

# What's the Modified DFT?

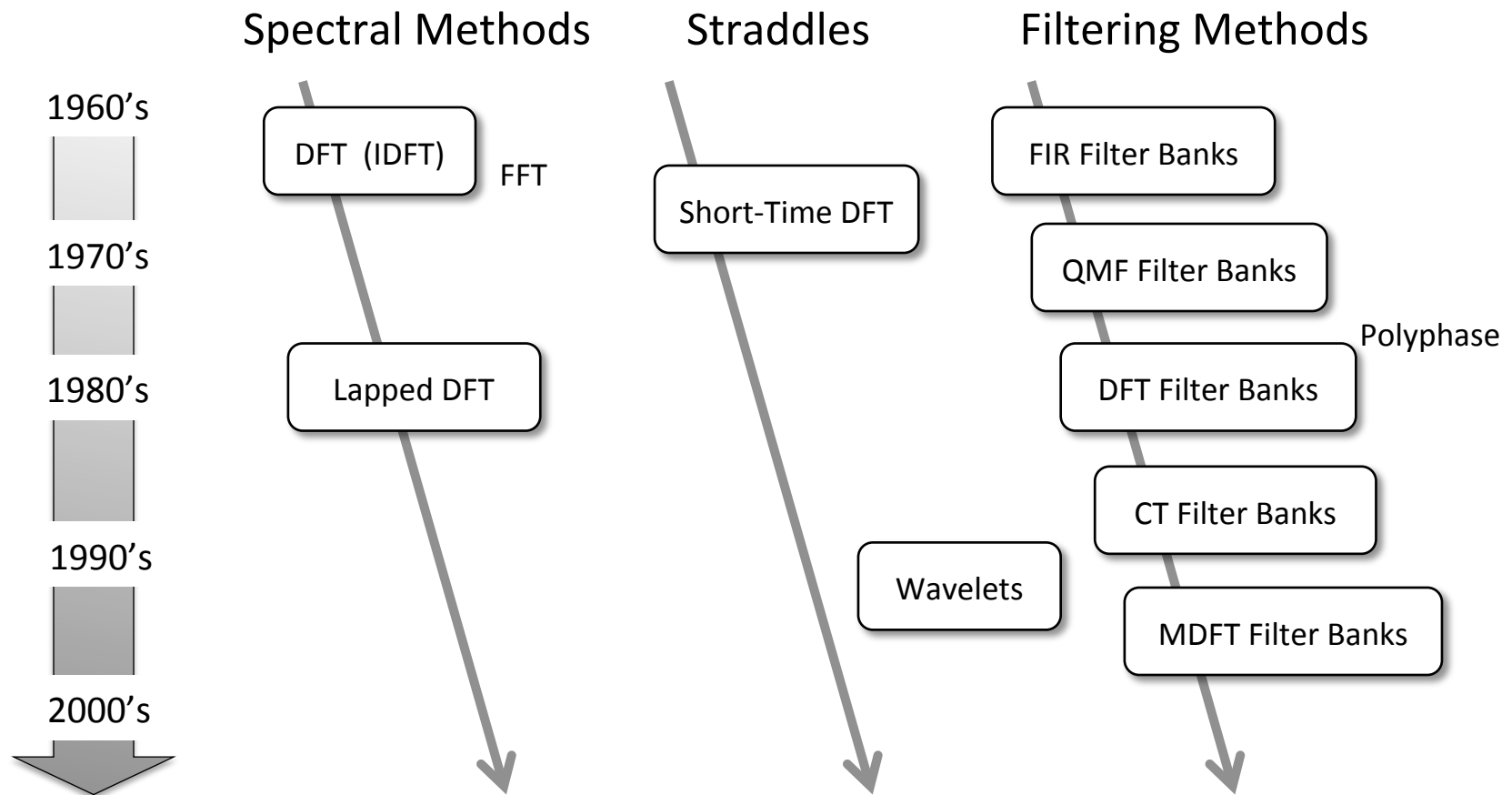
Keith Kumm  
Short Talk Topic

# Coverage

- Goal is for Everyone to Understand MDFT Basics
- Background Comments on Filter Banks in general
- Prime Motive: Perfect Reconstruction
- Warm-up: A few fundamental relations & memories
- QMF, a Two-Channel FB
- Polyphase Decomposition with DFT Matrix
- Uniform DFT, an  $M$ -Channel FB
- MDFT (finally!)
- Some Applications

# A “Natural History” of Analysis (Synthesis)

By 1990, Plenty of Clues Pointed to the MDFT



# MDFT Evolved from DFT Filter Banks

- **Grail: Perfect Reconstruction (PR) with (High) Resolution & Efficiency**

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

- Meaningful if Reconstruction is Part of the Design Objective
- Often, Modified Reconstruction is the Actual Objective
- *Also* Leads to High Resolution, High Accuracy Analysis (or Synthesis)
- Non- or Near-PR means

$$\hat{X}(z) = cz^{-k} X(z) \quad \hat{X}(z) = cz^{-k} T(z) X(z) + \text{aliasing terms}$$

- **Way Forward**
  - Obtain Filter Passband Flat Amplitude and Linear Phase
  - Obtain Passband Cross-over Equalization
  - Eliminate or Make Negligible Inter-Passband Aliasing
  - Exploit Complexity Reduction Inherent in Polyphase Realization

# Warm-Up 1

- z-Transform
  - If convergent on unit circle
$$X(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x(k)e^{-j\omega k} \quad z = e^{j\omega} \quad X(z) = \sum_{k=-\infty}^{\infty} x(k)z^{-k}$$
- Delay (by  $n$ )  $y(n) = x(n - n_0)$   $Y(z) = \sum_{k=-\infty}^{\infty} x(k - n_0)z^{-k} = z^{-n_0} \sum_{m=-\infty}^{\infty} x(m)z^{-m}$
- LTI Operations
  - Linear
  - Time Invariant
$$y(n) = x(n) * h(n) \quad \leftrightarrow \quad Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega})$$
$$y(n) = H(e^{j\omega_0})e^{j\omega_0 n} \quad \text{if } x(n) = e^{j\omega_0 n}$$
- LTV Operations
  - Linear
  - Time VaryingAnything that behaves differently from sample to sample

# Warm-Up 2

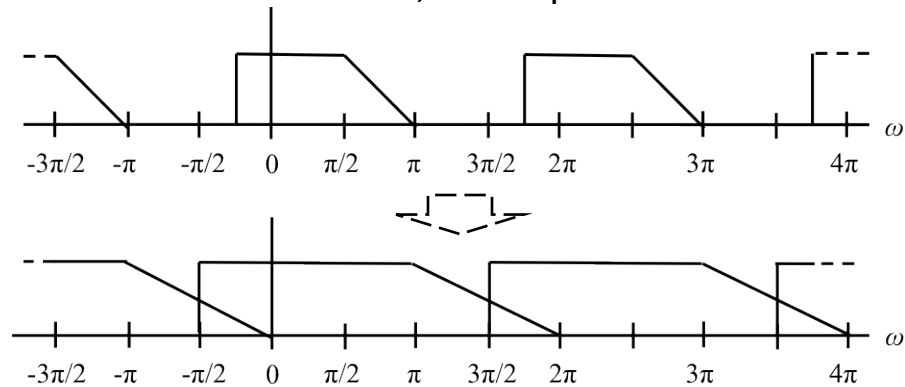
- Decimation (by  $M$ ) LTV !

$$\Omega_s = 2\pi / T_s \rightarrow \boxed{\downarrow M} \rightarrow \Omega_s \rightarrow \Omega_s / M$$

( deleted samples )

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

“stretch & shift,” example for  $M = 2$



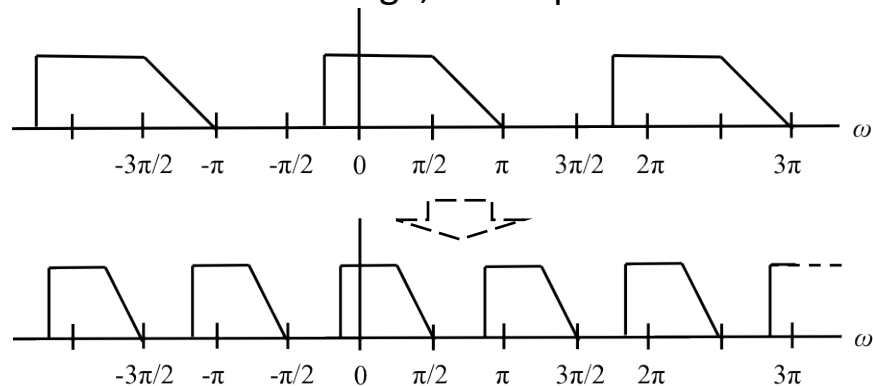
- Expansion (by  $L$ ) LTV !

$$\Omega_s = 2\pi / T_s \rightarrow \boxed{\uparrow L} \rightarrow \Omega_s \rightarrow L\Omega_s$$

( interstitial 0's )

