

# Two-Dimensional Non-Separable Block-Lifting Structure and Its Application to $M$ -Channel Perfect Reconstruction Filter Banks for Lossy-to-Lossless Image Coding

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**Abstract**—We propose a two-dimensional non-separable block-lifting structure (2D-NSBL) that is easily formulated from the one-dimensional separable block-lifting structure (1D-SBL) and 2D non-separable lifting structure (2D-NSL). The 2D-NSBL can be regarded as an extension of the 2D-NSL because a two-channel 2D-NSBL is completely equivalent to a 2D-NSL. We apply the 2D-NSBL to  $M$ -channel ( $M = 2^n$ ,  $n \in \mathbb{N}$ ) perfect reconstruction filter banks (PRFBs). The 2D-NSBL-based PRFBs outperform 1D-SBL-based PRFBs at lossy-to-lossless coding, whose image quality is scalable from lossless data to high compressed lossy data, because their rounding errors are reduced by merging many rounding operations.

**Index Terms**—Lossy-to-lossless image coding, perfect reconstruction filter bank (PRFB), two-dimensional non-separable block-lifting structure (2D-NSBL)

## I. INTRODUCTION

THE amount of video being sent over communications networks has been steadily increasing as a result of developments in multimedia devices and communication tools. Filter banks (FBs) [1] have been widely researched as a way to efficiently compress such signals. The polyphase matrices of  $M$ -channel ( $M = 2^n$ ,  $n \in \mathbb{N}$ ) FBs shown in Fig. 1 are presented as

$$\begin{aligned} & \begin{bmatrix} H_0(z) & H_1(z) & \cdots & H_{M-1}(z) \end{bmatrix}^T \\ &= \mathbf{E}(z^M) \begin{bmatrix} 1 & z^{-1} & \cdots & z^{-(M-1)} \end{bmatrix}^T \\ & \begin{bmatrix} F_0(z) & F_1(z) & \cdots & F_{M-1}(z) \end{bmatrix} \\ &= \begin{bmatrix} 1 & z^{-1} & \cdots & z^{-(M-1)} \end{bmatrix} \mathbf{R}(z^M), \end{aligned}$$

where  $H_i(z)$ ,  $F_i(z)$ ,  $z$ , and  $\cdot^T$  denote an analysis filter, a synthesis filter, a delay element, and matrix transposition, respectively. If  $\mathbf{E}(z)$  is invertible, the inverse of  $\mathbf{E}(z)$  can be chosen as a synthesis polyphase matrix  $\mathbf{R}(z)$ , and such FBs are called perfect reconstruction FBs (PRFBs). When  $\mathbf{R}(z) = \mathbf{E}^T(z^{-1})$ , the FBs are called paraunitary FBs (PUFBs) which are special classes of PRFBs. On the other hand, PRFBs that are not PUFBs are commonly called biorthogonal FBs (BOFBs). In particular, the JPEG series [2–4] and H.26x series [5], [6] of global standards use various classes of PRFBs, including the discrete cosine/sine transform (DCT and DST) [7], discrete wavelet transform (DWT) [8], and

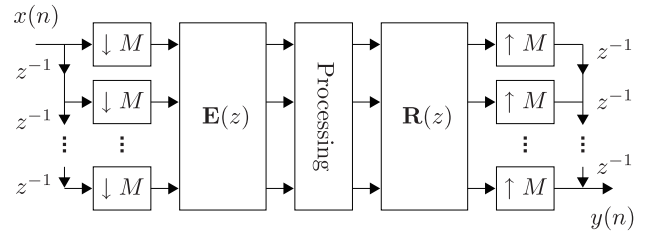


Fig. 1. Polyphase structure of  $M$ -channel FB.

TABLE I  
CLASSIFICATION OF LIFTING STRUCTURES.

	1D-SL [10–12]	1D-SBL [13]	2D-NSL [14], [15]	2D-NSBL Prop.
Block-Lifting	—	✓	—	✓
Non-Separable	—	—	✓	✓

hierarchical lapped transform (HLT) [9]. However, there is a growing need for better FBs in order to alleviate the burden on servers and free up communication bandwidth.

Lossy-to-lossless image coding, which merges two or more pieces of data into one piece of data of the same piece of content, i.e., “one source multi-use” image coding, has attracted attention from researchers as a possible way to meet this need. Reversible transforms that map integers to integers, called integer-to-integer transforms, are important tools for lossy-to-lossless image coding. Sweldens presented a lifting structure [10–12] with which to achieve integer-to-integer transforms, and this structure has been applied to many FBs [16–23]. Although JPEG XR [4] has scalability ranging from lossless to lossy as a result of using a lifting-based HLT, its coding performance is not sufficient especially for images with high-frequency components (texture).

The one-dimensional separable block-lifting structure (1D-SBL) of BOFB was proposed in [13] for the purpose of designing lifting-based FBs with higher coding performance. Usually, the design parameters and structure of lifting-based FBs are constrained when factorizing the original FB into lifting structures, whereas the 1D-SBL-based BOFBs presented in [13] do not constrain them except in the initial block.<sup>1</sup> The 1D-SBL is better at lossy-to-lossless image coding because it uses fewer rounding operations in comparison with the

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<sup>1</sup>The determinant of the initial block is constrained to be an integer value.

standard 1D separable lifting structure (1D-SL). Furthermore, the two-dimensional non-separable lifting structure (2D-NSL) for DWTs proposed in [14], [15] performs even better at coding because it uses fewer rounding operations than the 1D-SL.

Here, we propose a 2D non-separable block-lifting structure (2D-NSBL) that is easily formulated from the 1D-SBL and 2D-NSL methods. The 2D-NSBL can be regarded as an extension of the 2D-NSL because a 2D-NSBL with  $M = 2$  is completely equivalent to a 2D-NSL. We apply the 2D-NSBL to  $M$ -channel PRFBs and show that the PRFBs perform better at lossy-to-lossless coding than the conventional 1D-SBL-based PRFBs do because their rounding errors are reduced by merging many rounding operations.

The remaining part of this paper is organized as follows. We review and define the block-lifting structure, 1D-SBL-based PRFBs, and 2D-NSL-based DWTs in Section II. Section III presents the derivation of the 2D-NSBL and its application to  $M$ -channel PRFBs. Design examples, lossy-to-lossless image coding simulations, and comparisons with the conventional PRFBs are shown in Section IV. Section V concludes this paper.

