

## THE SOLVABILITY CONDITIONS FOR INVERSE EIGENVALUE PROBLEM OF ANTI-BISYMMETRIC MATRICES<sup>\*1)</sup>

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### Abstract

This paper is mainly concerned with solving the following two problems:

**Problem I.** Given  $X \in C^{n \times m}$ ,  $\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_m) \in C^{m \times m}$ . Find  $A \in ABSR^{n \times n}$  such that

$$AX = X\Lambda$$

where  $ABSR^{n \times n}$  is the set of all real  $n \times n$  anti-bisymmetric matrices.

**Problem II.** Given  $A^* \in R^{n \times n}$ . Find  $\hat{A} \in S_E$  such that

$$\|A^* - \hat{A}\|_F = \min_{A \in S_E} \|A^* - A\|_F,$$

where  $\|\cdot\|_F$  is Frobenius norm, and  $S_E$  denotes the solution set of Problem I.

The necessary and sufficient conditions for the solvability of Problem I have been studied. The general form of  $S_E$  has been given. For Problem II the expression of the solution has been provided.

*Key words:* Eigenvalue problem, Norm, Approximate solution.

### 1. Introduction

Inverse eigenvalue problem has widely been used in engineering. For example inverse eigenvalue method is a useful means in vibration design and vibration control of flyer. In recent years a serial of good conclusions have been made for inverse eigenvalue problem. However, inverse problems of anti-bisymmetric matrices have not be concerned yet. In this paper we will discuss this problem.

We denote the complex  $n \times m$  matrix space by  $C^{n \times m}$ , the real  $n \times m$  matrix space by  $R^{n \times m}$ , and  $R^n = R^{n \times 1}$ , the set of all matrices in  $R^{n \times m}$  with rank  $r$  by  $R_r^{n \times m}$ , the set of all  $n \times n$  orthogonal matrices by  $OR^{n \times n}$ , the set of all  $n \times n$  anti-symmetric matrices by  $ASR^{n \times n}$ , the column space, the null space and the Moore–Penrose generalized inverse of a matrix  $A$  by  $R(A)$ ,  $N(A)$ ,  $A^+$  respectively, the identity matrix of order  $n$  by  $I_n$ , the Frobenius norm of  $A$  by  $\|A\|_F$ . We define inner product in space  $R^{n \times m}$ ,  $(A, B) = \text{tr}(B^T A) = \sum_{i=1}^n \sum_{j=1}^m a_{ij} b_{ij}$ ,  $\forall A, B \in R^{n \times m}$ . Then  $R^{n \times m}$  is a Hilbert inner product space. The norm of a matrix produced by the

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inner product is Frobenius norm. Let  $S_k = (e_k, e_{k-1}, \dots, e_1) \in R^{k \times k}$  in which  $e_i$  is the  $i$ -th Cloumn of the identity matrix  $I_k$ .

**Definition 1.**  $A = (a_{ij}) \in R^{n \times n}$ , if

$$a_{ij} = -a_{ji}, \quad a_{ij} = -a_{n-j+1, n-i+1}, \quad i, j = 1, 2, \dots, n$$

then  $A$  is called a anti-bisymmetric matrix. The set of all anti-bisymmetric matrices is denoted by  $ABSR^{n \times n}$ .

Now we consider the following problems:

**Problem I.** Given  $X \in C^{n \times m}$ ,  $\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_m)$ . Find  $A \in ABSR^{n \times n}$  such that

$$AX = X\Lambda.$$

**Problem II.** Given  $A^* \in R^{n \times n}$ . Find  $\hat{A} \in S_E$  such that

$$\|A^* - \hat{A}\|_F = \min_{A \in S_E} \|A^* - A\|_F,$$

where  $S_E$  is the solution set of problem I.

At first, in this paper, we will discuss the character of eigenvector for anti-bisymmetric matrices. Then we will give the necessary and sufficient conditions for the solvability of Problem I and the expression of the general solution of Problem I in real number field, and point out  $S_E$  is a closed convex set. At last, we will prove that there exists a unique solution of Problem II and give an expression of the solution for Problem II.

## 2. The Solvability Conditions and General Form of the Solutions for Problem I in Real Number Field

At first we discuss the construction of  $ABSR^{n \times n}$  and the character of eigenvector for matrices in  $ABSR^{n \times n}$ .

Let

$$k = \lfloor \frac{n}{2} \rfloor, \quad [x] \text{ is integer number that is not greater than } x. \tag{2.1}$$

When  $n = 2k$ , 
$$D = \frac{1}{\sqrt{2}} \begin{pmatrix} I_k & I_k \\ S_k & -S_k \end{pmatrix}, \quad D^T D = I_n; \tag{2.2}$$

and when  $n = 2k + 1$ , 
$$D = \frac{1}{\sqrt{2}} \begin{pmatrix} I_k & 0 & I_k \\ 0 & \sqrt{2} & 0 \\ S_k & 0 & -S_k \end{pmatrix}, \quad D^T D = I_n. \tag{2.3}$$

**Lemma 1.**  $A \in ABSR^{n \times n}$  if and only if

$$A = S_n A S_n, \quad A = -A^T$$

**Theorem 1.**

$$ABSR^{2k \times 2k} = \left\{ \begin{pmatrix} M & HS_k \\ S_k H & S_k M S_k \end{pmatrix} \mid M, H \in ASR^{k \times k} \right\}. \tag{2.4}$$

$$ABSR^{(2k+1) \times (2k+1)} = \left\{ \begin{pmatrix} N & C & HS_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} \mid N, H \in ASR^{k \times k}, C \in R^k \right\}. \tag{2.5}$$

whether  $n$  is odd or even number, the general form of elements in  $ABSR^{n \times n}$  is

$$D \begin{pmatrix} A_{11} & 0 \\ 0 & A_{22} \end{pmatrix} D^T \tag{2.6}$$

Where  $A_{11}, A_{22}$  are anti-symmetric matrices,  $D$  is the same as (2.2) or (2.3).