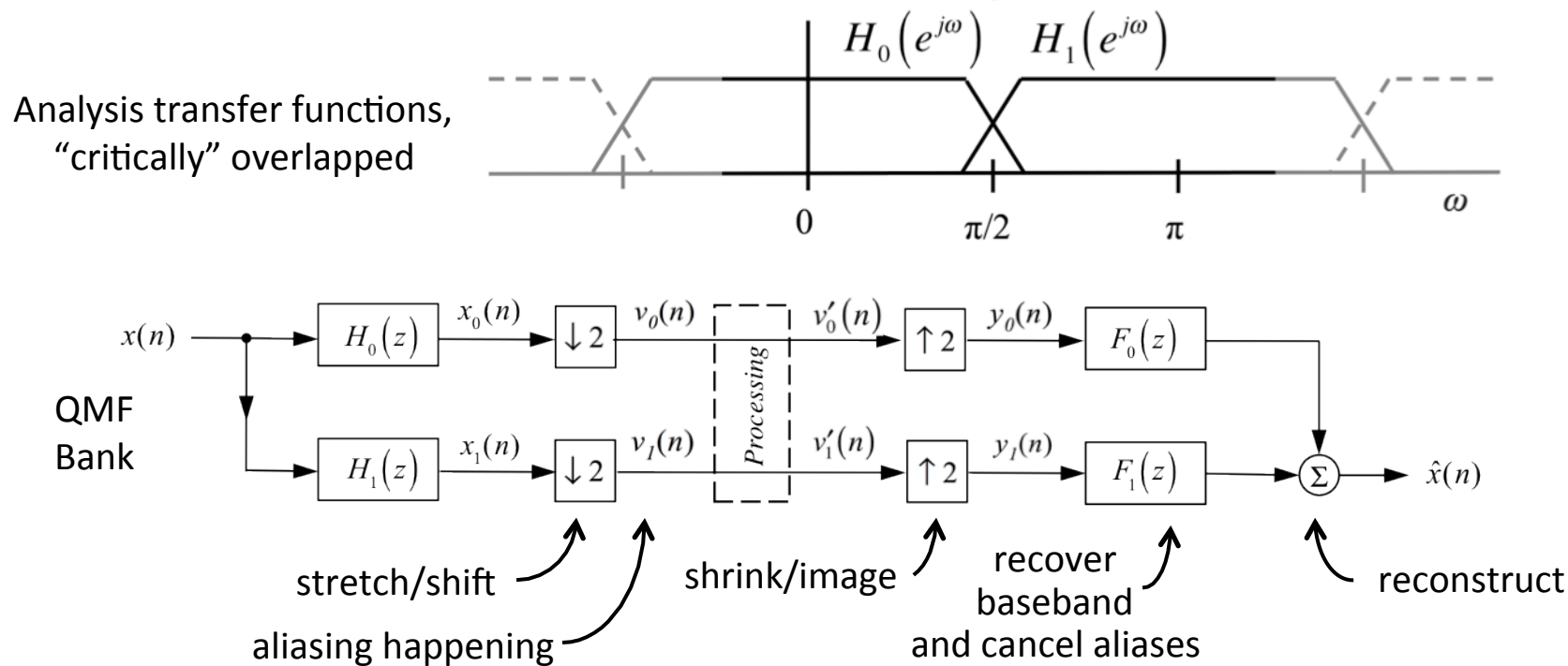
$$Y_E(e^{j\omega}) = X(e^{j\omega L})$$

“shrink & image,” example for  $L = 2$



# Quadrature Mirror Filter Bank (1970's)

- All  $x, v, y$  are time domain sequences
- $x(n), \hat{x}(n)$  are full bandwidth (e.g., for critical real sampling in  $[0, \pi)$ )



# QMF Alias Cancellation & PR

- By Careful Set-up of Synthesis Filters  $F_0(z), F_1(z)$
- Works for Any Mirror-Symmetric  $H_0(z), H_1(z)$  s.t.  $H_1(z) = H_0^*(-z) = H_0(-z)$
- $\hat{X}(z) = F_0(z)Y_0(z) + F_1(z)Y_1(z)$  Now, Let Synthesis Filter Design be:

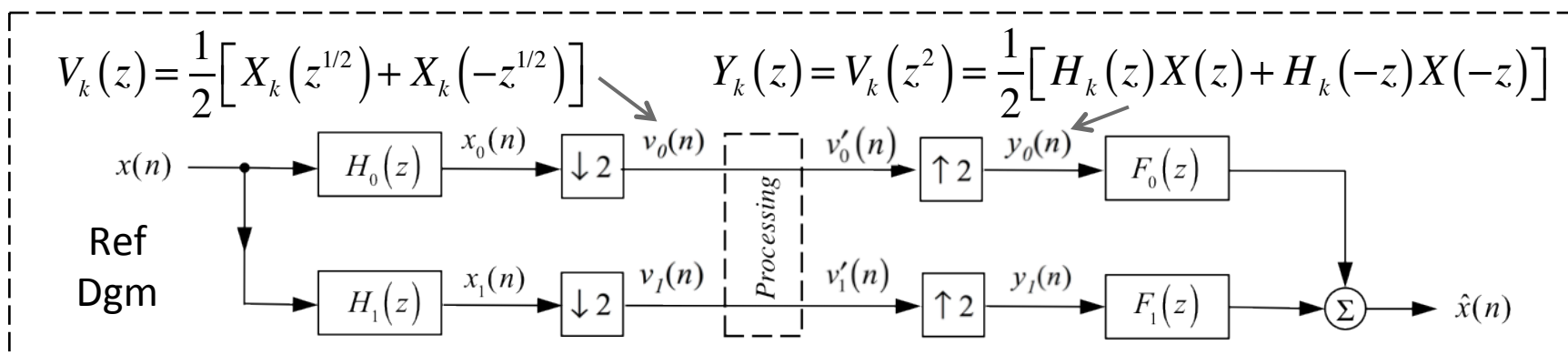
$$F_0(z) = H_1(-z), F_1(z) = -H_0(-z)$$

And via the Ref Dgm:

$$\hat{X}(z) = \frac{1}{2} [H_0(z)F_0(z) + H_1(z)F_1(z)] X(z) + \frac{1}{2} [H_0(-z)F_0(z) + H_1(-z)F_1(z)] X(-z)$$

alias!

$$\hat{X}(z) = T(z)X(z) = \frac{1}{2} [H_0(z)H_0^*(z) + H_1(z)H_1^*(z)] X(z) = \frac{|H_{0+1}(z)|^2}{2} z^{-2h} X(z) \quad \text{PR !}$$





# (Two Channel) Polyphase Decomposition

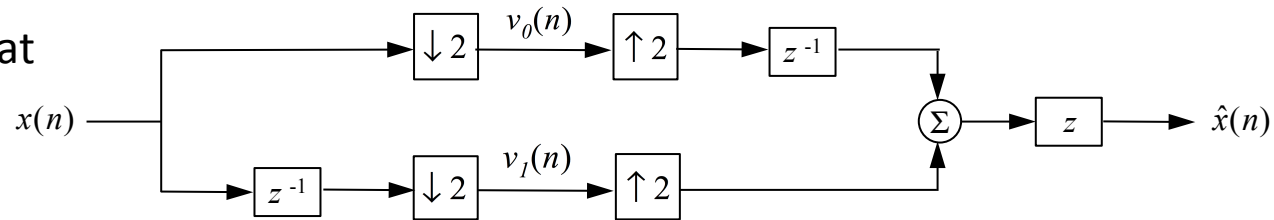
- Goal: eliminate redundancies, simplify computation
- “Modulation” matrix version of QMF *without*  $\downarrow 2$  and  $\uparrow 2$

$$\hat{X}(z) = \frac{1}{2} \begin{pmatrix} F_0(z) & F_1(z) \end{pmatrix} \begin{pmatrix} H_0(z) & H_0(-z) \\ H_1(z) & H_1(-z) \end{pmatrix} \begin{pmatrix} X(z) \\ X(-z) \end{pmatrix} = \frac{1}{2} \mathbf{F}_m(z) \mathbf{H}_m(z) \mathbf{x}_m(z)$$

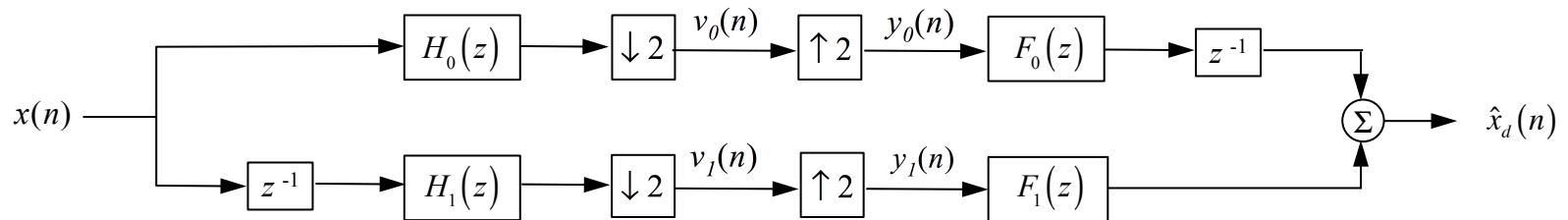
- Terms  $H_k(z)$ ,  $H_k(-z)$  have same coefficients and are redundant
- If we can eliminate the multiplication redundancies\*, we are ahead
  - \* even at the cost of some extra additions
- Very similar to the computation “gain” of the FFT over the DFT

# Polyphase QMF, 1

- Observe that

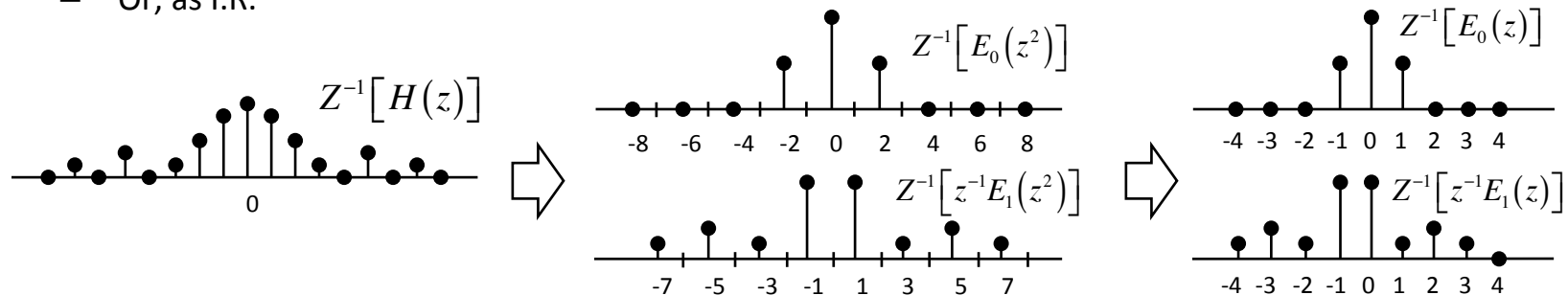


- Now Add Some Utility



- Consider for  $H$ , an FIR
  - Or, as I.R.

