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Research article

The generalized inverse eigenvalue problem of Hamiltonian matrices and its approximation

Lina Liu, Huiting Zhang and Yinlan Chen*

School of Mathematics and Statistics, Hubei Normal University, Huangshi, 435002, China

* Correspondence: Email: cylfzg@hbnu.edu.cn; Tel: +8618986595093.

Abstract: Let $J = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix} \in \mathbb{R}^{2n \times 2n}$. A matrix $A \in \mathbb{R}^{2n \times 2n}$ is said to be Hamiltonian if $(AJ)^{\top} = AJ$. In this paper, we first consider the following generalized inverse eigenvalue problem (GIEP): Given a pair of matrices (Λ, X) in the form $\Lambda = \text{diag}\{\lambda_1, \dots, \lambda_p\} \in \mathbb{C}^{p \times p}$ and $X = [\mathbf{x}_1, \dots, \mathbf{x}_p] \in \mathbb{C}^{2n \times p}$, where diagonal elements of Λ are all distinct with rank(X) = p, and both Λ and X are closed under complex conjugation in the sense that $\lambda_{2i} = \overline{\lambda}_{2i-1} \in \mathbb{C}$, $\mathbf{x}_{2i} = \overline{\mathbf{x}}_{2i-1} \in \mathbb{C}^{2n}$ for $i = 1, \dots, l$, and $\lambda_j \in \mathbb{R}$, $\mathbf{x}_j \in \mathbb{R}^{2n}$ for $j = 2l + 1, \dots, p$. Find Hamiltonian matrices A and B such that $AX\Lambda = BX$. Then, we consider the associated optimal approximation problem (OAP): Given $\tilde{A}, \tilde{B} \in \mathbb{R}^{2n \times 2n}$. Find $(\hat{A}, \hat{B}) \in \mathbb{S}_{\mathbb{B}}$ such that $\|\hat{A} - \tilde{A}\|^2 + \|\hat{B} - \tilde{B}\|^2 = \min_{(A,B) \in \mathbb{S}_{\mathbb{B}}} (\|A - \tilde{A}\|^2 + \|B - \tilde{B}\|^2)$, where $\mathbb{S}_{\mathbb{B}}$ is the solution set of Problem GIEP. Also, we obtain the unique optimal approximation solution (\hat{A}, \hat{B}) of Problem OAP.

Keywords: generalized inverse eigenvalue problem; Hamiltonian matrix; QR-decomposition; optimal approximation

Mathematics Subject Classification: 15A24, 65F18

1. Introduction

Throughout this paper, $\mathbb{C}^{m \times n}$, $\mathbb{R}^{m \times n}$, $\mathbb{OR}^{n \times n}$ and $\mathbb{SR}^{n \times n}$ stand for the sets of all $m \times n$ complex matrices, all $m \times n$ real matrices, all $n \times n$ orthogonal matrices and all $n \times n$ real-valued symmetric matrices, respectively. The symbol A^{\top} and tr(A) stand for the transpose and the trace of a matrix A, respectively. I_n represents the identity matrix of size n, and $\mathbb{HR}^{2n \times 2n}$ represents the set of all $2n \times 2n$ Hamiltonian matrices, that is, $\mathbb{HR}^{2n \times 2n} = \{A \mid (AJ)^{\top} = AJ, A \in \mathbb{R}^{2n \times 2n}\}$, where $J = \begin{bmatrix} 0 & I_n \\ -I_n & 0 \end{bmatrix} \in \mathbb{R}^{2n \times 2n}$.

Hamiltonian matrices are widely applied in Hamiltonian systems of differential equations [1, 2], optimal quadratic linear control [3] and H_{∞} optimization [4], etc. For example, Hamiltonian matrix

elements from a symmetric wave function are necessary to study the structure of deuterated molecules [5]. Also, the eigenvalue problems for Hamiltonian and skew-Hamiltonian matrices appear frequently in scientific and engineering applications. Such as to compute the conformal parameterization via a constrained energy minimization problem in the field of digital geometry processing [6, 7], quantum mechanical problems with time reversal symmetry [8, 9], the study of closed shell Hartree-Fock wave functions in response theory [10, 11] and total least squares problems with symmetric constraints [12].

Inverse eigenvalue problems emerge from many application areas, and have been studied by many Generalized inverse eigenvalue problems are concerned in structural scholars [13–18]. dynamics [19–21], parameter identification [22] and pole assignment [23], etc. Recently, Zhao and Zhang [24] derived the solvability conditions for the inverse eigenvalue problem of normal skew J-Hamiltonian matrices by the Moore-Penrose generalized inverse and the generalized singular value Zhang and Yuan [25] solved the generalized inverse eigenvalue problems of decomposition. J-Hamiltonian/skew-Hamiltonian matrices by applying the singular value Hermitian and decomposition and the spectral decomposition. However, the problem of OAP cannot be considered due to the complexity of the expression of the general solution. Very recently, Yuan and Chen [26] solved the inverse eigenvalue problem and the optimal approximation problem for Hamiltonian matrices by using the generalized singular value decomposition. Nevertheless, the generalized inverse eigenvalue problem of Hamiltonian matrices seems rarely to be discussed in the literatures, which motivates us to study such kind of inverse problem and the associated approximation problem. That is, in this paper, we will consider the following generalized inverse eigenvalue problem and the associated optimal approximation problem, which is a generalization of the problems discussed in [26].

Problem GIEP. Given a pair of matrices (Λ, X) in the form $\Lambda = \text{diag}\{\lambda_1, \dots, \lambda_p\} \in \mathbb{C}^{p \times p}$, and $X = [\mathbf{x}_1, \dots, \mathbf{x}_p] \in \mathbb{C}^{2n \times p}$, where diagonal elements of Λ are all distinct, X is of full column rank p, and both Λ and X are closed under complex conjugation in the sense that $\lambda_{2i} = \overline{\lambda}_{2i-1} \in \mathbb{C}$, $\mathbf{x}_{2i} = \overline{\mathbf{x}}_{2i-1} \in \mathbb{C}^{2n}$ for $i = 1, \dots, l$, and $\lambda_j \in \mathbb{R}$, $\mathbf{x}_j \in \mathbb{R}^{2n}$ for $j = 2l + 1, \dots, p$. Find $A, B \in \mathbb{HR}^{2n \times 2n}$ such that

$$AX\Lambda = BX. \tag{1.1}$$

Problem OAP. Given $\tilde{A}, \tilde{B} \in \mathbb{R}^{2n \times 2n}$. Find $(\hat{A}, \hat{B}) \in \mathbb{S}_{\mathbb{R}}$ such that

$$\|\hat{A} - \tilde{A}\|^{2} + \|\hat{B} - \tilde{B}\|^{2} = \min_{(A,B)\in\mathbb{S}_{\mathbb{B}}} \left(\|A - \tilde{A}\|^{2} + \|B - \tilde{B}\|^{2} \right),$$
(1.2)

where $\|\cdot\|$ is the Frobenius norm and $\mathbb{S}_{\mathbb{E}}$ is the solution set of Problem GIEP.