*Proof.* We only prove(2.4). If

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \in ABSR^{2k \times 2k}.$$

Then

$$A = -A^T, \quad \begin{pmatrix} 0 & S_k \\ S_k & 0 \end{pmatrix} \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} 0 & S_k \\ S_k & 0 \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}. \tag{2.7}$$

(2.7) is equivalent to

$$\begin{aligned} A_{11} &= -A_{11}^T, & A_{12} &= -A_{21}^T, & A_{22} &= -A_{22}^T, \\ A_{22} &= S_k A_{11} S_k & (A_{12} S_k)^T &= -A_{12} S_k \end{aligned} \tag{2.8}$$

Let  $A_{12} S_k = H, \quad A_{11} = M.$   
 Then  $H = -H^T, \quad M = -M^T, \quad A_{12} = H S_k, \quad A_{21} = S_k H, \quad A_{22} = S_k M S_k.$   
 It implies that

$$A = \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix}.$$

Then  $ABSR^{2k \times 2k} \subseteq \left\{ \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix} \mid M, H \in ASR^{k \times k} \right\}.$

Conversely, for every  $M, H \in ASR^{k \times k}$  it is easy to see

$$\begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix} = - \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix}^T$$

and

$$\begin{aligned} &\begin{pmatrix} 0 & S_k \\ S_k & 0 \end{pmatrix} \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix} \begin{pmatrix} 0 & S_k \\ S_k & 0 \end{pmatrix} \\ &= \begin{pmatrix} H & M S_k \\ S_k M & S_k H S_k \end{pmatrix} \begin{pmatrix} 0 & S_k \\ S_k & 0 \end{pmatrix} = \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix}. \end{aligned}$$

From Lemma 1, it follows that  $\begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix} \subseteq ABSR^{2k \times 2k}.$  Thus (2.4) holds. The form (2.5) can be obtained by the similar method.

Futhermore, when  $n = 2k$

$$D^T \begin{pmatrix} M & H S_k \\ S_k H & S_k M S_k \end{pmatrix} D = \begin{pmatrix} M + H & 0 \\ 0 & M - H \end{pmatrix} \tag{2.9}$$

we have the form (2.6). When  $n = 2k + 1$ ,  $D$  is the same as (2.3)

$$\begin{aligned} & D^T \begin{pmatrix} N & C & HS_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} D \\ &= \frac{1}{2} \begin{pmatrix} I_k & 0 & S_k \\ 0 & \sqrt{2} & 0 \\ I_k & 0 & -S_k \end{pmatrix} \begin{pmatrix} N & C & HS_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} \begin{pmatrix} I_k & 0 & I_k \\ 0 & \sqrt{2} & 0 \\ S_k & 0 & -S_k \end{pmatrix} \\ &= \begin{pmatrix} N+H & \sqrt{2}C & 0 \\ -\sqrt{2}C^T & 0 & 0 \\ 0 & 0 & N-H \end{pmatrix}. \end{aligned}$$

Hence

$$\begin{pmatrix} N & C & HS_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} = D \begin{pmatrix} N+H & \sqrt{2}C & 0 \\ -\sqrt{2}C^T & 0 & 0 \\ 0 & 0 & N-H \end{pmatrix} D^T \quad (2.10)$$

It implies that the elements in  $ABSR^{n \times n}$  have the form (2.6) when  $n = 2k + 1$ .

On the other hand, it can be directly verified that matrices in form (2.6) belong to  $ABSR^{n \times n}$  from Lemma 1

Next we consider the problem I in real number field.

When  $n = 2k$  in (2.9) suppose eigenvalues of  $M + H$  are  $\pm\lambda_1 i, \pm\lambda_2 i, \dots, \pm\lambda_{t_1} i, \underbrace{0, \dots, 0}_{k-2t_1}$ , where  $\lambda_j, j = 1, 2, \dots, t_1$  are real numbers, the corresponding eigenvectors are  $\alpha_1 \pm \beta_1 i, \alpha_2 \pm \beta_2 i, \dots, \alpha_{t_1} \pm \beta_{t_1} i, x_{2t_1+1}, \dots, x_k$ , where  $i^2 = -1$  and  $\alpha_j, \beta_j \in R^k, j = 1, \dots, t_1, x_s \in R^k, s = 2t_1 + 1, \dots, k$  and eigenvalues of  $M - H$  are  $\pm\mu_1 i, \pm\mu_2 i, \dots, \pm\mu_{t_2} i, \underbrace{0, \dots, 0}_{k-2t_2}, \mu_j \in R, j = 1, \dots, t_2$ ,