*Notations:* A classification of lifting structures is shown in Table I.  $\mathbf{I}_m$ ,  $\mathbf{0}$ ,  $\det(\cdot)$ , and  $\text{round}(\cdot)$  denote an  $m \times m$  identity matrix, a null matrix, determinant of a matrix, and a rounding operation, respectively.  $\mathbf{I}_m$  is simply expressed by  $\mathbf{I}$  if its size is clear. Indexes  $x$ ,  $y$ ,  $w$ , and  $2d$  in the matrices mean to operate horizontally, vertically, horizontally or vertically ( $w = x$  or  $y$ ), and horizontally and vertically, respectively. For example,

$$\mathbf{T}^x \mathbf{x} = \mathbf{x} \mathbf{T}^T, \quad \mathbf{T}^y \mathbf{x} = \mathbf{T} \mathbf{x}, \quad \text{and} \quad \mathbf{T}^{2d} \mathbf{x} = \mathbf{T} \mathbf{x} \mathbf{T}^T,$$

where  $\mathbf{T}$  and  $\mathbf{x}$  are a transform matrix and a 2D input signal, respectively.

## II. REVIEW AND DEFINITIONS

### A. Block-Lifting Structure

A lifting structure [10–12] is a transform that map integers to integers by implementing rounding operation in each lifting step; i.e., it is a means of lossy-to-lossless image coding. The elementary matrices are identity matrices with one single nonzero off-diagonal element. However, an FB with too many lifting steps cannot perform good coding because it generates a rounding error in each lifting step.

We proposed the block-lifting structure in [13], which is a special class of standard lifting structure (Fig. 2). It is good for lossy-to-lossless image coding because it reduces the rounding error by merging many rounding operations. In Fig. 2, the analysis input signal vectors  $\mathbf{x}_i$  and  $\mathbf{x}_j$ , the analysis output (synthesis input) signal vectors  $\mathbf{y}_i$  and  $\mathbf{y}_j$ , the synthesis output signal vectors  $\mathbf{z}_i$  and  $\mathbf{z}_j$ , and the lifting coefficient blocks  $\mathfrak{L}(z)$

and  $\mathfrak{U}(z)$  are related as follows:

$$\left. \begin{aligned} \mathbf{y}_j &= \mathbf{x}_j + \text{round}(\mathfrak{L}(z)\mathbf{x}_i) \\ \mathbf{y}_i &= \mathbf{x}_i + \text{round}(\mathfrak{U}(z)\mathbf{y}_j) \\ \mathbf{z}_i &= \mathbf{y}_i - \text{round}(\mathfrak{U}(z)\mathbf{y}_j) = \mathbf{x}_i \\ \mathbf{z}_j &= \mathbf{y}_j - \text{round}(\mathfrak{L}(z)\mathbf{x}_i) = \mathbf{x}_j \end{aligned} \right\} \text{(Case A)}$$

$$\left. \begin{aligned} \mathbf{y}_i &= \mathbf{x}_i + \text{round}(\mathfrak{U}(z)\mathbf{x}_j) \\ \mathbf{y}_j &= \mathbf{x}_j + \text{round}(\mathfrak{L}(z)\mathbf{y}_i) \\ \mathbf{z}_j &= \mathbf{y}_j - \text{round}(\mathfrak{L}(z)\mathbf{y}_i) = \mathbf{x}_j \\ \mathbf{z}_i &= \mathbf{y}_i - \text{round}(\mathfrak{U}(z)\mathbf{x}_j) = \mathbf{x}_i \end{aligned} \right\} \text{(Case B)}$$

In these cases, the matrices and their inverse matrices are expressed by

$$\begin{bmatrix} \mathbf{y}_i \\ \mathbf{y}_j \end{bmatrix} = \mathfrak{W}(z) \begin{bmatrix} \mathbf{x}_i \\ \mathbf{x}_j \end{bmatrix}, \quad \begin{bmatrix} \mathbf{z}_i \\ \mathbf{z}_j \end{bmatrix} = \mathfrak{W}^{-1}(z) \begin{bmatrix} \mathbf{y}_i \\ \mathbf{y}_j \end{bmatrix} = \begin{bmatrix} \mathbf{x}_i \\ \mathbf{x}_j \end{bmatrix},$$

where

$$\mathfrak{W}(z) = \begin{cases} \mathfrak{B}_U(z)\mathfrak{B}_L(z) & \text{(Case A)} \\ \mathfrak{B}_L(z)\mathfrak{B}_U(z) & \text{(Case B)} \end{cases}$$

$$\mathfrak{W}^{-1}(z) = \begin{cases} \mathfrak{B}_L^{-1}(z)\mathfrak{B}_U^{-1}(z) & \text{(Case A)} \\ \mathfrak{B}_U^{-1}(z)\mathfrak{B}_L^{-1}(z) & \text{(Case B)} \end{cases}$$

$$\mathfrak{B}_U(z) = \begin{bmatrix} \mathbf{I} & \mathfrak{U}(z) \\ \mathbf{0} & \mathbf{I} \end{bmatrix}, \quad \mathfrak{B}_U^{-1}(z) = \begin{bmatrix} \mathbf{I} & -\mathfrak{U}(z) \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathfrak{B}_L(z) = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathfrak{L}(z) & \mathbf{I} \end{bmatrix}, \quad \mathfrak{B}_L^{-1}(z) = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ -\mathfrak{L}(z) & \mathbf{I} \end{bmatrix}.$$

Note that the rounding operations are actually implemented even if the lifting matrix expression omits the notation of them. The block-lifting structure for a 1D implementation is called “1D-SBL” to distinguish it from the “2D-NSBL” proposed in this paper. When  $M = 2$ , they will also be called “1D-SL” and “2D-NSL”.

### B. 1D-SBL-based PRFBs

The polyphase matrix  $\mathbf{E}(z)$  of an  $M$ -channel PRFB of filter length  $MK$  ( $K \in \mathbb{N}, K \geq 2$ ) is expressed as [24]

$$\mathbf{E}(z) = \prod_{k=K-1}^1 \{\mathbf{E}_k(z)\} \mathbf{G}_0, \quad (1)$$

where the initial block  $\mathbf{G}_0$  is an  $M \times M$  nonsingular matrix and the building block  $\mathbf{E}_k(z)$  is expressed by

$$\mathbf{E}_k(z) = \mathbf{I} - \mathbf{u}_k \mathbf{v}_k^T + z^{-1} \mathbf{u}_k \mathbf{v}_k^T.$$

