# Multi-rate Signal Processing 

Prof. Min Wu<br>University of Maryland, College Park<br>minwu@umd.edu

Updated: September 19, 2011

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu.
The LaTeX version includes contributions from Mr. Wei-Hong Chuang.

## Outline of Part-I: Multi-rate Signal Processing

§1.1 Building blocks and their properties
§1.2 Properties of interconnection of multi-rate building blocks
§1.3 Polyphase representation
§1.4 Multistage implementation
§1.5 Applications (brief): digital audio system; subband coding
§1.6 Quadrature mirror filter bank (2-channel)
§1.7 M-channel filter bank
§1.8 Perfect reconstruction filter bank
§1.9 Aliasing free filter banks
§1.10 Application: multiresolution analysis
Ref: Vaidyanathan tutorial paper (Proc. IEEE '90);
Book §1, §4, §5.

## Single-rate v.s. Multi-rate Processing

- Single-rate processing: the digital samples before and after processing correspond to the same sampling frequency with respect to (w.r.t.) the analog counterpart.
e.g.: LTI filtering can be characterized by the freq. response.
- The need of multi-rate:
- fractional sampling rate conversion in all-digital domain: e.g. 44.1 kHZ CD rate $\Longleftrightarrow 48 \mathrm{kHZ}$ studio rate
- The advantages of multi-rate signal processing:
- Reduce storage and computational cost
- e.g.: polyphase implementation
- Perform the processing in all-digital domain without using analog as an intermediate step that can:
- bring inaccuracies - not perfectly reproducible
- increase system design / implementation complexity


## Basic Multi-rate Operations: Decimation and Interpolation

- Building blocks for traditional single-rate digital signal processing: multiplier (with a constant), adder, delay, multiplier (of 2 signals)

- New building blocks in multi-rate signal processing:

M-fold decimator


L-fold expander


Readings: Vaidyanathan Book §4.1; tutorial Sec. II A, B

M-fold Decimator

$$
y_{D}[n]=x[M n], M \in \mathbb{N}
$$

e.g. $M=2$


Corresponding to the physical time scale, it is as if we sampled the original signal in a slower rate when applying decimation.

Questions:

- What potential problem will this bring?
- Under what conditions can we avoid it?
- Can we recover $x[n]$ ?


## L-fold Expander

$y_{E}[n]=\left\{\begin{array}{ll}x[n / L] & \text { if } n \text { is integer multiple of } L \in \mathbb{N} \\ 0 & \text { otherwise }\end{array} \quad x[n] \longrightarrow \uparrow L \longrightarrow Y[n]\right.$


Question: Can we recover $x[n]$ from $y_{E}[n]$ ? $\rightarrow$ Yes.

The expander does not cause loss of information.

Question: Are $\uparrow L$ and $\downarrow M$ linear and shift invariant?

## Transform-Domain Analysis of Expanders

Derive the Z-Transform relation between the Input and Output:

## Input-Output Relation on the Spectrum

$$
\mathbb{Y}_{E}(z)=\mathbb{X}\left(z^{L}\right)
$$

Evaluating on the unit circle, the Fourier Transform relation is:

$$
\mathbb{Y}_{E}\left(e^{j \omega}\right)=\mathbb{X}\left(e^{j \omega L}\right) \quad \Rightarrow \quad \mathbb{Y}_{E}(\omega)=\mathbb{X}(\omega L)
$$

i.e. $L$-fold compressed version of $\mathbb{X}(\omega)$ along $\omega$


## Periodicity and Spectrum Image

The Fourier Transform of a discrete-time signal has period of $2 \pi$. With expander, $\mathbb{X}(\omega L)$ has a period of $2 \pi / L$.

The multiple copies of the compressed spectrum over one period of $2 \pi$ are called images.

And we say the expander creates an imaging effect.

## Transform-Domain Analysis of Decimators

$$
\mathbb{Y}_{D}(z)=\sum_{n=-\infty}^{\infty} y_{D}[n] z^{-n}=\sum_{n=-\infty}^{\infty} x[n M] z^{-n}
$$

Define $x_{1}[n]=\left\{\begin{array}{ll}x[n] & \text { if } n \text { is integer multiple of } M \\ 0 & \text { O.W. }\end{array}\right.$, then we have

$$
\begin{aligned}
& \mathbb{Y}_{D}(z)=\mathbb{X}_{1}\left(z^{\frac{1}{M}}\right) \\
& \mathbb{X}_{1}(z)=\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(W_{M}^{k} z\right)
\end{aligned}
$$

## Transform-Domain Analysis of Decimators

Putting all together:

$$
\begin{aligned}
& \mathbb{Y}_{D}(z)=\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(W_{M}^{k} z^{\frac{1}{M}}\right) \\
& \mathbb{Y}_{D}(\omega)=\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(\frac{\omega-2 \pi k}{M}\right)
\end{aligned}
$$

## Frequency-Domain Illustration of Decimation

Interpretation of $\mathbb{Y}_{D}(\omega)$

Step-1: stretch $\mathbb{X}(\omega)$ by a factor of $M$ to obtain $\mathbb{X}(\omega / M)$

Step-2: create $M-1$ copies and shift them in successive amounts of $2 \pi$

Step-3: add all $M$ copies together and multiply by $1 / M$.


## Aliasing

- The stretched version $\mathbb{X}(\omega / M)$ can in general overlap with its shifted replicas. This overlap effect is called aliasing.
- When aliasing occurs, we cannot recover $x[n]$ from the decimated version $y_{D}[n]$, i.e. $\downarrow M$ can be a lossy operation.
- We can avoid aliasing by limiting the bandwidth of $x[n]$ to

$$
|\omega|<\pi / M
$$

- When no aliasing, we can recover $x[n]$ from the decimated version $y_{D}[n]$ by using an expander, followed by filtering of the unwanted spectrum images.


## Example of Recovery from Decimated Signal


$y[n]=x[n]$ where no aliasing occurs.

freq.-domain interpretation

Question: Is the bandlimit condition $|\omega|<\pi / M$ necessary?

Decimation Filters

The decimator is normally preceded by a lowpass filter called decimator filter.

Decimator filter ensures the signal to be decimated is bandlimited and controls the extent of aliasing.


Interpolation Filters

An interpolation filter normally follows an expander to suppress all the images in the spectrum.


time-domain interpretation

## Fractional Sampling Rate Conversion

So far, we have learned how to increase or decrease sampling rate in the digital domain by integer factors.

Question: How to change the rate by a rational fraction $L / M$ ?

$$
\text { (e.g.: audio } 44.1 \mathrm{k} \Longleftrightarrow 48 \mathrm{k} \text { ) }
$$

- Method-1: convert into an analog signal and resample
- Method-2: directly in digital domain by judicious combination of interpolation and decimation

Question: Decimate first or expand first? And why?


Fractional Rate Conversion


Use a low pass filter with passband greater than $\pi / 3$ and stopband edge before $2 \pi / 3$ to remove images



Equiv. to getting 2 samples out of every 3 original samples

- the signal now is critically sampled
- some samples kept are interpolated from $x[n]$

Time Domain Descriptions of Multirate Filters
Recall:


## Summary of Time Domain Description

Input-output relation in the time domain for three types of multirate filters:

$$
y[n]= \begin{cases}\sum_{k=-\infty}^{\infty} x[k] h[n M-k] & \text { M-fold decimation filter } \\ \sum_{k=-\infty}^{\infty} x[k] h[n-k L] & \text { L-fold interpolation filter } \\ \sum_{k=-\infty}^{\infty} x[k] h[n M-k L] & M / L \text {-fold decimation filter }\end{cases}
$$

Note: Systems involving expander and decimator (plus filters) are in general linear time-varying (LTV) systems.

Digital Filter Banks
A digital filter bank is a collection of digital filters, with a common input or a common output.


- $H_{i}(z)$ : analysis filters
- $x_{k}[n]$ : subband signals
- $F_{i}(z)$ : synthesis filters Analysis Bank Synthesis Bank
- Typical frequency response for analysis filters: can be

- marginally overlapping
- non-overlapping
- (substantially) overlapping


## Review: Discrete Fourier Transform

Recall:
M-point DFT
time-domain
discrete periodic
$\left\{\begin{array}{l}\text { DFT: } \mathbb{X}[k]=\sum_{n=0}^{M-1} x[n] W^{n k} \\ \text { IDFT: } x[n]=\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}[k] W^{-n k}\end{array}\right.$
$\Rightarrow \quad \begin{aligned} & \text { frequency-domain } \\ & \text { discrete periodic }\end{aligned}$

$$
\left(W=e^{-j 2 \pi / M}\right)
$$

- Subscript is often dropped from $W_{M}$ if context is clear
- The $M \times M$ DFT matrix $\mathbf{W}$ is defined as $[\mathbf{W}]_{k n}=W^{k n}$
- We use $\mathbf{W}^{*}$ to represent the conjugate of $\mathbf{W}$; also note $\mathbf{W}=\mathbf{W}^{T}$ (symmetric)


## DFT Filter Bank

Consider passing $x[n]$ through a delay chain to get $M$ sequences $\left\{s_{i}[n]\right\}: s_{i}[n]=x[n-i]$

i.e., treat $\left\{s_{i}[n]\right\}$ as a vector $\underline{s}[n]$, then apply $\mathbf{W}^{*} \underline{s}[n]$ to get $\underline{x}[n]$.

Question: What are the equiv. analysis filters?

## Input-Output Relation of DFT Filter Bank

## Relation between $H_{i}(z)$

Uniform DFT Filter Bank
A filter bank in which the filters are related by

$$
H_{k}(z)=H_{0}\left(z W^{k}\right)
$$

is called a uniform DFT filter bank.



The response of filters $\left|H_{k}(\omega)\right|$ have a large amount of overlap.

## Time-domain Interpretation of the Uniform DFT FB

## Time-domain Interpretation of the Uniform DFT FB

The DFT filter bank can be thought of as a spectrum analyzer

- The output $\left\{x_{k}[n]\right\}_{k=0}^{M-1}$ is the spectrum captured based on the most recent $M$ samples of the input sequence $x[n]$.
- The filters themselves are not very good: wide transition bands and poor stopband attenuation of only 13 dB - due to the simple rectangular sliding window $H_{0}(z)$.

Question: How can we improve the filters in the uniform DFT filter bank?

## Interconnection of Building Blocks: Basic Properties

Basic interconnection properties:


Readings: Vaidyanathan Book §4.2; tutorial Sec. II B

## Decimator-Expander Cascades



## Questions:

(1) Is $y_{1}[n]$ always equal to $y_{2}[n]$ ? Not always.
E.g., when $L=M, y_{2}[n]=x[n]$, but
$y_{1}[n]=x[n] \cdot c_{M}[n] \neq y_{2}[n]$, where $c_{M}[n]$ is a comb sequence
(2) Under what conditions $y_{1}[n]=y_{2}[n]$ ?

## Example of Decimator-Expander Cascades


$\mathrm{L}=3, \mathrm{M}=2$


x 1



## Condition for $y_{1}[n]=y_{2}[n]$

Examine the ZT of $y_{1}[n]$ and $y_{2}[n]$ : (details)

## Condition for $y_{1}[n]=y_{2}[n]$

Equiv. to examine the condition of $\left\{W_{M}^{k}\right\}_{k=0}^{M-1} \equiv\left\{W_{M}^{k L}\right\}_{k=0}^{M-1}$ : iff $M$ and $L$ are relatively prime.

Question: Prove it. (see homework).
$\Rightarrow$ Thus the outputs of the two decimator-expander cascades, $\mathbb{Y}_{1}(z)$ and $\mathbb{Y}_{2}(z)$, are identical and $(a) \equiv(b)$ iff $M$ and $L$ are relatively prime.

## The Noble Identities

Recall: the cascades of decimators and expanders with LTI systems appeared in decimation and interpolation filtering.

## Question:


$\Rightarrow$ Generally "No".
Observations:

$$
\text { (1) } \rightarrow \delta^{-1} \rightarrow \text { WM } \rightarrow \neq \rightarrow M \rightarrow \delta^{-1} \rightarrow(\text { for } M>1) \text { by slift vatiance. }
$$

The Noble Identities

Consider a LTI digital filter with a transfer function $G(z)$ :

(b)


Recall: the transfer function $G(z)$ of a LTI digital filter is rational for practical implementation, ie., a ratio of polynomials in $z$ or $z^{-1}$. There should not be terms with fractional power in $z$ or $z^{-1}$.

