

Module 5 - Multirate Signal Processing

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Digital Signal Processing – p.1/25

Contents

- Applications of multirate signal processing
- Fundamentals
 - decimation
 - interpolation
- Resampling by rational fractions
- Multirate identities
- Polyphase representations
- Maximally decimated filter banks
 - aliasing
 - amplitude and phase distortion
 - perfect reconstruction conditions

Introduction

- In single-rate DSP systems, all data is sampled at the same rate
 no change of rate within the system.
- In multirate DSP systems, sample rates are changed (or are different) within the system
- Multirate can offer several advantages
 - reduced computational complexity
 - reduced transmission data rate.

Example: Audio sample rate conversion

- recording studios use 192 kHz
 - CD uses 44.1 kHz
 - wideband speech coding using 16 kHz

master from studio must be rate-converted by a factor



Example: Oversampling ADC

Consider a Nyquist rate ADC in which the signal is sampled at the desired precision and at a rate such that Nyquist's sampling criterion is just satisfied.

Bandwidth for audio is 20 Hz < f < 20 kHz

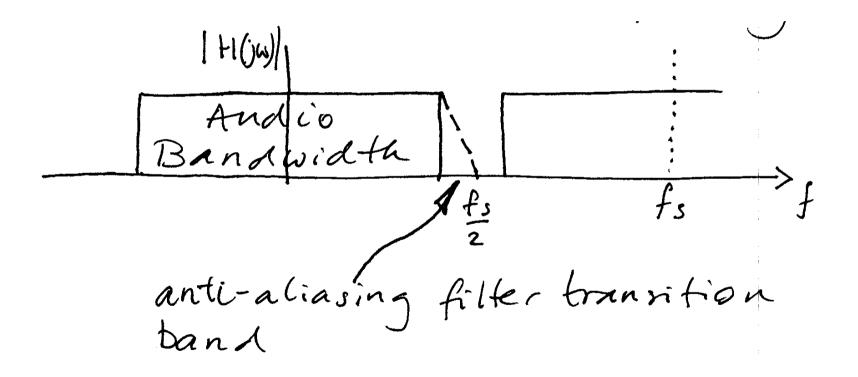
Antialiasing filter required has very demanding specification

$$|H(j\omega)| = 0 \text{ dB}, \ f < 20 \text{ kHz}$$
$$|H(j\omega)| < 96 \text{ dB}, \ f \ge \frac{44.1}{2} \text{ kHz}$$

Requires high order analogue filter such as elliptic filters that have very nonlinear phase characteristics

hard to design, expensive and bad for audio quality.

Nyquist Rate Conversion Anti-aliasing Filter.



Consider oversampling the signal at, say, 64 times the Nyquist rate but with lower precision. Then use multirate techniques to convert sample rate back to 44.1 kHz with full precision.

New (over-sampled) sampling rate is 44.1×64 kHz.