The  $M \times \gamma_k$  parameter matrices  $\mathbf{u}_k$  and  $\mathbf{v}_k$  satisfy

$$\mathbf{v}_k^T \mathbf{u}_k = \begin{bmatrix} 1 & \times & \cdots & \times \\ 0 & 1 & \ddots & \vdots \\ \vdots & \ddots & \ddots & \times \\ 0 & \cdots & 0 & 1 \end{bmatrix}_{\gamma_k \times \gamma_k} \triangleq \mathcal{W}_k,$$

where  $\times$  indicates possibly nonzero elements and  $\gamma_k$  is a McMillan degree ( $\gamma_k \in \mathbb{N}, 1 \leq \gamma_k \leq M - 1$ ).  $\mathcal{W}_k = \mathbf{I}$  when the filter lengths in analysis and synthesis banks are equal. A synthesis polyphase matrix  $\mathbf{R}(z)$  is defined as one that has the PR property  $\mathbf{R}(z)\mathbf{E}(z) = \mathbf{I}$ . Moreover,  $\mathbf{u}_k$  and  $\mathbf{v}_k$  are defined as  $\mathbf{u}_k = [\mathbf{u}_{k0}^T, \mathbf{u}_{k1}^T]^T$  and  $\mathbf{v}_k = [\mathbf{v}_{k0}^T, \mathbf{v}_{k1}^T]^T$ ,

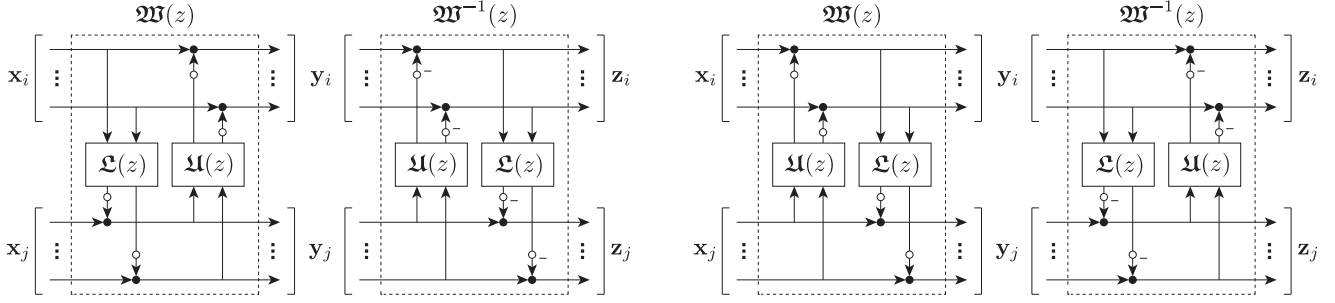


Fig. 2. Block-lifting structures. Black and white circles mean adders and rounding operations, respectively: (left) Case A and (right) Case B.

where  $\mathbf{u}_{k0}$  and  $\mathbf{v}_{k0}$  are  $(M - \gamma_k) \times \gamma_k$  matrices and  $\mathbf{u}_{k1}$  and  $\mathbf{v}_{k1}$  are  $\gamma_k \times \gamma_k$  square matrices. When  $\mathbf{V}_k = \mathbf{U}_k$ , the PRFB has the paraunitary property, i.e., PUFB. In this paper, we fix  $\mathbf{W}_k = \mathbf{I}$  and  $\gamma_k = M/2$ .

We factorized the PRFBs into the 1D-SBL in [13]. The 1D-SBL-based PRFBs represent  $\mathbf{E}_k(z)$  in Eq. (1) as

$$\mathbf{E}_k(z) = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ -\mathbf{L}_k & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{U}_k \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \mathbf{\Lambda}(z) \begin{bmatrix} \mathbf{I} & -\mathbf{U}_k \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{L}_k & \mathbf{I} \end{bmatrix},$$

where  $\mathbf{L}_k = \mathbf{v}_{k1}^{-T} \mathbf{v}_{k0}^T$ ,  $\mathbf{U}_k = \mathbf{u}_{k0} \mathbf{v}_{k1}^T$ , and

$$\mathbf{\Lambda}(z) = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{0} & z^{-1} \mathbf{I} \end{bmatrix}.$$

In addition,  $\det(\mathbf{G}_0)$  is constrained to be  $\det(\mathbf{G}_0) = \pm n$  ( $n \in \mathbb{N}$ ) for the purpose of making a lifting factorization. If paraunitariness is not required,  $\mathbf{L}_k$  and  $\mathbf{U}_k$  can be arbitrary  $M/2 \times M/2$  matrices. To improve coding performance, Eq. (1) can be rewritten as (Fig. 3)

$$\mathbf{E}(z) = \mathbf{W}_K(z) \prod_{k=K-1}^1 \{ \mathbf{\Lambda}(z) \mathbf{W}_k(z) \} \mathbf{G}_0, \quad (2)$$

where

$$\begin{aligned} \mathbf{W}_k(z) &= \begin{bmatrix} \mathbf{I} & \widehat{\mathbf{U}}_k(z) \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \widehat{\mathbf{L}}_k(z) & \mathbf{I} \end{bmatrix} \\ \widehat{\mathbf{U}}_k(z) &= \begin{cases} \mathbf{0} & (k = K) \\ (z^{-1} - 1) \mathbf{U}_k & (\text{otherwise}) \end{cases} \\ \widehat{\mathbf{L}}_k(z) &= \begin{cases} \mathbf{L}_1 & (k = 1) \\ -\mathbf{L}_{K-1} & (k = K) \\ \mathbf{L}_k - \mathbf{L}_{k-1} & (\text{otherwise}) \end{cases}. \end{aligned}$$

In comparison with the 1D-SBL-based PRFBs in Eq. (1), the 1D-SBL-based PRFBs are more effective at lossy-to-lossless image coding because they reduce the rounding error by merging more rounding operations.

### C. 2D-NSL-based DWTs

1D-SL-based 9/7-tap and 5/3-tap DWTs (9/7-DWT and 5/3-DWT) [8] are used in the JPEG 2000 [3] lossy and lossless modes, respectively. Let  $\mathbf{e}(z)$  be a polyphase matrix of 1D-SL-based DWTs, expressed as

$$\mathbf{e}(z) = \begin{bmatrix} s & \mathbf{0} \\ \mathbf{0} & s^{-1} \end{bmatrix} \prod_{k=N-1}^0 \mathbf{w}_k(z),$$

where

$$\mathbf{w}_k(z) = \begin{bmatrix} 1 & u_k(z) \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} 1 & \mathbf{0} \\ l_k(z) & 1 \end{bmatrix}.$$