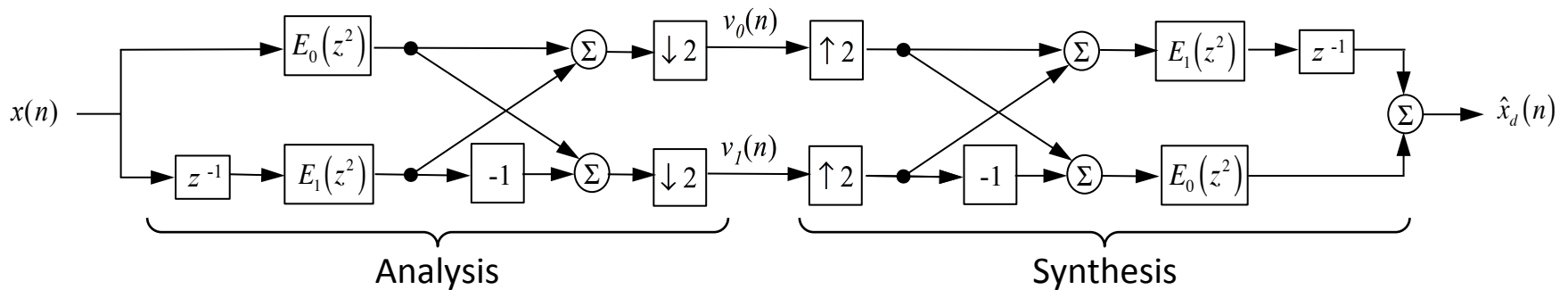
$$H(z) = E_{\text{even}}(z^2) + z^{-1} E_{\text{odd}}(z^2) \equiv E_0(z^2) + z^{-1} E_1(z^2)$$



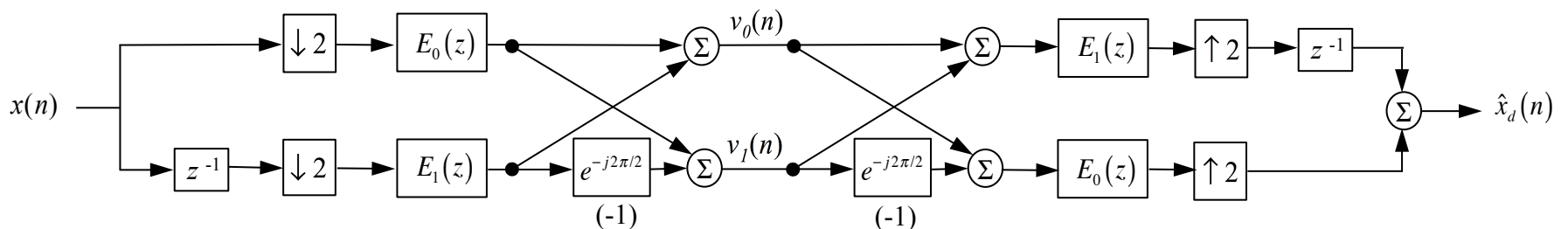
# Polyphase QMF, 2

- Now Recall for QMF  $H_1(z) = H_0(-z)$   $F_0(z) = H_1(-z)$ ,  $F_1(z) = -H_0(-z)$

- So on "Analysis" Side 
$$\begin{pmatrix} H_0(z) \\ H_1(z) \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} E_0(z^2) \\ z^{-1} E_1(z^2) \end{pmatrix}$$



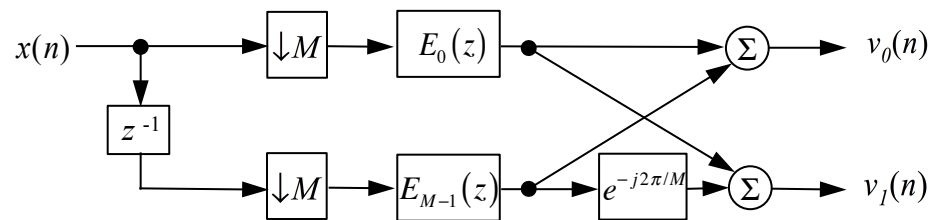
- Equivalently



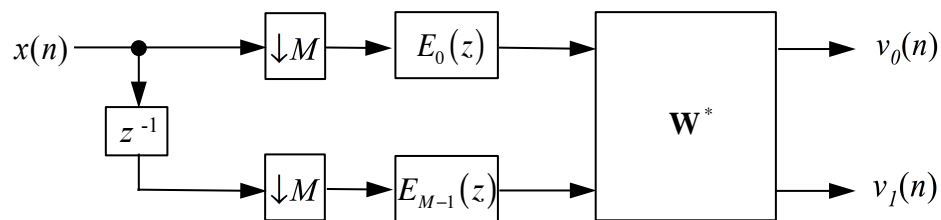
# Polyphase Generalization, 1

## with a Look-Ahead to DFT

- Consider Just Analysis Side of the QMF, where  $M = 2$



- Turns Out that, again  $M = 2$



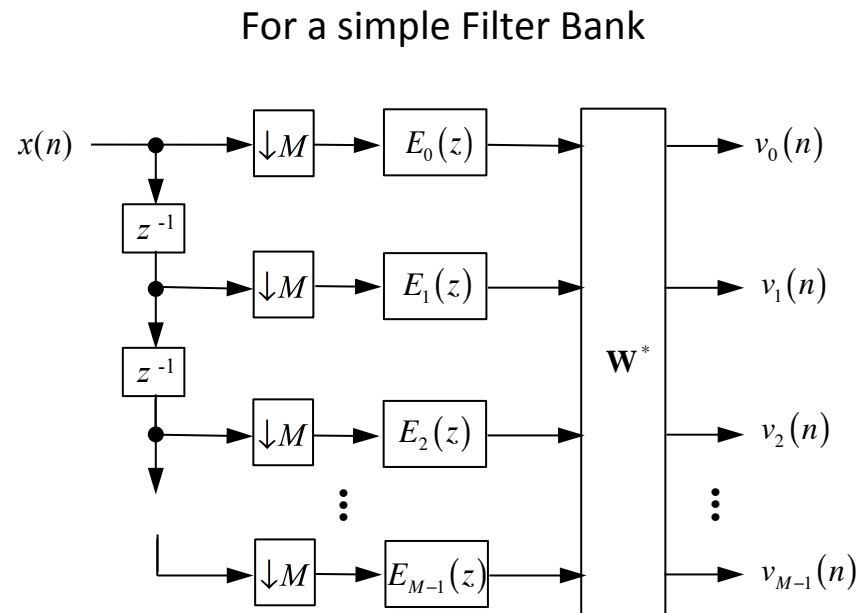
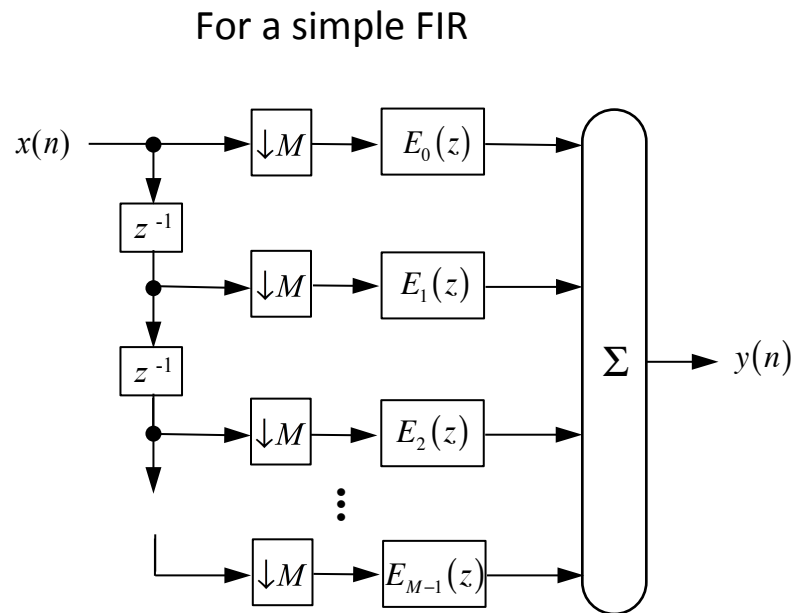
$$\mathbf{W}^* = \begin{bmatrix} W_M^{-km} \end{bmatrix}$$

$$W_M = e^{-j2\pi l/M}$$

# Polyphase Generalization, 2

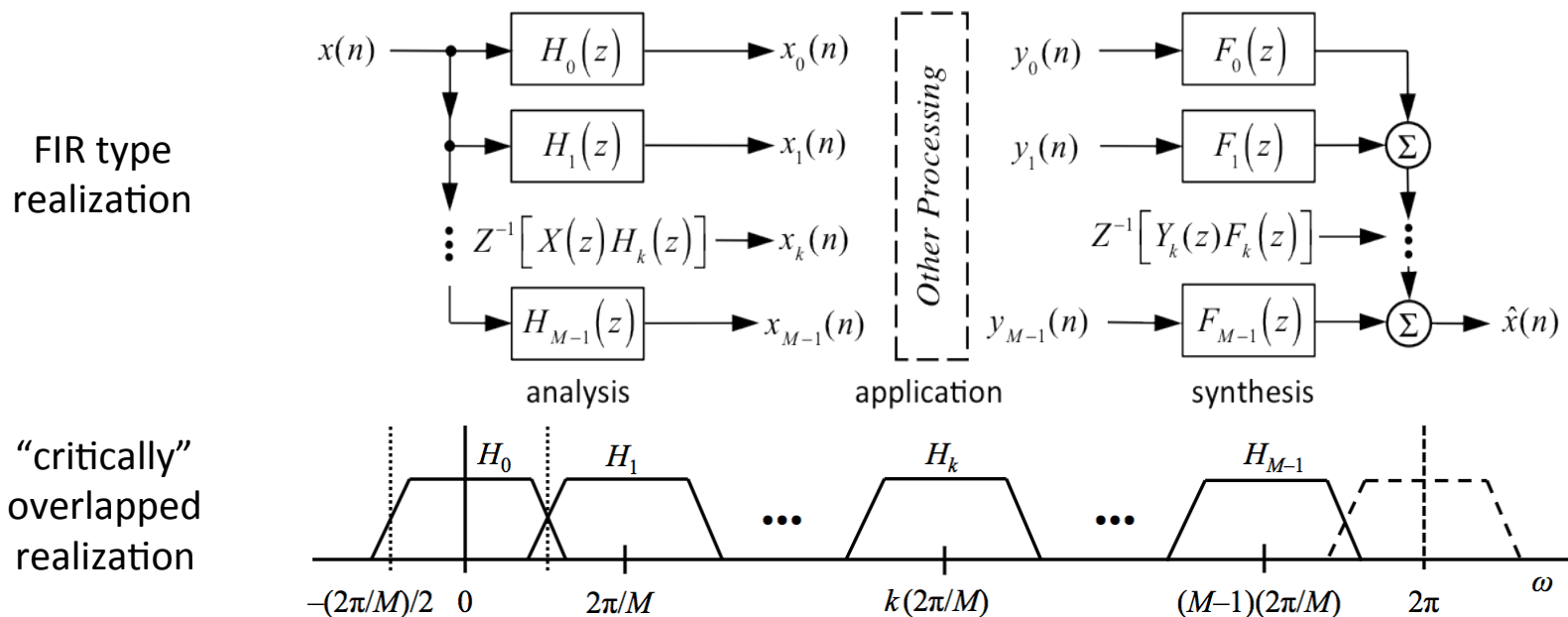
## with Sneak Peak at DFT Filter Bank

- We can Generalize to any  $M$  constrained only that  $\log_2 M = K$ , an integer



