# Orthogonal Matrix-Valued Wavelet Packets<sup>†</sup>

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Received June 29, 2004; Accepted (in revised version) March 12, 2006

Abstract. In this paper, we introduce matrix-valued multiresolution analysis and matrixvalued wavelet packets. A procedure for the construction of the orthogonal matrix-valued wavelet packets is presented. The properties of the matrix-valued wavelet packets are investigated. In particular, a new orthonormal basis of  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  is obtained from the matrix-valued wavelet packets.

Key words: Matrix-valued multiresolution analysis; matrix-valued scaling functions; matrix-valued wavelet packets; refinement equation.

AMS subject classifications: 42C40, 65T60

### 1 Introduction

Wavelet packets, due to their nice characteristics, have been applied to signal processing [1], image compression [2], integral equations [3] and so on. Coifman and Meyer [4] firstly introduced the concept of orthogonal wavelet packets. The introduction for biorthogonal wavelet packets was attributable to Cohen and Daubechies [5]. Furthermore, Yang and Cheng [6] constructed a-scale orthogonal multiwavelet packets which are more flexible in applications. Recently, the multiwavelets have become the focus of active research both in theory and application, such as signal processing [7], mainly because of their ability to offer properties like orthogonality and symmetry simultaneously. The matrix-valued wavelets are a class of generalized multiwavelets. Xia and Suter [8] introduced the concept of the matrix-valued wavelets and investigated its construction. Moreover, they showed that multiwavelets can be generated from the component functions of matrix-valued wavelets. However, the multiwavelets and matrix-valued wavelets are different in the following sense. For example, prefiltering is usually required for discrete multiwavelet transforms [9] but not necessary for discrete matrix-valued wavelet transforms. A typical example of such matrix-valued signals is video images. Hence, studying the matrix-valued wavelets is useful in representations of signals. It is necessary to extend the concept of orthogonal wavelet packets to the case of orthogonal matrix-valued wavelets. Based on an observation in

Numer. Math. J. Chinese Univ. (English Ser.)

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<sup>&</sup>lt;sup>†</sup>This work is partially supported by the Natural Science Foundation of Henan (0211044800).

[8] and some ideas from [5,6], we will give the definition for 3-scale orthogonal matrix-valued wavelet packets and investigate the properties of the orthogonal matrix-valued wavelet packets by using matrix theory and integral transform.

Throughout the paper, we use the following notations. Let  $\mathbb{R}$  and  $\mathbb{C}$  be sets of all real and complex numbers, respectively.  $\mathbb{Z}$  stands for all integers. Set  $s \in \mathbb{Z}$ ,  $s \geq 2$ , and  $\mathbb{Z}_+ = \{z : z \geq 0, z \in \mathbb{Z}\}$ . By  $\mathbf{I}_s$  and  $\mathbf{O}$ , we denote the  $s \times s$  identity matrix and zero matrix, respectively.

$$L^{2}(\mathbb{R}, \mathbb{C}^{s \times s}) := \left\{ h(t) := \begin{pmatrix} h_{11}(t) & h_{12}(t) & \cdots & h_{1s}(t) \\ h_{21}(t) & h_{22}(t) & \cdots & h_{2s}(t) \\ \cdots & \cdots & \cdots & \cdots \\ h_{s1}(t) & h_{s2}(t) & \cdots & h_{ss}(t) \end{pmatrix} : \begin{array}{c} t \in \mathbb{R}, \ h_{kl}(t) \in L^{2}(\mathbb{R}), \\ k, l = 1, 2, \cdots, s \end{pmatrix} \right\}$$

The signal space  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  is called a matrix-valued function space. Examples of matrix-valued signals are video images where  $h_{kl}(t)$  is the pixel on the kth row and the lth column at time t.

For each  $\hbar \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ ,  $||\hbar||$  represents the norm of operator  $\hbar$  as

$$||\hbar|| := \left(\sum_{k,\,l=1}^{s} \int_{\mathbb{R}} |h_{k,\,l}(t)|^2 dt\right)^{1/2}.$$
(1)

which is the norm used in this paper for the matrix-valued function spaces  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ .

For  $\hbar \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ , its integration  $\int_{\mathbb{R}} \hbar(t) dt$  is defined as  $\int_{\mathbb{R}} \hbar(t) dt := (\int_{\mathbb{R}} h_{k,l}(t) dt)_{k,l=1}^{s}$ , where  $\hbar(t)$  is the matrix-valued functions  $(h_{k,l}(t))_{k,l=1}^{s}$  to be defined below. The Fourier transform of  $\hbar(t)$  is defined by  $\hat{h}(\omega) := \int_{\mathbb{R}} \hbar(t) \exp\{-i\omega t\} dt$ ,  $\omega \in \mathbb{R}$ .

For two matrix-valued functions  $\hbar$ ,  $\Upsilon \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ , their symbol inner product is defined by  $[\hbar, \Upsilon] := \int_{\mathbb{R}} \hbar(t) \Upsilon(t)^* dt$ . Here and afterwards, \* means the transpose and the complex conjugate.