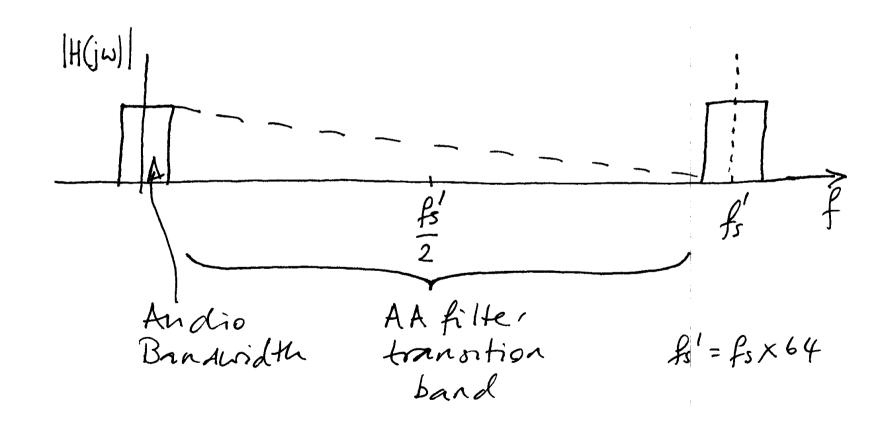
Requires simple antialiasing filter

$$|H(j\omega)| = 0 \text{ dB}, \ f < 20 \text{ kHz}$$

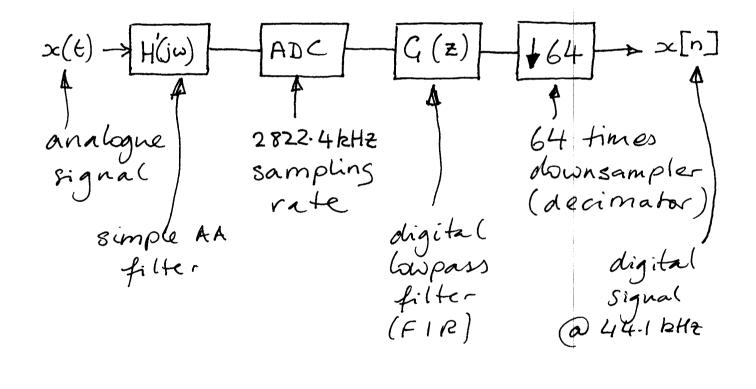
 $|H(j\omega)| < 96 \text{ dB}, \ f \ge (44.1 \times 64) - \frac{44.1}{2} \text{ kHz}$

- Could be implemented by simple filter (eg. RC network)
- Recover desired sampling rate by downsampling process.

Oversampled Conversion Antialiasing Filter



Overall System



This is a simplified version

In these lectures we will study blocks like G(z) and $\downarrow 64$

Example: Subband Coding

Consider quantizing the samples of a speech signal. How many bits are required?

- In general, 16 bits precision per sample is normally used for audio. This gives an adequate dynamic range.
- In practice, certain frequency bands are less important perceptually because they contain less significant information
 - bands with less information or lower perceptual importance may be quantized with lower precision - fewer bits.
- Divide the spectrum of the signal into several subbands then allocate bits to each band appropriately.

- 16 bits per sample, 10 kHz sampling frequency gives
 160 kbits/s
- Divide into 2 bands: high frequency and low frequency subbands.
 High frequencies of speech are less important to intelligibility.
 Therefore use only 8 bits per sample
- The sampling frequency can be reduced by a factor of 2 since bandwidth is halved, still satisfying Nyquist criterion.

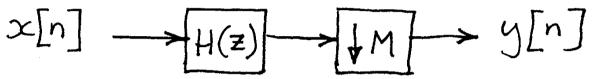
$$5 \times 16 + 5 \times 8 = 120$$
 kbits/s

4:3 compression

Reconstructed signal has no noticeable reduction is signal quality.

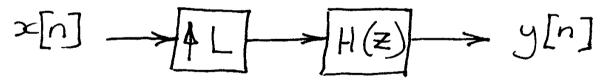
Fundamental Multirate Operations

Downsampling by a factor M
 filter and M-fold decimator



Upsampling by a factor *L*

L-fold expander and filter



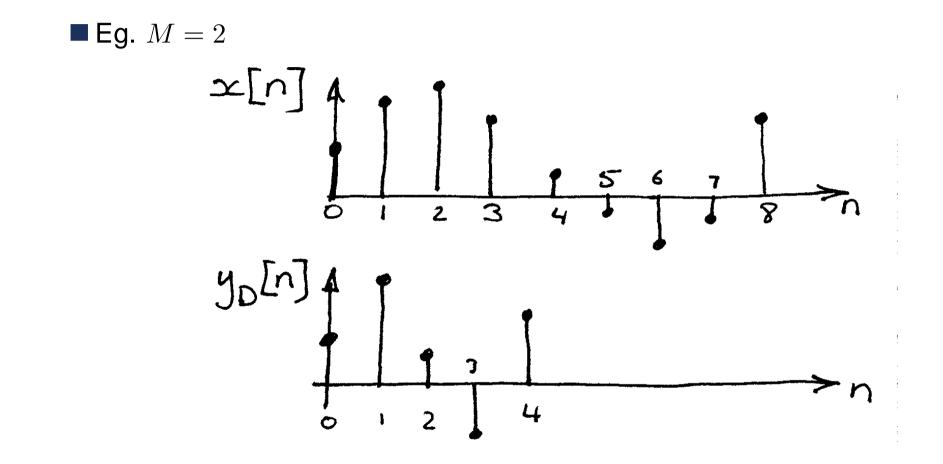
M-fold Decimator

For an input sequence x(n), select only the samples which occur at integer multiples of M. The other samples are thrown away.

$$x[n] \longrightarrow M \longrightarrow y_D[n]$$

 $y_D(n) = x(Mn)$

Aliasing will occur in y_D(n) unless x(n) is sufficiently bandlimited
 loss of information.