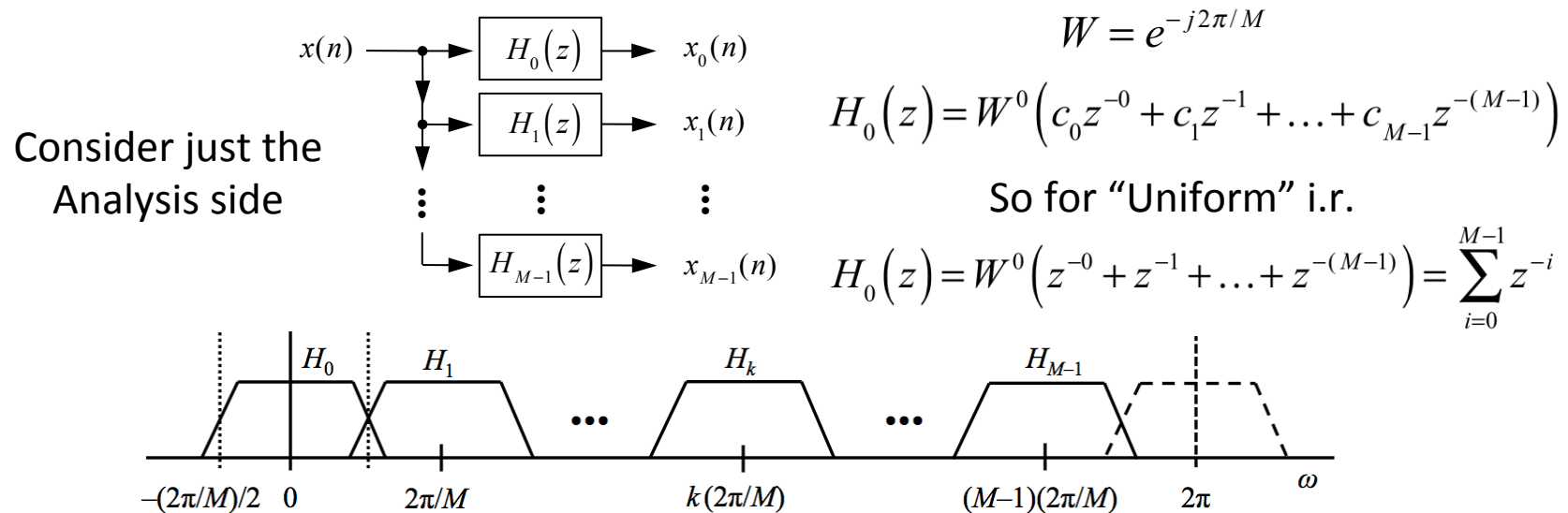
again  $\mathbf{W}^* = [\mathbf{W}_M^{-km}]$   $\mathbf{W}_M = e^{-j2\pi/M}$

# General Discrete (Time) Filter Bank (1970's)



- All  $x, y$  are time domain sequences
- $x(n), \hat{x}(n)$  are full bandwidth (e.g., for complex samples in  $[0, 2\pi)$ )
- $H_0(z), F_0(z)$  are baseband in  $\left[-(2\pi/M)/2, +(2\pi/M)/2\right]$
- $k = 1, M-1$ :  $H_k(z)$  are bandpass and downconvert,  $F_k(z)$  upconvert
- $x_k(n), y_k(n)$  are all baseband processes centered on 0 rad
- Aliasing happens in the Processing, and also is not correctable at  $\hat{x}(n)$

# Discrete Filter Bank – Uniform Case



Replica Filters on center freqs  $\omega_k = (k / M)2\pi$       so       $H_k(z) = H_0(zW^k)$

And for Uniform case       $h_0(n) = W^0 \delta(n-i) = \delta(n-i), \quad h_k(n) = W^{-ki} \delta(n-i)$

$H_0(z)$  will (eventually for MDFT) be called a "prototype" filter

$$X_k(z) = \sum_{i=0}^{M-1} H_k(z) X(z) = \sum_{i=0}^{M-1} H_0(zW^k) X(z) = \sum_{i=0}^{M-1} \left( \sum_{i=0}^{M-1} (zW^k)^{-i} \right) X(z) = \sum_{i=0}^{M-1} z^{-i} W^{-ki} X(z)$$

## Discrete Fourier Transform (DFT) & Inverse DFT (IDFT)

**$N \times N$  DFT Matrix**  $\mathbf{W}_N = [W_N^{km}]$ ,  $W_{(N)} = W = e^{-j2\pi/N}$  symmetric !

Commonly employed as  $X(k) = \sum_{m=0}^{N-1} W^{km} x(m)$

Observe that  $\mathbf{W}^\dagger \mathbf{W} = N \mathbf{I}$ ,  $\mathbf{W}^\dagger = (\mathbf{W}^T)^* = \mathbf{W}^*$

Meaning that  $\mathbf{W}^{-1} = \frac{\mathbf{W}^\dagger}{N} = \frac{\mathbf{W}^*}{N}$

$$\mathbf{W}_4 = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & +j \\ 1 & -1 & 1 & -1 \\ 1 & +j & -1 & -j \end{pmatrix}$$

**$N \times N$  IDFT Matrix**  $N^{-1} \mathbf{W}^* = N^{-1} [W_N^{-km}] = N^{-1} [(W_N^{km})^{-1}]$

Commonly employed as  $x(m) = \frac{1}{N} \sum_{k=0}^{N-1} W^{-km} X(k)$

or in matrices

$$\mathbf{x} = \mathbf{W}^* \mathbf{X}$$

$$\mathbf{W}_4^* = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & +j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & -1 & +j \end{pmatrix}$$



# Applying IDFT to Uniform Discrete FB

Recall

from IDFT (as a template)

from Uniform Discrete FB

Sticking with just the Analysis side

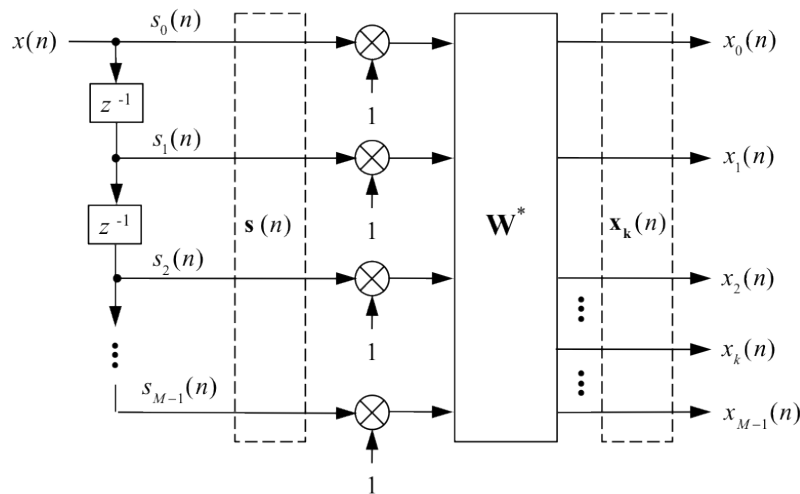
$$x(m) = \frac{1}{N} \sum_{k=0}^{N-1} W^{-km} X(k)$$

$$X_k(z) = \sum_{i=0}^{M-1} z^{-i} W^{-ki} X(z)$$

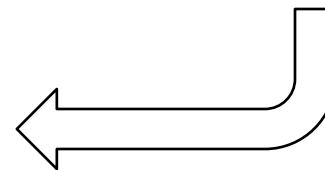
$$\mathbf{x} = \mathbf{W}^* \mathbf{X}$$

$$Z^{-1}[X_k(z)] = Z^{-1} \left[ \sum_{i=0}^{M-1} W^{-ki} z^{-i} X(z) \right]$$

$$x_k(n) = \sum_{i=0}^{M-1} W^{-ki} x(n-i) = \sum_{i=0}^{M-1} W^{-ki} s_i(n)$$



$$\mathbf{x}_k = \mathbf{W}^* \mathbf{x}$$

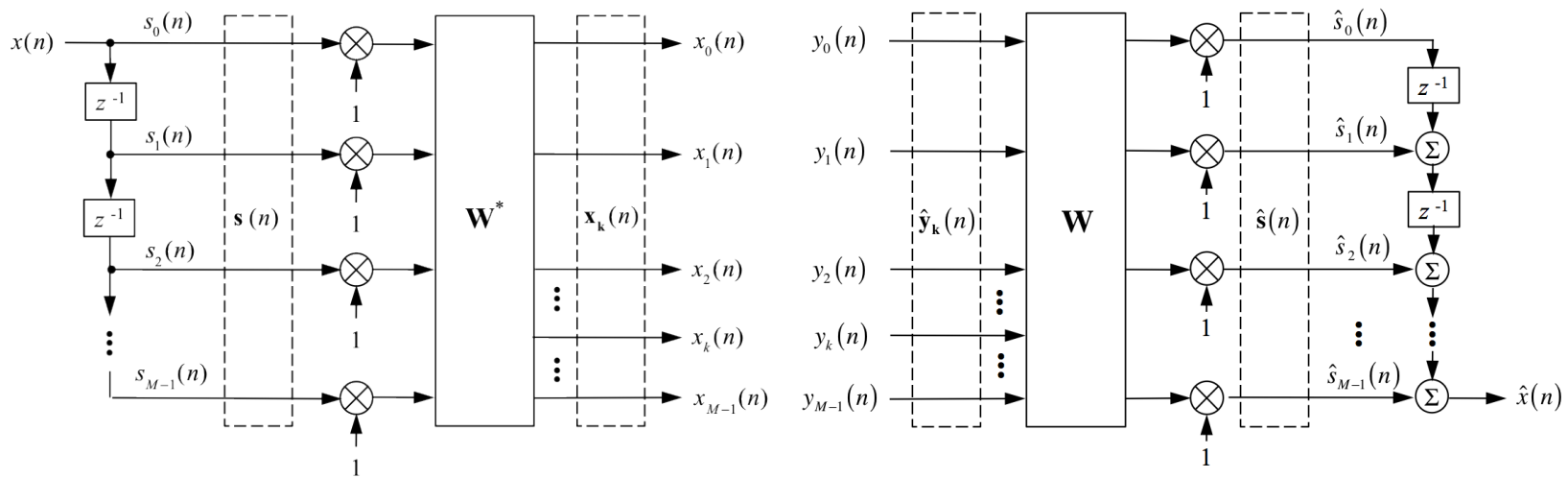


# Uniform Complex-Modulated DFT FB (1980's)

A similar thing happens on Synthesis side, but this time involving the DFT Matrix  $\mathbf{W}$

$$\text{(since of course)} \quad \mathbf{W}^\dagger \mathbf{W} = \mathbf{W}^* \mathbf{W} = (N/N) \mathbf{W}^* \mathbf{W} = N \mathbf{W}^{-1} \mathbf{W} = N \mathbf{I}$$

