

COMPLEX NONSEPARABLE OVERSAMPLED LAPPED TRANSFORM FOR SPARSE REPRESENTATION OF MILLIMETER WAVE RADAR IMAGE

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Abstract

This work generalizes an existing framework of nonseparable oversampled lapped transforms (NSOLTs) to effectively represent complex-valued images.

- NSOLT is a redundant transform with linear-phase, compact-supported and perfect-reconstruction property.
- Generalized structure of NSOLTs is proposed to cover complex-valued atomic images.
- Effectiveness of the structure is verified by evaluating the sparse approximation performance.

Key words – Complex dictionary, NSOLT, Parseval tight frame, Millimeter wave radar, Sparse approximation

Introduction

- Sparsity-based image restoration
 - denoising, deblurring, super-resolution, inpainting, compressive sensing, etc.
- NSOLT [1]
 - Effective for image restoration problem
 - Example-based design (dictionary learning)
- Problem
 - Existing NSOLT can only deal with real-valued signal.
 - Insufficient degrees of freedom for complex-valued images.
- Objective
 - Generalize existing real-valued NSOLT (RNSOLT) to complex number field.
 - Evaluate the effectiveness by sparse approximation performance of complex-valued images.

Complex Nonseparable Oversampled Lapped Transform

- Complex-valued redundant transform satisfying linear-phase, compact-supported and perfect-reconstruction property.
 - Type-I CNSOLT includes Type-I RNSOLT as a subset.
- Lattice structure of analysis bank is obtained by a cascade construction as

$$\mathbf{E}(\mathbf{z}) = \Phi \left(\prod_{k=1}^{K_v} \mathbf{G}_k^{\{v\}}(z_v) \right) \left(\prod_{k=1}^{K_h} \mathbf{G}_k^{\{h\}}(z_h) \right) \mathbf{E}_0(\mathbf{z}),$$

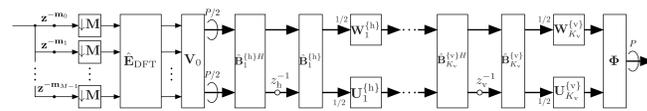


Figure 1: Lattice structure of analysis bank of 2D Type-I CNSOLT

Performance Evaluation

- Design example
 - RNSOLT and CNSOLT

Table 1: Common settings of 2-D Type-I NSOLTs

Downsampling factor \mathbf{M}	2×2
Polyphase order $\bar{\mathbf{n}}$	$[2, 2]^T$
#Channels P	6
#Tree levels τ	4
Redundancy R	$< \frac{5}{3}$

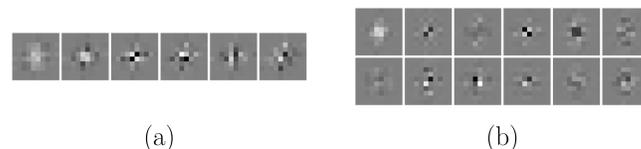


Figure 3: Atomic images of (a) RNSOLT, (b) CNSOLT

Sparse Approximation

- Problem settings

- Find a sparse vector $\hat{\mathbf{y}}$ that minimize:

$$\|\mathbf{x} - \mathbf{D}\hat{\mathbf{y}}\|_2^2 \text{ s.t. } \|\hat{\mathbf{y}}\|_0 \leq K, \quad (1)$$

- K : Number of nonzero elements

- Iterative hard thresholding (IHT) algorithm

- Heuristic algorithm for solving (1)

$$\mathbf{y}^{(i)} \leftarrow H_K(\mathbf{y}^{(i-1)} + \mathbf{D}^H(\mathbf{x} - \mathbf{D}\mathbf{y}^{(i-1)})) \quad (2)$$

- $H_K(\cdot)$: Hard thresholding operator.

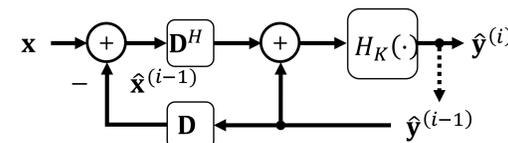


Figure 2: The block diagram of IHT.

Results

- Sparse approximate simulation

- Original image \mathbf{x} : millimeter wave radar complex image in Fig. 4.
- The intensity and hue show the magnitude and phase characteristics, respectively.

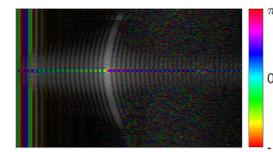


Figure 4: Original image of size 2,992 x 320 [pixels],

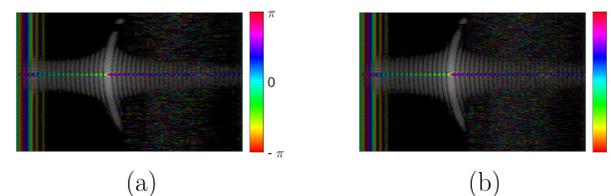


Figure 5: Sparse approximation results with 2-D NSOLT dictionaries through IHT.

(a) RNSOLT: $\text{MSE} = 2.36 \times 10^{-5}$

(b) CNSOLT: $\text{MSE} = 1.59 \times 10^{-5}$

Conclusion

- Proposed the lattice structure of 2-D CNSOLT
- Employed CNSOLT as a sparse modeling method for complex-valued images
- Effectiveness of the proposed method for millimeter wave radar images was examined by sparse approximation performance.

References

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