**Definition 1.1.** A sequence  $\{\hbar_k(t)\}_{k\in\mathbb{Z}} \subset \mathbf{X} \subset L^2(\mathbb{R}, \mathbb{C}^{s\times s})$  is called an orthonormal set in  $\mathbf{X}$ , if it satisfies

$$[\hbar_k, \hbar_l] = \delta_{k,l} \mathbf{I}_s, \qquad k, l \in \mathbb{Z}$$
<sup>(2)</sup>

where  $\delta_{k,l}$  is the Kronecker symbol, i.e.,  $\delta_{k,l} = 1$  as k = l and  $\delta_{k,l} = 0$  otherwise.

**Definition 1.2.** A matrix-valued function  $\hbar(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  is said to be orthonormal, if  $\{\hbar(t-k)\}_{k \in \mathbb{Z}}$  is an orthonormal set.

**Definition 1.3.** A sequence of matrix-valued functions  $\{\hbar_k(t)\}_{k\in\mathbb{Z}} \subset \mathbf{X} \subset L^2(\mathbb{R}, \mathbb{C}^{s\times s})$  is called an orthonormal basis of  $\mathbf{X}$  if it satisfies (2) and for any  $\Upsilon(t) \in \mathbf{X}$ , there exists a unique matrix sequence  $\{P_k\}_{k\in\mathbb{Z}}$  such that  $\Upsilon(t) = \sum_{k\in\mathbb{Z}} P_k \hbar_k(t), t \in \mathbb{R}$ .

This paper is organized as follows. In Section 2, we briefly recall the concepts relevant to the matrix-valued multiresolution analysis. In Section 3, we give our main result, and some properties of the matrix-valued wavelet packets.

### 2 Matrix-valued multiresolution analysis and wavelets

We begin with the generic setting of a matrix-valued multiresolution analysis of  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ . Let  $\mathbf{S}(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  satisfy the following refinement equation:

$$\mathbf{S}(t) = 3 \cdot \sum_{k \in \mathbb{Z}} A_k \, \mathbf{S}(3\,t-k),\tag{3}$$

where  $\{A_k\}_{k\in\mathbb{Z}}$  is a finitely supported sequence of  $s \times s$  constant matrix.

Define a closed subspace  $\mathbf{V}_j \subset L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  by

$$\mathbf{V}_{j} = \mathbf{clos}_{L^{2}(\mathbb{R},\mathbb{C}^{s\times s})} \langle \mathbf{S}(3^{j} \cdot -k) : k \in \mathbb{Z} \rangle, \ j \in \mathbb{Z}.$$

$$\tag{4}$$

**Definition 2.1.** We say that  $\mathbf{S}(t)$  in (3) generates a matrix-valued multiresolution analysis  $\{\mathbf{V}_j\}_{j\in\mathbb{Z}}$  of  $L^2(\mathbb{R}, \mathbb{C}^{s\times s})$ , if the sequence  $\{\mathbf{V}_j\}_{j\in\mathbb{Z}}$  defined by (4) satisfies:

- (1).  $\cdots \subset \mathbf{V}_{-1} \subset \mathbf{V}_0 \subset \mathbf{V}_1 \subset \cdots$ ;
- (2).  $\bigcap_{j \in \mathbb{Z}} \mathbf{V}_j = \{\mathbf{O}\}; \bigcup_{\mathbf{j} \in \mathbb{Z}} \mathbf{V}_{\mathbf{j}} \text{ is dense in } L^2(\mathbb{R}, \mathbb{C}^{s \times s});$
- (3).  $\hbar(\cdot) \in \mathbf{V}_0 \Longleftrightarrow \hbar(3^j \cdot) \in \mathbf{V}_j, \ \forall j \in \mathbb{Z};$
- (4).  $\exists \mathbf{S}(t) \in \mathbf{V}_0$  such that  $\mathbf{S}_k(t) := \mathbf{S}(t-k), k \in \mathbb{Z}$ , form an orthonarmal basis for  $\mathbf{V}_0$ .

A matrix-valued functions  $\mathbf{S}(t)$  in (3) is said to be a matrix-valued scaling function if it generates a matrix-valued multiresolution analysis. Equation (3) is called a refinement equation. Set  $\mathcal{A}(\omega) = \sum_{k \in \mathbb{Z}} A_k \cdot \exp\{-ik\omega\}, \quad \omega \in \mathbb{R}$ . Then, the frequency form of (3) is

$$\widehat{\mathbf{S}}(\omega) = \mathcal{A}(\omega/3) \,\widehat{\mathbf{S}}(\omega/3), \quad \omega \in \mathbb{R}.$$
(5)

In the following, without loss of generality we assume  $\widehat{\mathbf{S}}(\omega)$  is continuous at the origin and  $\widehat{\mathbf{S}}(0) = \mathbf{I}_s$ .