By using the QR-decomposition, the representation of the general solution of Problem GIEP is deduced and the unique optimal approximation solution (\hat{A}, \hat{B}) of Problem OAP is obtained. Finally, two numerical examples are presented to illustrate the efficiency of the results.

2. The solution of Problem GIEP

Define T_p as

$$T_p = \operatorname{diag} \left\{ \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -i \\ 1 & i \end{bmatrix}, \cdots, \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -i \\ 1 & i \end{bmatrix}, I_{p-2l} \right\} \in \mathbb{C}^{p \times p},$$

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where $i = \sqrt{-1}$. It is easy to verify that T_p is a unitary matrix, that is, $\overline{T}_p^{\top}T_p = I_p$. With this matrix, we have

$$\tilde{\Lambda} = \bar{T}_{p}^{\top} \Lambda T_{p} = \operatorname{diag} \left\{ \begin{bmatrix} \alpha_{1} & \beta_{1} \\ -\beta_{1} & \alpha_{1} \end{bmatrix}, \cdots, \begin{bmatrix} \alpha_{2l-1} & \beta_{2l-1} \\ -\beta_{2l-1} & \alpha_{2l-1} \end{bmatrix}, \lambda_{2l+1}, \cdots, \lambda_{p} \right\}$$

$$\triangleq \operatorname{diag} \{\tilde{\Lambda}_{1}, \cdots, \tilde{\Lambda}_{2l-1}, \lambda_{2l+1}, \cdots, \lambda_{p}\} \in \mathbb{R}^{p \times p}, \qquad (2.1)$$

$$\tilde{X} = XT_p = [\sqrt{2}\mathbf{y}_1, \sqrt{2}\mathbf{z}_1, \cdots, \sqrt{2}\mathbf{y}_{2l-1}, \sqrt{2}\mathbf{z}_{2l-1}, \mathbf{x}_{2l+1}, \cdots, \mathbf{x}_p] \in \mathbb{R}^{2n \times p},$$
(2.2)

where α_i and β_i are the real part and imaginary part of the complex number λ_i , and \mathbf{y}_i and \mathbf{z}_i are, respectively, the real part and imaginary part of the complex vector \mathbf{x}_i for $i = 1, 3, \dots, 2l - 1$. Then, Eq (1.1) can be equivalently written as

$$4\tilde{X}\tilde{\Lambda} = B\tilde{X},\tag{2.3}$$

clearly, Eq (2.3) is equivalent to

$$AJJ^{\mathsf{T}}\tilde{X}\tilde{\Lambda} = BJJ^{\mathsf{T}}\tilde{X}.$$
(2.4)

Since rank(\tilde{X}) = rank(\tilde{X}) = p, the QR-decomposition of $J^{\top}\tilde{X}$ is of the form

$$J^{\top}\tilde{X} = Q \begin{bmatrix} R \\ 0 \end{bmatrix} = [Q_1, Q_2] \begin{bmatrix} R \\ 0 \end{bmatrix}, \qquad (2.5)$$

where $Q = [Q_1, Q_2] \in \mathbb{OR}^{2n \times 2n}$ with $Q_1 \in \mathbb{R}^{2n \times p}$, and $R \in \mathbb{R}^{p \times p}$ is nonsingular. Partition the parameter matrices $Q^{\mathsf{T}}AJQ$ and $Q^{\mathsf{T}}BJQ$ into blocks:

$$Q^{\mathsf{T}}AJQ = \begin{bmatrix} A_{11} & A_{12} \\ A_{12}^{\mathsf{T}} & A_{22} \end{bmatrix} \qquad \begin{array}{c} p \\ 2n-p \\ p & 2n-p \end{array}, \qquad Q^{\mathsf{T}}BJQ = \begin{bmatrix} B_{11} & B_{12} \\ B_{12}^{\mathsf{T}} & B_{22} \\ p & 2n-p \end{bmatrix} \qquad \begin{array}{c} p \\ 2n-p \\ p & 2n-p \end{array}, \tag{2.6}$$

where A_{11} , A_{22} , B_{11} and B_{22} are real-valued symmetric matrices. By (2.5) and (2.6), Eq (2.4) is equivalent to

Then, it follows from Eq (2.7) that

$$A_{11}R\tilde{\Lambda} = B_{11}R,\tag{2.8}$$

$$A_{12}^{\top}R\tilde{\Lambda} = B_{12}^{\top}R.$$
(2.9)

By Eq (2.8), B_{11} is a symmetric matrix implies that

$$R^{\mathsf{T}}A_{11}R\tilde{\Lambda} = \tilde{\Lambda}^{\mathsf{T}}R^{\mathsf{T}}A_{11}R.$$
(2.10)

Write

$$C = R^{\mathsf{T}} A_{11} R, \tag{2.11}$$

then Eq (2.10) can be written as

$$C\tilde{\Lambda} = \tilde{\Lambda}^{\mathsf{T}}C, \text{ s. t. } C = C^{\mathsf{T}}.$$
 (2.12)

By direct calculation, we have

$$C = \operatorname{diag}\left\{ \begin{bmatrix} a_1 & b_1 \\ b_1 & -a_1 \end{bmatrix}, \cdots, \begin{bmatrix} a_{2l-1} & b_{2l-1} \\ b_{2l-1} & -a_{2l-1} \end{bmatrix}, c_{2l+1}, \cdots, c_p \right\} \in \mathbb{R}^{p \times p},$$
(2.13)

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where a_{2i-1} , b_{2i-1} , $i = 1, \dots, l$, and c_j , $j = 2l + 1, \dots, p$, are arbitrary real numbers. Thus

$$A_{11} = R^{-\top} C R^{-1}. \tag{2.14}$$

Combining (2.8) with (2.11), we find that

$$B_{11} = R^{-\top} C \tilde{\Lambda} R^{-1}.$$
 (2.15)

By Eq (2.9), we have

$$B_{12} = R^{-\top} \tilde{\Lambda}^{\top} R^{\top} A_{12}, \qquad (2.16)$$

where $A_{12} \in \mathbb{R}^{p \times (2n-p)}$ is an arbitrary matrix.