the corresponding eigenvectors are  $\theta_1 \pm \gamma_1 i, \theta_2 \pm \gamma_2 i, \dots, \theta_{t_1} \pm \gamma_{t_1} i, y_{2t_2+1}, \dots, y_k, \theta_j, \gamma_j \in R^k, j = 1, \dots, t_2, y_s \in R^k, s = 2t_2 + 1, \dots, k$ . Then eigenvalues of  $A$  are  $\pm\lambda_1 i, \pm\lambda_2 i, \dots, \pm\lambda_{t_1} i, \underbrace{0, \dots, 0}_{k-2t_1}, \pm\mu_1 i, \pm\mu_2 i, \dots, \pm\mu_{t_2} i, \underbrace{0, \dots, 0}_{k-2t_2}$ .

Let

$$\begin{aligned} z_1 &= \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_1 + \beta_1 i \\ S_k(\alpha_1 + \beta_1 i) \end{pmatrix}, \bar{z}_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_1 - \beta_1 i \\ S_k(\alpha_1 - \beta_1 i) \end{pmatrix}, \dots, z_{t_1} = \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_{t_1} + \beta_{t_1} i \\ S_k(\alpha_{t_1} + \beta_{t_1} i) \end{pmatrix}, \\ \bar{z}_{t_1} &= \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_{t_1} - \beta_{t_1} i \\ S_k(\alpha_{t_1} - \beta_{t_1} i) \end{pmatrix}, z_{2t_1+1} = \frac{1}{\sqrt{2}} \begin{pmatrix} x_{2t_1+1} \\ S_k x_{2t_1+1} \end{pmatrix}, \dots, z_k = \frac{1}{\sqrt{2}} \begin{pmatrix} x_k \\ S_k x_k \end{pmatrix}, \\ z_{k+1} &= \frac{1}{\sqrt{2}} \begin{pmatrix} \theta_1 + \gamma_1 i \\ -S_k(\theta_1 + \gamma_1 i) \end{pmatrix}, \bar{z}_{k+1} = \frac{1}{\sqrt{2}} \begin{pmatrix} \theta_1 - \gamma_1 i \\ -S_k(\theta_1 - \gamma_1 i) \end{pmatrix}, \dots, z_{k+2t_2+1} = \frac{1}{\sqrt{2}} \\ & \begin{pmatrix} y_{2t_2+1} \\ -S_k y_{2t_2+1} \end{pmatrix}, \dots, z_n = \frac{1}{\sqrt{2}} \begin{pmatrix} y_k \\ -S_k y_k \end{pmatrix}. \end{aligned}$$

Then  $z_1, z_2, \dots, z_n$  are the corresponding eigenvectors of  $A$ .

**Definition 2.** If  $x \in C^n$  satisfies  $S_n x = x$  then  $x$  is called a symmetric vector, if  $x \in C^n$  satisfies  $S_n x = -x$  then  $x$  is called an anti-symmetric vector.

It is easy to verify that  $S_n z_1 = z_1, S_n \bar{z}_1 = \bar{z}_1, \dots, S_n z_k = z_k, S_n z_{k+1} = -z_{k+1}, S_n \bar{z}_{k+1} = -\bar{z}_{k+1}, \dots, S_n z_n = -z_n$ . Therefore it is that there exist  $k$  symmetric vectors and  $k$  anti-symmetric vectors for every  $A \in ABSR^{2k \times 2k}$ .

On the other hand, if  $\lambda$  is an eigenvalue of  $A$  and  $u$  is a corresponding eigenvector, i.e.  $Au = \lambda u$ . By Lemma 1 we have

$$AS_{2k}u = S_{2k}Au = \lambda S_{2k}u$$

it implies that  $S_{2k}u$  is an eigenvector of  $A$  corresponding to  $\lambda$ . Then  $u \pm S_{2k}u$  is also an eigenvector corresponding to  $\lambda$ . Where  $u + S_{2k}u$  is a symmetric vector and  $u - S_{2k}u$  is an anti-symmetric vector. Hence we can obtain a group of symmetric eigenvectors and anti-symmetric eigenvectors from the given eigenvectors.

Because

$$A \begin{pmatrix} \alpha_j + \beta_j i \\ S_k(\alpha_j + \beta_j i) \end{pmatrix} = \lambda_j i \begin{pmatrix} \alpha_j + \beta_j i \\ S_k(\alpha_j + \beta_j i) \end{pmatrix}$$

$$A \begin{pmatrix} \alpha_j - \beta_j i \\ S_k(\alpha_j - \beta_j i) \end{pmatrix} = -\lambda_j i \begin{pmatrix} \alpha_j - \beta_j i \\ S_k(\alpha_j - \beta_j i) \end{pmatrix}$$

Hence

$$A \begin{pmatrix} \alpha_j \\ S_k \alpha_j \end{pmatrix} = -\lambda_j \begin{pmatrix} \beta_j \\ S_k \beta_j \end{pmatrix}$$

$$A \begin{pmatrix} \beta_j \\ S_k \beta_j \end{pmatrix} = \lambda_j \begin{pmatrix} \alpha_j \\ S_k \alpha_j \end{pmatrix}$$

$$A \begin{pmatrix} \alpha_j & \beta_j \\ S_k \alpha_j & S_k \beta_j \end{pmatrix} = \begin{pmatrix} \alpha_j & \beta_j \\ S_k \alpha_j & S_k \beta_j \end{pmatrix} \begin{pmatrix} 0 & \lambda_j \\ -\lambda_j & 0 \end{pmatrix}.$$

Suppose  $\pm\lambda_j i, j = 1, \dots, t_1, \pm\mu_j i, j = 1, \dots, t_2, \underbrace{0, \dots, 0}_{m-2(t_1+t_2)}$  are eigenvalues of  $A$ .