Let  $\mathbf{U}_j$ ,  $j \in \mathbb{Z}$  be the orthocomplement space of  $\mathbf{V}_j$  in  $\mathbf{V}_{j+1}$ . Assume there exist two matrixvalued functions  $W_1(t)$ ,  $W_2(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ , such that their translates and dilates form a *Riesz* basis of  $\mathbf{U}_j$ , i.e.,

$$\mathbf{U}_{j} = \mathbf{clos}_{L^{2}(\mathbb{R},\mathbb{C}^{s\times s})} \langle W_{i}(3^{j} \cdot -k) : i = 1, 2, \ k \in \mathbb{Z} \rangle, \quad j \in \mathbb{Z}.$$
(6)

Since  $W_1(t), W_2(t) \in \mathbf{U}_0 \subset \mathbf{V}_1$ , there exist two finitely supported sequences of  $s \times s$  matrix  $\{B_k^{(i)}\}_{k \in \mathbb{Z}}, i = 1, 2$  such that  $W_i(t) = 3 \sum_{k \in \mathbb{Z}} B_k^{(i)} \mathbf{S}(3t - k)$ . Taking Fourier transform for (6) gives

$$\widehat{W}_{i}(\omega) = \mathcal{B}^{(i)}(\omega/3)\,\widehat{\mathbf{S}}(\omega/3), \ i = 1, 2, \quad \omega \in \mathbb{R},$$
(7)

where

$$\mathcal{B}^{(i)}(\omega) = \sum_{k \in \mathbb{Z}} B_k \cdot \exp\{-ik\omega\}, \quad i = 1, 2.$$
(8)

We call  $\mathbf{S}(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  an orthonormal matrix-valued scaling function if it is a scaling function and satisfies

$$[\mathbf{S}(\cdot), \mathbf{S}(\cdot - n)] = \delta_{0, n} \mathbf{I}_s, \quad n \in \mathbb{Z}.$$
(9)

We say that  $W_1(t)$ ,  $W_2(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  are two orthonormal matrix-valued wavelet functions associated with an orthonormal matrix-valued scaling functions if it satisfies

$$[\mathbf{S}(\cdot), W_{i}(\cdot - n)] = \mathbf{O}, \quad i = 1, 2, \quad n \in \mathbb{Z};$$
(10)

$$[W_{i}(\cdot), W_{j}(\cdot - n)] = \delta_{i, j} \delta_{0, n} \mathbf{I}_{s}, \ i, j \in \{1, 2\}, \ n \in \mathbb{Z}.$$
(11)

**Lemma 2.1.** Let  $\hbar(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ . Then  $\hbar(t)$  is orthonormal if and only if

$$\sum_{k\in\mathbb{Z}}\widehat{\hbar}(\omega+2k\pi)\widehat{\hbar}(\omega+2k\pi)^* = \mathbf{I}_s.$$
(12)

**Proof** If  $\hbar(t)$  is an orthonormal matrix-valued functions, then we get from (2) that

$$\delta_{0,k} \mathbf{I}_{s} = [\hbar(\cdot), \hbar(\cdot - k)] = \frac{1}{2\pi} \int_{\mathbb{R}} \widehat{h}(\omega) \widehat{h}(\omega)^{*} \cdot \exp\{ik\omega\} d\omega$$
$$= \frac{1}{2\pi} \int_{0}^{2\pi} \sum_{l \in \mathbb{Z}} \widehat{h}(\omega + 2l\pi) \widehat{h}(\omega + 2l\pi)^{*} \cdot \exp\{ik\omega\} d\omega,$$

which implies that (12) holds. The converse is obvious.

By Lemma 2.1 and (5), (7), (9)-(11), we can obtain the following lemma.

**Lemma 2.2.** ([8]) Let  $\mathbf{S}(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  be an orthonormal matrix-valued scaling function. Assume  $W_1(t), W_2(t) \in L^2(\mathbb{R}, \mathbb{C}^{s \times s})$  are orthogonal matrix-valued wavelet functions associated with  $\mathbf{S}(t)$ . Then we have

$$\mathcal{A}(\omega)\mathcal{A}(\omega)^* + \mathcal{A}(\omega_1)\mathcal{A}(\omega_1)^* + \mathcal{A}(\omega_2)\mathcal{A}(\omega_2)^* = \mathbf{I}_s, \quad \omega \in \mathbb{R},$$
(13)

$$\mathcal{A}(\omega) \mathcal{A}(\omega)^{*} + \mathcal{A}(\omega_{1}) \mathcal{A}(\omega_{1})^{*} + \mathcal{A}(\omega_{2}) \mathcal{A}(\omega_{2})^{*} = \mathbf{I}_{s}, \quad \omega \in \mathbb{R},$$
(13)  
$$\mathcal{A}(\omega) \mathcal{B}^{(i)}(\omega)^{*} + \mathcal{A}(\omega_{1}) \mathcal{B}^{(i)}(\omega_{1})^{*} + \mathcal{A}(\omega_{2}) \mathcal{B}^{(i)}(\omega_{2})^{*} = \mathbf{O}, \quad i = 1, 2, \quad \omega \in \mathbb{R},$$
(14)  
$$\mathcal{B}^{(i)}(\omega) \mathcal{B}^{(j)}(\omega)^{*} + \mathcal{B}^{(i)}(\omega_{1}) \mathcal{B}^{(j)}(\omega_{1})^{*} + \mathcal{B}^{(i)}(\omega_{2}) \mathcal{B}^{(j)}(\omega_{2})^{*} = \delta_{i,j} \mathbf{I}_{s}, \quad i, j \in \{1, 2\},$$
(15)

$$\mathcal{B}^{(i)}(\omega) \mathcal{B}^{(j)}(\omega)^* + \mathcal{B}^{(i)}(\omega_1) \mathcal{B}^{(j)}(\omega_1)^* + \mathcal{B}^{(i)}(\omega_2) \mathcal{B}^{(j)}(\omega_2)^* = \delta_{i,j} \mathbf{I}_s, \quad i, j \in \{1, 2\}, \quad (15)$$

where  $\omega_1 = \omega + 2\pi/3$  and  $\omega_2 = \omega + 4\pi/3$ .