Summing up above discussion, we can obtain the following result.

Theorem 2.1. Suppose that $\Lambda = diag\{\lambda_1, \dots, \lambda_p\} \in \mathbb{C}^{p \times p}, X = [\mathbf{x}_1, \dots, \mathbf{x}_p] \in \mathbb{C}^{2n \times p}$, where diagonal elements of Λ are all distinct, X is of full column rank p, and both Λ and X are closed under complex conjugation. Let the real matrices $\tilde{\Lambda}$ and \tilde{X} be given by (2.1) and (2.2) and the QR-decomposition of $J^{\mathsf{T}}\tilde{X}$ be given by (2.5). Then the general solution of Problem GIEP can be expressed as

$$\mathbb{S}_{\mathbb{E}} = \left\{ (A, B) \middle| \begin{array}{c} A = Q \begin{bmatrix} R^{-\top} C R^{-1} & A_{12} \\ A_{12}^{\top} & A_{22} \end{bmatrix} Q^{\top} J^{\top}, \\ B = Q \begin{bmatrix} R^{-\top} C \tilde{\Lambda} R^{-1} & R^{-\top} \tilde{\Lambda}^{\top} R^{\top} A_{12} \\ A_{12}^{\top} R \tilde{\Lambda} R^{-1} & B_{22} \end{bmatrix} Q^{\top} J^{\top} \end{array} \right\},$$
(2.17)

where $A_{12} \in \mathbb{R}^{p \times (2n-p)}$, $A_{22} \in \mathbb{SR}^{(2n-p) \times (2n-p)}$ and $B_{22} \in \mathbb{SR}^{(2n-p) \times (2n-p)}$ are arbitrary matrices, and C is given by (2.13).

3. The solution of Problem OAP

According to (2.17), we know that the solution set $\mathbb{S}_{\mathbb{E}}$ is always nonempty and $\mathbb{S}_{\mathbb{E}}$ is a closed convex subset, which implies that Problem OAP has a unique solution $(\hat{A}, \hat{B}) \in \mathbb{S}_{\mathbb{E}}$ by the optimal approximation theorem (see Ref. [27]). For the given matrices $\tilde{A}, \tilde{B} \in \mathbb{R}^{2n \times 2n}$, write

$$Q^{\mathsf{T}}\tilde{A}JQ = \begin{bmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ \tilde{A}_{21} & \tilde{A}_{22} \end{bmatrix} \qquad \begin{array}{c} p \\ 2n-p \end{array}, \qquad Q^{\mathsf{T}}\tilde{B}JQ = \begin{bmatrix} \tilde{B}_{11} & \tilde{B}_{12} \\ \tilde{B}_{21} & \tilde{B}_{22} \end{bmatrix} \qquad \begin{array}{c} p \\ 2n-p \end{array}, \qquad (3.1)$$
$$p \quad 2n-p \qquad \qquad p \quad 2n-p \qquad \qquad \end{array}$$

then

$$\begin{split} & \|A - \tilde{A}\|^2 + \|B - \tilde{B}\|^2 \\ & = \left\| Q \begin{bmatrix} R^{-\top} C R^{-1} & A_{12} \\ A_{12}^{\top} & A_{22} \end{bmatrix} Q^{\top} J^{\top} - \tilde{A} \right\|^2 + \left\| Q \begin{bmatrix} R^{-\top} C \tilde{\Lambda} R^{-1} & R^{-\top} \tilde{\Lambda}^{\top} R^{\top} A_{12} \\ A_{12}^{\top} R \tilde{\Lambda} R^{-1} & B_{22} \end{bmatrix} Q^{\top} J^{\top} - \tilde{B} \right\|^2 \\ & = \|R^{-\top} C R^{-1} - \tilde{A}_{11}\|^2 + \|A_{12} - \tilde{A}_{12}\|^2 + \|A_{12}^{\top} - \tilde{A}_{21}\|^2 + \|A_{22} - \tilde{A}_{22}\|^2 \\ & + \|R^{-\top} C \tilde{\Lambda} R^{-1} - \tilde{B}_{11}\|^2 + \|R^{-\top} \tilde{\Lambda}^{\top} R^{\top} A_{12} - \tilde{B}_{12}\|^2 + \|A_{12}^{\top} R \tilde{\Lambda} R^{-1} - \tilde{B}_{21}\|^2 + \|B_{22} - \tilde{B}_{22}\|^2. \end{split}$$

Therefore, $||A - \tilde{A}||^2 + ||B - \tilde{B}||^2 = \min$ if and only if

$$f(C) = \|R^{-\top}CR^{-1} - \tilde{A}_{11}\|^2 + \|R^{-\top}C\tilde{\Lambda}R^{-1} - \tilde{B}_{11}\|^2 = \min,$$
(3.2)

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$$\|A_{12} - \tilde{A}_{12}\|^2 + \|R^{-\top}\tilde{\Lambda}^{\top}R^{\top}A_{12} - \tilde{B}_{12}\|^2 + \|A_{12}^{\top} - \tilde{A}_{21}\|^2 + \|A_{12}^{\top}R\tilde{\Lambda}R^{-1} - \tilde{B}_{21}\|^2 = \min, \quad (3.3)$$

$$\|A_{22} - \tilde{A}_{22}\|^2 = \min, \tag{3.4}$$

$$\|B_{22} - \tilde{B}_{22}\|^2 = \min.$$
(3.5)