Let

$$\Lambda = \text{diag}(B_1, \dots, B_{t_1}, 0 \cdot I_{l-2t_1}, C_1, \dots, C_{t_2}, 0 \cdot I_{m-l-2t_2}), \tag{2.11}$$

where  $B_j = \begin{pmatrix} 0 & \lambda_j \\ -\lambda_j & 0 \end{pmatrix}, C_j = \begin{pmatrix} 0 & \mu_j \\ -\mu_j & 0 \end{pmatrix} \in R^{2 \times 2}$

According to above analysis we obtain corresponding problem in real numbers field as follows:

**Problem  $I_0$ .** Given  $X \in R^{n \times m}, \Lambda$  is the same as (2.11). Find  $A \in ABSR^{n \times n}$  such that  $AX = X\Lambda$ ,

where  $ABSR^{n \times n}$  is the set of all real  $n \times n$  anti-bisymmetric matrices.

when  $n = 2k, X$  can be supposed the following form

$$X = \begin{pmatrix} X_1 & Y_1 \\ S_k X_1 & -S_k Y_1 \end{pmatrix}. \tag{2.12}$$

**Lemma 2.** <sup>[3]</sup> Suppose  $X = (X_1, X_2, \dots, X_t, X_{t+1}) \in R^{n \times m}$ , every column of  $X$  is nonzero vector,  $X_j \in R^{n \times 2}, j = 1, 2, \dots, t, X_{t+1} \in R^{n \times (m-2t)}, \text{rank}(X) = r, \Lambda = \text{diag}(B_1, \dots, B_t, 0 \cdot I_{m-2t})$ . Then there is  $A \in ASR^{n \times n}$  such that  $AX = X\Lambda$  if and only if

$$X_j^T X_l = 0, \quad B_j \neq B_l, j, l = 1, \dots, t + 1. \tag{2.13}$$

and the solutions of  $AX = X\Lambda$  can be represented as

$$A = X\Lambda X^+ + (X^+)^T \Lambda X^T - (X^+)^T \Lambda X^T X X^+ + U_2 Z U_2^T, \quad Z \in ASR^{n \times n}, \tag{2.14}$$

where  $U_2 \in R^{n \times (n-r)}, U_2^T U_2 = I_{n-r}, N(X^T) = R(U_2)$ .

**Theorem 2.** Given  $X \in R^{n \times m}$ , and  $X$  is the same as (2.12). Suppose  $X_1 = (X'_1 : X'_2 : \dots : X'_{t_1+1}) \in R^{k \times l}$ , every column of  $X_1$  is a non-vanishing vector,  $X'_j \in R^{k \times 2}$ , ( $j = 1, 2, \dots, t_1$ ),  $\text{rank}(X_1) = r_1$ ,  $X'_{t_1+1} \in R^{k \times (l-2t_1)}$ ,  $\Lambda_1 = \text{diag}(B_1, B_2, \dots, B_{t_1}, 0 \cdot I_{l-2t_1})$ ,  $B_j = \begin{pmatrix} 0 & \lambda_j \\ -\lambda_j & 0 \end{pmatrix}$ ,  $j = 1, 2, \dots, t_1$ ;  $Y_1 = (Y'_1 : Y'_2 : \dots : Y'_{t_2+1}) \in R^{k \times (m-l)}$ , every column of  $Y_1$  is a non-vanishing vector,  $Y'_j \in R^{k \times 2}$ , ( $j = 1, 2, \dots, t_2$ ),  $Y'_{t_2+1} \in R^{k \times (m-l-2t_2)}$ ,  $\text{rank}(Y_1) = r_2$ ,  $\Lambda_2 = \text{diag}(C_1, C_2, \dots, C_{t_2}, 0 \cdot I_{m-l-2t_2})$ ,  $C_j = \begin{pmatrix} 0 & \mu_j \\ -\mu_j & 0 \end{pmatrix}$ ,  $j = 1, 2, \dots, t_2$ . Let  $\Lambda = \begin{pmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{pmatrix}$ . Then there exists  $A \in \text{ABSR}^{n \times n}$  ( $n = 2k$ ) such that  $AX = X\Lambda$  if and only if

$$X'_j{}^T X'_l = 0, \quad \lambda_l \neq \lambda_j, l, j = 1, 2, \dots, t_1 + 1 \tag{2.15}$$

$$Y'_j{}^T Y'_l = 0, \quad \mu_l \neq \mu_j, l, j = 1, 2, \dots, t_2 + 1 \tag{2.16}$$

and the general solution can be represented as

$$A = A_0 + D \begin{pmatrix} U_2 G_1 U_2^T & 0 \\ 0 & P_2 G_2 P_2^T \end{pmatrix} D^T \tag{2.17}$$

where

$$A_0 = D \begin{pmatrix} X_1 \Lambda_1 X_1^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} D^T \tag{2.18}$$

$$\forall G_1 \in \text{ASR}^{(k-r_1) \times (k-r_1)}, \quad \forall G_2 \in \text{ASR}^{(k-r_2) \times (k-r_2)}$$

$$U_2 \in R^{k \times (k-r_1)}, \quad U_2^T U_2 = I_{k-r_1}, \quad N(X_1^T) = R(U_2),$$

$$P_2 \in R^{k \times (k-r_2)}, \quad P_2^T P_2 = I_{k-r_2}, \quad N(Y_1^T) = R(P_2),$$

$D$  is the same as (2.2).