For example, 5/3-DWT has  $N = 1$ ,  $s = 1$ ,  $u_0(z) = (1+z)/4$  and  $l_0(z) = -(1+z^{-1})/2$ . If an image is transformed by the 2D-NSL-based DWT polyphase matrix  $\mathbf{e}_k^{2d}(z_{2d})$ , one can write

$$\begin{aligned} & \begin{bmatrix} \mathbf{Y}_{LL}^T & \mathbf{Y}_{HL}^T & \mathbf{Y}_{LH}^T & \mathbf{Y}_{HH}^T \end{bmatrix}^T \\ &= \mathbf{e}_k^{2d}(z_{2d}) \begin{bmatrix} \mathbf{X}_{LL}^T & \mathbf{X}_{HL}^T & \mathbf{X}_{LH}^T & \mathbf{X}_{HH}^T \end{bmatrix}^T, \end{aligned}$$

where  $X_{LL}$ ,  $X_{HL}$ ,  $X_{LH}$ , and  $X_{HH}$  are the top-left, top-right, bottom-left, and bottom-right pixels in  $2 \times 2$  blocks composing the image,  $Y_{LL}$ ,  $Y_{HL}$ ,  $Y_{LH}$ , and  $Y_{HH}$  are their output pixels, and

$$\mathbf{e}_k^{2d}(z_{2d}) = \begin{bmatrix} s^2 & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_2 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & s^{-2} \end{bmatrix} \prod_{k=N-1}^0 \mathbf{w}_k^{2d}(z_{2d}).$$

$\mathbf{w}_k^{2d}(z_{2d})$  in  $\mathbf{e}^{2d}(z_{2d})$  is represented as [14], [15]

$$\begin{aligned} \mathbf{w}_k^{2d}(z_{2d}) &= \begin{bmatrix} 1 & [u_k^x(z_x) & u_k^y(z_y) & -u_k^{2d}(z_{2d})] \\ \mathbf{0} & \mathbf{I}_3 \end{bmatrix} \\ &\cdot \begin{bmatrix} 1 & \mathbf{0} & \mathbf{0} \\ [l_k^x(z_x) & \mathbf{I}_2 & [u_k^y(z_y) \\ l_k^y(z_y) & \mathbf{0} & [u_k^x(z_x) \\ \mathbf{0} & \mathbf{0} & 1 \end{bmatrix} \\ &\cdot \begin{bmatrix} \mathbf{I}_3 & \mathbf{0} \\ [l_k^{2d}(z_{2d}) & l_k^y(z_y) & l_k^x(z_x) & 1 \end{bmatrix}. \quad (3) \end{aligned}$$

The 2D-NSL is also more effective at lossy-to-lossless image coding than the 1D-SL is because it uses fewer rounding operations.

## III. 2D-NSBL-BASED PRFBs

### A. Derivation of 2D-NSBL

We introduce 2D-NSBL in this subsection.

*Theorem:* Consider an image that has been 2D-transformed by the set of lower and upper block-lifting matrices in Fig. 2 as follows:

$$\begin{aligned} & \begin{bmatrix} \mathbf{Y}_{LL}^T & \mathbf{Y}_{HL}^T & \mathbf{Y}_{LH}^T & \mathbf{Y}_{HH}^T \end{bmatrix}^T \\ &= \mathbf{W}^{2d}(z_{2d}) \begin{bmatrix} \mathbf{X}_{LL}^T & \mathbf{X}_{HL}^T & \mathbf{X}_{LH}^T & \mathbf{X}_{HH}^T \end{bmatrix}^T, \quad (4) \end{aligned}$$

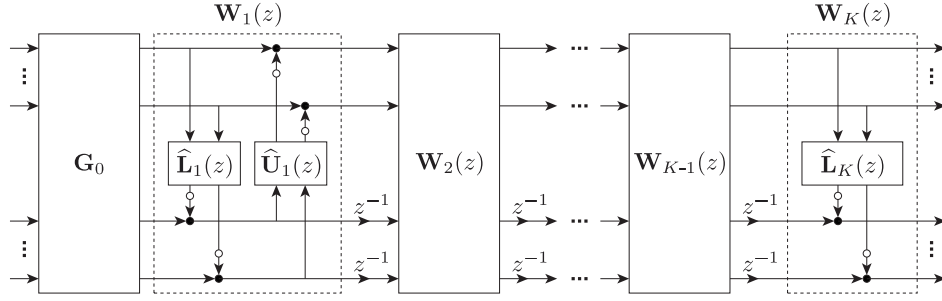


Fig. 3. 1D-SBL-based PRFB. Black and white circles mean adders and rounding operations, respectively.

where

$$\mathfrak{W}^{2d}(z_{2d}) = \begin{bmatrix} \mathfrak{W}^x(z_x) & \mathbf{0} \\ \mathbf{0} & \mathfrak{W}^y(z_y) \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathfrak{W}^y(z_y) & \mathbf{0} \\ \mathbf{0} & \mathfrak{W}^x(z_x) \end{bmatrix} \mathbf{P}$$

$$\mathfrak{W}^w(z_w) = \begin{cases} \mathfrak{B}_U^w(z_w) \mathfrak{B}_L^w(z_w) & \text{(Case A)} \\ \mathfrak{B}_L^w(z_w) \mathfrak{B}_U^w(z_w) & \text{(Case B)} \end{cases}$$

$$\mathbf{P} = \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix},$$

$\mathbf{X}_{LL}$ ,  $\mathbf{X}_{HL}$ ,  $\mathbf{X}_{LH}$ , and  $\mathbf{X}_{HH}$  are the top-left, top-right, bottom-left, and bottom-right  $M/2 \times M/2$  blocks of an  $M \times M$  image, and  $\mathbf{Y}_{LL}$ ,  $\mathbf{Y}_{HL}$ ,  $\mathbf{Y}_{LH}$ , and  $\mathbf{Y}_{HH}$  are their respective output blocks (Fig. 4).  $\mathfrak{W}^{2d}(z_{2d})$  in Eq. (4) can be factorized into three 2D-NSBL matrices, as follows (Fig. 5):

$$\mathfrak{W}^{2d}(z_{2d}) = \mathfrak{W}_2^{2d}(z_{2d}) \mathfrak{W}_1^{2d}(z_{2d}) \mathfrak{W}_0^{2d}(z_{2d}), \quad (5)$$

where

$$\mathfrak{W}_0^{2d}(z_{2d}) = \begin{cases} \begin{bmatrix} \mathbf{I}_{3M/2} & \mathbf{0} \\ \begin{bmatrix} \mathfrak{L}^{2d}(z_{2d}) & \mathfrak{L}^y(z_y) & \mathfrak{L}^x(z_x) \end{bmatrix} & \mathbf{I} \end{bmatrix} & \text{(Case A)} \\ \begin{bmatrix} \mathbf{I} & \begin{bmatrix} \mathfrak{U}^x(z_x) & \mathfrak{U}^y(z_y) & \mathfrak{U}^{2d}(z_{2d}) \end{bmatrix} \\ \mathbf{0} & \mathbf{I}_{3M/2} \end{bmatrix} & \text{(Case B)} \end{cases}$$

$$\mathfrak{W}_1^{2d}(z_{2d}) = \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \begin{bmatrix} \mathfrak{L}^x(z_x) \\ \mathfrak{L}^y(z_y) \end{bmatrix} & \mathbf{I}_M & \begin{bmatrix} \mathfrak{U}^y(z_y) \\ \mathfrak{U}^x(z_x) \end{bmatrix} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathfrak{W}_2^{2d}(z_{2d}) = \begin{cases} \begin{bmatrix} \mathbf{I} & \begin{bmatrix} \mathfrak{U}^x(z_x) & \mathfrak{U}^y(z_y) & -\mathfrak{U}^{2d}(z_{2d}) \end{bmatrix} \\ \mathbf{0} & \mathbf{I}_{3M/2} \end{bmatrix} & \text{(Case A)} \\ \begin{bmatrix} \mathbf{I}_{3M/2} & \mathbf{0} \\ \begin{bmatrix} -\mathfrak{L}^{2d}(z_{2d}) & \mathfrak{L}^y(z_y) & \mathfrak{L}^x(z_x) \end{bmatrix} & \mathbf{I} \end{bmatrix} & \text{(Case B)} \end{cases}.$$