We now present matrix-valued Meyer wavelets as a special family of the matrix-valued wavelets. For more about scalar-valued Meyer wavelets, see [10]. Let

$$\widehat{\mathbf{S}}(\omega) = \begin{cases} \mathbf{I}_s, & |\omega| < \frac{2\pi}{3}, \\ \cos\left[\frac{\pi}{2}f\left(\frac{3}{2\pi}\right)|\omega| - 1\right] \Gamma(\omega), & \frac{2\pi}{3} \le |\omega| \le \frac{4\pi}{3}, \\ 0, & \text{otherwise}, \end{cases}$$
(16)

where  $\Gamma(\omega)$  is paramitary and  $\Gamma(2\pi/3) = \Gamma(-2\pi/3) = \mathbf{I}_s$ , and f(t) is a scalar-valued smooth function such that

$$f(t) = \begin{cases} 1, & t \ge 1, \\ 0, & t \le 0, \end{cases} \quad \text{and} \quad f(t) + f(1-t) = 1, \quad \text{for}t \in (0, 1). \end{cases}$$

Then, after some computation, for  $\omega \in \mathbb{R}$ , we get that  $\sum_{k \in \mathbb{Z}} \widehat{\mathbf{S}}(\omega + 2k\pi) \widehat{\mathbf{S}}(\omega + 2k\pi)^* = \mathbf{I}_s$ .

By Lemma 2.1,  $\mathbf{S}(t)$  is an orthonormal matrix-valued scaling function. This implies that  $\mathbf{S}(t)$ defined by (16) is a matrix-valued scaling function. Similar to the scalar-valued Meyer wavelets ([10, p. 138]), the corresponding lowpass filter  $\mathcal{A}(\omega)$  is  $\mathcal{A}(\omega) = \sum_{k \in \mathbb{Z}} \widehat{\mathbf{S}}(2(\omega + 2k\pi)).$ 

By using paraunitary vector filter theory [11], we can obtain two filter functions  $\mathcal{B}^{(1)}(\omega)$  and  $\mathcal{B}^{(2)}(\omega)$  satisfying (14) and (15). Let  $\widehat{W}_{i}(\omega) = \mathcal{B}^{(i)}(\omega/3) \widehat{\mathbf{S}}(\omega/3), i = 1, 2$ . Then,  $W_{1}(t)$  and  $W_2(t)$  are two matrix-valued Meyer wavelets [8].

#### 3 Orthogonal matrix-valued wavelet packets

Xia and Suter [8] introduced the notion of matrix-valued wavelets and investigated their construction. In this section, we will give the definition of the matrix-valued wavelet packets and discuss some of their properties. First, we set

$$\Psi_0(t) = \mathbf{S}(t), \quad \Psi_i(t) = W_i(t); \quad \Omega_k^{(0)} = A_k, \quad \Omega_k^{(i)} = B_k^{(i)}, \ i = 1, 2, \quad k \in \mathbb{Z}.$$

**Definition 3.1.** The collection of the matrix-valued functions { $\Psi_{3n+\lambda}(t)$ ,  $n = 0, 1, \dots, \lambda = 0, 1, 2$ } is called a matrix-valued wavelet packet with respect to the orthogonal matrix-valued scaling function  $\mathbf{S}(t)$ , where

$$\Psi_{3n+\lambda}(t) = 3 \cdot \sum_{k \in \mathbb{Z}} \Omega_k^{(\lambda)} \Psi_n(3t-k), \quad \lambda = 0, 1, 2.$$
(17)

By implementing the Fourier transform for both sides of (17), we have

$$\widehat{\Psi}_{3n+\lambda}(\omega) = \mathbf{\Omega}^{(\lambda)}(\omega/3) \,\widehat{\Psi}_n(\omega/3), \quad \lambda = 0, 1, 2, \tag{18}$$

where

$$\mathbf{\Omega}^{(\lambda)}(\omega) = \sum_{k \in \mathbb{Z}} \Omega_k^{(\lambda)} \cdot \exp\{-ik\omega\}, \quad \lambda = 0, 1, 2, \quad \omega \in \mathbb{R}.$$
(19)

Thus,  $\Omega^{(0)}(\omega) = \mathcal{A}(\omega), \ \Omega^{(i)}(\omega) = \mathcal{B}^{(i)}(\omega), \ i = 1, 2$ . Formulas (13)-(15) can be written as

$$\sum_{\sigma=0}^{2} \mathbf{\Omega}^{(\lambda)} \left( \omega + \frac{2\pi\sigma}{3} \right) \mathbf{\Omega}^{(\mu)} \left( \omega + \frac{2\pi\sigma}{3} \right)^{*} = \delta_{\lambda,\mu} \mathbf{I}_{s}, \quad \lambda, \, \mu \in \{0, 1, 2\}, \quad \omega \in \mathbb{R}.$$
(20)

It is evident that (20) is equivalent to

$$\sum_{\sigma \in \mathbb{Z}} \Omega_{\sigma+3k}^{(\lambda)} \, (\Omega_{\sigma+3l}^{(\mu)})^* = \frac{1}{3} \delta_{\lambda,\,\mu} \delta_{k,\,l} \mathbf{I}_s, \, \lambda, \quad \mu = 0, \, 1, \, 2, \quad k, \, l \in \mathbb{Z}.$$

In the following, we will investigate the properties of the matrix-valued wavelet packets.