And we end up with the C-M DFT Filter Bank



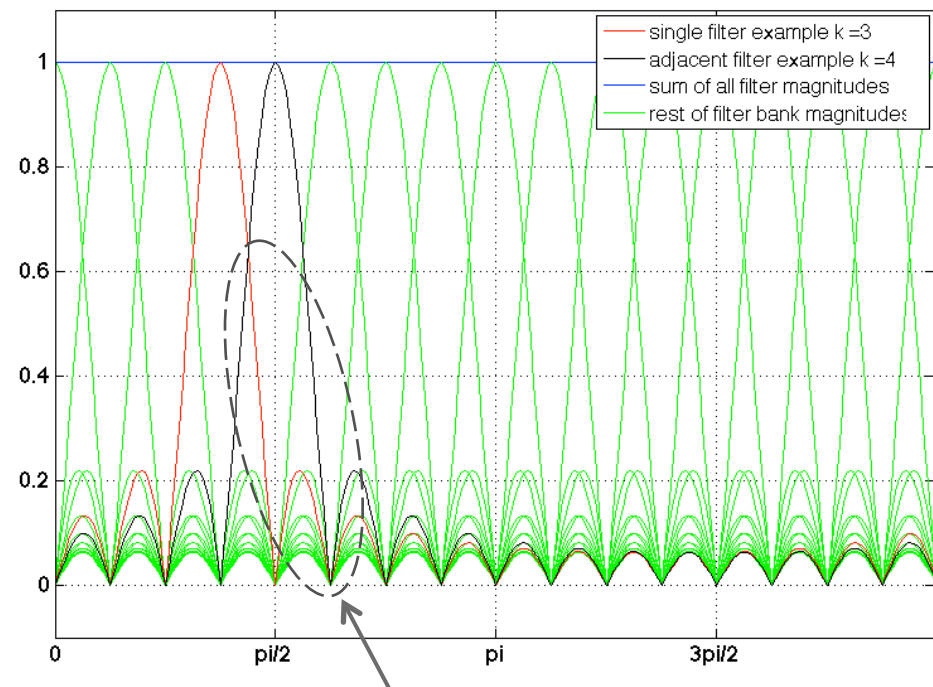
As is, the Uniform C-M DFT Filter Bank is PR !

# Uniform C-M DFT Analysis Freq Response

$$M = 16$$

$$\left| H_k(e^{j\omega}) \right|, \left| H_3(e^{j\omega}) \right|, \sum_{k=0}^{15} \left| H_k(e^{j\omega}) \right|$$

- Sum of individual filters (green & black) = 1 (blue) suggests PR !
- And indeed, entire bank (analysis + synthesis) can be made so
- But a single filter response (e.g., red at  $k = 3$ ) is **very poor** due to only using the uniform window
- Hardly flatness and selectivity of a good FIR, much sidelobe aliasing
- If only a better filter element could be had – a big suggestion – without giving up the relative simplicity of the DFT structure



Inter-filter aliasing is a very significant effect in the FB's "Processing" interior. This doesn't mean however that the entire bank can't be made PR !

# Uniform DFT FB Freq Response

## Is that |sinc| or |“sinq”| ?

- Freq Response of Block of 1's, i.e.  $H_0(z) = \sum_{i=0}^{M-1} z^{-i}$

Recall  $\sum_{i=0}^{M-1} q^i = \frac{(q^M - 1)}{q - 1}$       So  $H_0(q^{-1}) = \sum_{i=0}^{M-1} q^i = \frac{(q^M - 1)}{q - 1}$

$$H_0(e^{j\omega}) = \frac{(e^{-j\omega M} - 1)}{(e^{-j\omega} - 1)} = \left( \frac{e^{-jM\omega/2}}{e^{-j\omega/2}} \right) \frac{(e^{-jM\omega/2} - e^{jM\omega/2}) / (2j)}{(e^{-jM\omega/2} - e^{jM\omega/2}) / (2j)} = e^{-j(M-1)\omega/2} \frac{\sin(M\omega/2)}{\sin(\omega/2)}$$

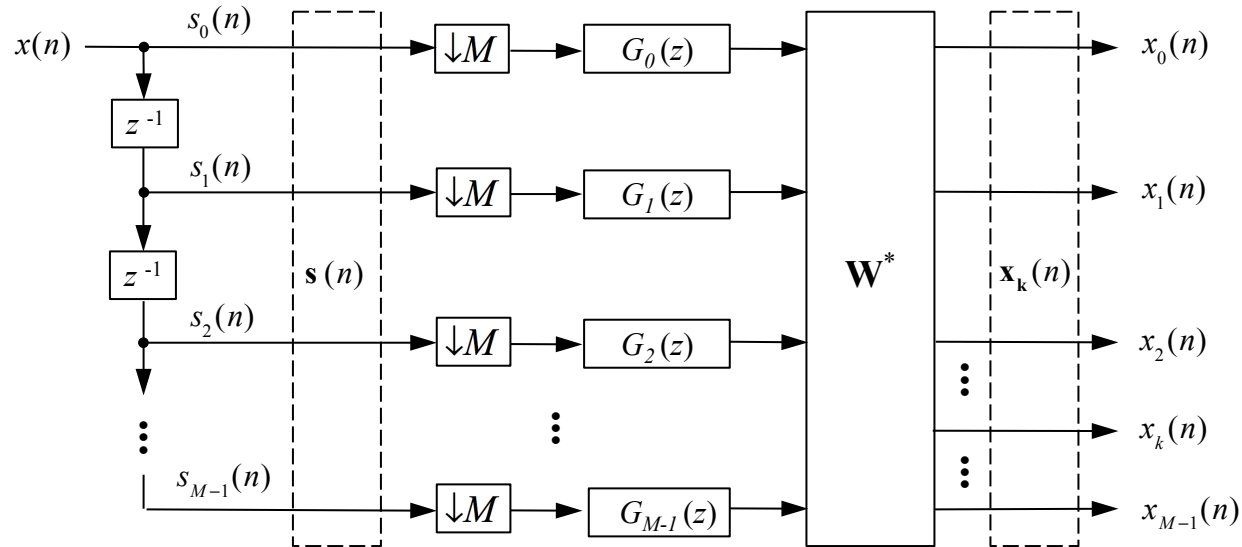
$$|H_0(e^{j\omega})| = \left| e^{j(M-1)\omega/2} \frac{\sin(M\omega/2)}{\sin(\omega/2)} \right| = \left| \frac{\sin(M\omega/2)}{\sin(\omega/2)} \right|$$

$$\angle H_0(e^{j\omega}) = \angle e^{j(M-1)\omega/2} + 0 = (M-1)\omega/2$$

that's no sinc !  
Call it a “sinq”

# General Complex-Modulated DFT FB with Polyphase Filtering, Analysis Side

## Add a Better Filter, But There's a Catch



General C-M DFT Filter Bank is the Most Computationally Efficient FB Structure

**But, General C-M DFT Filter Bank is No Longer PR !**

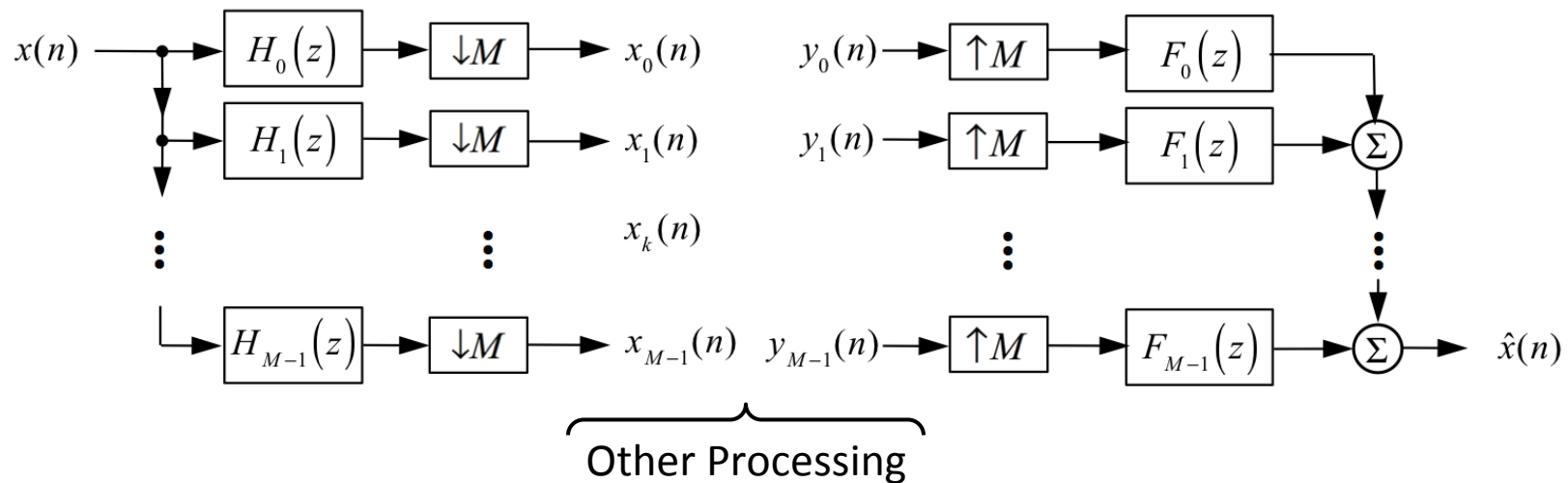
# Credit

- Essential to Acknowledge Karp & Fliege at this Point
  - See Bibliography
  - As Far as I Know, They Discovered the MDFT
  - Most Following Diagrams and Observations follow Theirs
  - Some Streamlining and Focusing in This Short Topic Talk
  - There is NO NEW ORIGINAL WORK here
- We are Left to Exploit this Technology Wisely
  - No Advice is Offered on Patents and IP in MDFT

# Modified DFT (MDFT)

(Not Yet Polyphase)