It is clear that the 2D-NSBL is an extension of the 2D-NSL in [14], [15] because the 2D-NSBL with  $M = 2$  in Eq. (5) (Case A) is completely equivalent to the 2D-NSL in Eq. (3). Furthermore, a 2D-NSBL with the McMillan degree  $\gamma_k \neq M/2$  can be easily obtained.

*Proof:* When a matrix  $\mathfrak{T} = \mathbf{T}_{n-1} \cdots \mathbf{T}_0$  ( $n \in \mathbb{N}$ ) is applied to a 2D input signal  $\mathbf{x}$  in the horizontal and vertical directions, the output signal  $\mathbf{y}$  is expressed as [25]

$$\mathbf{y} = \mathfrak{T} \mathbf{x} \mathfrak{T}^T = \mathbf{T}_{n-1} \cdots \mathbf{T}_0 \mathbf{x} \mathbf{T}_0^T \cdots \mathbf{T}_{n-1}^T. \quad (6)$$

This Eq. (6) means that the 2D implementation of  $\mathbf{T}_k$  is performed after that of  $\mathbf{T}_{k-1}$  ( $1 \leq k \leq n-1$ ), i.e., the two block-lifting matrices  $\mathfrak{B}_L^w(z_w)$  and  $\mathfrak{B}_U^w(z_w)$  in Eq. (4) can be operated separately. The resulting representation of  $\mathfrak{W}^{2d}(z_{2d})$  is

$$\mathfrak{W}^{2d}(z_{2d}) = \begin{cases} \mathfrak{B}_U^{2d}(z_{2d}) \mathfrak{B}_L^{2d}(z_{2d}) & \text{(Case A)} \\ \mathfrak{B}_L^{2d}(z_{2d}) \mathfrak{B}_U^{2d}(z_{2d}) & \text{(Case B)}, \end{cases} \quad (7)$$

where

$$\mathfrak{B}_L^{2d}(z_{2d}) = \begin{bmatrix} \mathfrak{B}_L^x(z_x) & \mathbf{0} \\ \mathbf{0} & \mathfrak{B}_L^y(z_y) \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathfrak{B}_L^y(z_y) & \mathbf{0} \\ \mathbf{0} & \mathfrak{B}_L^x(z_x) \end{bmatrix} \mathbf{P}$$

$$= \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathfrak{L}^x(z_x) & \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathfrak{L}^y(z_y) & \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathfrak{L}^{2d}(z_{2d}) & \mathfrak{L}^y(z_y) & \mathfrak{L}^x(z_x) & \mathbf{I} \end{bmatrix}$$

$$\mathfrak{B}_U^{2d}(z_{2d}) = \begin{bmatrix} \mathfrak{B}_U^x(z_x) & \mathbf{0} \\ \mathbf{0} & \mathfrak{B}_U^y(z_y) \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathfrak{B}_U^y(z_y) & \mathbf{0} \\ \mathbf{0} & \mathfrak{B}_U^x(z_x) \end{bmatrix} \mathbf{P}$$

$$= \begin{bmatrix} \mathbf{I} & \mathfrak{U}^x(z_x) & \mathfrak{U}^y(z_y) & \mathfrak{U}^{2d}(z_{2d}) \\ \mathbf{0} & \mathbf{I} & \mathbf{0} & \mathfrak{U}^y(z_y) \\ \mathbf{0} & \mathbf{0} & \mathbf{I} & \mathfrak{U}^x(z_x) \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix}.$$

Since the lifting matrix will have inevitably generated rounding operations in a process, as described in Section II-A, we separate each of  $\mathfrak{B}_L^{2d}(z_{2d})$  and  $\mathfrak{B}_U^{2d}(z_{2d})$  into two 2D-NSBL matrices:

$$\mathfrak{B}_L^{2d}(z_{2d}) = \begin{cases} \mathfrak{B}_{L1}^{2d}(z_{2d}) \mathfrak{B}_{L0}^{2d}(z_{2d}) & \text{(Case A)} \\ \mathfrak{B}_{L2}^{2d}(z_{2d}) \mathfrak{B}_{L1}^{2d}(z_{2d}) & \text{(Case B)} \end{cases} \quad (8)$$

$$\mathfrak{B}_U^{2d}(z_{2d}) = \begin{cases} \mathfrak{B}_{U2}^{2d}(z_{2d}) \mathfrak{B}_{U1}^{2d}(z_{2d}) & \text{(Case A)} \\ \mathfrak{B}_{U1}^{2d}(z_{2d}) \mathfrak{B}_{U0}^{2d}(z_{2d}) & \text{(Case B)}, \end{cases} \quad (9)$$

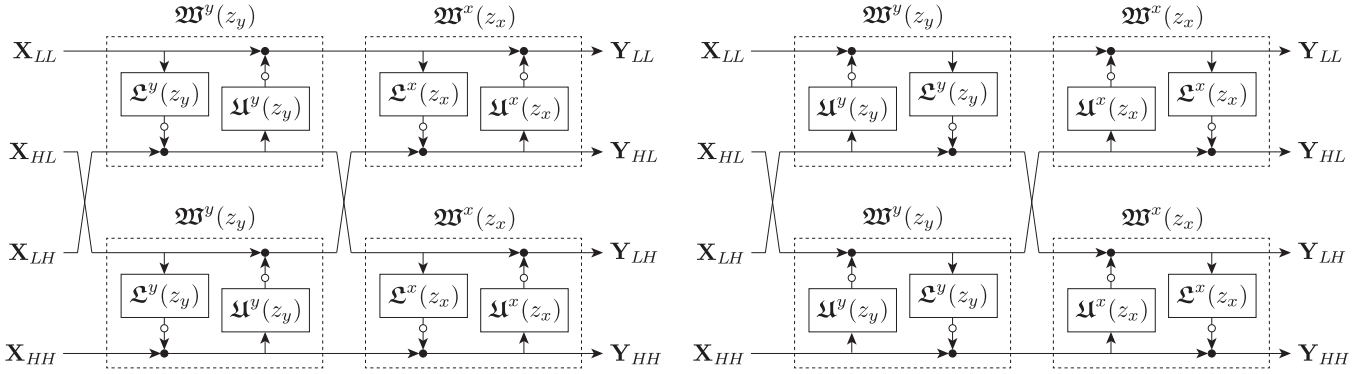


Fig. 4. 2D implementation of 1D-SBL. Black and white circles mean adders and rounding operations, respectively: (left) Case A and (right) Case B.

