cdot k}{M} \right)
\]

\[
y_k(m) = \sum_{r=0}^{\lambda-1} \sum_{\rho=0}^{M-1} h_\rho(r) \cdot \exp\left( j2\pi \cdot \frac{\rho \cdot k}{M} \right) \cdot x_\rho(m - r)
\]

\[
y_k(m) = \sum_{\rho=0}^{M-1} \exp\left( j2\pi \cdot \frac{\rho \cdot k}{M} \right) \cdot \sum_{r=0}^{\lambda-1} h_\rho(r) \cdot x_\rho(m - r)
\]

Note, \( k \) for one frequency, use an IFFT for all frequencies in \( k \).

The complex weighting and summation of \( M \lambda \)-tap filters.

The computation is performed once every \( M \) input samples.
Polyphase Implementation

\[ y_k(m) = \sum_{\rho=0}^{M-1} \sum_{r=0}^{\lambda-1} h_{\rho}(r) \cdot \exp \left( j2\pi \cdot \frac{\rho \cdot k}{M} \right) \cdot x_{\rho}(m-r) \]

A Single Frequency Digital Channel Output.
Polyphase Implementation

\[ y_k(m) = \sum_{\rho=0}^{M-1} \exp\left( j 2\pi \cdot \frac{\rho \cdot k}{M} \right) \cdot \sum_{r=0}^{M-1} h_\rho(r) \cdot x_\rho(m-r) \]

A Multichannel “Digital Channelizer” or “Analysis” Implementation
Interpolate-Filter (1)

• Deriving a computational structure

\[ w(n) = \begin{cases} 
  x\left(\frac{n}{L}\right), & n = m = 0, \pm 1, \pm 2 \ldots \\
  0, & \text{otherwise} 
\end{cases} \]

\[ y(n) = \sum_{p=-\infty}^{\infty} h(p) \cdot w(n - p) \]

\[ y(n) = \sum_{p=-\infty}^{\infty} h(p) \cdot x\left(\frac{n-p}{L}\right) \]

\[ p = r \cdot M + \rho \quad \quad n = s \cdot M + p \]
Interpolate-Filter (2)

\[ y(n) = \sum_{p=-\infty}^{\infty} h(p) \cdot x\left(\frac{n-p}{L}\right) \]

\[ p = r \cdot L + \rho \]

\[ y(n) = \sum_{\rho=0}^{L-1} \sum_{r=-\infty}^{\infty} h(r \cdot L + \rho) \cdot x\left(\frac{n-r \cdot L - \rho}{L}\right) \]

\[ n = s \cdot L + p \]

\[ y(s \cdot L + p) = \sum_{\rho=0}^{L-1} \sum_{r=-\infty}^{\infty} h(r \cdot L + \rho) \cdot x\left(\frac{s \cdot L + p - r \cdot L - \rho}{L}\right) \]

Note that the rho summation only exists for \( p = \rho \) making \( y \) exist as signal outputs from the \( r \) summation

\[ y_{\rho}(s \cdot L) = \sum_{r=-\infty}^{\infty} h(r \cdot L + \rho) \cdot x(s-r) \]
Interpolate-Filter (3)

- Using a causal filter of length $\lambda L$
  \[ h(k) = \begin{cases} 
  0, & \text{for } k < 0 \\
  h(k), & \text{for } 0 \leq k \leq \lambda L - 1 \\
  0, & \text{for } \lambda L - 1 < k
  \end{cases} \]

Implementation:
1. Generate polyphase elements
2. Output is individual elements

\[
y_{\rho}(s \cdot L) = \sum_{r=0}^{L-1} h(r \cdot L + \rho) \cdot x(s - r) \quad \text{for} \quad \rho = 0 : L - 1
\]

\[
y(s \cdot L + 0 : s \cdot L + \rho) = y_{\rho}(s \cdot L)
\]

Coefficient sets

\[
h_{\rho}(r) = h(r \cdot L + \rho) \quad \text{for} \quad r = 0 : \lambda - 1
\]

\[
y_{\rho}(s \cdot L) = \sum_{r=0}^{L-1} h_{\rho}(r) \cdot x(s - r) \quad \text{for} \quad \rho = 0 : L - 1
\]
Polyphase Interpolator-Filter

\[ y(s \cdot L + \rho) = \sum_{r=0}^{L-1} h_{\rho}(r) \cdot x(s - r) \]
Matrix Implementation

\[
y(s \cdot L + \rho) = \sum_{r=0}^{L-1} h_\rho(r) \cdot x(s-r)
\]
MATLAB Polyphase Interpolate Filter