## Proof of Noble Identities

## Detailed Derivations

Transform-Domain Analysis of Expanders

Z-Transform Relation between the Input and Output:

$$
\mathbb{Y}_{E}(z)=\mathbb{X}\left(z^{L}\right)
$$

Proof:

$$
\begin{aligned}
Y_{E}(z) & =\sum_{n=-\infty}^{+\infty} y_{E}[n] z^{-n}=\sum_{n=k L, K \in Z} y_{E}[n] z^{-n} \\
& =\sum_{k=-\infty}^{+\infty} y_{E}[k L] z^{-K L} \\
& =\sum_{k=-\infty}^{+\infty} x[K]\left(z^{L}\right)^{-K}=\bar{X}\left(z^{L}\right)
\end{aligned}
$$

Transform-Domain Analysis of Decimator

$$
\begin{aligned}
& \bar{Y}(z)=\sum_{n=-\infty}^{+\infty} y_{D}^{[n]} z^{-n}=\sum_{n=-\infty}^{+\infty} x[n M] z^{-n} \\
& \stackrel{x[n]}{\longrightarrow} \xrightarrow{\substack{\text { multiply } \\
\text { with } c_{m}[n]}} \xrightarrow{x_{1}[n]} \text { "inverse" } \\
& \text { expansion }
\end{aligned}
$$

Define $X_{1}[n]= \begin{cases}X[n] & \text { if } n \text { is integer multiple of } M \\ 0 & 0, W .\end{cases}$
We have $Y_{D}(z)=\sum_{K=n M, n \in Z} x[k] z^{-K / M}=\sum_{K=-\infty}^{+\infty} x[k]\left[z^{1 / M}\right)^{-K}=\bar{X}_{1}\left(z^{1 / M}\right)$ change variable

Transform-Domain Analysis of Decimator

To establish the transform-dowain relation between $X_{1}(z)$ and $X(z)$ :
we note $x_{1}[n]$ can be written as

$$
X_{1}[n]=C_{m}[n] X[n]
$$

where $C M[n]= \begin{cases}1 & \text { if } n \text { is integer multiple of } M \\ 0 & 0, w\end{cases}$ ("com b"sequence) 10 ow.

Thick Using the $M^{\text {th root of unity Wm defined as }}$

$$
W_{M}=e^{-j^{2 \pi / M}}
$$

We have

$$
C_{M}[n]=\frac{1}{M} \sum_{K=0}^{M-1} W_{M}^{-k n}
$$

Transform-Domain Analysis of Decimators

By the definition of $Z T$ :

$$
\left.\begin{array}{rl}
\bar{X}_{1}(z) & =\sum_{n=-\infty}^{+\infty} x_{1}[n] z^{-n}=\sum_{n=-\infty}^{+\infty} C M[n] X[n] z^{-n} \\
& =\sum_{n=-\infty}^{+\infty} \frac{1}{M} \sum_{k=0}^{M-1} w_{M}^{-k n} x[n] z^{-n} \\
& =\frac{1}{M} \sum_{k=0}^{M-1} \sum_{n=-\infty}^{+\infty} x[n]\left(z \cdot w_{M}^{k}\right)^{-n} \\
& =\frac{1}{M} \sum_{k=0}^{M-1} X\left(W_{M}^{k} \cdot z\right) \\
\therefore F_{D}(z) & =\bar{X}_{1}\left(z^{1 / M}\right)
\end{array}=\frac{1}{M} \sum_{k=0}^{M-1} X\left(w_{\mu}^{k} z^{1 / M}\right)\right] .
$$

Transform-Domain Analysis of Decimators

$$
\begin{aligned}
\text { Fourier Spectrum } & =\text { set } z=e^{j \omega} \\
T_{D}\left(e^{j \omega}\right) & =\frac{1}{M} \sum_{k=0}^{M-1} X\left(e^{-j 2 \pi k / M} e^{j \omega / M}\right) \\
& =\frac{1}{M} \sum_{k=0}^{M-1} X\left(e^{j(\omega-2 \pi k) / M}\right) \\
\therefore T_{D}(\omega) & =\frac{1}{M} \sum_{k=0}^{M-1} X\left(\frac{\omega-2 \pi k}{M}\right)
\end{aligned}
$$

Time Domain Descriptions of Multirate Filters
Recall:

1


$$
\begin{aligned}
W[n]= & \sum_{k} h[k] x[n-k]=\sum_{k} h[n-k] x[k] \\
y[m]=W[M m] & =\sum_{k} h[k] \times[M m-k] \\
& =\sum_{k} h[M m-k] x[k]
\end{aligned}
$$

(2)

$w[m]= \begin{cases}x[m / L] & \text { if } m \text { is multiple of } L \\ 0 & \text { on. }\end{cases}$
only keep nonzero $\mathrm{W}[\mathrm{m}]$ for all $K=K^{\prime}$ ᄂ

Input-Output Relation of DFT Filter Bank
$\qquad$
Define this as $H_{K}(z)$, we have: (the transfer functions)

Relation between $H_{i}(z)$

$$
H_{0}(z)=\sum_{i=0}^{M-1} z^{-i} \text { ie. } z T o f a \text { rectangular Window }
$$



$$
\begin{aligned}
& \longrightarrow\left|H_{0}(\omega)\right|=\left|\frac{\sin (M \omega / 2)}{\sin (\omega / 2)}\right| \\
H_{k}(z) & =H_{0}\left(\omega^{k} z\right)=H_{0}\left(e^{-j 2 \pi k / M z)}\right. \\
& \xrightarrow[z=j^{j \omega}]{\longrightarrow} H_{k}(\omega)
\end{aligned}=H_{0}\left(\omega-\frac{2 \pi k}{M}\right) .
$$

ie. uniformly shifted version of $H_{0}(\omega)$ spectmin

Time-domain Interpretation of the Uniform DFT FB

$$
X_{k}[n+M-1]=\sum_{i=0}^{M-1} x[n+M-1-i] W^{-k i} \quad l l l=M-1-i
$$

Here we consider

$$
\begin{aligned}
& \begin{array}{l}
\text { Here we consider } \\
\text { a delayed } \\
\text { version of } \times-k[] \\
\text { for convenience. }
\end{array}=\sum_{\ell=0}^{M-1} \times[n+l] W^{K l} \times W^{-K(M-1)} \quad \text { Note } W^{M}=1
\end{aligned}
$$

$$
=\underbrace{W^{k}} \sum_{l=0}^{M-1} x[n+l] W^{k l}
$$

phase shift $\underbrace{}_{k^{\text {th }}}$ point of the M-point DFT (Limearphase term reflects of $\{x[n], \ldots, x[n+M-1]\}$ time delay: $\left.e^{-j \omega}\right|_{\omega=\frac{2 \pi}{M} k}$ )

Condition for $y_{1}[n]=y_{2}[n]$

Examine the ZT of $y_{1}[n]$ and $y_{2}[n]$ :

$$
\left.\begin{array}{l}
\nabla_{1}(z)=Z_{1}\left(z{ }^{2}\right) \\
\mathbb{Z}_{1}(z)=\frac{1}{M} \sum_{k=1}^{1} X\left(W_{M}^{k} z^{1 / M}\right)
\end{array}\right\} \Rightarrow \bar{Y}_{1}(z)=\frac{1}{M} \sum_{k=0}^{M-1} \Psi\left(W_{M}^{k} z^{L / M}\right)
$$

$W_{M}^{K}=e^{-j \frac{2 \pi K}{M}}, K=0, \ldots M-1$ are $M$ distinct $M^{\text {th }}$ roots of unity
$W_{M}^{K L}=e^{-j \frac{2 \pi K L}{M}}, K=0, \ldots M-1$ may not represent $M$ distinct numbers when $L$ and $M$ share common factors.

Proof of Noble Identities

Proof:

$$
\begin{aligned}
& \bar{Y}_{2}(z)=\frac{1}{M} \sum_{k=0}^{M-1} \bar{X}_{2}\left(W_{M}^{k} z^{1 / M}\right) \\
& X_{2}(z)=G\left(z^{M}\right) X(\gamma) \Rightarrow X_{2}\left(W_{M}^{k} z^{1 / M}\right)=G\left(W_{M}^{M k} z\right) \cdot X\left(W_{M}^{k} z^{\prime / M}\right) \\
& \therefore \bar{T}_{2}(z)=\frac{1}{M} \sum_{K=0}^{M-1} G(z) X\left(W_{M}^{k} z^{1 / M}\right) \\
& =G(z) \nabla_{1}(z)=\nabla_{1}(z)
\end{aligned}
$$

(b)

$$
\begin{aligned}
\bar{Y}_{4}(z) & =G\left(z^{L}\right) X_{4}(z)=G\left(z^{L}\right) \Phi\left(z^{L}\right) \\
\left.\begin{array}{rl}
Y_{3}(z) & =\Phi_{3}\left(z^{L}\right) \\
X_{3}(z) & =G(z) \Phi(z)
\end{array}\right\} \Rightarrow \bar{Y}_{3}(z) & =G\left(z^{L}\right) X\left(z^{L}\right) \\
& =Y_{4}(z)
\end{aligned}
$$

## Equiv. to examine for $\left\{W_{M}^{k}\right\}_{k=0}^{M-1} \equiv ? \equiv\left\{W_{M}^{k L}\right\}_{k=0}^{M-1}$

$\left\{W_{M}^{k}\right\}_{k=0}^{M-1} \equiv\left\{W_{M}^{k L}\right\}_{k=0}^{M-1}$ ff $M$ and $L$ are relatively prime.
(1) " $\Rightarrow$ ". Prove by contradiction, i.e. let $L$ and $M$ have a common factor $\alpha \geqslant 2$ st. $M \equiv \alpha m$ for some integer $m<M$, and $L \equiv \alpha l$ for some $l<L$.
$m L=m \alpha l=M \cdot l \Rightarrow m L \bmod M=0$
i.e. the set $\{0, L, 2 L, \ldots(M-1) L\}$ mod $M$ has at most $M-1$ distinct elements, thus $\neq\{0,1, \ldots,(M-1)\}$ Contradict with the given condition
(2) " $\Leftrightarrow$ ". Prove by contradiction, i.e. suppose the two sets are different, where $\exists K_{1} \& K_{2}$ sit. $0 \leq K_{2}<K_{1} \leq M-1$ and $K_{1} L \bmod M=K_{2} L \bmod M$.
This means $\exists$ some integers $a_{1}$ and $a_{2}$, we have $\begin{aligned} & K_{1} L-a_{1} M=K_{2} L-a_{2} M \text { and }\left\{\begin{array}{l}\left(K_{1}-1\right) L<a_{1} M \leqslant K_{1} L \\ \left(K_{2}-1\right) L<a_{2} M \leqslant K_{2} L\end{array}\right. \\ & \Downarrow\end{aligned}$
$\because\left(a_{1}-a_{2}\right)$ is an integer $\therefore\left(k_{1}-k_{2}\right) L$ should be multiples of $M$
$\because 0<\left(K_{1}-K_{2}\right) \leq M-1 \therefore L$ must have Some common factors with $M$. $\Rightarrow$ Contradiction

# Multi-rate Signal Processing 3. The Polyphase Representation 

Prof. Min Wu<br>University of Maryland, College Park<br>minwu@umd.edu

Updated: September 19, 2011

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu.
The LaTeX version includes contributions from Mr. Wei-Hong Chuang.

## Polyphase Representation: Basic Idea

Example: FIR filter $H(z)=1+2 z^{-1}+3 z^{-2}+4 z^{-3}$
Group even and odd indexed coefficients, respectively:
$\Rightarrow H(z)=\left(1+3 z^{-2}\right)+z^{-1}\left(2+4 z^{-2}\right)$,

More generally: Given a filter $H(z)=\sum_{n=-\infty}^{\infty} h[n] z^{-n}$, by grouping the odd and even numbered coefficients, we can write

$$
H(z)=\sum_{n=-\infty}^{\infty} h[2 n] z^{-2 n}+z^{-1} \sum_{n=-\infty}^{\infty} h[2 n+1] z^{-2 n}
$$

## Polyphase Representation: Definition

$H(z)=\sum_{n=-\infty}^{\infty} h[2 n] z^{-2 n}+z^{-1} \sum_{n=-\infty}^{\infty} h[2 n+1] z^{-2 n}$
Define $E_{0}(z)$ and $E_{1}(z)$ as two polyphase components of $H(z)$ :

$$
\begin{gathered}
E_{0}(z)=\sum_{n=-\infty}^{\infty} h[2 n] z^{-n} \\
E_{1}(z)=\sum_{n=-\infty}^{\infty} h[2 n+1] z^{-n}
\end{gathered}
$$

We have

$$
H(z)=E_{0}\left(z^{2}\right)+z^{-1} E_{1}\left(z^{2}\right)
$$

- These representations hold whether $H(z)$ is FIR or IIR, causal or non-causal.
- The polyphase decomposition can be applied to any sequence, not just impulse response.


## FIR and IIR Example

(1) FIR filter: $H(z)=1+2 z^{-1}+3 z^{-2}+4 z^{-3}$

$$
\begin{aligned}
& \because H(z)=\left(1+3 z^{-2}\right)+z^{-1}\left(2+4 z^{-2}\right) \\
& \therefore E_{0}(z)=1+3 z^{-1} ; \quad E_{1}(z)=2+4 z^{-1}
\end{aligned}
$$

(2) IIR filter: $H(z)=\frac{1}{1-\alpha z^{-1}}$.