*Proof. Necessity:* Suppose  $A$  is a solution of Problem  $I_0$ . From Theorem 1 it is verified that there exist  $A_{11}, A_{22} \in \text{ASR}^{k \times k}$  such that

$$A = D \begin{pmatrix} A_{11} & 0 \\ 0 & A_{22} \end{pmatrix} D^T, \quad AX = X\Lambda$$

i. e

$$D \begin{pmatrix} A_{11} & 0 \\ 0 & A_{22} \end{pmatrix} D^T \begin{pmatrix} X_1 & Y_1 \\ S_k X_1 & -S_k Y_1 \end{pmatrix} = \begin{pmatrix} X_1 & Y_1 \\ S_k X_1 & -S_k Y_1 \end{pmatrix} \begin{pmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{pmatrix} \tag{2.19}$$

(2.19) is equivalent to

$$A_{11} X_1 = X_1 \Lambda_1, \quad A_{11} \in \text{ASR}^{k \times k} \tag{2.20}$$

$$A_{22} Y_1 = Y_1 \Lambda_2, \quad A_{22} \in \text{ASR}^{k \times k} \tag{2.21}$$

From Lemma 2 we know (2.20), (2.21) have a solution if and only if

$$X'_j{}^T X'_l = 0, \quad \lambda_j \neq \lambda_l, j, l = 1, 2, \dots, t_1 + 1 \tag{2.22}$$

$$Y_j'^T Y_l' = 0, \quad \mu_j \neq \mu_l, j, l = 1, 2, \dots, t_2 + 1 \tag{2.23}$$

and there are respectively  $G_1 \in ASR^{(k-r_1) \times (k-r_1)}$ ,  $G_2 \in ASR^{(k-r_2) \times (k-r_2)}$  such that

$$A_{11} = X_1 \Lambda_1 X_1^+ + U_2 G_1 U_2^T \tag{2.24}$$

$$A_{22} = Y_1 \Lambda_2 Y_1^+ + P_2 G_2 P_2^T \tag{2.25}$$

where  $U_2 \in R^{k \times (k-r_1)}$ ,  $U_2^T U_2 = I_{k-r_1}$ ,  $N(X_1^T) = R(U_2)$ ,  $P_2 \in R^{k \times (k-r_2)}$ ,  $P_2^T P_2 = I_{k-r_2}$ ,  $N(Y_1^T) = R(P_2)$ .

From (2.2), (2.6), (2.24) and (2.25) we have

$$A = D \begin{pmatrix} X_1 \Lambda_1 X_1^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} D^T + D \begin{pmatrix} U_2 G_1 U_2^T & 0 \\ 0 & P_2 G_2 P_2^T \end{pmatrix} D^T, \tag{2.26}$$

for some  $G_1 \in ASR^{(k-r_1) \times (k-r_1)}$ ,  $G_2 \in ASR^{(k-r_2) \times (k-r_2)}$ .

**Sufficiency:** In (2.26) taking  $G_1 = 0 \in ASR^{(k-r_1) \times (k-r_1)}$ ,  $G_2 = 0 \in ASR^{(k-r_2) \times (k-r_2)}$ . We have  $A \in ABSR^{2k \times 2k}$

$$\begin{aligned} AX &= D \begin{pmatrix} X_1 \Lambda_1 X_1^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} D^T \begin{pmatrix} X_1 & Y_1 \\ S_k X_1 & -S_1 Y_1 \end{pmatrix} \\ &= D \begin{pmatrix} X_1 \Lambda_1 X_1^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} \begin{pmatrix} \sqrt{2} X_1 & 0 \\ 0 & \sqrt{2} Y_1 \end{pmatrix} \end{aligned}$$

From  $U_2^T X_1 = 0$ ,  $P_2^T Y_1 = 0$  and (2.15), (2.16) we obtain [4]

$$X_1 \Lambda_1 X_1^+ X_1 = X_1 \Lambda_1, \quad Y_1 \Lambda_2 Y_1^+ Y_1 = Y_1 \Lambda_2.$$

Hence

$$AX = \sqrt{2} D \begin{pmatrix} X_1 \Lambda_1 & 0 \\ 0 & Y_1 \Lambda_2 \end{pmatrix} = X \Lambda.$$

Since  $(X_1 \Lambda_1 X_1^+)^T = -X_1 \Lambda_1 X_1^+$ ,  $(Y_1 \Lambda_2 Y_1^+)^T = -Y_1 \Lambda_2 Y_1^+$  [3], then  $A \in ABSR^{2k \times 2k}$ . Therefore  $A$  is a solution of Problem  $I_0$ .