**Theorem 3.1.** If  $\{\Psi_n(t)\}$  is a matrix-valued wavelet packets with respect to the orthogonal matrix-valued scaling function  $\mathbf{S}(t)$ , then for every  $n \in \mathbb{Z}_+$ , we have

$$\left[\Psi_{n}(\cdot - j), \Psi_{n}(\cdot - k)\right] = \delta_{j,k} \mathbf{I}_{s}, \quad j, k \in \mathbb{Z}.$$
(22)

**Proof** (Induction) (i) The result (22) follows from (9) as n = 0. (ii) Assume that (22) holds when  $0 \le n < 3^{\mathcal{L}}$ , where  $\mathcal{L}$  is a positive integer. Then, as  $3^{\mathcal{L}} \le n < 3^{\mathcal{L}+1}$ , we have  $3^{\mathcal{L}-1} \le [n/3] < 3^{\mathcal{L}}$  where  $[\rho] = \max\{\nu \in \mathbb{Z}, \nu \le \rho\}$ . Thus, order  $n = 3[n/3] + \lambda, \lambda = 0, 1, 2$ . By the induction assumption and Lemma 2.1, we obtain

$$\left[\Psi_{\left[\frac{n}{3}\right]}(\cdot-j),\,\Psi_{\left[\frac{n}{3}\right]}(\cdot-k)\right] = \delta_{j,\,k}\mathbf{I}_{s} \Longleftrightarrow \sum_{l\in\mathbb{Z}}\widehat{\Psi}_{\left[\frac{n}{3}\right]}(\omega+2l\pi)\widehat{\Psi}_{\left[\frac{n}{3}\right]}(\omega+2l\pi)^{*} = \mathbf{I}_{s}.$$
 (23)

It follows from (18), (20) and (23) that

$$\begin{split} &\sum_{l\in\mathbb{Z}}\widehat{\Psi}_{n}(\omega+2l\pi)\,\widehat{\Psi}_{n}(\omega+2l\pi)^{*} \\ &=\sum_{l\in\mathbb{Z}}\Omega^{(\lambda)}\left(\frac{\omega+2l\pi}{3}\right)\widehat{\Psi}_{\left[\frac{n}{3}\right]}\left(\frac{\omega+2l\pi}{3}\right)\widehat{\Psi}_{\left[\frac{n}{3}\right]}\left(\frac{\omega+2l\pi}{3}\right)^{*}\Omega^{(\lambda)}\left(\frac{\omega+2l\pi}{3}\right)^{*} \\ &=\sum_{\sigma=0}^{2}\Omega^{(\lambda)}\left(\frac{\omega+2\sigma\pi}{3}\right)\left\{\sum_{\kappa\in\mathbb{Z}}\widehat{\Psi}_{\left[\frac{n}{2}\right]}\left(\frac{\omega+2\sigma\pi}{3}+2\kappa\pi\right)\widehat{\Psi}_{\left[\frac{n}{2}\right]}\left(\frac{\omega+2\sigma\pi}{3}+2\kappa\pi\right)^{*}\right\}\Omega^{(\lambda)}\left(\frac{\omega+2\sigma\pi}{3}\right)^{*} \\ &=\sum_{\sigma=0}^{2}\Omega^{(\lambda)}\left(\frac{\omega+2\sigma\pi}{3}\right)\Omega^{(\lambda)}\left(\frac{\omega+2\sigma\pi}{3}\right)^{*} = \mathbf{I}_{s} \end{split}$$

Therefore, by Lemma 2.1, the result (22) follows.

**Theorem 3.2.** If  $\{\Psi_n(t)\}$  is a matrix-valued wavelet packets with respect to the orthogonal matrix-valued scaling function  $\mathbf{S}(t)$ , then for every  $n \in \mathbb{Z}_+$ , we have

$$[\Psi_{3n+\lambda}(\cdot), \Psi_{3n+\mu}(\cdot - k)] = \delta_{\lambda,\mu} \,\delta_{0,k} \,\mathbf{I}_s, \ \lambda, \mu \in \{0, 1, 2\}, \ k \in \mathbb{Z}.$$
(24)

**Proof** By (18) and (21) and Theorem 3.1, we obtain

$$\begin{split} \left[ \Psi_{3n+\lambda}(\cdot) , \Psi_{3n+\mu}(\cdot-k) \right] &= \frac{1}{2\pi} \int_{\mathbb{R}} \Omega^{(\lambda)} \left(\frac{\omega}{3}\right) \widehat{\Psi}_n \left(\frac{\omega}{3}\right)^* \Omega^{(\mu)} \left(\frac{\omega}{3}\right)^* \cdot e^{ik\omega} \, d\omega \\ &= \frac{1}{2\pi} \int_0^{6\pi} \Omega^{(\lambda)} \left(\frac{\omega}{3}\right) \left\{ \sum_{l \in \mathbb{Z}} \widehat{\Psi}_n \left(\frac{\omega}{3} + 2l\pi\right) \, \widehat{\Psi}_n \left(\frac{\omega}{3} + 2l\pi\right)^* \right\} \Omega^{(\mu)} \left(\frac{\omega}{3}\right)^* \cdot \exp\{ik\omega\} d\omega \\ &= \frac{1}{2\pi} \int_0^{2\pi} \sum_{\sigma=0}^2 \Omega^{(\lambda)} \left(\frac{\omega + 2\pi\sigma}{3}\right) \, \Omega^{(\mu)} \left(\frac{\omega + 2\pi\sigma}{3}\right)^* \exp\{ik\omega\} d\omega \\ &= \frac{1}{2\pi} \int_0^{2\pi} \delta_{\lambda,\mu} \mathbf{I}_s \cdot \exp\{ik\omega\} \, d\omega = \delta_{\lambda,\mu} \delta_{0,k} \mathbf{I}_s. \end{split}$$