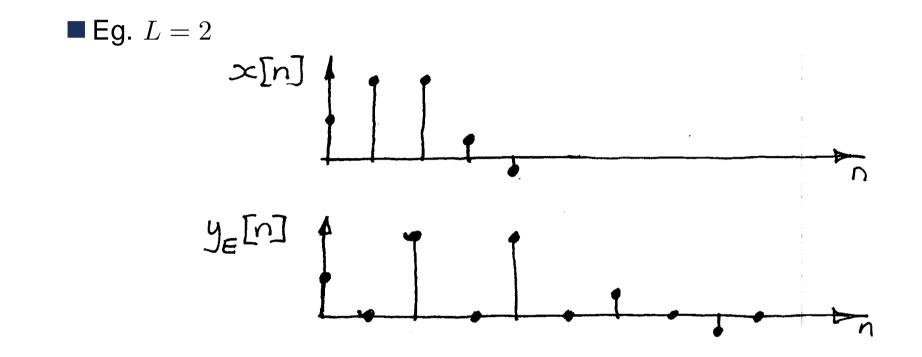


L-fold Expander

For an input sequence x(n), insert L-1 zeros between each sample.

$$\mathcal{P}_{E}[n] \longrightarrow [n] \longrightarrow \mathcal{P}_{E}[n]$$
$$y_{E}(n) = x(Mn)$$

x(n) can always be recovered from y_E(n)
 no loss of information, no aliasing.



Frequency Domain View of the Expander

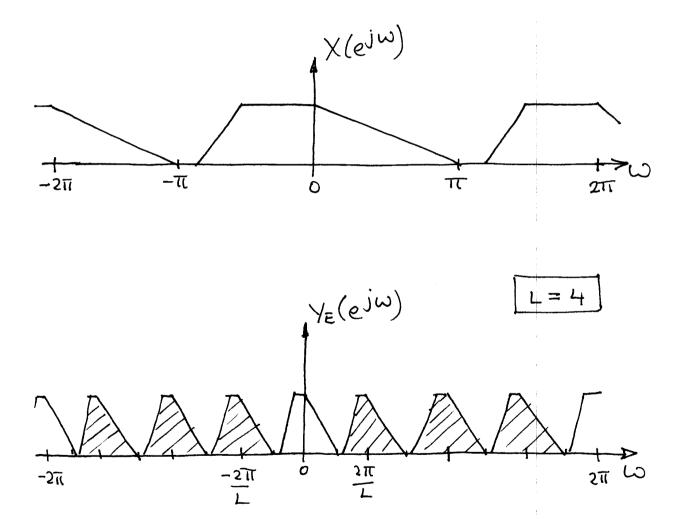
From the definition of the z-transform

$$Y_E(z) = \sum_{n=-\infty}^{\infty} y_E(n) z^{-n}$$
$$= \sum_{k=-\infty}^{\infty} y_E(kL) z^{-kL}$$
$$= \sum_{k=-\infty}^{\infty} x(k) z^{-kL} = X(z^L)$$

For frequency response write $z = e^{j\omega}$ giving

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

- \blacksquare Y_E is a compressed version of X
- In Multiple images of $X(e^{j\omega})$ are created in $Y_E(e^{j\omega})$ between $\omega = 0$ and $\omega = 2\pi$



To use the expander for interpolation, a lowpass filter is applied after the expander to remove the images (shaded).

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Digital Signal Processing – p.18/25

Frequency Domain View of the Decimator

From the definition of the z-transform

$$Y_D(z) = \sum_{n=-\infty}^{\infty} y_D(n) z^{-n} = \sum_{n=-\infty}^{\infty} x(Mn) z^{-n}$$

Let

$$x_1(n) = \begin{cases} x(n) & \text{if } n \text{ is an integer multiple of } M \\ 0 & \text{otherwise} \end{cases}$$

Then

$$Y_D(z) = \sum_{n = -\infty}^{\infty} x_1(Mn) z^{-n} = \sum_{k = -\infty}^{\infty} x_1(k) z^{-k/M}$$

since $x_1(k) = 0$ unless k is a multiple of M.