Write into the form of $H(z)=E_{0}\left(z^{2}\right)+z^{-1} E_{1}\left(z^{2}\right)$ :

$$
\begin{aligned}
\because H(z) & =\frac{1}{1-\alpha z^{-1}} \times \frac{1+\alpha z^{-1}}{1+\alpha z^{-1}}=\frac{1+\alpha z^{-1}}{1-\alpha^{2} z^{-2}} \\
& =\frac{1}{1-\alpha^{2} z^{-2}}+z^{-1} \frac{\alpha}{1-\alpha^{-2} z^{-2}}
\end{aligned}
$$

$\therefore E_{0}(z)=\frac{1}{1-\alpha^{2} z^{-1}} ; \quad E_{1}(z)=\frac{\alpha}{1-\alpha^{-2} z^{-1}}$

## Extension to $M$ Polyphase Components

For a given integer $M$ and $H(z)=\sum_{n=-\infty}^{\infty} h[n] z^{-n}$, we have:

$$
\begin{aligned}
H(z)= & \sum_{n=-\infty}^{\infty} h[n M] z^{-n M}+z^{-1} \sum_{n=-\infty}^{\infty} h[n M+1] z^{-n M} \\
& +\cdots+z^{-(M-1)} \sum_{n=-\infty}^{\infty} h[n M+M-1] z^{-n M}
\end{aligned}
$$

## Type-1 Polyphase Representation

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

where the $\ell$-th polyphase components of $H(z)$ given $M$ is

$$
E_{\ell}(z) \triangleq \sum_{n=-\infty}^{\infty} e_{\ell}[n] z^{-n}=\sum_{n=-\infty}^{\infty} h[n M+\ell] z^{-n}
$$

Note: $0 \leq \ell \leq(M-1)$; strictly we may denote as $E_{\ell}^{(M)}(z)$.

Example: $M=3$
eg.





$z^{\ell}$ : time advance
(there is a delay term when putting together the polyphase components)

ENEE630 Lecture Part-1

## Alternative Polyphase Representation

If we define $R_{\ell}(z)=E_{M-1-\ell}(z), 0 \leq \ell \leq M-1$, we arrive at the

## Type-2 polyphase representation

$$
H(z)=\sum_{\ell=0}^{M-1} z^{-(M-1-\ell)} R_{\ell}\left(z^{M}\right)
$$

Type-1: $E_{k}(z)$ is ordered consistently with the number of delays in the input

Type-2: reversely order the filter $R_{k}(z)$ with respect to the delays


## Issues with Direct Implementation of Decimation Filters

Decimation Filters:


Question: Any wasteful effort in the direct implementation?

- The filtering is applied to all original signal samples, even though only every $M$ filtering output is retained finally.
- Even if we let $H(z)$ operates only for time instants multiple of $M$ and idle otherwise, all multipliers/adders have to produce results within one step of time.
- Can $\downarrow M$ be moved before $H(z)$ ?

Only when $H(z)$ is a function of $z^{M}$, we can apply the noble identities to switch the order.

Efficient Structure for Decimation Filter

Apply Type-1 polyphase representation:


## Computational Cost

For FIR filter $H(z)$ of length $N$ :

- The total cost of $N$ multipliers and $(N-1)$ adders is unchanged.
- Considering multiplications per input unit time (MPU) and additions per input unit time (APU),
$E_{k}(z)$ now operates at a lower rate: only $N / M$ MPU and $(N-1) / M$ APU are required.
- This is as opposed to $N$ MPU and $(N-1)$ APU at every $M$ instant of time and system idling at other instants, which leads to inefficient resource utilization.
(i.e., requires use fast additions and multiplications but use them only $1 / M$ of time)


## Polyphase for Interpolation Filters



Observe: the filter is applied to a signal at a high rate, even though many samples are zero when coming out of the expander.

Using the Type-2 polyphase decomposition:

$H(z)=z^{-1} R_{0}\left(z^{2}\right)+R_{1}\left(z^{2}\right):$

- 2 polyphase components
- $R_{k}(z)$ is half length of $H(z)$

The complexity of the system is $N$ MPU and $(N-2)$ APU.

## General Cases

In general, for FIR filters with length $N$ :
$M$-fold decimation:

$$
\mathrm{MPU}=\frac{N}{M}, \mathrm{APU}=\frac{N-1}{M}
$$


filtering is performed at a lower data rate

## $L$-fold decimation:

$\mathrm{MPU}=N, \mathrm{APU}=N-L$


$$
\mathrm{APU}=\left(\frac{N}{L}-1\right) \times L
$$

## Fractional Rate Conversion



- Typically $L$ and $M$ should be chosen to have no common factors greater than 1 (o.w. it is wasteful as we make the rate higher than necessary only to reduce it down later)
- $H(z)$ filter needs to be fast as it operates in high data rate.
- The direct implementation of $H(z)$ is inefficient:
$\left\{\begin{array}{l}\text { there are } L-1 \text { zeros in between its input samples } \\ \text { only one out of } M \text { samples is retained }\end{array}\right.$

3 The Polyphase Representation Appendix: Detailed Derivations
3.1 Basic Ideas
3.2 Efficient Structures
3.3 Commutator Model
3.4 Discussions: Multirate Building Blocks \& Polyphase Concept

## Example: $L=2$ and $M=3$

(1) Use Type-1 polyphase decomposition (PD) for decimator:

(2) Use Type-2 PD for interpolator:


## Example: $L=2$ and $M=3$

(3) Try to take advantage of both:

Question: What's the lowest possible data rate to process? $f / M$

Challenge: Can't move $\uparrow 2$ further to the right and $\downarrow 3$ to the left across the delay terms.

## Trick to enable interchange of $\uparrow L$ and $\downarrow M$

$$
z^{-1}=z^{-3} \cdot z^{2}
$$

- $z^{-3}$ and $z^{2}$ can be considered as filters in $z^{-M}$ and $z^{+L}$
- Noble identities can be applied:

$\sqrt{V}$

can be interchanged as they are relatively prime

3 The Polyphase Representation Appendix: Detailed Derivations
3.1 Basic Ideas
3.2 Efficient Structures
3.3 Commutator Model
3.4 Discussions: Multirate Building Blocks \& Polyphase Concept

## Overall Efficient Structure

Now it becomes

can move decimation earlier by Type-1 PD of $R_{k}(z)$
Finally,

$$
f<\pi \longrightarrow \frac{1}{3} f<\| \quad \frac{2}{3} f
$$

$$
\begin{aligned}
& R_{0}(z)= \\
& R_{00}\left(z^{3}\right)+z^{-1} R_{01}\left(z^{3}\right)+z^{-2} R_{02}\left(z^{3}\right)
\end{aligned}
$$

$$
R_{1}(z)=
$$

$$
R_{10}\left(z^{3}\right)+z^{-1} R_{11}\left(z^{3}\right)+z^{-2} R_{12}\left(z^{3}\right)
$$

## Observations

- For $N$-th order $H(z):$ MPU $=(N+1) / M \Rightarrow$ independent of $L$
- The final structure is the most efficient:
\{ Decimators are moved to the left of all computational units
Expanders are moved to the right of all computational units
Thus the computation is operated at the lowest possible rate.
- The above scheme works for arbitrary integers $L$ and $M$ as long as they are relatively prime.
Under this condition, we have:
(1) $\exists n_{0}, n_{1} \in \mathbb{Z}$ s.t. $n_{1} M-n_{0} L=1$ (Euclid's theorem) We can then decompose $z^{-1}=z^{n_{0} L} z^{-n_{1} M}$
(2) $\uparrow L$ and $\downarrow M$ are interchangeable

Commutator Model: A Delay Chain followed by Decimators

Polyphase implementation is often characterized by
(1) A delay chain followed by a set of decimators,
counterclock wise commutator


Commutator Model: Expanders followed by A Delay Chain
(2) A set of expanders followed by a delay chain


Commutator/switch model is an appealing conceptual tool to visualize these operations

## Discussions: Linear Periodically Time Varying Systems

Some multirate systems that we have seen are linear periodically time varying (LPTV) systems.


$$
\begin{aligned}
y[n] & = \begin{cases}x[n] & \text { if } n \text { is multiple of } M \\
0 & \text { otherwise }\end{cases} \\
& =x[n] \cdot c[n]
\end{aligned}
$$

$c[n]$ is a comb function: takes 1 for $n$ is multiple of $M$ and 0 o.w.
$\Rightarrow$ This is a linear system with periodically time varying response coefficients, and the period is $M$.

## Time-invariant System with Decimator / Expander

Even though $\uparrow L$ and $\downarrow M$ are time-varying, a cascaded system having them as building blocks may become time-invariant.


This structure is the same as a fractional decimation system with $L=M$.

3 The Polyphase Representation Appendix: Detailed Derivations
3.1 Basic Ideas
3.2 Efficient Structures
3.3 Commutator Model
3.4 Discussions: Multirate Building Blocks \& Polyphase Concept

## Time-invariant System with $\uparrow M \& \downarrow M$



## details

Recall: $[\mathbb{X}(z)]_{\downarrow M}=$
$\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(W_{M}^{k} z^{1 / M}\right)$

## Perfect Reconstruction (PR) Systems



The above system is said to be a perfect reconstruction system if $\hat{x}[n]=c x\left[n-n_{0}\right]$ for some $c \neq 0$ and integer $n_{0}$, i.e., the output is identical to the input, except a constant multiplicative factor and some fixed delay.

## Special Time-invariant System with $\uparrow M \& \downarrow M$



## (back)

Recall: $[\mathbb{X}(z)]_{\downarrow M}=$ $\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(W_{M}^{k} z^{1 / M}\right)$

$$
\begin{aligned}
\mathbb{Y}(z) & =\left[\mathbb{X}\left(z^{M}\right) H(z)\right]_{\downarrow_{M}} \\
& =\frac{1}{M} \sum_{k=0}^{M-1} \mathbb{X}\left(W_{M}^{M k} z\right) H\left(W_{M}^{k} z^{1 / M}\right)=\mathbb{X}(z)[H(z)]_{\downarrow M}
\end{aligned}
$$

$[H(z)]_{\downarrow M}$ implies decimating the impulse response $h[n]$ by $M$-fold, corresponding to the 0 -th polyphase component of $H(z)$.

$$
\Rightarrow \mathbb{Y}(z)=\mathbb{X}(z) E_{0}(z) \text {, i.e., } \xrightarrow{x[n]} E_{0}(\xi) \xrightarrow{y[n]} \text {, an LTI system. }
$$

# Multi-rate Signal Processing 4. Multistage Implementations 5. Multirate Application: Subband Coding 

Prof. Min Wu<br>University of Maryland, College Park<br>minwu@umd.edu

Updated: September 21, 2011

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu.
The LaTeX version includes contributions from Mr. Wei-Hong Chuang.

## Preliminaries: Filter's magnitude response



## Filter design theory

A linear phase FIR filter that satisfies this specification has order $N=g\left(\delta_{1}, \delta_{2}, \Delta W\right)$

- as a function of $\delta_{1}, \delta_{2}$, and $\Delta f$
- $\Delta_{f} \approx \frac{\Delta W}{2 \pi}$
(normalized transition b.w. $\in[0,1]$ )

For fixed ripple size, $N \propto \frac{1}{\Delta f}$ :
$\Delta f \uparrow \rightarrow N \downarrow$ (computation $\downarrow$ )

## Doubling Filter Transition Band

Consider an original LPF implementation


If we have a LPF with transition band $2 \Delta f$, we may reduce the order by about half.


Double transition band leads to half of the required order for the filter.

## Interpolated FIR (IFIR)

## Questions:

- With passband and stopband also doubled, what will be the response of a new filter that is an expanded version of the impulse response for $G(z)$, i.e., $G\left(z^{2}\right)$ ?
- What else is needed to get the same system response as $H(z)$ ?

New Interpolated FIR Design:


## Multistage Decimation / Expansion

With what we have in IFIR design, reconsider now the efficient implementation of multirate filters:
e.g.,


$$
M=50
$$

- Narrow passband for $H(z)$ $\Rightarrow$ long filter needed
- Using polyphase representation $\Rightarrow$ need many decomposition components for large $M$ !

How about?


Multistage implementation can be more efficient (in terms of computations per unit time).

## Multistage Decimation / Expansion

Similarly, for interpolation,


## Summary

By implementing in multistage, not only the number of polyphase components reduces, but most importantly, the filter specification is less stringent and the overall order of the filters are reduced.

## Exercises:

- Close book and think first how you would solve the problems.
- Sketch your solutions on your notebook.
- Then read V-book Sec. 4.4.


## IFIR Design

Original system:


New system:

(omit ripples in the sketches)

Doubled transition band leads to half of the required order for the filter

Note the undesired spectrum image

Wide transition band $\Rightarrow$ $I(z)$ can have very low order

* $G(2 \omega) \times I(\omega) \approx H(\omega)$


## Discussions

The complexity of the two-stage implementation is much less than that of the direct implementation.