This completes the proof of this theorem.

**Theorem 3.3.** For any  $m, n \in \mathbb{Z}_+$  and  $k \in \mathbb{Z}$ , we have

$$\left[\Psi_{m}(\cdot), \Psi_{n}(\cdot - k)\right] = \delta_{m,n} \,\delta_{0,k} \mathbf{I}_{s}.$$
<sup>(25)</sup>

**Proof** For m = n, (25) follows by Theorem 3.1. Without loss of generality, we suppose m > n in case of  $m \neq n$ . Rewrite m, n as  $m = 3[m/3] + \lambda_1$ ,  $n = 3/[n/3] + \mu_1$ , where  $\lambda_1$ ,  $\mu_1 \in \{0, 1, 2\}$ . Case 1. If [m/3] = [n/3], then  $\lambda_1 \neq \mu_1$ . By (18), (20) and (23),

$$\begin{split} \left[ \Psi_m \left( \cdot \right), \, \Psi_n \left( \cdot - k \right) \right] &= \frac{1}{2\pi} \int_{\mathbb{R}} \mathbf{\Omega}^{(\lambda_1)} \left( \frac{\omega}{3} \right) \, \widehat{\Psi}_{\left[\frac{m}{3}\right]} \left( \frac{\omega}{3} \right)^* \mathbf{\Omega}^{(\mu_1)} \left( \frac{\omega}{3} \right)^* \cdot \exp\{ik\omega\} \, d\omega \\ &= \frac{3}{2\pi} \int_0^{2\pi} \mathbf{\Omega}^{(\lambda_1)} (\omega) \left\{ \sum_{l \in \mathbb{Z}} \widehat{\Psi}_{\left[\frac{n}{3}\right]} (\omega + 2l\pi) \, \widehat{\Psi}_{\left[\frac{n}{3}\right]} (\omega + 2l\pi)^* \right\} \, \mathbf{\Omega}^{(\mu_1)} (\omega)^* \cdot \exp\{3ik\omega\} \, d\omega \\ &= \frac{3}{2\pi} \int_0^{\frac{2\pi}{3}} \sum_{\sigma=0}^2 \mathbf{\Omega}^{(\lambda_1)} \left( \omega + \frac{2\pi\sigma}{3} \right) \, \mathbf{\Omega}^{(\mu_1)} \left( \omega + \frac{2\pi\sigma}{3} \right) \cdot \exp\{3ik\omega\} \, d\omega \\ &= \frac{3}{2\pi} \int_0^{\frac{2\pi}{3}} \delta_{\lambda_1, \, \mu_1} \mathbf{I}_s \cdot \exp\{3ik\omega\} \, d\omega = \mathbf{O}, \end{split}$$

which implies that (25) holds in this case.

Case 2. If  $[\frac{m}{3}] \neq [\frac{n}{3}]$ , then set  $[m/3] = 3[[m/3]/3] + \lambda_2$ ,  $[n/3] = 3[[n/3]/3] + \mu_2$ ,  $\lambda_2, \mu_2 \in \{0, 1, 2\}$ . If [[m/3]/3] = [[n/3]/3], then (25) can be established similar to Case 1. If  $[[m/3]/3] \neq [[n/3]/3]$ , then we again set  $[[m/3]/3] = 3[[[m/3]/3] + \lambda_3, [[n/3]/3] = 3[[[n/3]/3] + \mu_3, \lambda_3, \mu_3 \in \{0, 1, 2\}$ . Thus, after taking finite times steps (denoted by  $\kappa$ ), we obtain

$$a_{\kappa} = b_{\kappa} = 1, \quad \text{or} \quad a_{\kappa} = b_{\kappa} = 2,$$

$$(26)$$

where

$$a_{\kappa} = \overbrace{[\cdots]}^{\kappa} m/2 \cdots ]/2], \quad b_{\kappa} = \overbrace{[\cdots]}^{\kappa} n/2 \cdots ]/2].$$

 $a_{\kappa} = 1, \ b_{\kappa} = 0, \quad \text{or} \ a_{\kappa} = 2, \ b_{\kappa} = 1, \quad \text{or} \ a_{\kappa} = 2, \ b_{\kappa} = 0, \quad \lambda_{\kappa}, \ \mu_{\kappa} \in \{0, 1, 2\}.$  (27)

For the case (26), the result (25) follows similarly to Case 1. For the case (27), we have from (10) and (11) that

$$\sum_{l\in\mathbb{Z}}\widehat{\Psi}_{a_{\kappa}}(\omega+2l\pi)\,\widehat{\Psi}_{b_{\kappa}}(\omega+2l\pi)^{*}=\mathbf{O},\qquad\omega\in\mathbb{R}.$$