Let

$$R^{-1} = \begin{bmatrix} R_1 \\ R_2 \end{bmatrix}, \tag{3.6}$$

where

$$R_{1} = \begin{bmatrix} R_{1,1} \\ \vdots \\ R_{1,2l-1} \end{bmatrix}, R_{2} = \begin{bmatrix} R_{2,2l+1} \\ \vdots \\ R_{2,p} \end{bmatrix},$$

and $R_{1,2i-1} \in \mathbb{R}^{2 \times p}$, $R_{2,j} \in \mathbb{R}^{1 \times p}$ $(i = 1, \dots, l, j = 2l + 1, \dots, p)$. Furthermore, let

$$\begin{cases} D_{2i-1} = R_{1,2i-1}^{\top} F_1 R_{1,2i-1}, \\ D_{2i} = R_{1,2i-1}^{\top} F_2 R_{1,2i-1}, \\ D_j = R_{2,j}^{\top} R_{2,j}, \\ E_{2i-1} = R_{1,2i-1}^{\top} F_1 \tilde{\Lambda}_{2i-1} R_{1,2i-1}, \\ E_{2i} = R_{1,2i-1}^{\top} F_2 \tilde{\Lambda}_{2i-1} R_{1,2i-1}, \\ E_j = R_{2,j}^{\top} \lambda_j R_{2,j}, \\ i = 1, \cdots, l, j = 2l + 1, \cdots, p, \end{cases}$$

$$(3.7)$$

where

$$F_1 = \left[\begin{array}{cc} 1 & 0 \\ 0 & -1 \end{array} \right], \ F_2 = \left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right].$$

Then the relation of (3.2) is equivalent to

$$\begin{aligned} f(C) &= f(a_1, b_1, \cdots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \cdots, c_p) \\ &= \|a_1 D_1 + b_1 D_2 + \cdots + a_{2l-1} D_{2l-1} + b_{2l-1} D_{2l} + c_{2l+1} D_{2l+1} + \cdots + c_p D_p - \tilde{A}_{11}\|^2 \\ &+ \|a_1 E_1 + b_1 E_2 + \cdots + a_{2l-1} E_{2l-1} + b_{2l-1} E_{2l} + c_{2l+1} E_{2l+1} + \cdots + c_p E_p - \tilde{B}_{11}\|^2 = \min, \end{aligned}$$

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that is,

$$\begin{array}{l} f(a_1,b_1,\cdots,a_{2l-1},b_{2l-1},c_{2l+1},\cdots,c_p) \\ = & \operatorname{tr}[(a_1D_1^\top+b_1D_2^\top+\cdots+a_{2l-1}D_{2l-1}^\top+b_{2l-1}D_{2l}^\top+c_{2l+1}D_{2l+1}^\top+\cdots+c_pD_p^\top-\tilde{A}_{11}^\top) \\ & (a_1D_1+b_1D_2+\cdots+a_{2l-1}D_{2l-1}+b_{2l-1}D_{2l}+c_{2l+1}D_{2l+1}+\cdots+c_pD_p^\top-\tilde{A}_{11}^\top) \\ & + & (a_1E_1^\top+b_1E_2^\top+\cdots+a_{2l-1}E_{2l-1}^\top+b_{2l-1}E_{2l}^\top+c_{2l+1}E_{2l+1}^\top+\cdots+c_pE_p^\top-\tilde{B}_{11}^\top) \\ & (a_1E_1+b_1E_2+\cdots+a_{2l-1}E_{2l-1}+b_{2l-1}E_{2l}+c_{2l+1}E_{2l+1}^\top+\cdots+c_pE_p^\top-\tilde{B}_{11})] \\ = & a_1^2g_{1,1}+2a_1b_1g_{1,2}+\cdots+2a_1a_{2l-1}g_{1,2l-1}+2a_1b_{2l-1}g_{1,2l}+2a_1c_{2l+1}g_{1,2l+1}+\cdots \\ & + & 2a_1c_pg_{1,p}-2a_1h_1 \\ & + & b_1^2g_{2,2}+\cdots+2a_{2l-1}b_1g_{2,2l-1}+2b_1b_{2l-1}g_{2,2l}+2b_1c_{2l+1}g_{2,2l+1}+\cdots+2b_1c_pg_{2,p} \\ & - & 2b_1h_2 \\ & + & \cdots,\cdots \\ & + & a_{2l-1}^2g_{2l-1,2l-1}+2a_{2l-1}b_{2l-1}g_{2l-1,2l}+2a_{2l-1}c_{2l+1}g_{2l-1,2l+1}+\cdots+2a_{2l-1}c_pg_{2l-1,p} \\ & - & 2a_{2l-1}h_{2l-1} \\ & + & b_{2l-1}^2g_{2,2l}+2b_{2l-1}c_{2l+1}g_{2l,2l+1}+\cdots+2b_{2l-1}c_pg_{2l,p}-2b_{2l-1}h_{2l} \\ & + & c_{2l+1}^2g_{2l+1,2l+1}+\cdots+2c_{2l+1}c_pg_{2l+1,p}-2c_{2l+1}h_{2l+1} \\ & + & \cdots,\cdots \\ & + & c_p^2g_{p,p}-2c_ph_p+e, \end{array}$$

where $g_{m,n} = \text{tr}(D_m^{\top}D_n) + \text{tr}(E_m^{\top}E_n), \ h_m = \text{tr}(D_m^{\top}\tilde{A}_{11}) + \text{tr}(E_m^{\top}\tilde{B}_{11}), \ e = \text{tr}(\tilde{A}_{11}^{\top}\tilde{A}_{11}) + \text{tr}(\tilde{B}_{11}^{\top}\tilde{B}_{11}), \ m, n = 1, \cdots, p.$ Consequently,

$$\begin{array}{l} \frac{\partial f(a_{1},b_{1},\cdots,a_{2l-1},b_{2l-1},c_{2l+1},\cdots,c_{p})}{\partial a_{1}} &= 2a_{1}g_{1,1} + 2b_{1}g_{1,2} + \cdots + 2a_{2l-1}g_{1,2l-1} + 2b_{2l-1}g_{1,2l} + 2c_{2l+1}g_{1,2l+1} \\ &+ \cdots + 2c_{p}g_{1,p} - 2h_{1}, \\ \frac{\partial f(a_{1},b_{1},\cdots,a_{2l-1},b_{2l-1},c_{2l+1},\cdots,c_{p})}{\partial b_{1}} &= 2a_{1}g_{2,1} + 2b_{1}g_{2,2} + \cdots + 2a_{2l-1}g_{2,2l-1} + 2b_{2l-1}g_{2,2l} + 2c_{2l+1}g_{2,2l+1} \\ &+ \cdots + 2c_{p}g_{2,p} - 2h_{2}, \\ \end{array}$$

Clearly, $f(a_1, b_1, \dots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \dots, c_p) = \min$ if and only if

$$\frac{\partial f(a_1, b_1, \cdots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \cdots, c_p)}{\partial a_1} = 0, \cdots, \frac{\partial f(a_1, b_1, \cdots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \cdots, c_p)}{\partial c_p} = 0.$$