- $G(z)$ : the model filter (designed according to the "scaled" specification of $H(z)$ )
- I(z): image suppressor
- Number of adders: $N_{i}+N_{g} \ll N$
- Number of multipliers: $\left(N_{i}+1\right)+\left(N_{g}+1\right) \ll(N+1)$


## Principle of IFIR Design

$\Rightarrow$ Motivated multistage design from an efficient design technique of narrowband LPF known as IFIR.

- Applicable for designing any narrowband FIR filter (by itself not tied with $\uparrow L$ or $\downarrow M)$

Readings: Vaidyanathan's Book Sec. 4.4

## Extension to $M \geq 2$

In general, it is possible to stretch more, by an amount $M \geq 2$,


- so that the transition band of $G(z)$ can be even wider $(\approx M \Delta f)$ and further reduces the order $N_{g}$
- Stopband edge in $G(z): M \omega_{s} \leq \pi$
$\Rightarrow M \leq\left\lfloor\frac{\pi}{\omega_{s}}\right\rfloor$


## Extension to $M \geq 2$ : Tradeoff



Tradeoff of the total cost: $\quad M \uparrow$

- $G(z)$ : transition b.w. $\uparrow \rightarrow$ order $\downarrow$
- $I(z)$ : transition b.w. $\downarrow$ (could become very narrow) $\rightarrow$ order $\uparrow$
$\Rightarrow$ can search for optimal $M$.


## Multistage Design of Decimation Filter


polyphase implementation each stage
$M=M_{1} M_{2}:$
Choice of $M_{1}$ can be cast as an optimization problem

Rule of thumb: choose $M_{1}$ larger to reduce the computation complexity \& data rate early on

## Multistage Design Example: (1) Direct Design


polyphase implementation each stage
e.g., $M=50$ fold decimation of an 8 kHz signal
$H(z): \delta_{1}=0.01, \delta_{2}=0.001$, passband edge $=70 \mathrm{~Hz}$, stopband edge $=80 \mathrm{~Hz}$
$\sim$ normalized $\Delta f=\frac{10}{8 k}=\frac{1}{800}$
the order of direct equiripple filter design $\Rightarrow N=2028$

## Multistage Design Example: (2) Two-stage Design

$$
\begin{aligned}
& M_{1}=25, M_{2}=2 \\
& G(z): \Delta f=25 \times \frac{1}{800} \\
& \omega_{p}=0.4375 \pi, \omega_{s}=0.5 \pi, \\
& \delta_{1}=0.005, \delta_{2}=0.001 \\
& \Rightarrow N_{g}=\mathbf{9 0} \\
& I(z): \Delta f=17 \times \frac{1}{800} \\
& \omega_{p}=0.0175 \pi, \omega_{s}=0.06 \pi, \\
& \delta_{1}=0.005, \delta_{2}=0.001 \\
& \Rightarrow N_{i}=\mathbf{1 3 9}
\end{aligned}
$$

higher order than $G(z)$ due to narrower transition

See spectrum sketch in Vaidyanathan's Book, Fig. 4.4-6.

## Interpolation Filter


$L_{1}$ should be small to avoid too much increase in data rate and filter computation at early stage
e.g., $L=50: L_{1}=2, L_{2}=25$

## Summary

By implementing in multistage, not only the number of polyphase components reduces, but most importantly, the filter specification is less stringent and the overall order of the filters are reduced.

## How to compress a signal?

- Tradeoff between bit rate and fidelity
- Many aspects to explore:
use bits wisely; exploit redundancy; discard unimportant parts;
- Allocate bit rate strategically: equal allocation vs. focused effort


## Compression Tool \#1 (lossless if free from aliasing): Downsample a signal of limited bandwidth

(From what we learned about decimation in §1.1)
If a discrete-time signal is bandlimited with bandwidth smaller than $2 \pi$, the signal can be decimated by an appropriate factor without losing information.

- i.e., we don't need to keep that many samples
- Recall the example in $\S 1.1 .1:|\omega|<\frac{2}{3} \pi$ $\Rightarrow$ can change data rate to $\frac{2}{3}$ of original
- If signal spectrum support is in $\left(\omega_{1}, \omega_{1}+\frac{2 \pi}{M}\right)$, we can decimate the signal by $M$ fold without introducing aliasing.
(Decimated signal may extend to entire $2 \pi$ spectrum range)


## Compression Tool \#2 (lossy): Quantization

Dynamic range $A$ of a signal:
the
value
range


- Use a finite number of bits to represent a continuous valued sample via scalar quantization:
partition $A$ into $N$ intervals, pick $N$ representative values and use $\log _{2} N$ bits to represent each value.
$\rightarrow$ Simple quantization: uniform quantization


## Compression Tool \#2 (lossy): Quantization

- Quantify the "imprecision" between original and quantized:
- maximum error $\max _{x}|x-\hat{x}|$
- mean squared error $\mathbb{E}\left[(x-\hat{x})^{2}\right]$ : easy to differential in an optimization formulation
- For a fixed amount of average error, signal with large dynamic range requires more bits in representation. e.g., uniform quantizer: max error $=A /(2 N)$ $\Rightarrow$ dynamic range $A \uparrow$ or $\#$ intervals $N \downarrow$ lead to higher error
- Non-uniform quantizer: may consider a few aspects
(1) keep relative error low (smaller stepsize in low value range)
(2) take account of signal's probability distribution and keep the expected error low (reduce error in most seen values) e.g., MMSE / Lloyd-Max quantizer


## Non-bandlimited Signals

We often encounter signals that are not bandlimited, but have dominant frequency bands.

Question: How to use fewer bits to represent the signal and keep the imprecision low?

e.g. $x[n]: 10 \mathrm{kHz}$ sampled signal, 16 bits/sample (to cover the dynamic range) $\Rightarrow$ data bit rate 160 kbps

## Subband Coding



(1) $x_{0}[n]$ and $x_{1}[n]$ are bandlimited and can be decimated
(2) $\mathbb{X}_{1}(\omega)$ has smaller power s.t. $x_{1}[n]$ has smaller dynamic range, thus can be

Suppose now to represent each subband signal, we need $x_{0}[n]: 16$ bits / sample $x_{1}[n]$ : 8 bits / sample
$\therefore 16 \times \frac{10 k}{2}+8 \times \frac{10 k}{2}=120 k b p s$ represented with fewer bits

## Filter Bank for Subband Coding



Role of $F_{k}(z)$ :

- eliminate spectrum images introduced by $\uparrow 2$
- If $\left\{H_{k}(z)\right\}$ is not perfect, the decimated subband signals may have aliasing.
- $\left\{F_{k}(z)\right\}$ should be chosen carefully so that the aliasing gets canceled at the synthesis stage (in $\hat{x}[n]$ ).


## Applications in Digital Audio Systems

- During A/D conversion: Oversampling to alleviate the stringent requirements on the analog anti-aliasing filter
- During D/A conversion: Filter to remove spectrum images
- Fractional sampling rate conversion: Studio 48 KHz vs. CD 44.1 KHz

Readings to explore more: Vaidynathan Tutorial Sec. III-A.

Warm-up Exercise: Two-Channel Filter Bank

Under what conditions does a filter bank preserve information?

Derive the input-output relation in Z-domain.


# Multi-rate Signal Processing 6. Quadrature Mirror Filter (QMF) Bank 

Electrical \& Computer Engineering
University of Maryland, College Park

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu. The LaTeX slides were made by Prof. Min Wu and Mr. Wei-Hong Chuang.
Contact: minwu@umd.edu. Updated: September 29, 2011.
6.1 Errors Created in the QMF Bank 6.2 A Simple Alias-Free QMF System 6.A Look Ahead

Review: Two-channel Filter Bank

Recall: the 2-band QMF bank example in subband coding


Typical magnitude response



Overlapping filter response across $\pi / 2$ may cause aliased subband signals

### 6.1 Errors Created in the QMF Bank

The reconstructed signal $\hat{x}[n]$ can differ from $x[n]$ due to
(1) aliasing
(2) amplitude distortion
(3) phase distortion
(9) processing of the decimated subband signal $v_{k}[n]$

- quantization, coding, or other processing
- inherent in practical implementation and/or depends on applications
$\Rightarrow$ ignored in this section.
Readings: Vaidynathan Book 5.0-5.2; Tutorial Sec.VI.


## Input-Output Relation



Examine the input-output relation:

## Input-Output Relation

$$
\begin{aligned}
\hat{X}(z)= & \frac{1}{2}\left[H_{0}(z) F_{0}(z)+H_{1}(z) F_{1}(z)\right] X(z)+ \\
& \frac{1}{2}\left[H_{0}(-z) F_{0}(z)+H_{1}(-z) F_{1}(z)\right] X(-z)
\end{aligned}
$$

In matrix-vector form: details

## What is $X(-z)$ ?

- $\left.X(-z)\right|_{z=e^{j} \omega}=X(\omega-\pi)$, i.e., shifted version of $X(\omega)$ Referred to as the "alias term".


If $X(\omega)$ is not bandlimited by $\pi / 2$, then $X(-z)$ may overlap with $X(z)$ spectrum.
In the reconstructed signal $\hat{x}[n]$, this alias term reflects aliasing due to downsampling and residue imaging due to expansion.

## Linear Periodically Time Varying (LPTV) Viewpoint

details Write $\hat{X}(z)$ expression as: $\hat{X}(z)=T(z) X(z)+A(z) X(-z)$
i.e., alternatingly taking output from one of the two LTI subsystems (note: input and ouput have the same rate)

## Linear Periodically Time Varying (LPTV) Viewpoint



If aliasing is cancelled (i.e., $A(z)=0$ ), this will become LTI with transfer function $T(z)$.

Questions: Why we may want to permit some aliasing?

- To avoid excessive attenuation of input signal around $\omega=\frac{\pi}{2}$ and expensive $H_{k}(z)$ filters for sharp transition band, we permit some aliasing in the decimated analysis bank instead of trying to completely avoid it.
- We then choose synthesis filters so that the alias components in the two branches can cancel out each other.


## Alias Cancellation

To cancel aliasing for all possible inputs $x[n]$ s.t.

$$
H_{0}(-z) F_{0}(z)+H_{1}(-z) F_{1}(z)=0
$$

we can choose

$$
\left\{\begin{array}{l}
F_{0}(z)=H_{1}(-z) \\
F_{1}(z)=-H_{0}(-z)
\end{array}\right.
$$

(a sufficient condition)

Example: sketch intermediate spectrums step-by-step


## Alias Cancellation in the Spectrum

P.P. Vaidyanathan: "Multirate digital filters, filter banks, polyphase networks, andapplications: a tutorial",
Proceedings of the IEEE, Jan 1990, Volume: 78, Issue:
1 , pages 56-93. DOI: $10.1109 / 5.52200$
ig. 23. Ilustration of various Fourier transforms in two-channel QMF bank. Here horzontal axis represents $\omega$. (a) Typical input. (b) Transform. (c) Aliasing effect. (d) Imaging effect. (e) Using $x_{1}$. (f) Using $v_{1}$. (g) Using $y_{1}$. (h) Alias-term at output of $F_{0}(z)$. (i) Alias-term at output of $F_{1}(z)$.


(a)
$x_{0}\left(e^{j \omega}\right)$

(b)
$v_{0}\left(e^{j \omega}\right)$

(c)
$y_{0}\left(e^{j \omega}\right)$

(d)

Alias-term at Output
of $\mathrm{F}_{0}(2)$

(h)

(a)

(e)

(f)
$y_{1}\left(e^{j \omega}\right)$


Alias-term
at Output of $F_{1}(z)$

(i)

Alias Cancellation in the Spectrum
(sketch)
Assume $H_{0}(z)$ and $H_{1}(z)$ have some overlap and across $\pi / 2$


$\Downarrow \sqrt{ } \sqrt{2}$


$\Downarrow$ T2 squeeze along $\omega$.




possible to choose $F_{k}(z)$ to make these terms cancel each other out

## Amplitude and Phase Distortions

## Distortion Transfer Function

For an aliasing-free QMF bank, $\hat{X}(z)=T(z) X(z)$,
where $T(z)=\frac{1}{2}\left[H_{0}(z) F_{0}(z)+H_{1}(z) F_{1}(z)\right]$

$$
=\frac{1}{2}\left[H_{0}(z) H_{1}(-z)-H_{1}(z) H_{0}(-z)\right]
$$

This is called the distortion transfer function, or the overall transfer function of the alias-free system.

Let $T(\omega)=|T(\omega)| e^{j \phi(\omega)}$
To prevent amplitude distortion and phase distortion, $T(\omega)$ must be allpass (i.e. $|T(\omega)|=\alpha \neq 0$ for all $\omega, \alpha$ is a constant) and linear phase (i.e., $\phi(\omega)=a+b \omega$ for constants $a, b$ )

## Properties of $T(z)$

- Perfect reconstruction (PR) property: if a QMF bank is free from aliasing, amplitude distortion and phase distortion, i.e., $T(z)=c z^{-n 0} \Rightarrow \hat{x}[n]=c x\left[n-n_{0}\right]$
- With our above alias-free choice of $F_{k}(z)$, $T(z)$ is in the form of $T(z)=W(z)-W(-z)$, where $W(z)=H_{0}(z) H_{1}(-z)$.
$\Rightarrow T(z)$ has only odd power of $z$ (as the even powers get cancelled), i.e., $T(z)=z^{-1} S\left(z^{2}\right)$ for some $S(z)$.