- Chap5_3.m
  ```matlab
  xivector = zeros(lambda,1);
  xshift2 = [zeros(1,lambda); eye(lambda-1) zeros(lambda-1,1) ] ;

  for ii = 1:numblocks
    xivector = xshift2 * xivector;
    Tindex = 1+((ii-1)*M:(ii*M-1))';
    xivector(1,1) = yout1(ii);
    ymatrix(:,ii) = Pfilter1 * xivector;
  end

  yout2 = M*reshape((ymatrix),num_samples,1);
  ```
Vector Interpretation

\[ y(n) = \sum_{k=0}^{\lambda L - 1} h(k) \cdot x \left( \frac{n - k}{L} \right) \]

\[ y(sL + t) = \sum_{r=0}^{\lambda - 1} h_t(r) \cdot x(s - r) \]

x has \( \lambda \) rows of length 1 columns, with rows numbered top-to-bottom in time as \( r \) increases (negative time).

h has \( \lambda \) columns of \( L \) rows, with columns numbered left-to-right as \( r \) increases and top-to-bottom as \( t \) increases.

To perform vector-matrix multiplication, if h is stored left-to-right and top-to-bottom it will line up correctly for vector multiplication with x which is inverted in time. (Note: if x is flipud, then fliplr, flipud h and flipud the output vector).

Extract the output \( y \) from the column based result top-to-bottom.

This appears as a “matrix” multiplication of the sample vectors.

Interpolation-Filter
Polyphase Implementation

Commutator Base

\[ x(m) \rightarrow h_0(m) \rightarrow \cdots \rightarrow h_{M-1}(m) \rightarrow y(n) \]
What about Mixing (Up-conversion) After The Interpolation Filter

\[ y(n) = \exp\left(j2\pi \frac{n \cdot k}{N}\right) \cdot \sum_{\rho=0}^{L-1} \sum_{r=0}^{\lambda-1} h(rL + \rho) \cdot x_k\left(\frac{n - \rho}{L} - r\right) \cdot w_{M}^{-nk} \]

Let \( n = sL + t \)

\[ y(sL + t) = \exp\left(j2\pi \cdot \frac{(sL + t) \cdot k}{N}\right) \cdot \sum_{\rho=0}^{L-1} \sum_{r=0}^{\lambda-1} h(rL + \rho) \cdot x_k\left(\frac{sL + t - \rho}{L} - r\right) \]

\[ y(sL + t) = \exp\left(j2\pi \cdot \frac{sL \cdot k}{N}\right) \cdot \exp\left(j2\pi \cdot \frac{t \cdot k}{N}\right) \cdot \sum_{r=0}^{\lambda-1} h(rL + t) \cdot x_k(s - r) \]

Let \( L = N \)

\[ y(sL + t) = \sum_{r=0}^{\lambda-1} \left[h(rL + t) \cdot \exp\left(j2\pi \cdot \frac{t \cdot k}{N}\right)\right] \cdot x_k(s - r) \]

Complex filter equivalence
Mixing Polyphase Implementation

\[ y(sL + t) = \sum_{r=0}^{\lambda-1} h(rL + t) \cdot \left[ x_k(s - r) \cdot \exp \left( j2\pi \cdot \frac{t \cdot k}{N} \right) \right] \]

\[ y(sL + t) = \sum_{r=0}^{\lambda-1} h^{BPF}(r) \cdot x_k(s - r) \]

\( x_k \) placed in Nyquist region \( k \).

Filter becomes a complex bandpass filter.
Multiple Input Signal Implementation

Each x is mixed to a different bin/band and summed with all the other k prior to polyphase interpolation.

\[
y(sL + t) = \sum_{k=0}^{N-1} \sum_{r=0}^{\lambda-1} h(rL + t) \cdot \left[ x_k(s-r) \cdot \exp\left( j2\pi \cdot \frac{t \cdot k}{N} \right) \right]
\]
“Cascaded” Elements

• FFT interpolation-filtering
  – Nyquist rate TDM samples to FDM frequency
  – Complex TDM symbols to FDM output
    (similar to OFDM symbol generation)

• Filter-decimation with FFT
  – FDM frequencies to narrowband TDM at the Nyquist rate!
  – FDM symbols to Complex TDM output
    (similar to OFDM symbol reception)
FDM Generation and Processing

Forming Wideband and Reforming Narrowband

Filters narrower than the Nyquist regions are used for generating FDM waveforms. A guard band between adjacent frequencies is typically used.
Quadrature Mirror Filter Processing

Observing Narrowband and Reforming Wideband

Significant filter restriction are required if the output is required to approximate the input!

Quadrature Mirror Filter Definition and Requirements

Analysis

Synthesis
QMF Applications

- Frequency domain filtering or equalization
- Time-Spectral Analysis with reconstruction

- Arbitrarily take signals apart and then reconstruct them
  - Partial-Band Synthesis to one or more arbitrary bandwidths (universal base station receiver)
  - Partial-Band Analysis with frequency domain summation and full-band synthesis (universal base station transmitter)
  - Applications: cellular telephone base stations, satellite relay stations, etc.