Design of Optimal Quincunx Filter Banks for Image Coding

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May 23, 2006



- Quincunx Filter Banks
- Optimal Design Algorithm
- 4 Design Examples

Introduction of Quincunx Filter Banks

• Two-dimensional two-channel nonseparable filter banks



Quincunx lattice



$$oldsymbol{M} = egin{bmatrix} 1 & 1 \ 1 & -1 \end{bmatrix}$$

Motivation

• Desirable properties for image coding

- Perfect reconstruction (PR)
- Linear phase
- High coding gain
- Vanishing moments
- Good frequency selectivity
- Existing design methods
 - Transformation of variables
 - Direct optimization
 - Two-step lifting structure

Lifting Realization - Structure

• Analysis Side



Synthesis Side



Lifting Realization - Transfer Functions

• Analysis filter transfer functions $H_0(z)$ and $H_1(z)$

$$H_k(\mathbf{z}) = H_{k,0}\left(\mathbf{z}^{\mathbf{M}}\right) + z_0 H_{k,1}\left(\mathbf{z}^{\mathbf{M}}\right),$$

$$\begin{bmatrix} H_{0,0}(\boldsymbol{z}) & H_{0,1}(\boldsymbol{z}) \\ H_{1,0}(\boldsymbol{z}) & H_{1,1}(\boldsymbol{z}) \end{bmatrix} = \prod_{k=1}^{\lambda} \left(\begin{bmatrix} 1 & A_{2k}(\boldsymbol{z}) \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ A_{2k-1}(\boldsymbol{z}) & 1 \end{bmatrix} \right)$$

• Synthesis filter transfer functions $G_0(z)$ and $G_1(z)$

$$G_k(\mathbf{z}) = (-1)^{1-k} z_0^{-1} H_{1-k}(-\mathbf{z})$$

Lifting Realization - Advantages

- PR is satisfied automatically.
- ② Linear phase property can be imposed structurally.

Theorem

If each lifting filter A_k is symmetric with its group delay c_k satisfying

$$\boldsymbol{c}_k = (-1)^k \left[\frac{1}{2} \frac{1}{2} \right]^T,$$

then the analysis filters H_0 and H_1 are symmetric with group delays $\begin{bmatrix} 0 & 0 \end{bmatrix}^T$ and $\begin{bmatrix} -1 & 0 \end{bmatrix}^T$, respectively.

Reversible integer-to-integer transforms

Outline Introduction Quincunx Filter Banks Optimal Design Algorithm Design Examples Conclusion

Octave-Band Filter Banks

• N-level octave-band filter bank: analysis side



• Equivalent one-level analysis filters {H'_i}

$$H_{i}'(\boldsymbol{z}) = \begin{cases} \prod_{k=0}^{N-1} H_{0}\left(\boldsymbol{z}^{\boldsymbol{M}^{k}}\right) & i = 0\\ H_{1}\left(\boldsymbol{z}^{\boldsymbol{M}^{N-i}}\right) \prod_{k=0}^{N-i-1} H_{0}\left(\boldsymbol{z}^{\boldsymbol{M}^{k}}\right) & 1 \le i \le N-1\\ H_{1}\left(\boldsymbol{z}\right) & i = N. \end{cases}$$

Coding Gain

- Measure of the energy compaction ability of a filter bank
- Coding gain G_{SBC} for an N-level octave-band filter bank

$$G_{SBC} = \prod_{k=0}^{N} (A_k B_k / \alpha_k)^{-\alpha_k},$$
$$A_k = \sum_{\boldsymbol{m} \in \mathbb{Z}^2} \sum_{\boldsymbol{n} \in \mathbb{Z}^2} h'_k[\boldsymbol{m}] h'_k[\boldsymbol{n}] r[\boldsymbol{m} - \boldsymbol{n}], B_k = \alpha_k \sum_{\boldsymbol{n} \in \mathbb{Z}^2} g'^2_k[\boldsymbol{n}],$$
$$\alpha_0 = 2^{-N}, \alpha_k = 2^{-(N+1-k)} \text{ for } k = 1, 2, \dots, N,$$

Autocorrelation r

$$r[n_0, n_1] = egin{cases}
ho^{|n_0|+|n_1|} & ext{for separable model} \
ho^{\sqrt{n_0^2+n_1^2}} & ext{for isotropic model}, \end{cases}$$

where ρ is the correlation coefficient (typically, 0.90 $\leq \rho \leq$ 0.95).