So $|T(\omega)|$ has period of $\pi$ (instead of $2 \pi$ ).
And for real-coefficient filters, this implies $|T(\omega)|$ is symmetric w.r.t. $\pi / 2$ for $0 \leq \omega<\pi$.

### 6.2 A Simple Alias-Free QMF System

Consider the analysis filters are related as

$$
H_{1}(z)=H_{0}(-z)
$$

For real filter coefficients, this means $\left|H_{1}(\omega)\right|=\left|H_{0}(\pi-\omega)\right|$.
$\because\left|H_{0}(\omega)\right|$ symmetric w.r.t. $\omega=0 ;\left|H_{1}(\omega)\right| \sim$ shift $\left|H_{0}(\omega)\right|$ by $\pi$.
i.e., $\left|H_{1}(\omega)\right|$ is a mirror image of $\left|H_{0}(\omega)\right|$ w.r.t. $\omega=\pi / 2=2 \pi / 4$, the "quadrature frequency" of the normalized sampling frequency.

If $H_{0}(z)$ is a good LPF, then $H_{1}(z)$ is a good HPF.


## (1) QMF Choice and Alias-free Condition

With QMF choice of $H_{1}(z)=H_{0}(-z)$, now the alias-free condition becomes

$$
\left\{\begin{array} { l } 
{ F _ { 0 } ( z ) = H _ { 1 } ( - z ) } \\
{ F _ { 1 } ( z ) = - H _ { 0 } ( - z ) }
\end{array} \Rightarrow \left\{\begin{array}{l}
F_{0}(z)=H_{0}(z) \\
F_{1}(z)=-H_{1}(1 z)
\end{array}\right.\right.
$$

All four filters are completely determined by a single filter $H_{0}(z)$.
The distortion transfer function becomes

$$
T(z)=\frac{1}{2}\left[H_{0}^{2}(z)-H_{1}^{2}(z)\right]=\frac{1}{2}\left[H_{0}^{2}(z)-H_{0}^{2}(-z)\right]
$$

## (2) Polyphase Representation of QMF

* beneficial both computationally and conceptually

$$
\text { Let } H_{0}(z)=E_{0}\left(z^{2}\right)+z^{-1} E_{1}\left(z^{2}\right) \quad(\text { Type-1 PD })
$$

Then $H_{1}(z)=H_{0}(-z)=E_{0}\left(z^{2}\right)-z^{-1} E_{1}\left(z^{2}\right)$
In matrix/vector form,

$$
\left[\begin{array}{l}
H_{0}(z) \\
H_{1}(z)
\end{array}\right]=\left[\begin{array}{cc}
1 & 1 \\
1 & -1
\end{array}\right]\left[\begin{array}{c}
E_{0}\left(z^{2}\right) \\
z^{-1} E_{1}\left(z^{2}\right)
\end{array}\right]
$$

Similarly, for synthesis filters,

$$
\begin{aligned}
{\left[\begin{array}{ll}
F_{0}(z) & F_{1}(z)
\end{array}\right] } & =\left[\begin{array}{ll}
H_{0}(z) & -H_{1}(z)
\end{array}\right] \\
& =\left[\begin{array}{cc}
z^{-1} E_{1}\left(z^{2}\right) & E_{0}\left(z^{2}\right)
\end{array}\right]\left[\begin{array}{cc}
1 & 1 \\
1 & -1
\end{array}\right]
\end{aligned}
$$

Polyphase Representation: Signal Flow Diagram

$$
\begin{aligned}
& {\left[\begin{array}{l}
H_{0}(z) \\
H_{1}(z)
\end{array}\right]=\left[\begin{array}{cc}
1 & 1 \\
1 & -1
\end{array}\right]\left[\begin{array}{c}
E_{0}\left(z^{2}\right) \\
z^{-1} E_{1}\left(z^{2}\right)
\end{array}\right]} \\
& {\left[\begin{array}{ll}
F_{0}(z) & F_{1}(z)
\end{array}\right]=\left[\begin{array}{ll}
z^{-1} E_{1}\left(z^{2}\right) & E_{0}\left(z^{2}\right)
\end{array}\right]\left[\begin{array}{cc}
1 & 1 \\
1 & -1
\end{array}\right]}
\end{aligned}
$$

## Polyphase Representation: Efficient Structure

Rearrange using nobel identities to obtain efficient implementation:


For $H_{0}(z)$ of length $N \Rightarrow E_{k}(z)$ has length $N / 2$

- Analysis bank: N/2 MPU, N/2 APU; same for synthesis bank
- Total: N MPU \& APU
$\because H_{0}^{2}(z)=E_{0}^{2}\left(z^{2}\right)+E_{1}^{2}\left(z^{2}\right) z^{-2}+2 z^{-1} E_{0}\left(z^{2}\right) E_{1}^{2}\left(z^{2}\right)$
So the distortion transfer function becomes

$$
T(z)=\frac{1}{2}\left[H_{0}^{2}(z)-H_{0}^{2}(-z)\right]=2 z^{-1} E_{0}\left(z^{2}\right) E_{1}\left(z^{2}\right)
$$

## Polyphase Representation: Matrix Form



In matrix form: (with MIMO transfer function for intermediate stages)


* Note: Multiplication is from left for each stage when intermediate signals are in column vector form.


## Observations

## The distortion transfer function of QMF

$T(z)=2 z^{-1} E_{0}\left(z^{2}\right) E_{1}\left(z^{2}\right)$

- If $H_{0}(z)$ is FIR, so are $E_{0}(z), E_{1}(z)$ and $T(z)$.
- For $H_{0}(z)$ FIR and $H_{1}(z)=H_{0}(-z)$, the amplitude distortion can be eliminated iff $E_{0}(z)$ and $E_{1}(z)$ represent a delay:

$$
\left\{\begin{array}{l}
E_{0}(z)=c_{0} z^{-n_{0}} \\
E_{1}(z)=c_{1} z^{-n_{1}}
\end{array}\right.
$$

## Observations

For $E_{0}(z)$ and $E_{1}(z)$ each representing a delay, we can only have analysis filters in the form of

$$
\left\{\begin{array}{l}
H_{0}(z)=c_{0} z^{-2 n_{0}}+c_{1} z^{-\left(2 n_{1}+1\right)} \\
H_{1}(z)=c_{0} z^{-2 n_{0}}-c_{1} z^{-\left(2 n_{1}+1\right)}
\end{array}\right.
$$

Such filters don't have sharp cutoff and good stopband attenuations.

Therefore $H_{1}(z)=H_{0}(-z)$ is not a good choice to build FIR perfect reconstruction QMF systems for such applications as subband coding.

## (3) Eliminating Phase Distortions with FIR Filters

If $H_{0}(z)$ has linear phase, then we can show that

$$
T(z)=\frac{1}{2}\left[H_{0}^{2}(z)-H_{0}^{2}(-z)\right]
$$

also has linear phase (thus eliminating phase distortion).
Let $H_{0}(z)=\sum_{n=0}^{N} h_{0}[n] z^{-n}$ with $h_{0}[n]$ real. The linear phase and low pass conditions lead to $h_{0}[n]=h_{0}[N-n]$ (symmetric).

We can write $H_{0}(\omega)=e^{-j \omega \frac{N}{2}} \underbrace{R(\omega)}_{\text {real valued }}$

## (3) Eliminating Phase Distortions with FIR Filters

$T(\omega)$ now becomes:

Note: $\left|H_{0}(\omega)\right|=|R(\omega)|$ and
$\left|H_{0}(\omega)\right|$ is even symmetric
$\Rightarrow T(\omega)=\frac{e^{-j \omega N}}{2}\left[\left|H_{0}(\omega)\right|^{2}-(-1)^{N}\left|H_{0}(\pi-\omega)\right|^{2}\right]$
If $N$ is even, $\left.T(\omega)\right|_{\omega=\frac{\pi}{2}}=0$, which brings severe amplitude distortion around $\omega=\pi / 2$.

To avoid this, the filter order $N$ should be odd (or length is even)
so that $T(\omega)=\frac{e^{-j \omega N}}{2}\left[\left|H_{0}(\omega)\right|^{2}+\left|H_{0}(\pi-\omega)\right|^{2}\right]$

## (4) Minimizing Amplitude Distortion with FIR Filters

- Recall: after choosing $H_{1}(z)=H_{0}(-z)$, the amplitude distortion can be removed iff $H_{0}(z)$ 's two polyphase components are pure delay.
But such $H_{0}(z)$ doesn't have good low-pass response.
- For more flexible choices of $H_{0}(z)$ while eliminating aliasing and phase distortion, there will be some amplitude distortion.
- What we can do is to adjust the coefficients in $H_{0}(z)$ to minimize the amplitude distortion, i.e., to make $T(\omega)$ approximately constant:

$$
\left|H_{0}(\omega)\right|^{2}+\left|H_{1}(\omega)\right|^{2} \approx 1
$$

## (4) Minimizing Amplitude Distortion with FIR Filters

\(\xrightarrow[\pi / 2]{\substack{\left.1 H_{0}(w)\right|^{2} <br>

+\left|H_{1}(\omega)\right|^{2}}}\)| $1-$ too much overlap b/w $H_{0}$ and $H_{1}$ |
| :--- |
| 2 - too little overlap |
| $3-$ good choice |
| (can be obtained by trial and error or by |
| optimization formulation) |

- Recall $T(z)$ has only odd power of $z$. For real-coeff. filter, $|T(\omega)|$ is symmetric w.r.t. $\pi / 2$ for $0 \leq \omega<\pi$.
- By quadrature mirror condition, $|T(\omega)|$ is almost constant in the passbands of $H_{0}(z)$ and $H_{1}(z)$ if $H_{0}(z)$ has good passband and stopband responses.
- The main problem is with the transition band. The degree of overlap between $H_{0}(z)$ and $H_{1}(z)$ is crucial in determining this distortion.

See Vaidyanathan's Book $\S 5.2 .2$ for details and examples

## (5) Eliminating Amplitude Distortion with IIR Filters

How about IIR filters?

- The choice of $E_{1}(z)=\frac{1}{E_{0}(z)}$ can lead to perfect reconstruction and provide more room for designing $H(z)$.
But the filters $H_{k}(z)$ would become IIR and may not provide desirable response.
- To completely eliminate amplitude distortion, $T(z)$ must be all-pass (which is IIR).
- Review: a 1st-order all-pass filter $G(z)=\frac{a^{*}+z^{-1}}{1+a z^{-1}}$

$$
\Rightarrow|G(\omega)|=1 ; \text { zero }=-1 / a^{*}, \text { pole }=-a(\text { conjugate reciprocal }) .
$$

## (5) Eliminating Amplitude Distortion with IIR Filters

One way to make $T(z)$ allpass is to choose $E_{0}(z)$ and $E_{1}(z)$ to be IIR and allpass.

Let $E_{0}(z)=\frac{a_{0}(z)}{2}$ and $E_{1}(z)=\frac{a_{1}(z)}{2}$ where $a_{0}(z)$ and $a_{1}(z)$ are allpass with $\left|a_{0}(\omega)\right|=\left|a_{0}(\omega)\right|=1$.

The analysis filter becomes
$H_{0}(z)=E_{0}\left(z^{2}\right)+z^{-1} E_{1}\left(z^{2}\right)=\frac{a_{0}\left(z^{2}\right)+z^{-1} a_{1}\left(z^{2}\right)}{2}$
$\Rightarrow$ possible to have good $H(\omega)$ response with such all-pass polyphase form.
Explore PPV book 5.3

The overall distortion transfer function is allpass:
$T(z)=\frac{z^{-1}}{2} a_{0}\left(z^{2}\right) a_{1}\left(z^{2}\right)$

## Phase Distortion with IIR Filters

- This design of QMF bank is free from amplitude distortion and aliasing, regardless of the details of the allpass filters $a_{0}(z)$ and $a_{1}(z)$.
- But the phase distortion remains due to the IIR components. The phase distortion is governed by the phase responses of $a_{0}(z)$ and $a_{1}(z)$.

Question: Can $a_{0}(z)$ and $a_{1}(z)$ be designed to cancel out phase distortion?

Note the difficulty in designing filters to meet many constraints.

## Summary

Many "wishes" to consider toward achieving alias-free P.R. QMF:
(0) alias free, (1) phase distortion, (2) amplitude distortion,
(3) desirable filter responses.

Can't satisfy them all at the same time, so often meet most of them and try to approximate/optimize the rest.

A particular relation of synthesis-analysis filters to cancel alias:
$\left\{\begin{array}{l}F_{0}(z)=H_{1}(-z) \\ F_{1}(z)=-H_{0}(-z)\end{array}\right.$
s.t. $H_{0}(-z) F_{0}(z)+H_{1}(-z) F_{1}(z)=0$.