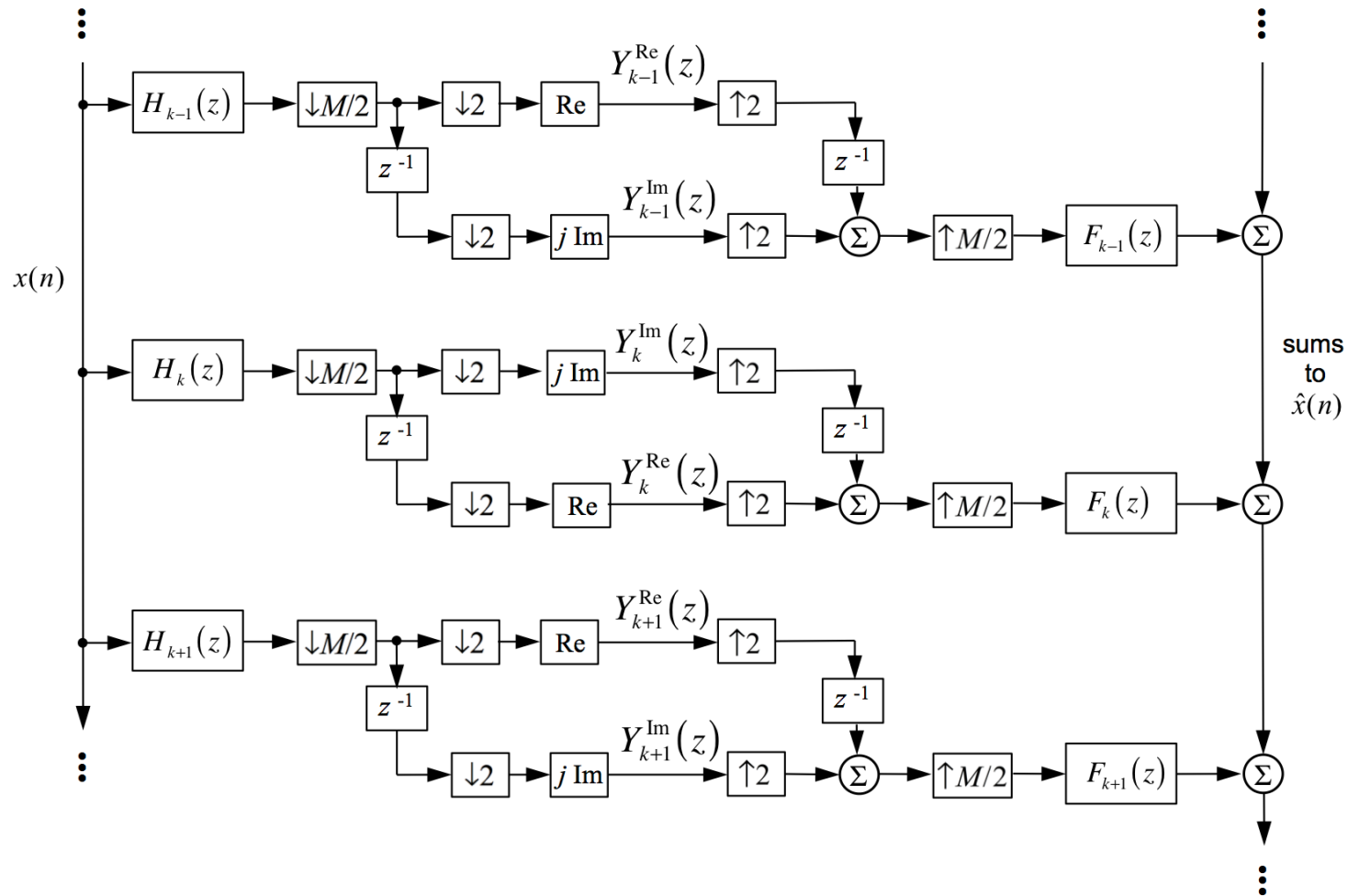
Start with our old friend the Discrete Filter Bank,  
 Add decimation (then expansion) to obtain high  
 computational efficiency for any Other Processing,  
 Then take the plunge into Alias Cancellation



# Modified DFT (MDFT)

A Most Talented Stroke (Karp & Fliege, ~1993)

redraw Discrete FB this way, *almost* exact

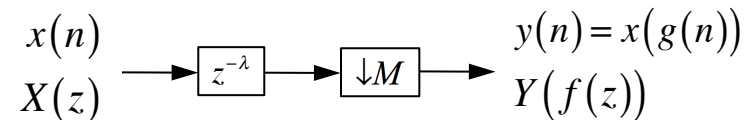




# Modified DFT (MDFT)

## Supporting Concept

z-transform of a *delayed*, decimated signal



On the delay side\*

$$y(n) = x(n \cdot M - \lambda), \lambda \geq 0$$

$$z^{-1} \leftrightarrow e^{1 \cdot (+j2\pi/M)} = W^{-1}, W = e^{-j2\pi/M}$$

$$z^{-\lambda} \leftrightarrow e^{\lambda \cdot (+j2\pi/M)} = W^{-\lambda}$$

On the decimation side

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

$$Y_D(z) = \frac{1}{M} \sum_{k=0}^{M-1} X(z^{1/M} W^k), z = e^{j\omega}$$

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

\* including  $\downarrow M$  effect

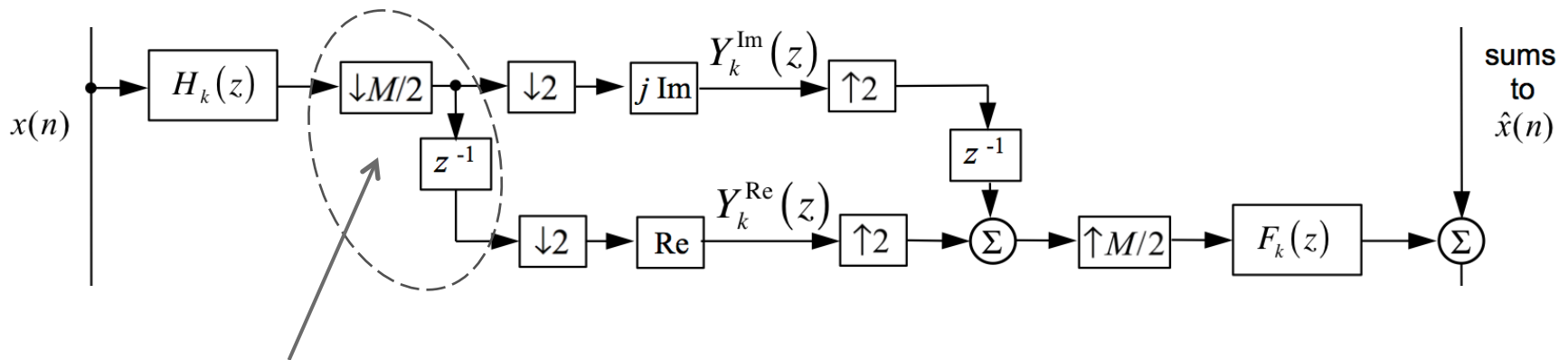
Put them together

$$z^{-\lambda} Y_D(z^M) = \frac{1}{M} \sum_{k=0}^{M-1} X(z W^{k(1-\lambda)}) = \frac{1}{M} \sum_{k=0}^{M-1} X(z W^k) W^{-\lambda k}$$

# Modified DFT (MDFT)

## Path to “Structure-based” Anti-Alias Cancellation

new structure gets some redundancy with a Twist



Due to decimation followed by delay, the lower path always gets  $W_M^{-kM/2} = e^{jk\pi}$  extra phase in the  $l$  th lobe of the frequency response of  $H_k(z)$

For  $l$  even, this is just 1, but for  $l$  odd, it is  $\pm j$ , alternating with  $k$

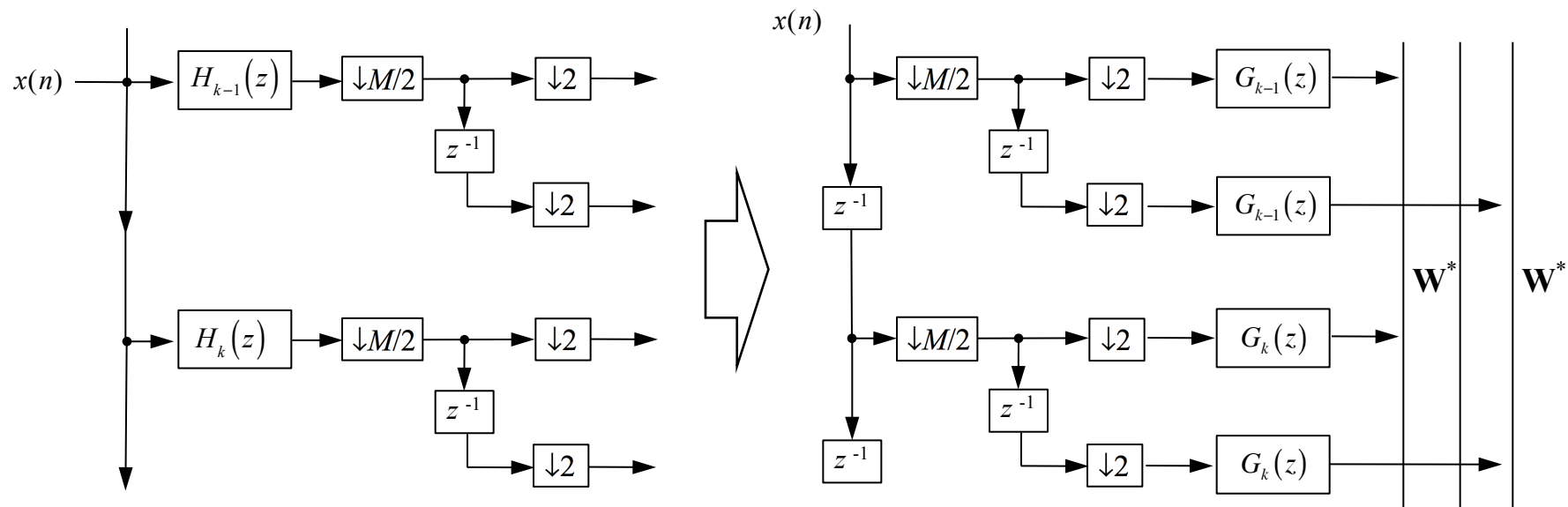
It is this very characteristic that forces odd sidelobe cancellation in the MDFT !