where

$$\begin{aligned} \mathbf{B}_{L0}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{L}^{2d}(z_{2d}) & \mathbf{I}_{3M/2} & \mathbf{0} \\ \mathbf{L}^y(z_y) & \mathbf{L}^x(z_x) & \mathbf{I} \end{bmatrix} \\ \mathbf{B}_{L1}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{L}^x(z_x) & \mathbf{I}_M & \mathbf{0} \\ \mathbf{L}^y(z_y) & \mathbf{0} & \mathbf{I} \end{bmatrix} \\ \mathbf{B}_{L2}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I}_{3M/2} & \mathbf{0} \\ -\mathbf{L}^{2d}(z_{2d}) & \mathbf{L}^y(z_y) & \mathbf{L}^x(z_x) \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \\ \mathbf{B}_{U0}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I} & \mathbf{U}^x(z_x) & \mathbf{U}^y(z_y) & \mathbf{U}^{2d}(z_{2d}) \\ \mathbf{0} & \mathbf{I}_{3M/2} & \mathbf{0} & \mathbf{0} \end{bmatrix} \\ \mathbf{B}_{U1}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_M & \begin{bmatrix} \mathbf{U}^y(z_y) \\ \mathbf{U}^x(z_x) \end{bmatrix} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix} \\ \mathbf{B}_{U2}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I} & \mathbf{U}^x(z_x) & \mathbf{U}^y(z_y) & -\mathbf{U}^{2d}(z_{2d}) \\ \mathbf{0} & \mathbf{I}_{3M/2} & \mathbf{0} & \mathbf{0} \end{bmatrix}. \end{aligned}$$

Consequently,  $\mathbf{W}^{2d}(z_{2d})$  is expressed as

$$\mathbf{W}^{2d}(z_{2d}) = \begin{cases} \mathbf{B}_{U2}^{2d}(z_{2d})\mathbf{B}_{UL}^{2d}(z_{2d})\mathbf{B}_{L0}^{2d}(z_{2d}) & \text{(Case A)} \\ \mathbf{B}_{L2}^{2d}(z_{2d})\mathbf{B}_{LU}^{2d}(z_{2d})\mathbf{B}_{U0}^{2d}(z_{2d}) & \text{(Case B)}, \end{cases}$$

where

$$\mathbf{B}_{UL}^{2d}(z_{2d}) = \mathbf{B}_{LU}^{2d}(z_{2d}) = \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{L}^x(z_x) & \mathbf{I}_M & \mathbf{L}^y(z_y) \\ \mathbf{L}^y(z_y) & \mathbf{0} & \mathbf{L}^x(z_x) \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix}$$

from Eqs. (7)-(9). The resulting equation is completely the same as Eq. (5).  $\square$

### B. Application to PRFBs

Here, we will apply the 2D-NSBL in Eq. (5) to the conventional 1D-SBL-based PRFBs in Eq. (2). Let  $\mathbf{E}^{2d}(z_{2d})$  be a 2D separable polyphase matrix based on a 1D separable polyphase matrix  $\mathbf{E}(z)$  in Eq. (2). Since the 2D implementation of the separable block transform allows us to change the order in which the blocks are operated on, the polyphase matrix  $\mathbf{E}^{2d}(z_{2d})$  can be expressed as

$$\mathbf{E}^{2d}(z_{2d}) = \mathbf{W}_K^{2d}(z_{2d}) \prod_{k=K-1}^1 \{ \mathbf{\Lambda}^{2d}(z_{2d}) \mathbf{W}_k^{2d}(z_{2d}) \} \mathbf{G}_0^{2d},$$

where

$$\begin{aligned} \mathbf{\Lambda}^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{\Lambda}(z_x) & \mathbf{0} \\ \mathbf{0} & \mathbf{\Lambda}(z_y) \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathbf{\Lambda}(z_y) & \mathbf{0} \\ \mathbf{0} & \mathbf{\Lambda}(z_x) \end{bmatrix} \mathbf{P} \\ \mathbf{W}_k^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{W}_k^x(z_x) & \mathbf{0} \\ \mathbf{0} & \mathbf{W}_k^y(z_y) \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathbf{W}_k^y(z_y) & \mathbf{0} \\ \mathbf{0} & \mathbf{W}_k^x(z_x) \end{bmatrix} \mathbf{P} \\ \mathbf{G}_0^{2d} &= \begin{bmatrix} \mathbf{G}_0^x & \mathbf{0} \\ \mathbf{0} & \mathbf{G}_0^y \end{bmatrix} \mathbf{P} \begin{bmatrix} \mathbf{G}_0^y & \mathbf{0} \\ \mathbf{0} & \mathbf{G}_0^x \end{bmatrix} \mathbf{P}. \end{aligned}$$

Applying Case A of the proposed 2D-NSBL in Eq. (5) to  $\mathbf{W}_k^{2d}(z_{2d})$  yields

$$\begin{aligned} \mathbf{W}_k^{2d}(z_{2d}) &= \begin{bmatrix} \mathbf{I} & \begin{bmatrix} \hat{\mathbf{U}}_k^x(z_x) & \hat{\mathbf{U}}_k^y(z_y) & -\hat{\mathbf{U}}_k^{2d}(z_{2d}) \end{bmatrix} \\ \mathbf{0} & \mathbf{I}_{3M/2} \end{bmatrix} \\ &\cdot \begin{bmatrix} \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \begin{bmatrix} \hat{\mathbf{L}}_k^x(z_x) \\ \hat{\mathbf{L}}_k^y(z_y) \end{bmatrix} & \mathbf{I}_M & \begin{bmatrix} \hat{\mathbf{U}}_k^y(z_y) \\ \hat{\mathbf{U}}_k^x(z_x) \end{bmatrix} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix} \\ &\cdot \begin{bmatrix} \mathbf{I}_{3M/2} & \mathbf{0} \\ \hat{\mathbf{L}}_k^{2d}(z_{2d}) & \hat{\mathbf{L}}_k^y(z_y) & \hat{\mathbf{L}}_k^x(z_x) \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{bmatrix}. \end{aligned}$$

For  $\mathbf{G}_0^{2d}$ , a 1D-SL factorization is used. As is done in [13], we use the single-row elementary reversible matrix (SERM) presented in [16] for each initial block  $\mathbf{G}_0^w$ , where any other 1D-SL factorization can be applied to  $\mathbf{G}_0^w$ .