Vanishing Moments

- \tilde{N} dual vanishing moments \Rightarrow \tilde{N} th order zero at $\begin{bmatrix} 0 & 0 \end{bmatrix}^T$ of $\hat{h}_1(\boldsymbol{\omega})$
- N primal vanishing moments \Rightarrow Nth order zero at $[\pi \ \pi]^T$ of $\hat{h}_0(\boldsymbol{\omega})$
- Linear phase filter H with group delay $\textbf{\textit{c}} \in \mathbb{Z}^2$

$$\frac{\partial^{m_0+m_1}\hat{h}}{\partial\omega_0^{m_0}\partial\omega_1^{m_1}} = \begin{cases} \sum_{\boldsymbol{n}\in\mathbb{Z}^2} h[\boldsymbol{n}] (\boldsymbol{n}-\boldsymbol{c})^{\boldsymbol{m}} \cos\left(\boldsymbol{\omega}^T (\boldsymbol{n}-\boldsymbol{c})\right) & \text{for } |\boldsymbol{m}| \text{ even} \\ -\sum_{\boldsymbol{n}\in\mathbb{Z}^2} h[\boldsymbol{n}] (\boldsymbol{n}-\boldsymbol{c})^{\boldsymbol{m}} \sin\left(\boldsymbol{\omega}^T (\boldsymbol{n}-\boldsymbol{c})\right) & \text{otherwise,} \end{cases}$$

where $\boldsymbol{m} = [m_0 \ m_1]^T$ and $\boldsymbol{m} = m_0 + m_1$. • \tilde{N} th order zero at $\boldsymbol{\omega} = [0 \ 0]^T$

$$\sum_{\boldsymbol{n}\in\mathbb{Z}^2}h[\boldsymbol{n}]\,(\boldsymbol{n}-\boldsymbol{c})^{\boldsymbol{m}}=0\quad\text{for all even }|\boldsymbol{m}|\text{ such that }|\boldsymbol{m}|<\tilde{N}.$$

Frequency Selectivity

• Error function of a linear phase filter H

$$e_{h}=\int_{\left[-\pi,\ \pi
ight]^{2}}W(oldsymbol{\omega})\left|\hat{h}_{s}(oldsymbol{\omega})-D\hat{h}_{d}(oldsymbol{\omega})
ight|^{2}\mathrm{d}oldsymbol{\omega}$$

• Ideal frequency responses and weighting function



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- Lifting parameterization of linear-phase filter banks
- Maximize coding gain subject to vanishing moments and frequency ٩ response constraints
- Iterative second-order cone programming

minimize $\boldsymbol{b}^T \boldsymbol{x}$ subject to: $\|\boldsymbol{A}_{i}^{T}\boldsymbol{x} + \boldsymbol{c}_{i}\| \leq \boldsymbol{b}_{i}^{T}\boldsymbol{x} + d_{i}$ for $i = 1, \dots, q$.

Linear/quadratic approximations

Two Lifting Steps - Problem Formulation (1)

- Lifting filter coefficients **x**
- Vanishing moments
 - Constraint: an underdetermined linear system Ax = b
 - Solutions: $\mathbf{x} = \mathbf{x}_s + \mathbf{V}_r \boldsymbol{\phi}$
- Coding gain
 - Define $G = -10 \log_{10} G_{SBC}$
 - ▶ For a given ϕ , seek a small perturbation δ_{ϕ} such that $G(\phi + \delta_{\phi})$ is reduced relative to $G(\phi)$
 - $\bullet \ \|\boldsymbol{\delta_{\phi}}\| \text{ is small} \Rightarrow G(\boldsymbol{\phi} + \boldsymbol{\delta_{\phi}}) \approx G(\boldsymbol{\phi}) + \boldsymbol{g}^{\mathsf{T}} \boldsymbol{\delta_{\phi}}$
 - ▶ Iteratively minimize $m{g}^{ op}m{\delta_{\phi}}$, update $m{\phi}$ until $|G(m{\phi}+m{\delta_{\phi}})-G(m{\phi})|<arepsilon$