Therefore

$$Y_D(z) = X_1(z^{1/M}) = \frac{1}{M} \sum_{k=0}^{M-1} X\left(z^{1/M} W_M^k\right)$$

as will be shown on the next slide and using $W^k_M = e^{-j2\pi k/M}$

For frequency response write $z = e^{j\omega}$ to give

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

We arrive at the previous expression for $Y_D(z)$ by considering a new sequence

 $c_M(n) = \begin{cases} 1 & \text{if } n \text{ is an integer multiple of } M \\ 0 & \text{otherwise} \end{cases}$

and then writing

$$x_1(n) = c_M(n)x(n)$$

Further consideration of $c_M(n)$ tells us that $c_M(n)$ is the inverse Fourier transform of unity and can be written

$$c_M(n) = \frac{1}{M} \sum_{k=0}^{M-1} W_M^{-kn}$$

Then

$$X_{1}(z) = \frac{1}{M} \sum_{k=0}^{M-1} \sum_{n=-\infty}^{\infty} x(n) W_{M}^{-kn} z^{-n}$$
$$= \frac{1}{M} \sum_{k=0}^{M-1} \sum_{n=-\infty}^{\infty} x(n) (W_{M}^{k} z)^{-n}$$
$$= \frac{1}{M} \sum_{k=0}^{M-1} X (z W_{M}^{k})$$

from the definition of the z-transform. So finally

$$Y_D(z) = \frac{1}{M} \sum_{k=0}^{M-1} X\left(z^{1/M} W_M^k\right)$$

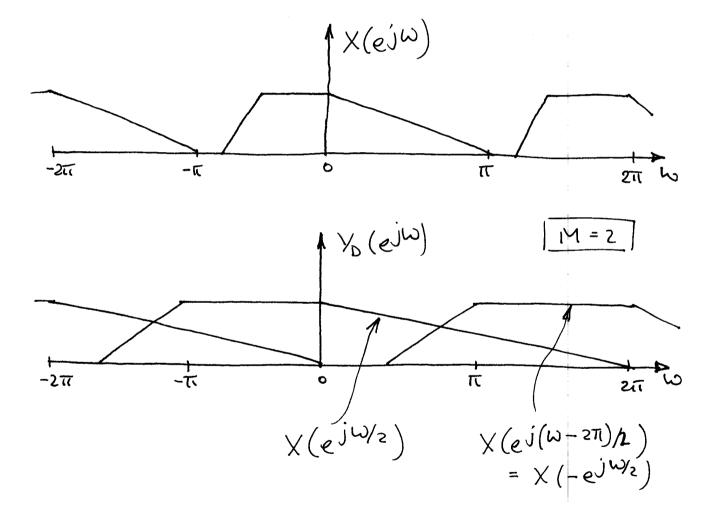
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Digital Signal Processing – p.22/25

• What does $Y_D(z) = \frac{1}{M} \sum_{k=0}^{M-1} X(z^{1/M} W_M^k)$ represent?

stretching of $X(e^{j\omega})$ to $X(e^{j\omega/M})$

- \blacksquare creating M-1 copies of the stretched versions
- shifting each copy by successive multiples of 2π and superimposing (adding) all the shifted copies
- \blacksquare dividing the result by M



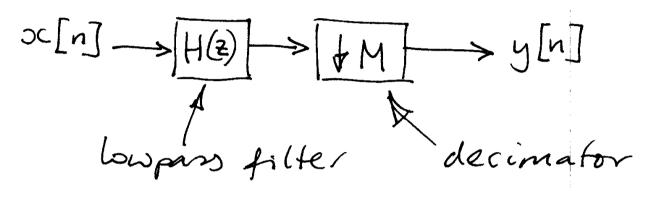
To use a decimation process we must first bandlimit the signal to $|\omega| < \frac{\pi}{M}$.