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$$a_{1}g_{1,1} + b_{1}g_{1,2} + \dots + a_{2l-1}g_{1,2l-1} + b_{2l-1}g_{1,2l} + c_{2l+1}g_{1,2l+1} + \dots + c_{p}g_{1,p} = h_{1},$$

$$a_{1}g_{2,1} + b_{1}g_{2,2} + \dots + a_{2l-1}g_{2,2l-1} + b_{2l-1}g_{2,2l} + c_{2l+1}g_{2,2l+1} + \dots + c_{p}g_{2,p} = h_{2},$$

$$\dots, \dots,$$

$$a_{1}g_{2l-1,1} + b_{1}g_{2l-1,2} + \dots + a_{2l-1}g_{2l-1,2l-1} + b_{2l-1}g_{2l-1,2l} + c_{2l+1}g_{2l-1,2l+1} + \dots + c_{p}g_{2l-1,p} = h_{2l-1},$$

$$a_{1}g_{2l,1} + b_{1}g_{2l,2} + \dots + a_{2l-1}g_{2l,2l-1} + b_{2l-1}g_{2l,2l} + c_{2l+1}g_{2l,2l+1} + \dots + c_{p}g_{2l,p} = h_{2l},$$

$$a_{1}g_{2l+1,1} + b_{1}g_{2l+1,2} + \dots + a_{2l-1}g_{2l+1,2l-1} + b_{2l-1}g_{2l+1,2l} + c_{2l+1}g_{2l+1,2l+1} + \dots + c_{p}g_{2l+1,p} = h_{2l+1},$$

$$\dots, \dots,$$
(3.8)

 $a_1g_{p,1} + b_1g_{p,2} + \dots + a_{2l-1}g_{p,2l-1} + b_{2l-1}g_{p,2l} + c_{2l+1}g_{p,2l+1} + \dots + c_pg_{p,p} = h_p.$

If let

$$G = \begin{bmatrix} g_{1,1} & g_{1,2} & \cdots & g_{1,2l-1} & g_{1,2l} & g_{1,2l+1} & \cdots & g_{1,p} \\ g_{2,1} & g_{2,2} & \cdots & g_{2,2l-1} & g_{2,2l} & g_{2,2l+1} & \cdots & g_{2,p} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ g_{2l-1,1} & g_{2l-1,2} & \cdots & g_{2l-1,2l-1} & g_{2l-1,2l} & g_{2l-1,2l+1} & \cdots & g_{2l-1,p} \\ g_{2l,1} & g_{2l,2} & \cdots & g_{2l,2l-1} & g_{2l,2l} & g_{2l,2l+1} & \cdots & g_{2l,p} \\ g_{2l+1,1} & g_{2l+1,2} & \cdots & g_{2l+1,2l-1} & g_{2l+1,2l} & g_{2l+1,2l+1} & \cdots & g_{2l+1,p} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ g_{p,1} & g_{p,2} & \cdots & g_{p,2l-1} & g_{p,2l} & g_{p,2l+1} & \cdots & g_{p,p} \end{bmatrix} \\ T = \begin{bmatrix} a_1 \\ b_1 \\ \vdots \\ a_{2l-1} \\ b_{2l-1} \\ c_{2l+1} \\ \vdots \\ c_p \end{bmatrix}, \quad H = \begin{bmatrix} h_1 \\ h_2 \\ \vdots \\ h_{2l-1} \\ h_{2l} \\ h_{p} \end{bmatrix},$$

where G is symmetric matrix. Then Eq (3.8) is equivalent to

$$GT = H, \tag{3.9}$$

and the solution of Eq (3.9) is

$$T = G^{-1}H.$$
 (3.10)

Substituting (3.10) into (2.13), we can obtain C explicitly. Similarly, Eq (3.3) is equivalent to

$$f(A_{12}) = tr[(A_{12}^{\top} - \tilde{A}_{12}^{\top})(A_{12} - \tilde{A}_{12})] + tr[(A_{12}^{\top} P - \tilde{B}_{12}^{\top})(P^{\top}A_{12} - \tilde{B}_{12})] + tr[(A_{12} - \tilde{A}_{21}^{\top})(A_{12}^{\top} - \tilde{A}_{21})] + tr[(P^{\top}A_{12} - \tilde{B}_{21}^{\top})(A_{12}^{\top} P - \tilde{B}_{21})].$$

Thus,

$$\frac{\partial f(A_{12})}{\partial A_{12}} = 2A_{12} - 2\tilde{A}_{12} + 2PP^{\mathsf{T}}A_{12} - 2P\tilde{B}_{12} + 2A_{12} - 2\tilde{A}_{21}^{\mathsf{T}} + 2PP^{\mathsf{T}}A_{12} - 2P\tilde{B}_{21}^{\mathsf{T}},$$

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setting $\frac{\partial f(A_{12})}{\partial A_{12}} = 0$, we obtain

$$A_{12} = \frac{1}{2} (I_p + PP^{\mathsf{T}})^{-1} (\tilde{A}_{12} + P\tilde{B}_{12} + \tilde{A}_{21}^{\mathsf{T}} + P\tilde{B}_{21}^{\mathsf{T}}), \qquad (3.11)$$

where $P = R\tilde{\Lambda}R^{-1}$. A_{22} , B_{22} are symmetric matrices implies that the relations of (3.4) and (3.5) are equivalent to

$$\|A_{22} - \tilde{A}_{22}\|^{2} = \|A_{22} - \frac{1}{2}(\tilde{A}_{22} + \tilde{A}_{22}^{\mathsf{T}})\|^{2} + \|\frac{1}{2}(\tilde{A}_{22} - \tilde{A}_{22}^{\mathsf{T}})\|^{2},$$

$$\|B_{22} - \tilde{B}_{22}\|^{2} = \|B_{22} - \frac{1}{2}(\tilde{B}_{22} + \tilde{B}_{22}^{\mathsf{T}})\|^{2} + \|\frac{1}{2}(\tilde{B}_{22} - \tilde{B}_{22}^{\mathsf{T}})\|^{2},$$

therefore, we have

$$A_{22} = \frac{1}{2} (\tilde{A}_{22} + \tilde{A}_{22}^{\top}), \ B_{22} = \frac{1}{2} (\tilde{B}_{22} + \tilde{B}_{22}^{\top}).$$
(3.12)

Theorem 3.1. Given $\tilde{A}, \tilde{B} \in \mathbb{R}^{2n \times 2n}$, then the Problem OAP has a unique solution and the unique solution of Problem OAP is

$$\hat{A} = Q \begin{bmatrix} R^{-T} C R^{-1} & A_{12} \\ A_{12}^{\top} & A_{22} \end{bmatrix} Q^{\top} J^{\top}, \ \hat{B} = Q \begin{bmatrix} R^{-T} C \tilde{\Lambda} R^{-1} & R^{-T} \tilde{\Lambda}^{\top} R^{\top} A_{12} \\ A_{12}^{\top} R \tilde{\Lambda} R^{-1} & B_{22} \end{bmatrix} Q^{\top} J^{\top},$$
(3.13)

where A_{12} and A_{22} , B_{22} are given by (3.11) and (3.12), and $a_1, b_1, \dots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \dots, c_p$ are given by (3.10), respectively.