Two Lifting Steps - Problem Formulation (2)

- Frequency selectivity
 - Analysis highpass filter frequency response

$$\hat{h}_1(oldsymbol{\omega}) = \hat{a}_1(oldsymbol{M}^{ op}oldsymbol{\omega}) + e^{j\omega_0}$$

where $\hat{a}_1(\pmb{M}^T \pmb{\omega})$ is linear in $\pmb{\phi}$

Error function

$$e_{h_1} = \boldsymbol{\phi}^T \boldsymbol{H}_{\boldsymbol{\phi}} \boldsymbol{\phi} + \boldsymbol{\phi}^T \boldsymbol{s}_{\boldsymbol{\phi}} + C_{\boldsymbol{\phi}}$$

Frequency response constraint is a second-order cone

$$\left\| \tilde{\boldsymbol{H}}_k \boldsymbol{\delta_{\phi}} + \tilde{\boldsymbol{s}}_k \right\| \leq \delta'_{h_1}$$

Introduction

Two Lifting Steps - Design Algorithm



Two Lifting Steps - Comments

- β : upper bound of $\|\boldsymbol{\delta_\phi}\|$
 - Too large: $\mathbf{g}^{T} \boldsymbol{\delta}_{\boldsymbol{\phi}}$ cannot correctly reflect the actual reduction in G
 - Too small: the solution to the SOCP subproblem is restricted to an unnecessarily small region around \u03c6k
 - Should be chosen such that

$$\boldsymbol{g}^{\mathsf{T}}\boldsymbol{\delta} pprox \mathcal{G}(\boldsymbol{\phi} + \boldsymbol{\delta}) - \mathcal{G}(\boldsymbol{\phi}) \quad ext{for} \quad \|\boldsymbol{\delta}\| = eta$$

- δ_{h_1} : upper bound of the error function e_{h_1}
 - Too small: feasible region may be empty
 - Chosen to be a scaled version of e_{h_1} evaluated at ϕ_k

$$\delta_{h_1} = d \left(\boldsymbol{\phi}_k^{\mathsf{T}} \boldsymbol{H}_{\boldsymbol{\phi}} \boldsymbol{\phi}_k + \boldsymbol{\phi}_k^{\mathsf{T}} \boldsymbol{s}_{\boldsymbol{\phi}} + c_{\boldsymbol{\phi}}
ight) \quad ext{for some} \quad 0 < d \leq 1$$

• Error e_{h_1} is reduced after each iteration.

Outline

More Than Two Lifting Steps - Problem Formulation

- Lifting filter coefficients **x**
- Coding gain: linear approximation

$$G(\mathbf{x} + \boldsymbol{\delta}_{\mathbf{x}}) = G(\mathbf{x}) + \mathbf{g}^{T} \boldsymbol{\delta}_{\mathbf{x}}$$

Vanishing moments

Introduction

- Polynomial equations in x
- Approximated by

$$oldsymbol{A}_koldsymbol{\delta}_{oldsymbol{x}}=oldsymbol{b}_k$$

- Moments are nearly vanishing
- Frequency selectivity
 - Frequency response: polynomial in x
 - Error function e_{h_1} : approximated by $\boldsymbol{\delta}_{\boldsymbol{x}}^T \boldsymbol{H}_k \boldsymbol{\delta}_{\boldsymbol{x}} + \boldsymbol{\delta}_{\boldsymbol{x}}^T \boldsymbol{s}_k + C_k$
 - Constraint: approximated by the second-order cone

$$\left\| \tilde{\boldsymbol{H}}_k \boldsymbol{\delta}_{\boldsymbol{x}} + \tilde{\boldsymbol{s}}_k \right\| \leq \delta'_{h_1}$$