Consequently,

$$\begin{split} \left[ \Psi_{m}(\cdot), \Psi_{n}(\cdot-k) \right] &= \frac{1}{2\pi} \int_{\mathbb{R}} \widehat{\Psi}_{m}(\omega) \widehat{\Psi}_{n}(\omega)^{*} \cdot \exp\{ik\omega\} d\omega \\ &= \frac{1}{2\pi} \int_{\mathbb{R}} \Omega^{(\lambda_{1})} \left(\frac{\omega}{3}\right) \widehat{\Psi}_{[m/3]} \left(\frac{\omega}{3}\right) \widehat{\Psi}_{[n/3]} \left(\frac{\omega}{3}\right)^{*} \Omega^{(\mu_{1})} \left(\frac{\omega}{3}\right)^{*} \cdot \exp\{ik\omega\} d\omega = \cdots \cdots \\ &= \frac{1}{2\pi} \int_{\mathbb{R}} \prod_{\sigma=1}^{\kappa} \Omega^{(\lambda_{\sigma})} \left(\frac{\omega}{3^{\sigma}}\right) \widehat{\Psi}_{a_{\kappa}} \left(\frac{\omega}{3^{\kappa}}\right) \widehat{\Psi}_{b_{\kappa}} \left(\frac{\omega}{3^{\kappa}}\right)^{*} \left(\prod_{\sigma=1}^{\kappa} \Omega^{(\mu_{\sigma})} \left(\frac{\omega}{3^{\sigma}}\right)\right)^{*} \cdot \exp\{ik\omega\} d\omega \\ &= \frac{1}{2\pi} \int_{0}^{3^{\kappa+1}\pi} \prod_{\sigma=1}^{\kappa} \Omega^{(\lambda_{\sigma})} \left(\frac{\omega}{3^{\sigma}}\right) \left(\sum_{l \in \mathbb{Z}} \widehat{\Psi}_{a_{\kappa}} \left(\frac{\omega}{3^{\kappa}} + 2l\pi\right) \widehat{\Psi}_{b_{\kappa}} \left(\frac{\omega}{3^{\kappa}} + 2l\pi\right)^{*} \right) \\ &\quad \cdot \left(\prod_{\sigma=1}^{\kappa} \Omega^{(\mu_{\sigma})} \left(\frac{\omega}{3^{\sigma}}\right)\right)^{*} \cdot e^{ik\omega} d\omega \\ &= \frac{1}{2\pi} \int_{0}^{3^{\kappa+1}\pi} \prod_{\sigma=1}^{\kappa} \Omega^{(\lambda_{\sigma})} \left(\frac{\omega}{2^{\sigma}}\right) \cdot \mathbf{O} \cdot \left(\prod_{\sigma=1}^{\kappa} \Omega^{(\mu_{\sigma})} \left(\frac{\omega}{3^{\sigma}}\right)\right)^{*} \cdot \exp\{ik\omega\} d\omega = \mathbf{O}. \end{split}$$

Therefore, for any  $m, n \in \mathbb{Z}_+$  and  $k \in \mathbb{Z}$ , (25) holds.

**Lemma 3.1.** If  $\{\Psi_n(t), n = 0, 1, 2, \dots\}$  is a matrix-valued wavelet packets with respect to the orthonormal matrix-valued scaling functions  $\mathbf{S}(t)$ , then for every  $n \in \mathbb{Z}+$ , we have

$$\Psi_n(3t-k) = \frac{1}{3} \sum_{\sigma=0}^2 \sum_{l \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^* \Psi_{3n+\sigma}(t-l), \ k \in \mathbb{Z}.$$
 (28)

**Proof** Observe

$$\frac{1}{3} \sum_{\sigma=0}^{2} \sum_{l \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^{*} \Psi_{3n+\sigma}(t-l) = \sum_{\sigma=0}^{2} \sum_{l \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^{*} \sum_{j \in \mathbb{Z}} \Omega_{j}^{(\sigma)} \Psi_{n}(3t-3l-j)$$

$$= \sum_{\sigma=0}^{2} \sum_{l \in \mathbb{Z}} \sum_{m \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^{*} \Omega_{m-3l}^{(\sigma)} \Psi_{n}(3t-m) = \sum_{m \in \mathbb{Z}} \left\{ \sum_{\sigma=0}^{2} \sum_{l \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^{*} \Omega_{m-3l}^{(\sigma)} \right\} \Psi_{n}(3t-m)$$

$$= \sum_{m \in \mathbb{Z}} \delta_{k,m} I_{s} \Psi_{n}(3t-m) = \Psi_{n}(3t-k).$$

This completes the proof of Lemma 3.1.  $\hfill\blacksquare$ 

We shall discuss the orthogonal decomposition relation for  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ . Let

$$\mathbf{Y}_{j}^{n} = \mathbf{clos}_{L^{2}(\mathbb{R}, \mathbb{C}^{s \times s})} \langle \Psi_{n}(3^{j} \cdot -k) : k \in \mathbb{Z} \rangle, \ n \in \mathbb{Z}_{+}, \ j \in \mathbb{Z}.$$

$$(29)$$

**Theorem 3.4.** Let  $n \in \mathbb{Z}_+$  and  $\bigoplus$  denote orthogonal direct sum. We have

$$\mathbf{Y}_{j+1}^{n} = \mathbf{Y}_{j}^{3n} \bigoplus \mathbf{Y}_{j}^{3n+1} \bigoplus \mathbf{Y}_{j}^{3n+2}, \ j \in \mathbb{Z}.$$
(30)