4. Numerical examples

According to Theorems 2.1 and 3.1, we have the following algorithm for solving Problem OAP.

Algorithm 4.1.

- 1). Input Λ , X, J, \tilde{A} , \tilde{B} .
- 2). Compute real-valued matrices $\tilde{\Lambda}$, \tilde{X} by (2.1) and (2.2), respectively.
- 3). Compute the QR-decomposition of the matrix $J^{\mathsf{T}}\tilde{X}$ by (2.5).
- 4). Compute \tilde{A}_{ij} , \tilde{B}_{ij} by (3.1), i, j = 1, 2.
- 5). Compute R^{-1} by (3.6) to form R_1, R_2 .
- 6). Compute $D_{2i-1}, D_{2i}, D_j, E_{2i-1}, E_{2i}$ and E_j $(i = 1, \dots, l, j = 2l + 1, \dots, p)$ by (3.7).
- 7). Compute $g_{m,n} = tr(D_m^{\top}D_n) + tr(E_m^{\top}E_n)$ and $h_m = tr(D_m^{\top}\tilde{A}_{11}) + tr(E_m^{\top}\tilde{B}_{11}) (m, n = 1, \dots, p)$.
- 8). Compute $a_1, b_1, \dots, a_{2l-1}, b_{2l-1}, c_{2l+1}, \dots, c_p$ by (3.10).
- 9). Compute A_{12} by (3.11).
- 10). Compute *A*₂₂ and *B*₂₂ by (3.12).
- 11). Compute \hat{A} and \hat{B} by (3.13).

Remark 4.1. After statistics, we find that the amount of computations required by Algorithm 1 is about $p^5 + p^4 + \frac{19}{3}p^3 + 14np^2 + 64n^3$ flops.

Example 4.1. Let n = 5, p = 5, and the matrices Λ , X, \tilde{A} and \tilde{B} be given by

 $\Lambda = \text{diag} \{-0.2218 + 2.0231i, -0.2218 - 2.0231i, -0.1617 + 0.5721i, -0.1617 - 0.5721i, 2.7670\},\$

$$X = \begin{bmatrix} -0.6377 - 0.1444i & -0.6377 + 0.1444i & -0.0405 + 0.3341i & -0.0405 - 0.3341i & -1.0000 \\ 0.2678 - 0.0983i & 0.2678 + 0.0983i & -0.2615 - 0.3973i & -0.2615 + 0.3973i & 0.1151 \\ 0.4260 + 0.5740i & 0.4260 - 0.5740i & 0.1348 - 0.2946i & 0.1348 + 0.2946i & 0.0278 \\ -0.2032 - 0.0489i & -0.2032 + 0.0489i & -0.5238 + 0.4762i & -0.5238 - 0.4762i & 0.0357 \\ 0.2111 - 0.1510i & 0.2111 + 0.1510i & 0.7982 - 0.1668i & 0.7982 + 0.1668i & 0.6793 \\ -0.5233 + 0.1151i & -0.5233 - 0.1151i & -0.3033 - 0.4132i & -0.3033 + 0.4132i & -0.1091 \\ 0.4820 - 0.2541i & 0.4820 + 0.2541i & 0.1398 - 0.0715i & 0.1398 + 0.0715i & 0.8550 \\ 0.3183 - 0.3431i & 0.3183 + 0.3431i & -0.2716 - 0.3411i & -0.2716 + 0.3411i & 0.6436 \\ 0.1376 + 0.3007i & 0.1376 - 0.3007i & -0.1391 + 0.0981i & -0.1391 - 0.0981i & 0.0883 \\ -0.0672 + 0.0189i & -0.0672 - 0.0189i & 0.3143 + 0.1462i & 0.3143 - 0.1462i & -0.2828 \end{bmatrix}$$

$$\tilde{A} = \begin{bmatrix} 1.8351 & 3.0635 & 9.3900 & 1.9476 & 9.7975 & 1.1742 & 7.3033 & 6.2406 & 2.6187 & 9.0372 \\ 3.6848 & 5.0851 & 8.7594 & 2.2592 & 4.3887 & 2.9668 & 4.8861 & 6.7914 & 3.3536 & 8.9092 \\ 6.2562 & 5.1077 & 5.5016 & 1.7071 & 1.1112 & 3.1878 & 5.7853 & 3.9552 & 6.7973 & 3.3416 \\ 7.8023 & 8.1763 & 6.2248 & 2.2766 & 2.5806 & 4.2417 & 2.3728 & 3.6744 & 1.3655 & 6.9875 \\ 0.8113 & 7.9483 & 5.8704 & 4.3570 & 4.0872 & 5.0786 & 4.5885 & 9.8798 & 7.2123 & 1.9781 \\ 9.2939 & 6.4432 & 2.0774 & 3.1110 & 5.9490 & 0.8552 & 9.6309 & 0.3774 & 1.0676 & 0.3054 \\ 7.7571 & 3.7861 & 3.0125 & 9.2338 & 2.6221 & 2.6248 & 5.4681 & 8.8517 & 6.5376 & 7.4407 \\ 4.8679 & 8.1158 & 4.7092 & 4.3021 & 6.0284 & 8.0101 & 5.2114 & 9.1329 & 4.9417 & 5.0002 \\ 4.3586 & 5.3283 & 2.3049 & 1.8482 & 7.1122 & 0.2922 & 2.3159 & 7.9618 & 7.7905 & 4.7992 \\ 4.4678 & 3.5073 & 8.4431 & 9.0488 & 2.2175 & 9.2885 & 4.8890 & 0.9871 & 7.1504 & 9.0472 \end{bmatrix}$$