When  $n = 2k + 1$ ,  $D$  is the same as (2.3). Then

$$D^T A D = D^T \begin{pmatrix} N & C & H S_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} D = \begin{pmatrix} N + H & \sqrt{2} C & 0 \\ \sqrt{2} C^T & 0 & 0 \\ 0 & 0 & N - H \end{pmatrix} \tag{2.27}$$

Suppose eigenvalues of main submatrix  $\begin{pmatrix} N + H & \sqrt{2} C \\ -\sqrt{2} C^T & 0 \end{pmatrix}$  are  $\pm \lambda_1 i, \pm \lambda_2 i, \dots, \pm \lambda_{t_1} i, \underbrace{0, \dots, 0}_{k+1-2t_1}$ ,

$\lambda_j, j = 1, 2, \dots, t_1$  are real numbers and the corresponding eigenvectors are  $\begin{pmatrix} \alpha_1 \pm \beta_1 i \\ e_1 \pm f_1 i \end{pmatrix}$ ,

$\dots, \begin{pmatrix} \alpha_{t_1} \pm \beta_{t_1} i \\ e_{t_1} \pm f_{t_1} i \end{pmatrix}, \begin{pmatrix} x_{2t_1+1} \\ e_{2t_1+1} \end{pmatrix}, \dots, \begin{pmatrix} x_{k+1} \\ e_{k+1} \end{pmatrix}, \alpha_j, \beta_j \in R^k, j = 1, 2, \dots, t_1, x_l \in R^k, l = 2t_1 + 1, \dots, k + 1, e_j, j = 1, \dots, k + 1$  and  $f_j, j = 1, \dots, t_1$  are real numbers; eigenvalues of  $N - H$  are  $\pm \mu_1 i, \pm \mu_2 i, \dots, \pm \mu_{t_2} i, \underbrace{0, \dots, 0}_{k-2t_2}$ ,  $\mu_j, j = 1, \dots, t_2$  are real numbers and the corresponding eigen-

vectors are  $\theta_1 \pm \gamma_1 i, \theta_2 \pm \gamma_2 i, \dots, \theta_{t_2} \pm \gamma_{t_2} i, y_{2t_2+1}, \dots, y_k, \theta_j, \gamma_j \in R^k, j = 1, 2, \dots, t_2, y_l \in R^k$ ,

$l = 2t_2 + 1, \dots, k$ . Then eigenvalues of  $A$  are  $\pm\lambda_1 i, \pm\lambda_2 i, \dots, \pm\lambda_{t_1} i, \underbrace{0, \dots, 0}_{k+1-2t_1}, \pm\mu_1 i, \pm\mu_2 i, \dots, \pm\mu_{t_2} i, \underbrace{0, \dots, 0}_{k-2t_2}$ , whereas symmetric vectors  $\begin{pmatrix} \alpha_1 \pm \beta i \\ e_1 \pm \sqrt{2}f_1 i \\ S_k(\alpha_1 \pm \beta i) \end{pmatrix}, \dots, \begin{pmatrix} \alpha_{t_1} \pm \beta_{t_1} i \\ e_{t_1} \pm f_{t_1} i \\ S_k(\alpha_{t_1} \pm \beta_{t_1} i) \end{pmatrix}, \begin{pmatrix} x_{2t_1+1} \\ e_{2t_1+1} \\ S_k x_{2t_1+1} \end{pmatrix}, \dots, \begin{pmatrix} x_{k+1} \\ e_{k+1} \\ S_k x_{k+1} \end{pmatrix}$  and anti-symmetric vectors  $\begin{pmatrix} \theta_1 \pm \gamma_1 i \\ w_1 \pm q_1 i \\ -S_k(\theta_1 \pm \gamma_1 i) \end{pmatrix}, \dots, \begin{pmatrix} \theta_{t_2} \pm \gamma_{t_2} i \\ w_{t_2} \pm q_{t_2} i \\ -S_k(\theta_{t_2} \pm \gamma_{t_2} i) \end{pmatrix}, \begin{pmatrix} y_{2t_2+1} \\ 0 \\ -S_k y_{2t_2+1} \end{pmatrix}, \dots, \begin{pmatrix} y_k \\ 0 \\ -S_k y_k \end{pmatrix}$  are eigenvectors corresponding to  $\pm\lambda_1 i, \pm\lambda_2 i, \dots, \pm\lambda_{t_1} i, \underbrace{0, \dots, 0}_{k+1-2t_1}, \pm\mu_1 i, \dots, \pm\mu_2 i, \dots, \pm\mu_{t_2} i, \underbrace{0, \dots, 0}_{k-2t_2}$ . Hence it is easy to see that exist  $k + 1$  symmetric eigenvectors and  $k$  anti-symmetric eigenvectors for every  $A \in ABSR^{(2k+1) \times (2k+1)}$ .

Similarly to case  $n = 2k$  in Problem  $I_0$  we can suppose

$$X = \begin{pmatrix} X_1 & Y_1 \\ e^T & 0^T \\ S_k X_1 & -S_k Y_1 \end{pmatrix} \tag{2.28}$$

where  $X_1 \in R^{k \times l}, Y_1 \in R^{k \times (m-l)}, e^T = (\sqrt{2}e_1, \sqrt{2}e_2, \dots, \sqrt{2}e_l) \in R^{1 \times l}, 0^T = (0, 0, \dots, 0) \in R^{1 \times (m-l)}$ .