## IV. EXPERIMENTAL RESULTS

### A. Filter Design

By following the method presented in [13],  $8 \times 16$  and  $8 \times 24$  BOFBs with order-1 building blocks were designed by using the cost function  $\Phi$  and `fminunc.m` in Optimization Toolbox of MATLAB.  $\Phi$  was a weighted linear combination of the coding gain  $C_{CG}$  and the stopband attenuation values of analysis and synthesis filters  $C_{SAa}$  and  $C_{SAs}$ :

$$\Phi = (w_1 C_{SAa} + w_2 C_{SAs}) - w_3 C_{CG},$$

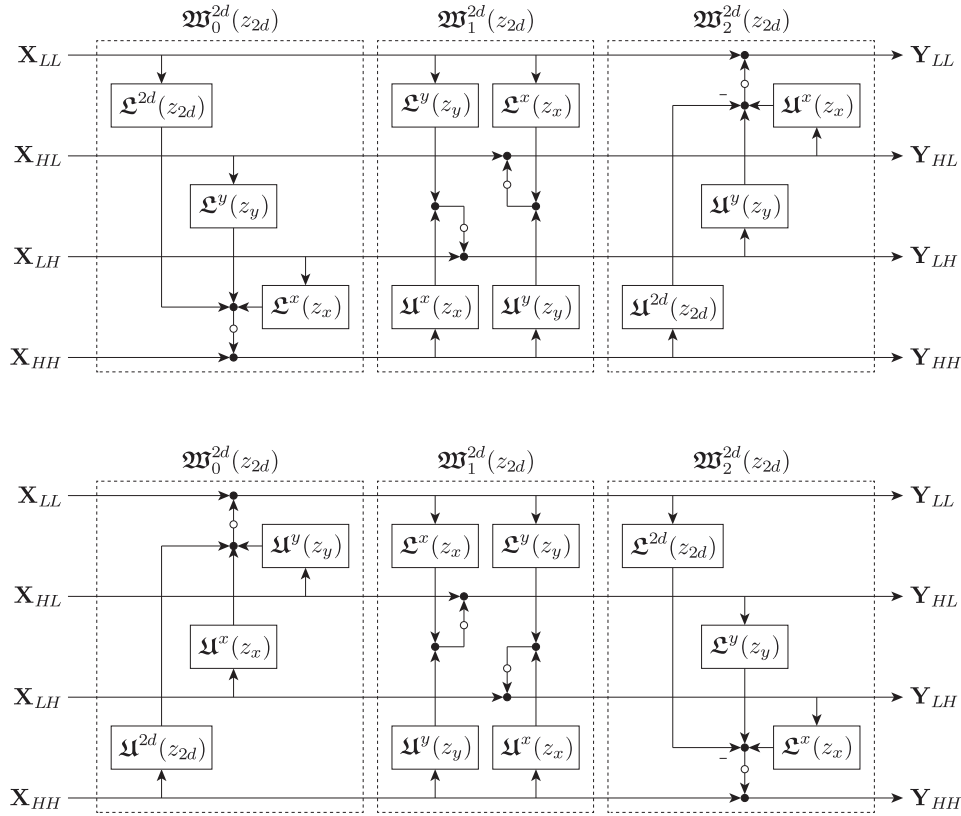


Fig. 5. 2D-NSBL. Black and white circles mean adders and rounding operations, respectively: (top) Case A and (bottom) Case B.

TABLE II  
LOSSLESS IMAGE CODING RESULTS (LBR [bpp]).

Test Images	Conventional Methods				Proposed Method			
	Not Lifting 8 × 8 DCT [26]	1D-NSL		1D-SBL [13]		2D-NSBL		
	8 × 8 DCT [26]	4 × 8 HLT [9]	5/3-DWT [14], [15]	8 × 16 BOFB	8 × 24 BOFB	8 × 16 BOFB	8 × 24 BOFB	
Barbara	—	4.81	4.86	4.79	4.76	4.76	<b>4.75</b>	
Boat	—	5.13	5.09	5.09	5.10	<b>5.08</b>	5.09	
Finger	—	5.71	5.83	5.66	5.65	5.65	<b>5.64</b>	
Grass	—	<b>6.05</b>	6.06	<b>6.05</b>	<b>6.05</b>	<b>6.05</b>	<b>6.05</b>	
Lena	—	4.61	<b>4.48</b>	4.63	4.62	4.61	4.61	
Pepper	—	4.96	<b>4.85</b>	4.93	4.92	4.92	4.92	
Bridge	—	<b>4.63</b>	4.65	4.65	4.67	<b>4.63</b>	<b>4.63</b>	
Deer	—	4.77	4.76	4.76	4.76	<b>4.74</b>	<b>4.74</b>	
Arri	—	11.28	11.40	<b>11.27</b>	<b>11.27</b>	<b>11.27</b>	<b>11.27</b>	
Face	—	10.33	10.37	<b>10.28</b>	<b>10.28</b>	<b>10.28</b>	<b>10.28</b>	

where  $w_k$ s are weighted coefficients.  $C_{CG}$ ,  $C_{SAa}$ , and  $C_{SAs}$  are

$$C_{CG} = 10 \log_{10} \frac{\sigma_x^2}{\prod_{k=0}^{M-1} \sigma_{x_i}^2 \|f_i\|^2}$$

$$C_{SAa} = \sum_{k=0}^{M-1} \int_{\omega \in \Omega_i} W_i^a |H_i(e^{j\omega})|^2 d\omega$$

$$C_{SAs} = \sum_{k=0}^{M-1} \int_{\omega \in \Omega_i} W_i^s |F_i(e^{j\omega})|^2 d\omega,$$

where  $\sigma_x^2$ ,  $\sigma_{x_i}^2$ ,  $\|f_i\|^2$ ,  $W_i^a$ ,  $W_i^s$ , and  $\Omega_i$  are the variance of the input signal, the variance of the  $i$ -th subbands, the norm of the  $i$ -th synthesis filter, weighting functions for the stopband attenuation of the analysis, synthesis bank, and the stopband

region of  $H_i(z)$  and  $F_i(z)$ , respectively. The input signal  $x(n)$  is the AR(1) process with an intersample autocorrelation coefficient  $\rho = 0.95$  in common use.