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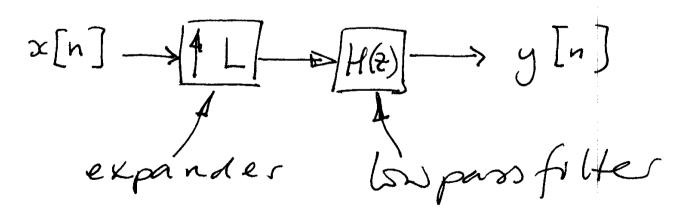
Digital Signal Processing – p.24/25

Summary

Downsampling

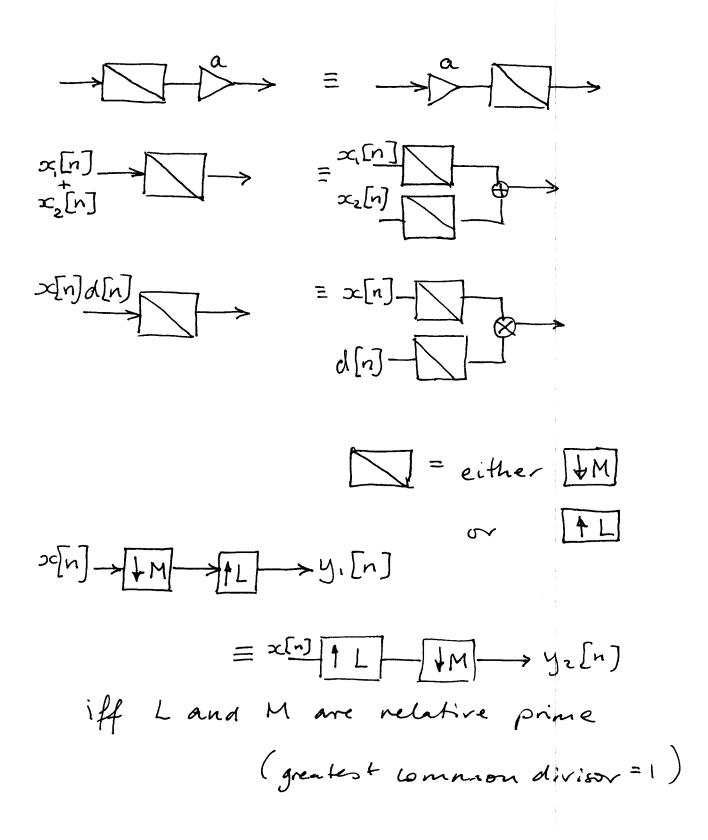


Upsampling

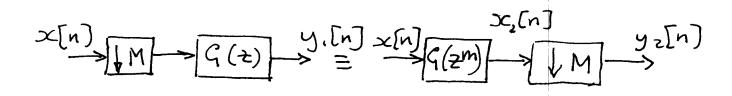


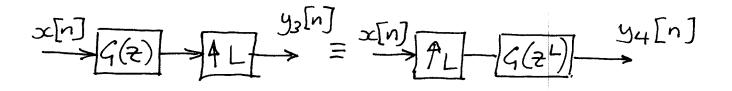
Resampling by a Rational Fraction L/M -> Interpolate >> decimate >> by L by M $x[n] \rightarrow [L] \rightarrow H_{L}(z) \rightarrow H_{d}(z) \rightarrow JM \rightarrow y[n]$ this block an be simplified $x[n] \rightarrow H(z) \rightarrow H(z) \rightarrow y[n]$ H(z) is designed to attenuate the images of x[n] created by the expander If L>M, the system will not introduce aliasing (H(Z) designed as stated) If M > L, x[n] must be bandlimited to the "new" nyquist rate - either intrinsically or by H(2) Sufficient conditions on H(2) new $|H(e^{j\omega})|=0, W >, W \leq n$ Sampting freg. 17

Some Multirate Identities



The Noble Identitie,





since

$$Y_{i}(z) = \frac{1}{M} \sum_{k=0}^{M-1} \chi(z^{1/m} \omega^{k}) G(z)$$

Holyphase Representation 4 Filters

Motivation:

- efficient realisation of filters for multicate DSP.

Consider a classical FIR filter followed by a down-by-two decimator $\Sigma[n] = \frac{1}{2} \frac{$

Consider the fire example

$$\frac{Y(2)}{X(2)} = H(2) = 1 + h(1) \overline{z}^{-1} + h(2) \overline{z}^{-2} + h(3) \overline{z}^{-3}$$

$$y[n] = x[n] + h(1)x[n-1] + h(2)x[n-2] + h(3)x[n-3]$$

$$y[n+1] = x[n+1] + h(1)x[n] + h(2)x[n-1] + h(3]x[n-2)$$

$$y[n+2] = x[n+2] + h(1)x[n+1] + h(2)x[n] + h(3)x[n-1]$$

$$y[n+3] = x[n+3] + h(1)x[n+2] + h(2)x[n+1] + h(2)x[n]$$

$$y[n+4] = x[n+4] + h(1)x[n+3] + h(2)x[n+1] + h(3)x[n+1]$$

$$\vdots$$

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. . .