**Theorem 3.** Given  $X \in R^{(2k+1) \times m}$ , and  $X$  is the same as (2.28). Suppose  $X' = \begin{pmatrix} \sqrt{2}X_1 \\ e^T \end{pmatrix} = (Z'_1 : Z'_2 : \dots : Z'_{t_1+1}) \in R^{(k+1) \times l}$ , every column of  $X'$  is a non-vanishing vector,  $Z'_j \in R^{(k+1) \times 2}, j = 1, 2, \dots, t_1, Z'_{t_1+1} \in R^{(k+1) \times (l-2t_1)}, rank(X') = r_1, \Lambda_1 = diag(B_1, B_2, \dots, B_{t_1}, 0 \cdot I_{l-2t_1}), B_j = \begin{pmatrix} 0 & \lambda_j \\ -\lambda_j & 0 \end{pmatrix}, j = 1, 2, \dots, t_1; Y_1 = (Y'_1 : Y'_2 : \dots : Y'_{t_2+1}) \in R^{k \times (m-l)}$ , every column of  $Y_1$  is a non-vanishing vector,  $Y'_j \in R^{k \times 2}, (j = 1, 2, \dots, t_2), Y'_{t_2+1} \in R^{k \times (m-l-2t_2)}, \sum_{j=1}^{t_2+1} m_j = m - l, rank(Y_1) = r_2, \Lambda_2 = diag(C_1, C_2, \dots, C_{t_2}, 0 \cdot I_{k-2t_2}), C_j = \begin{pmatrix} 0 & \mu_j \\ -\mu_j & 0 \end{pmatrix}, j = 1, 2, \dots, t_2$ . Let  $\Lambda = \begin{pmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{pmatrix}$ . Then there exists  $A \in ABSR^{n \times n} (n = 2k + 1)$  such that  $AX = X\Lambda$  if and only if

$$Z_j'^T Z_l' = 0, \quad \lambda_l \neq \lambda_j, l, j = 1, 2, \dots, t_1 + 1 \tag{2.29}$$

$$Y_j'^T Y_l' = 0, \quad \mu_l \neq \mu_j, l, j = 1, 2, \dots, t_2 + 1 \tag{2.30}$$

and the general solution can be represented as

$$A = A'_0 + D \begin{pmatrix} U'_2 E_1 U'^T_2 & 0 \\ 0 & P'_2 E_2 P'^T_2 \end{pmatrix} D^T \tag{2.31}$$

where

$$A'_0 = D \begin{pmatrix} X' \Lambda_1 X'^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} D^T \tag{2.32}$$



$$\begin{aligned} \forall E_1 \in ASR^{(k+1-r_1) \times (k+1-r_1)}, \quad \forall E_2 \in ASR^{(k-r_2) \times (k-r_2)} \\ U'_2 \in R^{(k+1) \times (k+1-r_1)}, \quad U'^T_2 U'_2 = I_{k+1-r_1}, \quad N(X'^T) = R(U'_2), \\ P'_2 \in R^{k \times (k-r_2)}, \quad P'^T_2 P'_2 = I_{k-r_2}, \quad N(Y'^T_1) = R(P'_2), \end{aligned}$$

$D$  is the same as (2.3).

**Proof. Necessity:** Suppose  $A$  is a solution of problem  $I_0$ . From Theorem 1 we know that exist  $N, H \in ASR^{k \times k}, C \in R^n$ , such that

$$\begin{pmatrix} N & C & HS_k \\ -C^T & 0 & -C^T S_k \\ S_k H & S_k C & S_k N S_k \end{pmatrix} \begin{pmatrix} X_1 & Y_1 \\ e^T & 0 \\ S_k X_1 & -S_k Y_1 \end{pmatrix} = \begin{pmatrix} X_1 & Y_1 \\ e^T & 0 \\ S_k X_1 & -S_k Y_1 \end{pmatrix} \begin{pmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{pmatrix} \quad (2.33)$$

i. e

$$\begin{cases} NX_1 + Ce^T + HX_1 & = X_1 \Lambda_1 \\ -2C^T X_1 & = e^T \Lambda_1 \\ (N - H)Y_1 & = Y_1 \Lambda_2 \end{cases} \quad (2.34)$$

(2.34) is equivalent to

$$\begin{cases} \begin{pmatrix} N + H & \sqrt{2}C \\ -\sqrt{2}C^T & 0 \end{pmatrix} \begin{pmatrix} \sqrt{2}X_1 \\ e^T \end{pmatrix} = \begin{pmatrix} \sqrt{2}X_1 \\ e^T \end{pmatrix} \Lambda_1 \\ (N - H)Y_1 & = Y_1 \Lambda_2 \end{cases} \quad (2.35)$$