### B. Lossy-to-Lossless Image Coding

The resulting BOFBs were implemented with a rounding operation at each lifting step and compared in terms of the lossless bitrate (LBR) [bpp] in lossless image coding:

$$\text{LBR [bpp]} = \frac{\text{Total number of bits [bit]}}{\text{Total number of pixels [pixel]}}$$

TABLE III  
LOSSY IMAGE CODING RESULTS (PSNR [dB]).

Test Images	Bitrate [bpp]	Conventional Methods					Proposed Method	
		Not Lifting 8 × 8 DCT [26]	4 × 8 HLT [9]	1D-NSL 9/7-DWT [14], [15]	1D-SBL [13] 8 × 16 BOFB 8 × 24 BOFB		2D-NSBL 8 × 16 BOFB 8 × 24 BOFB	
<i>Barbara</i>	0.25	26.96	26.85	27.24	28.04	28.64	28.05	<b>28.65</b>
	0.50	30.40	30.43	30.46	31.63	32.18	31.67	<b>32.20</b>
	1.00	34.98	35.05	34.85	35.88	36.26	35.94	<b>36.38</b>
<i>Boat</i>	0.25	27.85	27.62	28.45	28.26	28.62	28.25	<b>28.63</b>
	0.50	30.87	30.87	31.38	31.35	31.61	31.36	<b>31.63</b>
	1.00	34.39	34.31	34.48	34.66	34.85	34.70	<b>34.91</b>
<i>Finger</i>	0.25	22.78	22.96	23.49	23.52	<b>23.86</b>	23.51	<b>23.86</b>
	0.50	25.42	25.56	25.98	26.43	26.93	26.43	<b>26.95</b>
	1.00	29.17	29.01	29.07	30.06	30.78	30.07	<b>30.81</b>
<i>Grass</i>	0.25	24.00	23.99	24.35	24.27	<b>24.50</b>	24.28	<b>24.50</b>
	0.50	25.94	25.86	26.09	26.30	26.61	26.30	<b>26.62</b>
	1.00	28.70	28.68	28.68	29.08	29.42	29.10	<b>29.44</b>
<i>Lena</i>	0.25	30.55	31.65	<b>32.50</b>	32.20	31.85	32.20	31.86
	0.50	34.43	35.03	35.49	35.41	35.65	35.44	<b>35.73</b>
	1.00	38.87	38.65	38.63	38.68	38.76	38.78	<b>38.95</b>
<i>Pepper</i>	0.25	29.84	31.20	<b>31.90</b>	31.45	31.08	31.45	31.10
	0.50	32.83	33.95	<b>34.45</b>	33.97	33.70	34.01	33.76
	1.00	35.75	35.59	36.08	35.92	36.01	35.97	<b>36.11</b>
<i>Bridge</i>	0.25	31.11	31.14	31.79	31.93	32.19	31.94	<b>32.22</b>
	0.50	33.76	34.06	34.25	34.57	34.54	34.60	<b>34.63</b>
	1.00	36.64	37.35	36.83	37.24	37.27	37.38	<b>37.42</b>
<i>Deer</i>	0.25	<b>34.14</b>	33.96	34.10	33.88	34.02	33.90	34.06
	0.50	35.10	34.83	34.88	34.99	35.06	35.03	<b>35.13</b>
	1.00	<b>37.45</b>	36.97	36.83	37.11	37.04	37.20	37.27
<i>Arri</i>	0.25	31.92	33.22	33.28	33.63	<b>34.14</b>	33.63	<b>34.14</b>
	0.50	36.82	36.75	37.30	38.26	<b>38.98</b>	38.26	<b>38.98</b>
	1.00	41.80	42.28	41.97	43.87	<b>43.93</b>	43.87	<b>43.93</b>
<i>Face</i>	0.25	45.01	45.49	45.96	46.27	<b>46.67</b>	46.27	<b>46.67</b>
	0.50	47.85	48.40	48.72	48.98	<b>49.31</b>	48.98	<b>49.31</b>
	1.00	50.87	51.47	51.68	51.89	<b>52.22</b>	51.89	<b>52.22</b>

and the peak signal-to-noise ratio (PSNR) [dB] in lossy image coding:

$$\text{PSNR [dB]} = 10 \log_{10} \left( \frac{\text{MAX}_p^2}{\text{MSE}} \right),$$

where  $\text{MAX}_p$  and MSE are the maximum possible pixel value of the image and the mean squared error, respectively. To evaluate transform performance fairly, we employed two-, three-, six-, and two-level decompositions, respectively, on the eight-channel DCT [26] (H.265/HEVC)<sup>2</sup>, HLT [9] (JPEG XR), 1D-NSL-based DWTs without adaptive directionalities [14], [15], and eight-channel BOFBs. The 1D-SBL-based BOFBs had the same transfer function as the proposed FBs. The image set included six  $512 \times 512$  eight-bit standard grayscale images in [27], two  $2048 \times 2048$  eight-bit clipped grayscale images in [28], and two  $2816 \times 1600$  16-bit grayscale images in [29]. A quadtree-based embedded image coder EZW-IP [30] was used to encode the transformed images. EZW-IP is more suited to block transforms than are the popular zerotree-based coders, e.g., EZW [31] and SPIHT [32]. A periodic extension was used in the image boundary processing of the BOFBs, whereas the extensions used in JPEG XR and JPEG 2000 were used as the respective boundary processings of the HLT and DWTs.

Tables II, III, and Fig. 6 show lossless and lossy image coding results. Although the conventional methods sometimes

performed better on images with many low frequency components, overall, the 2D-NSBL-based BOFBs outperformed the conventional methods. These results are considered to be due to the merging (reducing) of many rounding operations in the 2D-NSBL-based BOFBs. Comparing Figs. 4 and 5, it is clear that the number of rounding operations of the 2D non-separable implementation is the almost half that of the 1D separable implementation. However, there were no differences between the 1D-SBL and 2D-NSBL of BOFBs in 16-bit images. In the future, we should solve the problem in high bit images.

## V. CONCLUSION

We devised a 2D-NSBL and applied it to  $M$ -channel PRFBs in lossy-to-lossless image coding. The 2D-NSBL is easily formulated from the 1D-SBL and 2D-NSL methods and can be regarded as an extension of the 2D-NSL because it is completely equivalent to a 2D-NSL when  $M = 2$ . A lossy-to-lossless image coding experiment confirmed the improvements that could be had with 2D-NSBL.

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<sup>2</sup>Since the DCT in H.265/HEVC is not composed of lifting structures, it is unsuitable for lossless image coding.



Fig. 6. Comparison of a particular area of an image *Barbara* (bitrate: 0.25 [bpp]): (left-right)  $8 \times 8$  DCT,  $4 \times 8$  HLT, 2D-NSL-based 9/7-DWT,  $8 \times 24$  1D-SBL-based BOFB [13], and  $8 \times 24$  2D-NSBL-based BOFB.

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