$$\therefore Y(2) = \sum_{k} \chi(2) \overline{z}^{2k} h(2k) + \sum_{k} \chi(2) \overline{z}^{(2k+1)} h(2k+1)$$

where $Y(2) = Z \{ y[2n] \}$
 $\chi(2) = Z \{ x[2n] \}$

$$H(z) = \frac{Y(z)}{X(z)} = \sum_{k} h(2k) z^{-2k} + z^{-1} \sum_{k} h(2k+1) z^{-2k}$$
$$= E_0(z^2) + z^{-1} E_1(z^2)$$

This is the two-phase decomposition of M(2).

Summary and Generalisation
Twophone?
For any filter
$$H(z) = \sum_{k=-\infty}^{\infty} h(k) z^{-k}$$

We can write
 $H(z) = \sum_{k=-\infty}^{\infty} h(2k) z^{2k} + z^{-1} \sum_{k=-\infty}^{\infty} h(2k+1) z^{-2k}$
 $= E_6(z^2) + z^{-1} E_1(z^2)$
 $f_{rr} E_6(z) = \sum_{k=-\infty}^{\infty} h(2k) z^{-k}$
 $E_1(z) = \sum_{k=-\infty}^{\infty} h(2k+1) z^{-k}$

$$\frac{M - h^{2} h^{2}}{M - h^{2}} + 2^{-1} \sum_{k=-\infty}^{\infty} h(kM) z^{-kM} + 2^{-1} \sum_{k=-\infty}^{\infty} h(kM + 1) z^{-kM} + 2^{-(M-1)} \sum_{k=-\infty}^{\infty} h(kM + 1) z^{-M} + 2^{-(M-1)} \sum_{k=-\infty}^{\infty} h(kM + M - 1) z^{-M}$$

or more compactly:

$$H(z) = \sum_{l=0}^{M-1} z^{-l} E_{l}(z^{M})$$

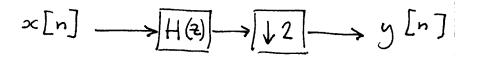
$$E_{l}(z) = \sum_{k=-\infty}^{\infty} e_{l}(k) z^{-k}$$

$$e_l(n) = h(kM + l) \quad o_i \leq M - i$$

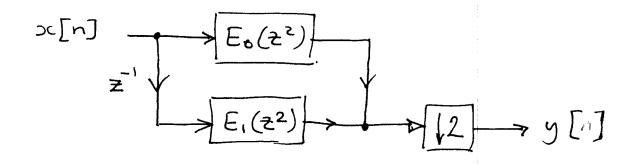
- called Type 1 polyphase representation.

Efficient Polyphane Structures

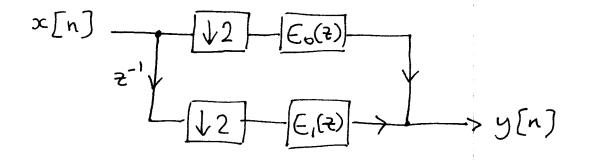
Consider the decimation filter for M=2



In polyphase representation $H(z) = E_0(z^2) + z^2 E_1(z^2)$

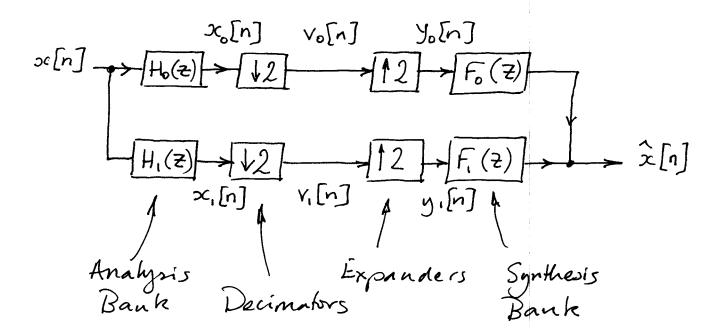


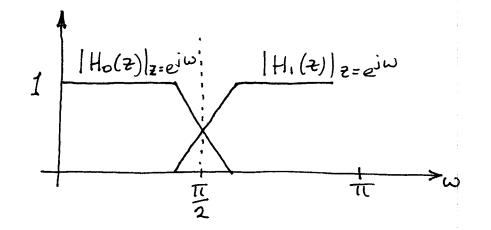
From the Noble Identities we obtain



- the filtering is performed at the lower rate.