**MDFT Filter Bank is “Almost Perfect” Reconstruction !**

# Modified DFT (MDFT)

## Adding in Polyphase Efficiency, Step 1

using same transformation as for General DFT FB, Focus on Analysis Side



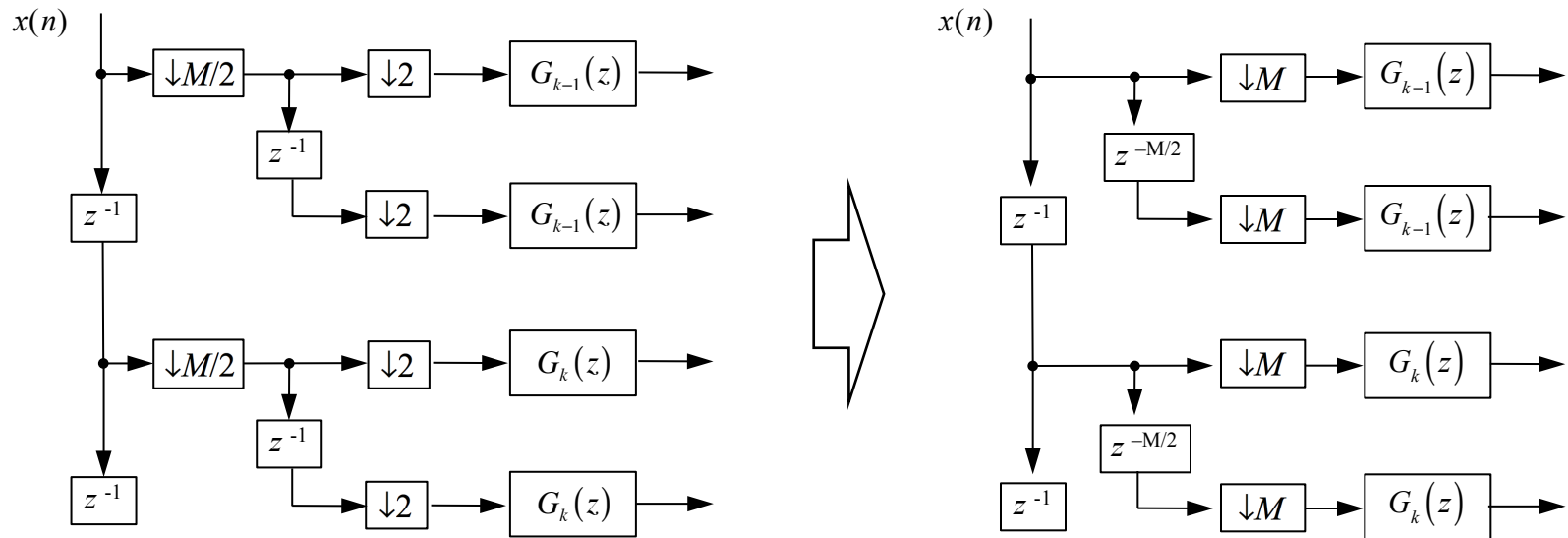
$G_k(z)$  is  $1/M$  the length of  $H_k(z)$

**Polyphase MDFT Filter Bank will Achieve Both Highest Efficiency and Near PR !**

# Modified DFT (MDFT)

## Intermediate Step 2

“pull through” the Unit Delay into an  $M/2$  Delay, Analysis Side  
 this transformation is exact