Introduction

More Than Two Lifting Steps - Design Algorithm

Algorithm 2 **O** Select an initial point x_0 For the kth iteration, solve minimize $\boldsymbol{g}^{T}\boldsymbol{\delta}_{\boldsymbol{v}}$ subject to: $A_k \delta_x = b_k$ $\left\| \tilde{\boldsymbol{H}}_k \boldsymbol{\delta}_{\boldsymbol{x}} + \tilde{\boldsymbol{s}}_k \right\| \leq \delta'_{h_1}$ $\|\boldsymbol{\delta}_{\mathbf{x}}\| < \beta$ update **x** by $\mathbf{x}_{k+1} = \mathbf{x}_k + \boldsymbol{\delta}_{\mathbf{x}}$ Solution When $|G(\mathbf{x}_{k+1}) - G(\mathbf{x}_k)| < \varepsilon$, output and stop

- $\bullet\,$ Isotropic image model with $\rho=$ 0.95 for six levels of decomposition
- CAL1: two 6×6 lifting filters
- CAL2: three 4×4 lifting filters

Comparison with existing filter banks

Filter	Support of	Coding	Vanishing moments		
banks	analysis filters	gain(dB)	Ñ	Ν	Max. order
CAL1	13 imes13, $7 imes7$	12.06	2	2	0
CAL2	9 imes 9, $13 imes 13$	12.23	2	2	10^{-12}
KS	13 imes13, $7 imes7$	11.95	6	6	0
9/7	9×9 , 7×9 , 9×7 , 7×7	12.09	4	4	0

Optimal Design Algorithm

Design Examples

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Frequency Responses of CAL1



Optimal Design Algorithm

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Frequency Responses of CAL2



Scaling and Wavelet Functions for CAL1



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Scaling and Wavelet Functions for CAL2



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Image Coding Results

- Image coder: separable/nonseparable based on the lifting scheme
- Reversible integer-to-integer mappings
- Test images: grayscale images in the JPEG-2000 test set
- Coding
 - Lossy coding at various bit rates
 - Six/three levels of decomposition for quincunx/separable transforms
 - Difference measured in terms of PSNR
- Coding results: CAL1 and CAL2 outperform KS in 80% cases

Introduction

Experimental Results for finger

CR [†]	PSNR (dB)					
	CAL1	CAL2	KS	9/7		
128	19.88	19.95	19.67	19.98		
64	21.70	21.75	21.53	21.72		
32	24.52	24.39	24.36	24.20		
16	27.75	27.83	27.65	27.61		
[†] compression ratio						

Test image: finger

- CAL1 and CAL2 outperform the KS filter bank.
- CAL1 and CAL2 outperform the 9/7 filter bank except at the lowest bit rate.

Conclusion

- New optimization-based design method is proposed.
- This method yields linear-phase PR quincunx filter banks with high coding gain, good analysis/synthesis filter frequency responses, and prescribed vanishing moments properties.
- Effectiveness is demonstrated by the experimental results.

Optimal Design for a Particular Image

- Optimize with the autocorrelation function of the finger image
- CAL1f: same filter support as CAL1
- CAL2f: same filter support as CAL2
- Coding gains for the finger image

CAL1f	CAL2f	CAL1	CAL2	KS	9/7
12.76	12.35	12.17	12.04	12.27	12.05

Coding results

CR [†]	PSNR (dB)						
	CAL1f	CAL2f	CAL1	CAL2	KS	9/7	
128	19.92	19.35	19.88	19.95	19.67	19.98	
64	21.82	21.37	21.70	21.75	21.53	21.72	
32	24.53	24.21	24.52	24.39	24.36	24.20	
16	27.84	27.63	27.75	27.83	27.65	27.61	

[†]compression ratio