**Proof** According to (17) and (29),  $\mathbf{Y}_{j}^{3n} \bigoplus \mathbf{Y}_{j}^{3n+1} \bigoplus \mathbf{Y}_{j}^{3n+2} \subset \mathbf{Y}_{j+1}^{n}$ . On the other hand,  $\mathbf{Y}_{j}^{3n}$ ,  $\mathbf{Y}_{j}^{3n+1}$  and  $\mathbf{Y}_{j}^{3n+2}$  are orthogonal to each other by Theorem 3.2. By Lemma 3.1, we have

$$\Psi_n(3^{j+1}t - k) = \frac{1}{3} \sum_{\sigma=0}^2 \sum_{l \in \mathbb{Z}} (\Omega_{k-3l}^{(\sigma)})^* \Psi_{n+\sigma}(3^j t - l), \ j, \ k \in \mathbb{Z}.$$

Hence, the basis of the space  $\mathbf{Y}_{j+1}^n$  can be linearly represented by the basis of the space  $\mathbf{Y}_{j}^{3n}$ ,  $\mathbf{Y}_{j}^{3n+1}$  and  $\mathbf{Y}_{j}^{3n+2}$ . Then, we have  $\mathbf{Y}_{j+1}^n \subset \mathbf{Y}_{j}^{3n} \bigoplus \mathbf{Y}_{j}^{3n+1} \bigoplus \mathbf{Y}_{j}^{3n+2}$ . This implies that (30) holds for every  $n \in \mathbb{Z}_+$ ,  $j \in \mathbb{Z}$ .

**Corollary 3.1.** For every  $j \ge 1$  and  $1 \le k \le j$ , we have

$$\mathbf{U}_{j} = \mathbf{Y}_{j-k}^{3^{k}} \bigoplus \mathbf{Y}_{j-k}^{3^{k+1}} \bigoplus \cdots \bigoplus \mathbf{Y}_{j-k}^{3^{k+1}-1}.$$
(31)

Moreover,

$$L^{2}(\mathbb{R}, \mathbb{C}^{s \times s}) = \bigoplus_{j \in \mathbb{Z}} \mathbf{U}_{j} = \cdots \bigoplus \mathbf{U}_{-2} \bigoplus \mathbf{U}_{-1} \bigoplus \mathbf{U}_{0} \bigoplus_{\kappa=3}^{\infty} \mathbf{Y}_{0}^{\kappa}.$$
 (32)

Finally, the family of matrix-valued functions

$$\{\Psi_1(3^j-k), \Psi_n(\cdot-k): j=\cdots, -2, -1, 0; n=3, 4, \cdots, k \in \mathbb{Z}\}$$

is an orthogonal basis of  $L^2(\mathbb{R}, \mathbb{C}^{s \times s})$ .

## Acknowledgments

The authors would like to thank the anonymous referees and the editors for their useful suggestions and comments which make this paper more readable.

### References

- Toliyat H A, Abbaszadeh K, Rahimian M M, Olson L.E. Rail defect diagnosis using wavelet packet decomposition. IEEE T. Ind. Appl., 2003, 39(3): 1454 - 1461.
- Martin M. B, Bell A E. New image compression technique using multiwavelet packets. IEEE T. Image Process., 2001, 10(4): 500-511.
- [3] Deng Hai, Ling Hao. Fast solution of electromagnetic integral equations using adaptive wavelet packet transform. IEEE T. Antenn. Propag., 1999, 47(4): 674-682.
- [4] Coifman R R, Meyer Y, Wickerhauser M V. Wavelet analysis and signal processing. In Wavelets and Their Applications, Beylkin G. (Ed.), Boston: Jones and Barlett, MA, 1992, pp. 153-178.
- [5] Cohen A, Daubeches I. On the instability of arbitrary biorthogonal wavelet packets. SIAM Math. Anal., 1993, 24(5): 1340-1354.
- [6] Yang S, Cheng Z. A-scale multiple orthogonal wavelet packets. Math. Appl. China, 2000, 13(1): 61-65.
- [7] Efromovich S., Lakey J., Pereyia M C, Tymes N Jr. Data-diven and optimal denoising of a signal and recovery of its derivation using multiwavelets. IEEE T. Signal Process., 2004, 52(3): 628-635.

- [8] Xia X G, Suter B W. Vector-valued wavelets and vector filter banks. IEEE T. Signal Process., 1996, 44(3): 508-518.
- [9] Xia X G, Geronimo J S, Hardin D P, Suter B W. Design of prefilters for discrete multiwavelet transforms. IEEE T. Signal Process., 1996, 44(1): 25-35.
- [10] Daubechies I. Ten Lectures on Wavelets. Academic, New York, 1992.
- [11] Xia X G, Suter B W. FIR paraunitary filter banks given several analysis filters: Factorizations and constructions. IEEE T. Signal Process., 1996, 44(3): 720-723.