$$\tilde{B} = \begin{bmatrix} 6.0987 & 1.6793 & 0.9673 & 4.5380 & 3.9926 & 1.0622 & 4.2284 & 6.6653 & 3.6892 & 1.2061 \\ 6.1767 & 9.7868 & 8.1815 & 4.3239 & 5.2688 & 3.7241 & 5.4787 & 1.7813 & 4.6073 & 5.8951 \\ 8.5944 & 7.1269 & 8.1755 & 8.2531 & 4.1680 & 1.9812 & 9.4274 & 1.2801 & 9.8164 & 2.2619 \\ 8.0549 & 5.0047 & 7.2244 & 0.8347 & 6.5686 & 4.8969 & 4.1774 & 9.9908 & 1.5640 & 3.8462 \\ 5.7672 & 4.7109 & 1.4987 & 1.3317 & 6.2797 & 3.3949 & 9.8305 & 1.7112 & 8.5552 & 5.8299 \\ 1.8292 & 0.5962 & 6.5961 & 1.7339 & 2.9198 & 9.5163 & 3.0145 & 0.3260 & 6.4476 & 2.5181 \\ 2.3993 & 6.8197 & 5.1859 & 3.9094 & 4.3165 & 9.2033 & 7.0110 & 5.6120 & 3.7627 & 2.9044 \\ 8.8651 & 0.4243 & 9.7297 & 8.3138 & 0.1549 & 0.5268 & 6.6634 & 8.8187 & 1.9092 & 6.1709 \\ 0.2867 & 0.7145 & 6.4899 & 8.0336 & 9.8406 & 7.3786 & 5.3913 & 6.6918 & 4.2825 & 2.6528 \\ 4.8990 & 5.2165 & 8.0033 & 0.6047 & 1.6717 & 2.6912 & 6.9811 & 1.9043 & 4.8202 & 8.2438 \end{bmatrix}$$

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By applying Algorithm 4.1, we can obtain the unique solution (\hat{A}, \hat{B}) of Problem OAP as follows:

[3.0180	-0.4225	-0.1555	0.8402	0.9125	4.0071	3.8840	3.2121	4.0178	6.8173]
	-0.3558	-1.7726	-0.3149	2.4300	-2.4421	3.8840	1.9734	5.0988	2.4334	6.2185	
	1.8766	-3.0189	-0.9503	-1.1732	-1.8790	3.2121	5.0988	2.6896	6.6750	5.4058	
	1.9510	-0.1294	-0.2763	-0.7668	-1.9008	4.0178	2.4334	6.6750	2.9195	5.2316	
$\hat{\lambda}$ –	3.0211	0.2273	-0.6075	-0.9217	-2.7818	6.8173	6.2185	5.4058	5.2316	3.9665	
A –	6.3731	6.9288	5.5322	4.4397	5.4356	-3.0180	0.3558	-1.8766	-1.9510	-3.0211	
	6.9288	4.5231	6.2946	7.2049	3.6851	0.4225	1.7726	3.0189	0.1294	-0.2273	
	5.5322	6.2946	4.8453	3.7706	4.4618	0.1555	0.3149	0.9503	0.2763	0.6075	
	4.4397	7.2049	3.7706	2.8249	6.6239	-0.8402	-2.4300	1.1732	0.7668	0.9217	
	5.4356	3.6851	4.4618	6.6239	2.8468	-0.9125	2.4421	1.8790	1.9008	2.7818	
	0.6709	-1.2381	-0.7557	-1.1218	-1.9835	-1.1000	3.7817	3.8263	0.6376	3.5484	
	1.3451	0.2456	-0.0257	-0.2862	0.2915	3.7817	5.3559	5.3224	3.6947	8.6240	
	3.0095	1.2171	-1.0933	0.9515	2.1292	3.8263	5.3224	1.2147	6.8013	2.4324	
	2.0994	-1.8754	3.1477	-0.1903	0.6793	0.6376	3.6947	6.8013	3.3691	7.2508	
ô	1.4047	0.3032	-1.6547	-2.5697	1.2202	3.5484	8.6240	2.4324	7.2508	6.0558	
D –	2.7999	2.6538	5.9368	4.0530	1.3808	-0.6709	-1.3451	-3.0095	-2.0994	-1.4047	,
	2.6538	5.3390	2.3523	2.6326	4.1208	1.2381	-0.2456	-1.2171	1.8754	-0.3032	
	5.9368	2.3523	8.4193	6.9480	4.5598	0.7557	0.0257	1.0933	-3.1477	1.6547	
	4.0530	2.6326	6.9480	8.0838	5.2373	1.1218	0.2862	-0.9515	0.1903	2.5697	
	1.3808	4.1208	4.5598	5.2373	3.4028	1.9835	-0.2915	-2.1292	-0.6793	-1.2202	

and

$$\|\hat{A}X\Lambda - \hat{B}X\| = 2.0806 \times 10^{-14},$$

which implies that $\hat{A}X\Lambda = \hat{B}X$ reproduces the desired eigenvalues and eigenvectors. **Example 4.2.** We consider an inverse problem for the spectral conformal parameterization (see Refs. [6,7]). Let n = 5, p = 4, and the matrices Λ , X, \tilde{B} and \tilde{L}_C be given by

 $\Lambda = \text{diag} \{-0.0822, -0.0250, 0, 0.0757\} \triangleq \text{diag} \{\lambda_1, \lambda_2, \lambda_3, \lambda_4\},\$

$$X = \begin{bmatrix} 0.5760 & 0.2684 & 0.0000 & 0.1340 \\ -0.1510 & -0.0673 & 0.0000 & -1.0000 \\ -0.3890 & -0.0610 & 0.0000 & -0.1246 \\ -0.0359 & -0.1401 & 0.0000 & 0.9907 \\ 0.3409 & 1.0000 & 1.0000 & 0.0366 \\ 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ -0.5672 & 0.0665 & 0.0000 & 0.3072 \\ 1.0000 & -0.1288 & 0.0000 & -0.1395 \\ -0.4328 & 0.0623 & 0.0000 & -0.1677 \\ 0.0650 & 0.2282 & 0.2000 & -0.1508 \end{bmatrix} \triangleq \operatorname{diag}\{f_1, f_2, f_3, f_4\},$$