Here we fours on the filterbanks, not the processing.





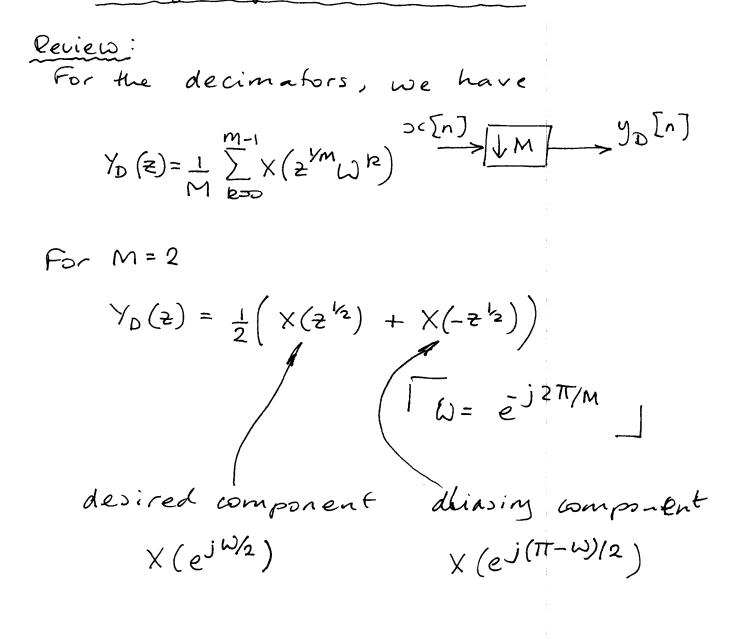
- Extension to M bands : H(2) bandpass.

When
$$|H_0(z)|$$
 and $|H_1(z)|$ are
mirror images around π/z , they
are said to be RMF (Quadrature
Mirror Filters)
 $H_1(z)$ can be designed to mirror
 $H_0(z)$ by setting
 $H_1(z) = H_0(-z)$
i.e. $h_1[n] = h_0[n] \cdot (-1)^n$

where h[n] is the impulse response of H(2).

Errors introduced by the filter banks. In a perfect reconstruction system: $\hat{c}[n] = cx[n-n_0]$ -reconstruction is exactly to within a multiplicative constant and a delay. Otherwise, 3 possible types of distortion: · aliasing - xo[n] and x.[n] are not sufficiently band limited · amplitude distortion $|T(e^{j\omega})| = \left|\frac{\hat{X}(z)}{X(z)}\right| \neq C$ · phase distortion (non-linear phase) $\left(T(ej\omega) = +an^{-1} \left(\frac{Im \left\{ \frac{\hat{X}(z)}{\chi(z)} \right\}_{z=e^{j\omega}}}{Re \left\{ \frac{\hat{X}(z)}{\chi(z)} \right\}_{z=e^{j\omega}}} \right) \neq a + b\omega$ $\left(\frac{Re \left\{ \frac{\hat{X}(z)}{\chi(z)} \right\}_{z=e^{j\omega}}}{Re \left\{ \frac{\chi(z)}{\chi(z)} \right\}_{z=e^{j\omega}}} \right) = a + b\omega$ $\left(\frac{Re \left\{ \frac{\chi(z)}{\chi(z)} \right\}_{z=e^{j\omega}}}{Re \left\{ \frac{\chi(z)}{\chi(z)} \right\}_{z=e^{j\omega}}} \right) = a + b\omega$

Analysis of the 2 band case



For the expander, we have

 $Y_{\varepsilon}(z) = X(z^{L}) \xrightarrow{f_{1}} \xrightarrow{f_{1}} y_{\varepsilon}[n]$

$$\begin{aligned} \text{Using these relationships:} \\ \chi_{k}(z) &= H_{k}(z) \; \chi(z) \qquad k = 0, 1 \\ V_{k}(z) &= \frac{1}{2} \left(\chi_{k}(z^{k}) + \chi_{k}(-z^{k}) \right) \qquad k = 0, 1 \\ V_{k}(z) &= V_{k}(z^{2}) = \frac{1}{2} \left(\chi_{k}(z) + \chi_{k}(-z) \right) \\ &= \frac{1}{2} \left(H_{k}(z) \chi(z) + H_{k}(-z) \chi(-z) \right) \qquad k = 0, 1. \end{aligned}$$

.....