We considered a specific relation between the analysis filters: $H 1(z)=H_{0}(-z)$ s.t. response symmetric w.r.t. $\omega=\pi / 2$ (QMF)

With polyphase structure: $T(z)=2 z^{-1} E_{0}\left(z^{2}\right) E_{1}\left(z^{2}\right)$

## Summary: $\quad T(z)=2 z^{-1} E_{0}\left(z^{2}\right) E_{1}\left(z^{2}\right)$

## Case-1 $H_{0}(z)$ is FIR:

- P.R.: require polyphase components of $H_{0}(z)$ to be pure delay s.t. $H_{0}(z)=c_{0} z^{-2 n_{0}}+c_{1} z^{-\left(2 n_{1}+1\right)}$ [cons] $H_{0}(\omega)$ response is very restricted.
- For more desirable filter response, the system may not be P.R., but can minimize distortion:
- eliminate phase distortion: choose filter order N to be odd, and $h_{0}[n]$ be symmetric (linear phase)
- minimize amplitude distortion: $\left|H_{0}(\omega)\right|^{2}+\left|H_{1}(\omega)\right|^{2} \approx 1$

Case-2 $H_{0}(z)$ is IIR:

- $E_{1}(z)=\frac{1}{E_{0}(z)}$ can get P.R. but restrict the filter responses.
- eliminate amplitude distortion: choose polyphase components to be all pass, s.t. $T(z)$ is all-pass, but may have some phase distortion


## Look Ahead: Simple FIR P.R. Systems

2-channel simple P.R. system:


How are $\hat{X}(z)$ and $X(z)$ related?
What are the equiv. $H_{k}(z)$ and $F_{k}(z)$ ?

Extend to M-channel:


How are $\hat{X}(z)$ and $X(z)$ related?
What are the equiv. $H_{k}(z)$ and $F_{k}(z)$ ?
Interpretation: demultiplex then multiplex again

## Look Ahead: Simple Filter Bank Systems



If all $S_{k}(z)$ are identical as $S(z)$, how are $\hat{X}(z)$ and $X(z)$ related? How is this related to the simple M-channel P.R. system on the last page?

Look Ahead: M-channel filter bank

Study more general conditions of alias-free and PR; examine $M$-channel filter bank:


Derive the input-output relation.

Appendix: Detailed Derivations

Input-Output Relation

Examine the input-output relation:
(1) Subband signals

$$
X_{k}(z)=H_{k}(z) S(z)
$$

$$
k=0,1 .
$$

(2)
aliasing occues if this put's spertrum woveraps, with $X\left(\left\{z^{2}\right)\right.$
(3)

$$
\begin{aligned}
\bar{T}_{k}(z) & =V_{k}\left(z^{2}\right)=\frac{1}{2} \mathbb{X}_{k}(z)+\frac{1}{2} \mathbb{X}_{k}(-z) \\
& =\frac{1}{2} H_{k}(z) \mathbb{X}(z)+\frac{1}{2} H_{k}(-z) \mathbb{X _ { k }}(-8) \quad k=0,1
\end{aligned}
$$

(4)

$$
\begin{aligned}
\hat{X}(z)= & F_{0}(z) Y_{0}(z)+F_{1}(z) Y_{1}(z) \\
= & \frac{1}{2}\left[H_{0}(z) F_{0}(z)+H_{1}(z) F_{1}(z)\right] X(z) \\
& +\frac{1}{2}\left[H_{0}(-z) F_{0}(z)+H_{1}(-z) F_{1}(z)\right] X(-z)
\end{aligned}
$$

$$
\begin{aligned}
& \begin{array}{l}
\text { Decimated } \\
\text { subband signals } \\
V_{k}
\end{array}(\xi)=\left.\left[\bar{X}_{k}(\xi)\right]\right|_{\downarrow 2} \\
& =\frac{1}{2} X_{k}\left(z^{1 / 2} W_{2}^{0}\right)+\frac{1}{2} X_{k}\left(z^{1 / 2} W_{2}^{1}\right) \\
& \begin{array}{l}
\text { reall } \\
W_{M}=e^{-j \frac{2 \pi}{M}}=\frac{1}{2} X_{k}\left(z^{1 / 2}\right)+\frac{1}{2} X_{k}\left(-z^{1 / 2}\right) \quad k=0,1 .
\end{array}
\end{aligned}
$$

Input-Output Relation

In matrix-vector form:

$$
\left.\begin{array}{rl}
2 \widehat{X}(z) & =\left[\begin{array}{ll}
\mathbb{X}(z) & \mathbb{X}(-z)
\end{array}\right] \underbrace{\left[\begin{array}{ll}
H_{0}(z) & H_{1}(z) \\
H_{0}(-z) & H_{r}(-z)
\end{array}\right]}_{\text {Define as He) }}\left[\begin{array}{l}
\text { "alias component matrix }
\end{array}\right] \\
F_{0}(z) \\
F_{1}(z)
\end{array}\right]
$$

LPTV (Linear Periodically Time Varying) Viewpoint

Write $\hat{X}(z)$ expression as:

$$
\begin{aligned}
& \hat{X}(z)=T(z) X(z)+A(z) X(-z) \\
& \Leftrightarrow \hat{X}[n]=\sum_{k}\left(t[k]+(-1)^{n-k} a[k]\right) x[n-k]
\end{aligned} \quad\left\{\begin{array} { l } 
{ X ( z ) = \sum _ { n = - \infty } ^ { + \infty } x [ n ] \delta ^ { - n } } \\
{ X ( - z ) = \sum _ { n = - \infty } ^ { + \infty } x [ n ] \times ( - 1 ) ^ { - n } } \\
{ \text { Define } }
\end{array} \quad \left\{\begin{array}{l}
g_{0}[k]=t[k]+(-1)^{k} a[k] \\
g_{1}[k]=t[k]-(-1)^{k} a[k]
\end{array} \quad \begin{array}{l}
\Rightarrow \hat{x}[n]= \begin{cases}g_{0}[n] * x[n] & \text { his even } \\
g_{1}[n] * x[n] & \text { nos odd }\end{cases}
\end{array}\right.\right.
$$

i.e., alternatingly taking output from one of the two LTI subsystems (note: input and ouput have the same rate)

## Eliminating Phase Distortions with FIR Filters

$T(\omega)$ now becomes

$$
\begin{aligned}
T(\omega) & =\frac{1}{2}\left[H_{0}^{2}(\omega)-H_{0}^{2}(\omega-\pi)\right] \\
& =\frac{1}{2}\left[e^{-j \omega N} R^{2}(\omega)-e^{-j(\omega-\pi) N} R^{2}(\omega-\pi)\right] \\
& =\frac{e^{-j \omega N}}{2}\left[R^{2}(\omega)-(-1)^{N} R^{2}(\pi-\omega)\right] \\
& =\frac{e^{-j \omega N}}{2}\left[\left|H_{0}(\omega)\right|^{2}-(-1)^{N}\left|H_{0}(\pi-\omega)\right|^{2}\right]
\end{aligned}
$$

also used here $\left|H_{0}(\omega)\right|=|R(\omega)|$ and $\left|H_{0}(\omega)\right|$ being even symmetric

If $N$ is even, $\left.T(\omega)\right|_{\omega=\frac{\pi}{2}}=0$, which brings severe amplitude distortion around $\omega=\pi / 2$.

To avoid this, $N$ should be odd so that $T(\omega)=\frac{e^{-j \omega N}}{2}\left[\left|H_{0}(\omega)\right|^{2}+\left|H_{0}(\pi-\omega)\right|^{2}\right]$

# Multi-rate Signal Processing <br> 7. M-channel Maximally Decmiated Filter Banks 

Electrical \& Computer Engineering<br>University of Maryland, College Park

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu. The LaTeX slides were made by Prof. Min Wu and Mr. Wei-Hong Chuang.
Contact: minwu@umd.edu. Updated: October 6, 2011.
7.1 The Reconstructed Signal and Errors Created
7.2 The Alias Component (AC) Matrix
7.3 The Polyphase Representation
7.4 Perfect Reconstruction Filter Bank
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$

M-channel Maximally Decimated Filter Bank
To study more general conditions of alias-free and P.R., it becomes more convenient to examine $M$-channel filter bank.


As each of the filter has passband of about $2 \pi / M$ wide, the subband signal output can be decimated up to $M$ without substantial aliasing. The filter bank is said to be "maximally decimated" if this maximal decimation factor is used.

7 M-channel Maximally Decimated Filter Bank

## The Reconstructed Signal and Errors Created

[ Readings: PPV Book 5.4, 5.5; Tutorial Sec. VIII ]

Relations between $\hat{X}(z)$ and $X(z)$ :
$\hat{X}(z)=\sum_{l=0}^{M-1} A_{\ell}(z) X\left(W^{\ell} z\right)$

- $A_{\ell}(z) \triangleq \frac{1}{M} \sum_{k=0}^{M-1} H_{k}\left(W^{\ell} z\right) F_{k}(z), 0 \leq \ell \leq M-1$.
- $\left.X\left(W^{\ell} z\right)\right|_{z=e^{i} \omega}=X\left(\omega-\frac{2 \pi \ell}{M}\right)$, i.e., shifted version from $X(\omega)$.
- $X\left(W^{\ell} z\right)$ : $\ell$-th aliasing term, $A_{\ell}(z)$ : gain for this aliasing term.
7.3 The Polyphase Representation
7.4 Perfect Reconstruction Filter Bank
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$


## Conditions for LPTV, LTI, and PR

- In general, the M-channel filter bank is a LPTV system with period $M$.
- The aliasing term can be eliminated for every possible input $x[n]$ iff $A_{\ell}(z)=0$ for $1 \leq \ell \leq M-1$. When aliasing is eliminated, the filter bank becomes an LTI system:

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

where $T(z) \triangleq A_{0}(z)=\frac{1}{M} \sum_{\ell=0}^{M-1} H_{k}(z) F_{k}(z)$ is the overall transfer function, or distortion function.

- If $T(z)=c z^{-n_{0}}$, it is a perfect reconstruction system (i.e., free from aliasing, amplitude distortion, and phase distortion).


## The Alias Component (AC) Matrix

From the definition of $A_{\ell}(z)$, we have in matrix-vector form:

$\mathcal{H}(z): M \times M$ matrix called the "Alias Component matrix"
The condition for alias cancellation is

$$
\mathcal{H}(z) \underline{\mathbb{t}}(z)=\mathbb{t}(z), \quad \text { where } \mathbb{t}(z)=\left[\begin{array}{c}
M A_{0}(z) \\
0 \\
\vdots \\
0
\end{array}\right]
$$

## The Alias Component (AC) Matrix

Now express the reconstructed signal as

$$
\hat{X}(z)=\mathcal{A}^{T}(z) \mathcal{X}(z)=\frac{1}{M} \underline{\mathbb{I}}^{T}(z) \mathcal{H}^{T}(z) \mathcal{X}(z)
$$

where $\mathcal{X}(z)=\left[\begin{array}{c}X(z) \\ X(z W) \\ \vdots \\ X\left(z W^{M-1}\right)\end{array}\right]$.
Given a set of analysis filters $\left\{H_{k}(z)\right\}$, if $\operatorname{det} \mathcal{H}(z) \neq 0$, we can choose synthesis filters as $\underline{\mathbb{t}}(z)=\mathcal{H}^{-1}(z) \mathbb{t}(z)$ to cancel aliasing and obtain P.R. by requiring

$$
\underline{\mathbb{t}}(z)=\left[\begin{array}{c}
c z^{-n_{0}} \\
0 \\
: \\
0
\end{array}\right]
$$

## Difficulty with the Matrix Inversion Approach

- $\mathcal{H}^{-1}(z)$ and thus the synthesis filters $\left\{F_{k}(z)\right\}$ can be IIR even if $\left\{H_{k}(z)\right\}$ are all FIR.
- Difficult to ensure $\left\{F_{k}(z)\right\}$ stability (all poles inside the unit circle)
- $\left\{F_{k}(z)\right\}$ may have high order even if the order of $\left\{H_{k}(z)\right\}$ is moderate
- ......
$\Rightarrow$ Take a different approach for P.R. design via polyphase representation.
7.1 The Reconstructed Signal and Errors Created
7.2 The Alias Component (AC) Matrix
7.3 The Polyphase Representation
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$

Type-1 PD for $H_{k}(z)$
Using Type-1 PD for $H_{k}(z)$ :

$$
H_{k}(z)=\sum_{\ell=0}^{M-1} z^{-\ell} E_{k \ell}\left(z^{M}\right)
$$

We have


$$
\underline{\ln }(z)=\mathbb{E}\left(z^{M}\right) \underline{e}(z)
$$

$\mathbb{E}\left(z^{M}\right): M \times M$ Type-1 polyphase component matrix for analysis bank
7.2 The Alias Component (AC) Matrix
7.3 The Polyphase Representation
7.4 Perfect Reconstruction Filter Bank
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$