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[0.7000	0.5483	0.5369	0.6919	0.6032	0	-0.8083	0.0306	-0.7715	0.0234	l
	0.5483	1.6617	1.3425	0.6037	1.4860	0.8083	0	0.4237	-0.5357	-0.9292	
$\tilde{B} =$	0.5369	1.3425	1.5075	0.9112	1.0372	-0.0306	-0.4237	0	0.4304	0.0488	
	0.6919	0.6037	0.9112	1.5583	0.9459	0.7715	0.5357	-0.4304	0	-0.0255	
	0.6032	1.4860	1.0372	0.9459	0.6742	-0.0234	0.9292	-0.0488	0.0255	0	
	0	0.8083	-0.0306	0.7715	-0.0234	0.7000	0.5483	0.5369	0.6919	0.6032	,
	-0.8083	0	-0.4237	0.5357	0.9292	0.5483	1.6617	1.3425	0.6037	1.4860	
	0.0306	0.4237	0	-0.4304	-0.0488	0.5369	1.3425	1.5075	0.9112	1.0372	
	-0.7715	-0.5357	0.4304	0	0.0255	0.6919	0.6037	0.9112	1.5583	0.9459	
	0.0234	-0.9292	0.0488	-0.0255	0	0.6032	1.4860	1.0372	0.9459	0.6742	
	F 1 = 000		2 52 60	4 60 4 0	0.0000	0		0.0(10	1 5 1 2 0	0.0460	Ъ
	1.7000	2.5483	3.5369	4.6919	-9.3968	0	-1.6166	0.0612	-1.5430	0.0468	
	2.5483	4.6617	5.3425	-9.3963	2.4860	1.6166	0	0.8474	-1.0714	-1.8584	
	3.5369	5.3425	-8.4925	1.9112	3.0372	-0.0612	-0.8474	0	0.8608	0.0976	
	4.6919	-9.3963	1.9112	3.5583	3.9459	1.5430	1.0714	-0.8608	0	-0.0510	
ĩ –	-9.3968	2.4860	3.0372	3.9459	4.6742	-0.0468	1.8584	-0.0976	0.0510	0	
$L_C =$	0	1.6166	-0.0612	1.5430	-0.0468	1.7000	2.5483	3.5369	4.6919	-9.3968	
	-1.6166	0	-0.8474	1.0714	1.8584	2.5483	4.6617	5.3425	-9.3963	2.4860	
	0.0612	0.8474	0	-0.8608	-0.0976	3.5369	5.3425	-8.4925	1.9112	3.0372	
	-1.5430	-1.0714	0.8608	0	0.0510	4.6919	-9.3963	1.9112	3.5583	3.9459	
	0.0468	-1.8584	0.0976	-0.0510	0	-9.3968	2.4860	3.0372	3.9459	4.6742	

By calculating, we can obtain the unique solution (\hat{B}, \hat{L}_C) of Problem OAP as follows:

[1.0866	-0.0847	-0.0520	0.0501	0.1093	0.0000	-0.0330	0.0833	-0.0503	0.0134	l
$\hat{B} =$	-0.1591	1.7380	-0.2522	-0.3467	-0.2414	0.0330	-0.0000	-0.0191	0.0087	0.0234	
	-0.1345	-0.2315	1.6597	-0.2656	-0.2222	-0.0833	0.0191	0.0000	0.0206	-0.0230	
	-0.0049	-0.3583	-0.2945	1.6496	-0.3472	0.0503	-0.0087	-0.0206	-0.0000	-0.1716	
	0.2724	-0.0519	-0.1132	-0.1159	0.8731	-0.0134	-0.0234	0.0230	0.1716	0.0000	
	0.0000	0.0864	-0.2014	0.1221	-0.0239	1.0866	-0.1591	-0.1345	-0.0049	0.2724	,
	-0.0864	-0.0000	0.0415	0.0244	0.1288	-0.0847	1.7380	-0.2315	-0.3583	-0.0519	
	0.2014	-0.0415	0.0000	-0.1586	-0.1553	-0.0520	-0.2522	1.6597	-0.2945	-0.1132	
	-0.1221	-0.0244	0.1586	-0.0000	-0.0000	0.0501	-0.3467	-0.2656	1.6496	-0.1159	
	0.0239	-0.1288	0.1553	0.0000	0.0000	0.1093	-0.2414	-0.2222	-0.3472	0.8731	
	3.0749	3.0182	2.9855	3.3985	-0.0221	-0.0000	-0.5788	0.0130	1.0166	0.1107]
	0.8672	0.7497	1.1429	0.2011	0.4437	0.5788	0.0000	-0.0656	-0.3528	-2.2185	
	0.7590	0.8360	0.5193	0.6574	0.1300	-0.0130	0.0656	0.0000	-0.0313	-0.6502	
	0.1065	0.3474	-0.4752	0.5081	-0.2263	-1.0166	0.3528	0.0313	0.0000	1.1314	
î _	-0.1139	0.6946	-0.6294	-0.1520	0.0000	-0.1107	2.2185	0.6502	-1.1314	0.0000	
L_C –	0.0000	-0.3435	0.7467	-0.4158	0.0228	3.0749	0.8672	0.7590	0.1065	-0.1139	ŀ
	0.3435	0.0000	0.9259	0.1743	-0.1389	3.0182	0.7497	0.8360	0.3474	0.6946	
	-0.7467	-0.9259	-0.0000	-1.3216	0.1259	2.9855	1.1429	0.5193	-0.4752	-0.6294	
	0.4158	-0.1743	1.3216	-0.0000	0.0304	3.3985	0.2011	0.6574	0.5081	-0.1520	
	-0.0228	0.1389	-0.1259	-0.0304	-0.0000	-0.0221	0.4437	0.1300	-0.2263	0.0000	

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Furthermore, we can obtain the following numerical results:

(λ_i, f_i)	(λ_1, f_1)	(λ_2, f_2)	(λ_3, f_3)	(λ_4, f_4)
$\ \lambda_i \hat{B} f_i - \hat{L}_C f_i\ $	1.1762×10^{-15}	1.1322×10^{-15}	1.0372×10^{-15}	9.7141×10^{-16} .

Table 4.1. Residuals of the eigenpairs (λ_i, f_i) .

Therefore, the new model $\hat{B}X\Lambda = \hat{L}_C X$ reproduces the desired eigenvalues and eigenvectors.

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Conflict of interest

The authors declare no conflict of interest.

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