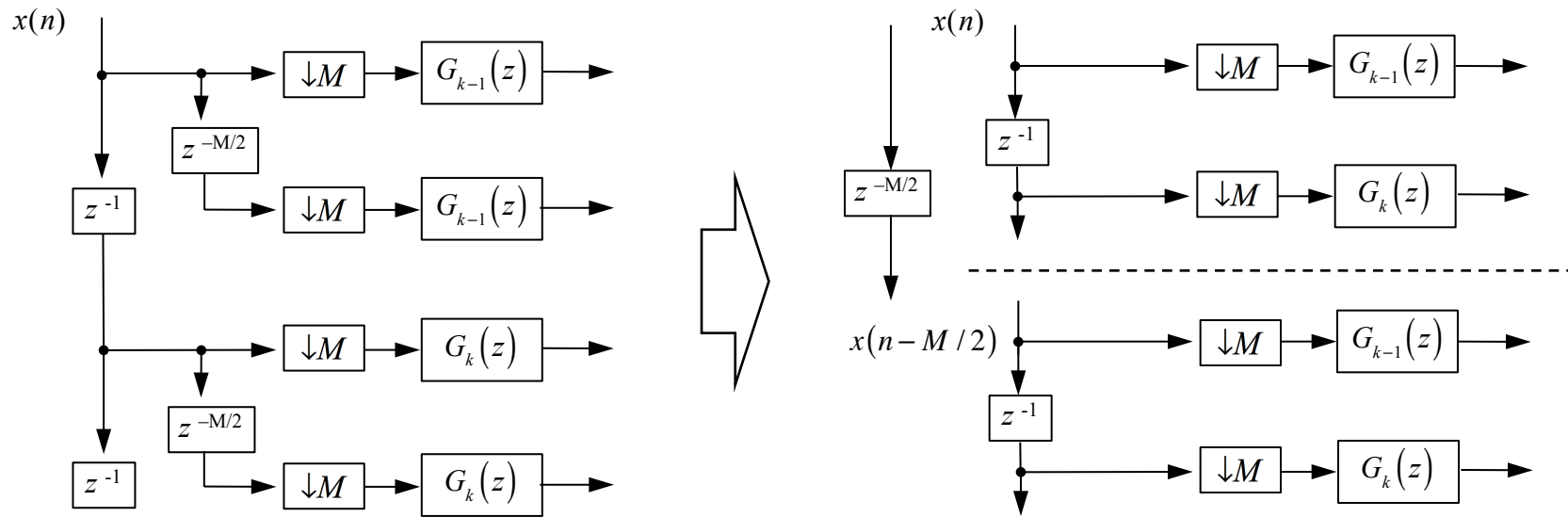


working toward a more regular repeated structure

# Modified DFT (MDFT)

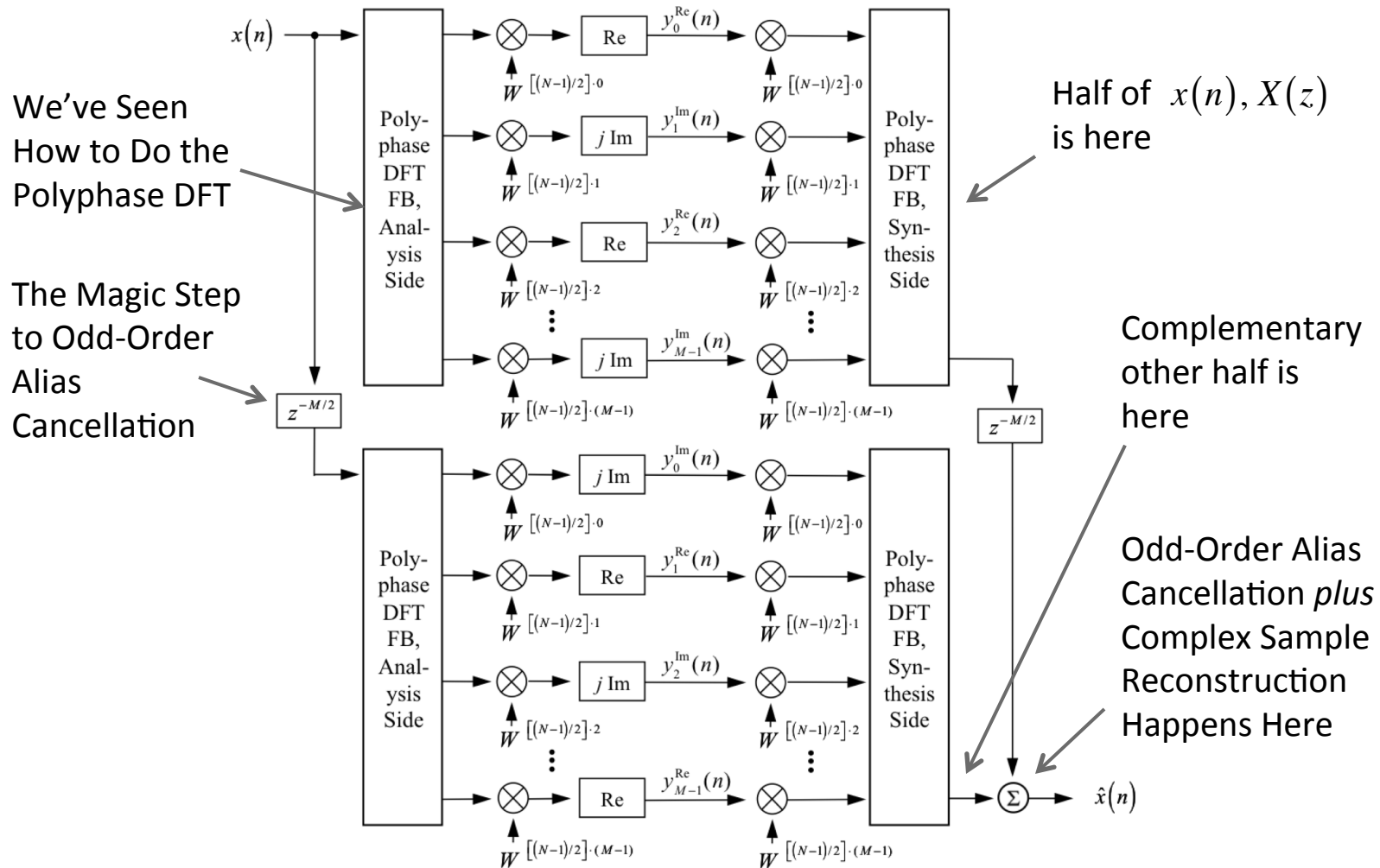
## Get Back to Minimum Memory, Step 3

“pull through” the entire  $M/2$ -delayed “Lower Path,” Analysis Side



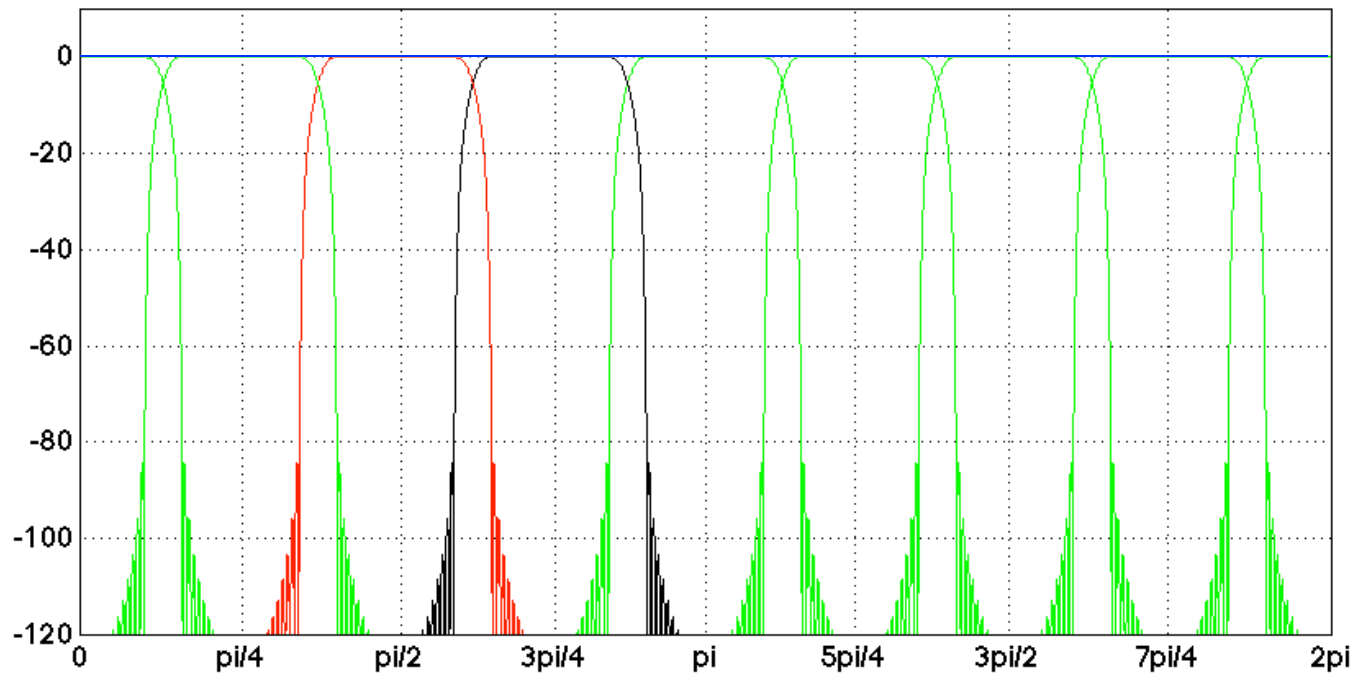
only take the  $M/2$  delay once

# Full-Up MDFT, “Dual-DFT” Configuration

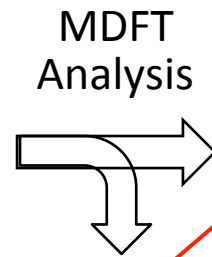
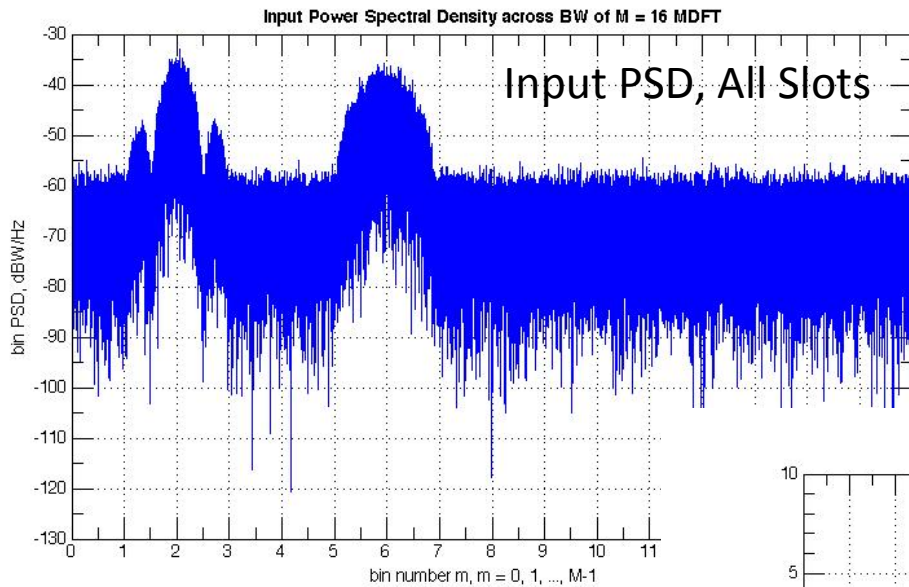


# MDFT Analysis Side Freq Response

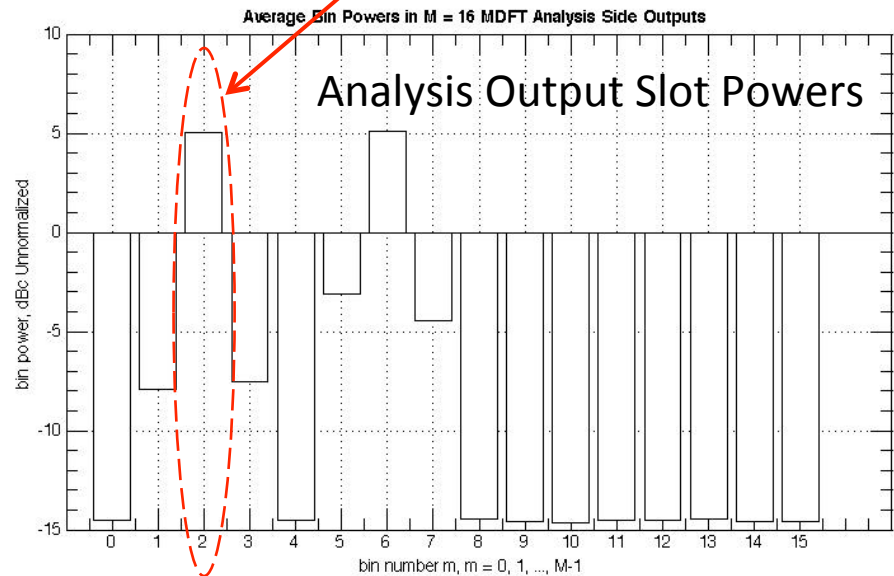
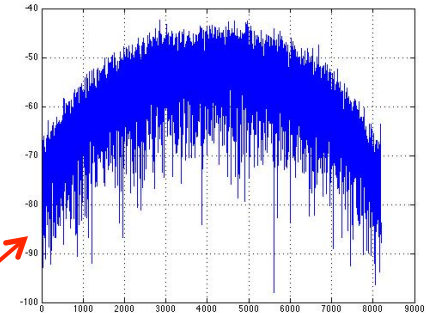
$M = 8, \alpha = 0.22$  RRC Prototype Filter



# Example, Analysis Side



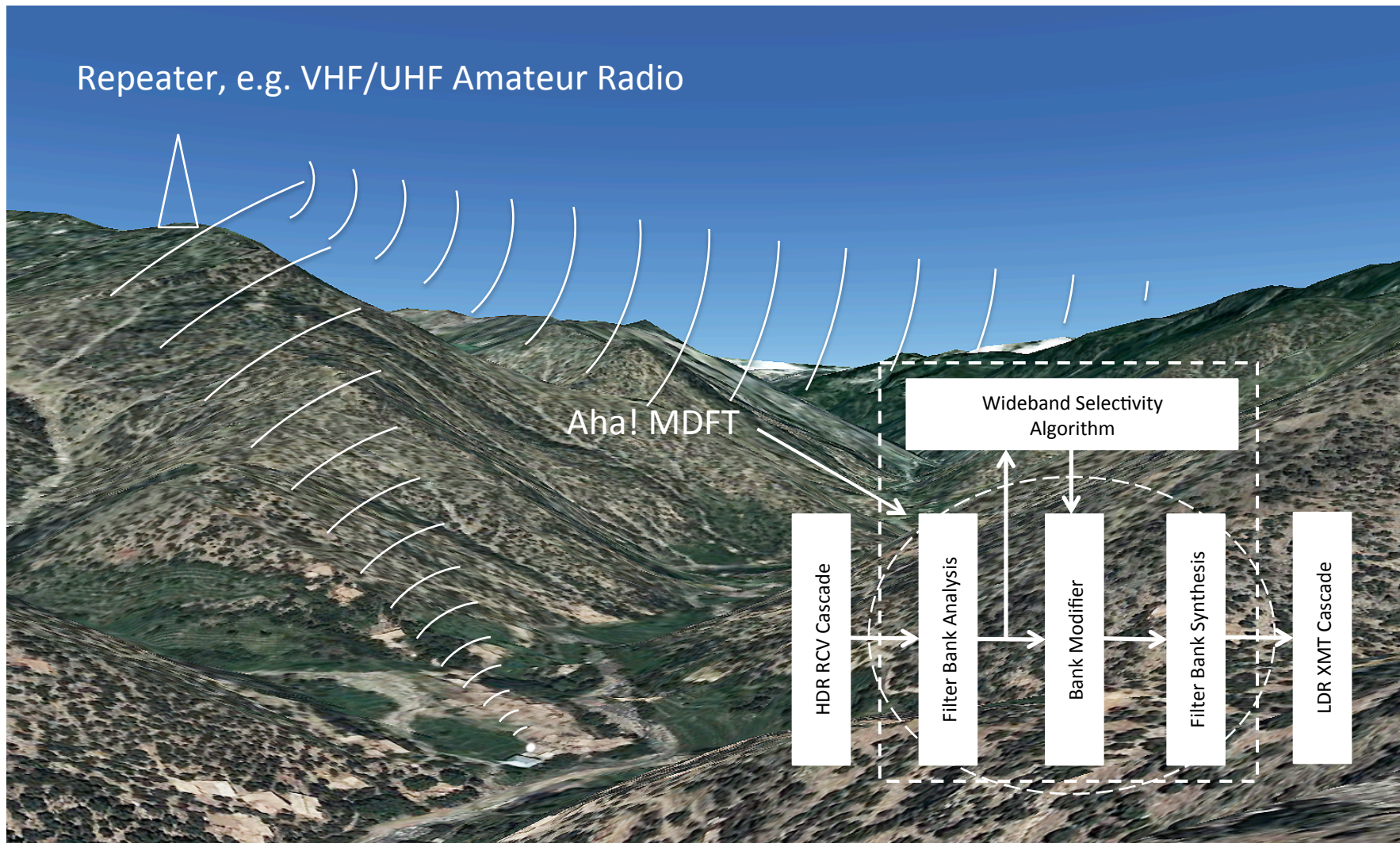
Output PSD, Slot 2





# An RF Type MDFT Application

## When You See Everything, The Problem is Selectivity



# Some Bibliography

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