## Type-2 PD for $F_{k}(z)$

Similarly, using Type-2 PD for $F_{k}(z)$ :

$$
F_{k}(z)=\sum_{\ell=0}^{M-1} z^{-(M-1-\ell)} R_{\ell k}\left(z^{M}\right)
$$

We have in matrix form:
$\left[F_{0}(z) \cdots F_{M-1}(z)\right]=\underbrace{\left[z^{-(M-1)}, z^{-(M-2)}, \ldots 1\right.}]\left[\begin{array}{c}R_{00}\left(z^{M}\right) \cdots R_{0, M-1}\left(z^{M}\right) \\ R_{10}\left(z^{M}\right) \cdots R_{1, M-1}\left(z^{M}\right) \\ \vdots \\ \left.R_{M-1, \sigma^{(z)}} z^{M}\right) \cdots R_{M-1, M-1}\left(z^{M}\right)\end{array}\right]$
$\Leftrightarrow \underline{\mathbb{f}}^{\top}(z)=\mathbb{e}_{B}^{\top}(z) \mathbb{R}\left(z^{M}\right)$
$\underline{\mathbb{e}}_{B}^{T}(z)$ : reversely ordered version of $\underline{\mathbb{e}}(z)$
$\mathbb{R}\left(z^{M}\right)$ : Type-2 polyphase component matrix for synthesis bank
7.2 The Alias Component (AC) Matrix
7.3 The Polyphase Representation
7.4 Perfect Reconstruction Filter Bank
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$

## Overall Polyphase Presentation



Combine polyphase matrices into one matrix: $\mathbb{P}(z)=\underbrace{\mathbb{R}(z) \mathbb{E}(z)}$ note the order!
7.1 The Reconstructed Signal and Errors Created
7.2 The Alias Component (AC) Matrix
7.3 The Polyphase Representation
7.4 Perfect Reconstruction Filter Bank
7.5 Relation between Polyphase Matrix $\mathbb{E}(z)$ and AC Matrix $\mathcal{H}(z)$

## Simple FIR P.R. Systems



$$
\hat{X}(z)=z^{-1} X(z)
$$

i.e., transfer function $T(z)=z^{-1}$

Extend to $M$ channels:
$H_{k}(z)=z^{-k}$
$F_{k}(z)=z^{-M+k+1}, 0 \leq k \leq M-1$
$\Rightarrow \hat{\mathbb{X}}(z)=z^{-(M-1)} \mathbb{X}(z)$
i.e. demultiplex then multiplex again


## General P.R. Systems

Recall the polyphase implementation of $M$-channel filter bank:


Combine polyphase matrices into one matrix: $\mathbb{P}(z)=\mathbb{R}(z) \mathbb{E}(z)$
If $\mathbb{P}(z)=\mathbb{R}(z) \mathbb{E}(z)=\mathbb{I}$, then the system is equivalent to the simple system $\Rightarrow H_{k}(z)=z^{-k}, F_{k}(z)=z^{-M+k+1}$

In practice, we can allow $\mathbb{P}(z)$ to have some constant delay, i.e., $\mathbb{P}(z)=c z^{-m_{0}} \mathbb{I}$, thus $T(z)=c z^{-\left(M m_{0}+M-1\right)}$.

## Dealing with Matrix Inversion

To satisfy $\mathbb{P}(z)=\mathbb{R}(z) \mathbb{E}(z)=\mathbb{I}$, it seems we have to do matrix inversion for getting the synthesis filters $\mathbb{R}(z)=(\mathbb{E}(z))^{-1}$.

Question: Does this get back to the same inversion problem we have with the viewpoint of the AC matrix $\underline{f}(z)=\mathcal{H}^{-1}(z) \mathbb{t}(z)$ ?

## Solution:

- $\mathbb{E}(z)$ is a physical matrix that each entry can be controlled. In contrast, for $\mathcal{H}(z)$, only 1st row can be controlled (thus hard to ensure desired $H_{k}(z)$ responses and $\mathbb{f}(z)$ stability
- We can choose $\operatorname{FIR} \mathbb{E}(z)$ s.t. $\operatorname{det} \mathbb{E}(z)=\alpha z^{-k}$ thus $\mathbb{R}(z)$ can be FIR (and has determinant of similar form).

Summary: With polyphase representation, we can choose $\mathbb{E}(z)$ to produce desired $H_{k}(z)$ and lead to simple $\mathbb{R}(z)$ s.t. $\mathbb{P}(z)=c z^{-k} \mathbb{I}$.

## Paraunitary

A more general way to simplify the need of matrix inversion:
Constrain $\mathbb{E}(z)$ to be paraunitary: $\quad \tilde{\mathbb{E}}(z) \mathbb{E}(z)=d \mathbb{I}$
Here $\tilde{\mathbb{E}}(z)=\mathbb{E}_{*}^{T}\left(z^{-1}\right)$, i.e. taking conjugate of the transfer function coeff., replace $z$ with $z^{-1}$ that corresponds to time reversely order the filter coeff., and transpose.

For further exploration: PPV Book Chapter 6.

7 M-channel Maximally Decimated Filter Bank

## Relation b/w Polyphase Matrix $\mathbb{E}(z)$ and $A C$ Matrix $\mathcal{H}(z)$

The relation between $\mathbb{E}(z)$ and $\mathcal{H}(z)$ can be shown as:

$$
\mathcal{H}(z)=\left[\mathbb{W}^{*}\right]^{T} \mathbb{D}(z) \quad \mathbb{E}^{T}\left(z^{M}\right)
$$

where $\mathbb{W}$ is the $M \times M$ DFT matrix, and a diagonal delay matrix
$\mathbb{D}(z)=\left[\begin{array}{llll}1 & & & \\ & z^{-1} & & \\ & & \ddots & \\ & & & z^{-(M-1)}\end{array}\right]$
(details) See also the homework.

## Detailed Derivations

The Reconstructed Signal and Errors Created
(1) $\quad X_{k}(z)=H_{k}(z) X(\xi)$ $\qquad$ $\Psi_{k}\left(\omega_{M}^{l} z^{1 / M}\right)$
(2)

$$
V_{K}(z)=\frac{1}{M} \sum_{l=0}^{M-1} H_{K}\left(W_{M}^{l} z^{1 / M}\right) \underset{\text { where } W_{M}=e^{j}\left(W_{M}^{l} z^{1 / M}\right)}{ }
$$

(3) $\quad U_{K}(z)=V_{k}\left(z^{M}\right)=\frac{1}{M} \sum_{l=0}^{M-1} H_{K}\left(W_{M}^{l} z\right) X\left(W_{M}^{\ell} z\right)$
(4)

$$
\begin{aligned}
& \hat{X}(z)=\sum_{k=0}^{M-1} F_{k}(z) U_{k}(z) \\
&=\sum_{l=0}^{M-1}[\underbrace{\left.\frac{1}{M} \sum_{k=0}^{M-1} H_{k}\left(w^{l} z\right) F_{k}(z)\right]} \begin{array}{l} 
\\
\\
\end{array}=\sum_{l=0}^{M-1} A_{l}^{l}(z) X\left(w^{l} z\right) \\
&\left.W^{l} z\right)
\end{aligned}
$$

- $A_{\ell}(z) \triangleq \frac{1}{M} \sum_{k=0}^{M-1} H_{k}\left(W^{\ell} z\right) F_{k}(z), 0 \leq \ell \leq M-1$.
- $\left.\mathbb{X}\left(W^{\ell} z\right)\right|_{z=e^{j \omega}}=\mathbb{X}\left(\omega-\frac{2 \pi \ell}{M}\right)$, ie., shifted version from $\mathbb{X}(\omega)$.
- $\mathbb{X}\left(W^{\ell} z\right): \ell$-th aliasing term, $A_{\ell}(z)$ : gain for this aliasing term.


## Review: Matrix Inversion

$\mathcal{H}(z)^{-1}=\frac{\operatorname{Adj}[\mathcal{H}(z)]}{\operatorname{det}[\mathcal{H}(z)]}$
Adjugate or classical adjoint of a matrix:
$\{\operatorname{Adj}[\mathcal{H}(z)]\}_{i j}=(-1)^{i+j} M_{j i}$
where $M_{j i}$ is the $(j, i)$ minor of $\mathcal{H}(z)$ defined as the determinant of the matrix by deleting the $j$-th row and $i$-th column.

## An Example of P.R. Systems

$$
\begin{aligned}
& H_{0}(z)=2+z^{-1}, H_{1}(z)=3+2 z^{-1} \\
& \mathbb{E}(z)=\left[\begin{array}{ll}
2 & 1 \\
3 & 2
\end{array}\right], \mathbb{E}^{-1}(z)=\frac{\operatorname{Adj} \mathbb{E}(z)}{\operatorname{det} \mathbb{E}(z)}=1 \times\left[\begin{array}{cc}
2 & -1 \\
-3 & 2
\end{array}\right]
\end{aligned}
$$

Choose $\mathbb{R}(z)=\mathbb{E}^{-1}(z)$ s.t. $\mathbb{P}(z)=\mathbb{R}(z) \mathbb{E}(z)=\mathbb{I}$,
$\therefore \mathbb{R}(z)=\left[\begin{array}{cc}2 & -1 \\ -3 & 2\end{array}\right]$
$\left[\begin{array}{ll}F_{0}(z) & F_{1}(z)\end{array}\right]=\left[\begin{array}{ll}z^{-1} & 1\end{array}\right] \mathbb{R}\left(z^{2}\right)=\left[\begin{array}{ll}2 z^{-1}-3, & -z^{-1}+2\end{array}\right]$
$\Rightarrow\left\{\begin{array}{l}F_{0}(z)=-3+2 z^{-1} \\ F_{1}(z)=2-z^{-1}\end{array}\right.$


## Relation b/w Polyphase Matrix $\mathbb{E}(z)$ and $A C$ Matrix $\mathcal{H}(z)$

From the definition of $\mathcal{H}(z)$ and $\underline{\mathbb{h}}(z)$, we have

$$
\begin{aligned}
\mathcal{H}^{T}(z) & =\left[\begin{array}{llll}
\underline{\mathbb{h}}(z) & \underline{\mathbb{h}}(z W) & \cdots & \underline{\mathfrak{h}}\left(z W^{M-1}\right)
\end{array}\right] \\
& =\mathbb{E}\left(z^{M}\right)\left[\begin{array}{llll}
\underline{\mathbb{E}}(z) & \underline{\mathbb{E}}(z W) & \cdots & \underline{\mathbb{E}}\left(z W^{M-1}\right)
\end{array}\right]
\end{aligned}
$$

Examine $\mathbb{e}\left(z W^{k}\right), k=0,1, \ldots,(M-1)$ :
$\underline{\mathbb{E}}\left(z W^{k}\right)=\left[\begin{array}{c}1 \\ \left(z W^{k}\right)^{-1} \\ \cdots \\ \left(z W^{k}\right)^{-(M-1)}\end{array}\right]=\underbrace{\left[\begin{array}{cccc}1 & & & \\ & z^{-1} & & \\ & & \ddots & \\ & & & z^{-(M-1)}\end{array}\right]}_{\text {define as } \mathbb{D}(z), \text { a diagonal delay matrix }}\left[\begin{array}{c}1 \\ W^{-k} \\ \cdots \\ W^{-(M-1) k}\end{array}\right]$

## Relation b/w Polyphase Matrix $\mathbb{E}(z)$ and $A C$ Matrix $\mathcal{H}(z)$

Put together: $\mathcal{H}^{T}(z)=\mathbb{E}\left(z^{M}\right) \mathbb{D}(z) \mathbb{W}^{*}$
Thus we arrive at the relation between $\mathbb{E}(z)$ and $\mathcal{H}(z)$ :

$$
\mathcal{H}(z)=\left[\mathbb{W}^{*}\right]^{T} \mathbb{D}(z) \mathbb{E}^{T}\left(z^{M}\right)
$$

Note: $\left[\mathbb{W}^{*}\right]^{T}$ is equal to $\mathbb{W}^{*}$ due to symmetry of $M \times M$ DFT matrix $\mathbb{W}$

# Multi-rate Signal Processing <br> 8. General Alias-Free Conditions for Filter Banks 9. Tree Structured Filter Banks and Multiresolution Analysis 

Electrical \& Computer Engineering
University of Maryland, College Park

Acknowledgment: ENEE630 slides were based on class notes developed by Profs. K.J. Ray Liu and Min Wu. The LaTeX slides were made by Prof. Min Wu and Mr. Wei-Hong Chuang.
Contact: minwu@umd.edu. Updated: October 6, 2011.