Output signal:

$$\hat{\chi}(z) = F_{o}(z)Y_{o}(z) + F_{i}(z)Y_{i}(z)$$

$$= \frac{1}{2}(H_{o}(z)F_{o}(z) + H_{i}(z)F_{i}(z))X(z)$$

$$+ \frac{1}{2}(H_{o}(z)F_{o}(z) + H_{i}(z)F_{i}(z))X(z)$$

Condition for Alias Cancella Fion.

We require

$$H_o(-z)F_o(z) + H_i(-z)F_i(z) = 0$$

 $\therefore F_o(z) = H_i(-z)$
 $F_i(z) = -H_o(-z)$

Amphitude and Phase Distortions.

If we write

$$\hat{X}(z) = T(z) X(z)$$

and assume the alias cancelling conditions are satisfied then

$$T(z) = \frac{1}{2} \left(H_{o}(z)F_{o}(z) + H_{i}(z)F_{i}(z) \right)$$
$$= \frac{1}{2} \left(H_{o}(z)H_{i}(z) - H_{i}(z)H_{o}(-z) \right)$$

For no amplitude distortion we require

$$|T(z)|_{z=e^{j\omega}} = d$$

is all pass

For no phase distortion we require

$$\left| \sum_{z=e^{j\omega}} = a + b\omega \right|$$

Perfect Reconstruction

- alias cancellation (or no alianing generated) - no amplitude distortion - no phase distortion

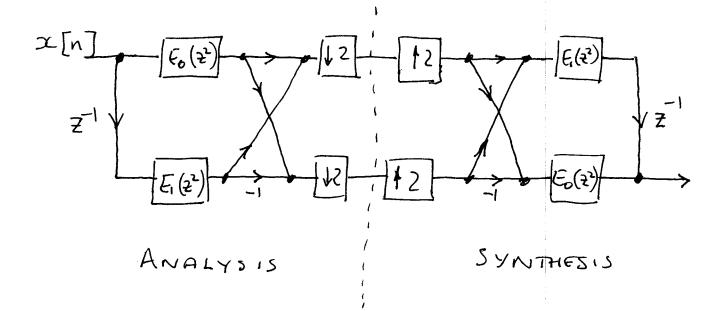
 $\hat{X}(z) = c z^{-n_o} X(z)$

Pohyphase Representation

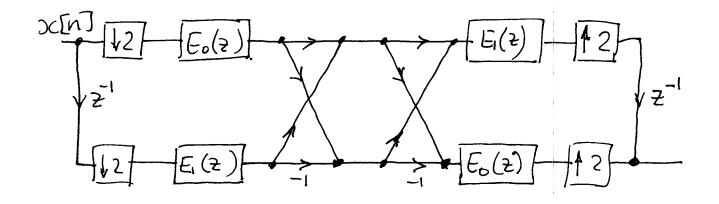
· now apply the polyphase representation to the RME filter bank.

 $H_{o}(z) = E_{o}(z^{2}) + z^{-1} E_{i}(z^{2})$ $H_{i}(z) = E_{o}(z^{2}) - z^{-1} E_{i}(z^{2})$ $F_{o}(z) = E_{o}(z^{2}) + z^{-1} E_{i}(z^{2})$ $F_{i}(z) = -E_{o}(z^{2}) + z^{-1} E_{i}(z^{2})$

In matrix form: $\begin{bmatrix} H_{0}(z) \\ H_{1}(z) \end{bmatrix} = \begin{bmatrix} I & I \\ I & -I \end{bmatrix} \cdot \begin{bmatrix} E_{0}(z^{2}) \\ z^{-1} E_{1}(z^{2}) \end{bmatrix}$ $\begin{bmatrix} F_{0}(z) \\ F_{1}(z) \end{bmatrix} = \begin{bmatrix} I & I \\ I & -I \end{bmatrix} \cdot \begin{bmatrix} z^{-1} E_{1}(z^{2}) \\ E_{0}(z^{2}) \end{bmatrix}$







Note that all filtening is done at the lower sampling rate.