## Recall: Simple Filter Bank Systems



If all $S_{k}(z)$ are identical as $S(z)$ :

$$
\begin{aligned}
& \mathbb{P}(z)=S(z) \mathbb{I} \\
& \Rightarrow \hat{X}(z)=z^{-(M-1)} S\left(z^{M}\right) X(z)
\end{aligned}
$$

Alias Free


## General Alias-free Condition

Recall from Section 7: The condition for alias cancellation in terms of $\mathcal{H}(z)$ and $\mathbb{f}(z)$ is

$$
\mathcal{H}(z) \underline{\mathbb{E}}(z)=\mathbb{t}(z)=\left[\begin{array}{c}
M A_{0}(z) \\
0 \\
\vdots \\
0
\end{array}\right]
$$

## Theorem

A M-channel maximally decimated filter bank is alias-free iff the matrix $\mathbb{P}(z)=\mathbb{R}(z) \mathbb{E}(z)$ is pseudo circulant.
[ Readings: PPV Book 5.7 ]

## Circulant and Pseudo Circulant Matrix

(right-)circulant matrix

$$
\left[\begin{array}{lll}
P_{0}(z) & P_{1}(z) & P_{2}(z) \\
P_{2}(z) & P_{0}(z) & P_{1}(z) \\
P_{1}(z) & P_{2}(z) & P_{0}(z)
\end{array}\right]
$$

Each row is the right circular shift of previous row.
pseudo circulant matrix

$$
\left[\begin{array}{ccc}
P_{0}(z) & P_{1}(z) & P_{2}(z) \\
z^{-1} P_{2}(z) & P_{0}(z) & P_{1}(z) \\
z^{-1} P_{1}(z) & z^{-1} P_{2}(z) & P_{0}(z)
\end{array}\right]
$$

Adding $z^{-1}$ to elements below the diagonal line of the circulant matrix.

- Both types of matrices are determined by the 1st row.
- Properties of pseudo circulant matrix (or as an alternative definition): Each column as up-shift version of its right column with $z^{-1}$ to the wrapped entry.


## Proof of the Theorem

## (Details)



Denote $\mathbb{P}(z)=\left[P_{s, \ell}(z)\right]$.

## Overall Transfer Function

The overall transfer function $T(z)$ after aliasing cancellation:
$\hat{X}(z)=T(z) X(z)$, where

$$
T(z)=z^{-(M-1)}\left\{P_{0,0}\left(z^{M}\right)+z^{-1} P_{0,1}\left(z^{M}\right)+\cdots+z^{-(M-1)} P_{0, M-1}\left(z^{M}\right)\right\}
$$

## Most General P.R. Conditions

Necessary and Sufficient P.R. Conditions

$$
\mathbb{P}(z)=c z^{-m_{0}}\left[\begin{array}{cc}
0 & \mathbb{I}_{M-r} \\
z^{-1} \mathbb{I}_{r} & 0
\end{array}\right] \text { for some } r \in 0, \ldots, M-1
$$

When $r=0, \mathbb{P}(z)=\mathbb{I} \cdot c z^{-m_{0}}$, as the sufficient condition seen in §1.7.3.
(Details)

## (Binary) Tree-Structured Filter Bank

A multi-stage way to build $M$-channel filter bank:
Split a signal into 2 subbands $\Rightarrow$ further split one or both subband signals into $2 \Rightarrow \cdots$


Question: Under what conditions is the overall system free from aliasing? How about P.R.?

## (Binary) Tree-Structured Filter Bank



- Can analyze the equivalent filters by noble identities.
- If a 2-channel QMF bank with $H_{0}^{(K)}(z), H_{1}^{(K)}(z), F_{0}^{(K)}(z)$, $F_{1}^{(K)}(z)$ is alias-free, the complete system above is also alias-free.
- If the 2-channel system has P.R., so does the complete system.
[ Readings: PPV Book 5.8 ]

Multi-resolution Analysis: Analysis Bank

Consider the variation of the tree structured filter bank (ie., only split one subband signals)


Multi-resolution Analysis: Synthesis Bank


Synthesis Bank

* $\left\{v_{k}[n]\right\}$ have the same sampling rate as $x[n]$ and $\hat{x}[n]$
$y_{3}[n] \longrightarrow \hat{r}_{2} \longrightarrow F_{3}(z) \xrightarrow{V_{3} n n} \xrightarrow[x]{ }[n]$ They are called the multinesolution components.


## Discussions

(1) The typical frequency response of the equivalent analysis and synthesis filters are:

(2) The multiresolution components $v_{k}[n]$ at the output of $F_{k}(z)$ :

- $v_{0}[n]$ is a lowpass version of $x[n]$ or a "coarse" approximation;
- $v_{1}[n]$ adds some high frequency details so that $v_{0}[n]+v_{1}[n]$ is a finer approximation of $x[n]$;
- $v_{3}[n]$ adds the finest ultimate details.


## Discussions

(3) If 2-ch QMF with $G(z), F(z), G_{s}(z), F_{s}(z)$ has P.R. with $\underline{\text { unit-gain }}$ and zero-delay, we have $x[n]=x[n]$.
(4) For compression applications: can assign more bits to represent the coarse info, and the remaining bits (if available) to finer details by quantizing the refinement signals accordingly.

## Brief Note on Subband vs Wavelet Coding

- The octave (dyadic) frequency partition can reflect the logarithmic characteristics in human perception.
- Wavelet coding and subband coding have many similarities (e.g. from filter bank perspectives)
- Traditionally subband coding uses filters that have little overlap to isolate different bands
- Wavelet transform imposes smoothness conditions on the filters that usually represent a set of basis generated by shifting and scaling (dilation) of a mother wavelet function
- Wavelet can be motivated from overcoming the poor time-domain localization of short-time FT
$\Rightarrow$ Explore more in Proj\#1. See PPV Book Chapter 11


## Detailed Derivations

## Details of the Proof (1)



Denote $\mathbb{P}(z)=\left[P_{s, \ell}(z)\right]$.

Details of the Proof (2)
(1)

Taking ZT on $C_{l}[n]$, we have :

$$
\begin{aligned}
& C_{l}(z)=\frac{1}{M} \sum_{K=0}^{M-1}\left(W_{M}^{K} z^{1 / M}\right)^{-l} X\left(W_{M}^{K} z^{1 / M}\right) \text { for } 0 \leqslant l \leqslant M-1 \\
& W_{M}=e^{j 2 \pi / M}
\end{aligned}
$$

For $Z T$ on $b_{e}[u]$, we have:

$$
\begin{aligned}
& B_{s}(z)=\sum_{l=0}^{M-1} P_{s_{1} l}(z) C_{l}(z) \\
& \begin{aligned}
\hat{X}(z) & =\sum_{s=0}^{M-1} z^{-(M-s-1)} B_{s}\left(z^{M}\right)=\sum_{s=0}^{M-1} \sum_{l=0}^{M-1} z^{-(M-s-1)} P_{s i l}\left(z^{M}\right) C_{l}\left(z^{M}\right) \\
& =\frac{1}{M} \sum_{k=0}^{M-1} X\left(W_{M}^{K} z\right) \sum_{l=0}^{M-1} W_{M}^{-K l} \underbrace{\sum_{s=0}^{M-1} z^{-l} z^{-(M-s-1)} P_{s_{r} l}\left(z^{M}\right)}_{0=0}
\end{aligned}
\end{aligned}
$$

We have alias-free iff $\sum_{l=0}^{M-1} W_{M}^{-K l} V_{l}(z)=0$ for $K \neq 0$.

Details of the Proof (3)
(2) The above nece-suff condition in matrix-vector form is:

$$
W^{H}\left[\begin{array}{c}
V_{0}(z) \\
V_{1}(z) \\
\vdots \\
V_{M-1}(z)
\end{array}\right]=\left[\begin{array}{c}
* \\
0 \\
\vdots \\
0
\end{array}\right]
$$

Where * represents a nou-zero entry, $A^{H}$ represents the transposeconjugate of a matrix $A$, and $W$ is the $M \times M$ DFT matrix: $[W]_{k m}=W_{M}^{K m}=e^{j 2 \pi \frac{k m}{M}}, 0 \leqslant K \leqslant M-1,0 \leqslant m \leqslant M-1, W^{H} W=W W^{H}=M I$. all-one entries

$$
\left.\therefore\left[\begin{array}{c}
V_{0}(z) \\
\vdots \\
V_{M-1}(z)
\end{array}\right]=W\left[\begin{array}{c}
* \\
0 \\
\vdots \\
0
\end{array}\right]=\text { W's first column }=\begin{array}{c}
* \\
* \\
\vdots \\
*
\end{array}\right]
$$

ie. $\quad V_{0}(\xi)=V_{1}(\xi)=\cdots=V_{M-1}(\xi) \triangleq V_{(\xi)}$

Details of the Proof (4)
(3) From the definition $V_{e}(z)=\sum_{s=0}^{M-1} z^{-l} z^{-(M-1-S)} P_{S_{1} l}\left(z^{M}\right)$, we have.

$$
\begin{aligned}
& V_{0}(z)=\sum_{s=0}^{M-1} \delta^{-(\mu-S)} P_{s_{1}}\left(z^{\mu}\right) \quad V_{1}(z)=z^{-} \sum_{s=0}^{\mu-1} z^{-(M-1-s) P_{S_{1}}\left(子^{\mu}\right)}
\end{aligned}
$$

Since $V_{0}(z)=V_{1}(z)$, we have

ie. each column is the up -shift version of its right column with $z^{-1}$ to the heaped entry

identical along diagonal directions

## Concluding the Proof

The same relation can be found for other columns.
Thus $\mathbb{P}(z)$ is pseudo circulant for an alias-free system as a necessary and sufficient condition.

We've done with the proof of the theorem.

Overall Transfer Function

The overall transfer function $T(z)$ after aliasing cancellation:

$$
\begin{aligned}
\hat{X}(z) & =T(z) X(z), \text { where } \\
T(z) & =\frac{1}{M} \sum_{\ell=0}^{M-1} V_{\ell}(z)=\sum_{\ell=0}^{M-1} \sum_{s=0}^{M-1} z^{-(M-1+\ell-s)} P_{s, \ell}\left(z^{M}\right)
\end{aligned}
$$

We can represent entries of pseudo-circulant matrix in terms of the 0-th row entries:


$$
P_{S, l}(z)=z^{-1} P_{0, M+l-s}(z) \text { for } l<s \text {; }
$$

Let $r \triangleq l-s$, we have

$$
z^{-(l-s)} P_{s, l}\left(z^{M}\right)=\left\{\begin{array}{l}
z^{-r} P_{0, r}\left(z^{M}\right) \text { for } 0 \leqslant r \leqslant M-1 \\
z^{-(M+r)} P_{0, M+r}\left(z^{M}\right) \text { for } r<0 \text { such that } 0<M+\Gamma \leqslant M-1
\end{array}\right.
$$

- for each $P_{0, r}\left(z^{M}\right), \quad 0 \leqslant r \leqslant M-1$, we have $M$ such terms in $\left.T(z)\right]$

Overall Transfer Function

$$
\begin{aligned}
\Rightarrow T(z) & =\frac{1}{M} z^{-(M-1)} \sum_{\Gamma=0}^{M-1} M \times z^{-r} P_{0, r}\left(z^{M}\right) \\
& =z^{-(M-1)} \sum_{r=0}^{M-1} z^{-r} P_{0, r}\left(z^{M}\right)<\text { Note } z^{M} \text { here! } \\
& \text { (owingto } \$-M])
\end{aligned}
$$

i.e. given the $0^{\text {th }}$ row of the $\mathbb{P}(z)$ matrix, the system's transfer function is

$$
T(z)=z^{-(M-1)}\left[P_{0,0}\left(z^{\mu}\right)+z^{-1} P_{0,1}\left(z^{\mu}\right)+\ldots+z^{-(\mu-1)} P_{0, M-1}\left(z^{M}\right)\right]
$$

$$
T(z)=z^{-(M-1)}\left[P_{0,0}\left(z^{M}\right)+z^{-1} P_{0,1}\left(z^{M}\right)+\cdots+z^{-(M-1)} P_{0, M-1}\left(z^{M}\right)\right]
$$

Most General P.R. Conditions (necessary and sufficient)

Recall $\S 1.7 .3$ : sufficient condition for P.R. is $\mathbb{P}(z)=c z^{-m_{0}} \mathbb{I}$.
(1) To be free from aliasing, $\mathbb{P}(z)$ must be pseudocirculant.
(2) $T(z)$ must be a pure delay with possibly a constant
multiplicative factor i.e. $T(z)=c z^{-d}$.
$\Leftrightarrow$ iff the top row of $\mathbb{P}(z)$ is in the form of

$$
\left[0,0 \ldots 0, c z^{-m_{0}}, 0, \ldots 0\right]
$$

Thus $\mathbb{P}(z)=c z^{-m_{0}}\left[\begin{array}{cc}0 & I_{M-r} \\ z^{-1} I_{r} & 0\end{array}\right]$ for some $r \in[0, M-i]$
When $r=0, \mathbb{P}(z)=I \cdot c z^{-m_{0}}$ as in $\xi 1.7 .3$
The overall transfer function is

$$
T(z)=c \cdot z^{-(M-1)} z^{-1} z^{-M_{0} M}
$$