Let

$$X' = \begin{pmatrix} \sqrt{2}X_1 \\ e^T \end{pmatrix} = (Z'_1 : Z'_2 : \dots : Z'_{t_1+1}) \in R^{(k+1) \times l} \quad (2.36)$$

$$Y_1 = (Y'_1 : Y'_2 : \dots : Y'_{t_2+1}) \in R^{k \times (m-l)}$$

(2.29) and (2.30) hold from Lemma 2. And there is  $E_1 \in ASR^{(k+1-r_1) \times (k+1-r_1)}, E_2 \in ASR^{(k-r_2) \times (k-r_2)}$  such that

$$\begin{pmatrix} N + H & \sqrt{2}C \\ -\sqrt{2}C^T & 0 \end{pmatrix} = X' \Lambda_1 X'^+ + U'_2 E_1 U'^T_2 \quad (2.37)$$

$$N - H = Y_1 \Lambda_2 Y_1^+ + P'_2 E_2 P'^T_2 \quad (2.38)$$

where  $U'_2 \in R^{(k+1) \times (k+1-r_1)}, U'^T_2 U'_2 = I_{k+1-r_1}, N(X'^T) = R(U'_2); P'_2 \in R^{k \times (k-r_2)}, P'^T_2 P'_2 = I_{k-r_2}, N(Y'^T_1) = R(P'_2)$

By (2.27), (2.37) and (2.38) we know

$$A = D \begin{pmatrix} X' \Lambda_1 X'^+ & 0 \\ 0 & Y_1 \Lambda_2 Y_1^+ \end{pmatrix} D^T + D \begin{pmatrix} U'_2 E_1 U'^T_2 & 0 \\ 0 & P'_2 E_2 P'^T_2 \end{pmatrix} D^T, \quad (2.39)$$

$$\forall E_1 \in ASR^{(k+1-r_1) \times (k+1-r_1)}, \quad \forall E_2 \in ASR^{(k-r_2) \times (k-r_2)}.$$

**Sufficiency:** In (2.39) taking  $E_1 = 0 \in ASR^{(k+1-r_1) \times (k+1-r_1)}, E_2 = 0 \in ASR^{(k-r_2) \times (k-r_2)}$ . Because  $U'^T_2 X' = 0, P'^T_2 Y_1 = 0$  and from (2.29), (2.30) we obtain [4]

$$X' \Lambda_1 X'^+ X' = X' \Lambda_1, \quad Y_1 \Lambda_2 Y_1^+ Y_1 = Y_1 \Lambda_2.$$

Similar to the demonstration of Theorem 2 it is directly verified that  $AX = X\Lambda$ . Because  $(X'\Lambda_1X'^+)^T = -X'\Lambda_1X'^+$ ,  $(Y_1\Lambda_2Y_1^+)^T = -Y_1\Lambda_2Y_1^+$ [3]. We have  $A \in ABSR^{(2k+1) \times (2k+1)}$ . Therefore  $A$  is a solution of Problem  $I_0$ .

### 3. The Expression of Solution for Problem II

When solution set of Problem  $I_0$  is nonempty it is easily verified that  $S_E$  is a closed convex set. Therefore when  $n$  is even number we have

**Theorem 4.** *Given  $A^* \in R^{n \times n}$ ,  $X \in R^{n \times m}$  ( $n = 2k$ ) and the notation of  $X, \Lambda$  and conditions are the same as Theorem 2. Then there is a unique solution  $\hat{A} \in S_E$  for Problem II and  $\hat{A}$  can be represented as*

$$\hat{A} = A_0 + D \begin{pmatrix} \frac{(I-X_1X_1^+)(\tilde{A}_{11}-\tilde{A}_{11}^T)(I-X_1X_1^+)}{2} & 0 \\ 0 & \frac{(I-Y_1Y_1^+)(\tilde{A}_{22}-\tilde{A}_{22}^T)(I-Y_1Y_1^+)}{2} \end{pmatrix} D^T \quad (3.1)$$

where  $A_0$  is the same as (2.18) and  $D$  is the same as (2.2)

$$\tilde{A}_{11} = \frac{1}{2} \begin{pmatrix} I_k & S_k \end{pmatrix} (A^* - A_0) \begin{pmatrix} I_k \\ S_k \end{pmatrix}, \quad \tilde{A}_{22} = \frac{1}{2} \begin{pmatrix} I_k & -S_k \end{pmatrix} (A^* - A_0) \begin{pmatrix} I_k \\ -S_k \end{pmatrix} \quad (3.2)$$

*Proof.* Because  $S_E$  is a closed convex set there is a unique solution  $\hat{A}$  for problem II. According to (2.17) every element  $A$  in  $S_E$  can be represented as

$$A = A_0 + D \begin{pmatrix} U_2G_1U_2^T & 0 \\ 0 & P_2G_2P_2^T \end{pmatrix} D^T$$

Taking  $U_1 \in R^{k \times (k-r_1)}$ ,  $P_1 \in R^{k \times (k-r_1)}$  such that  $U \triangleq (U_1:U_2) \in R^{k \times r_1}$ ,  $P \triangleq (P_1:P_2) \in R^{k \times r_2}$   
Let

$$\tilde{A} = D^T(A^* - A_0)D = \begin{pmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ \tilde{A}_{21} & \tilde{A}_{22} \end{pmatrix} \quad (3.3)$$

where

$$\tilde{A}_{11} = \frac{1}{2}(I_k \quad S_k)(A^* - A_0) \begin{pmatrix} I_k \\ S_k \end{pmatrix} \quad (3.4)$$

$$\tilde{A}_{12} = \frac{1}{2}(I_k \quad S_k)(A^* - A_0) \begin{pmatrix} I_k \\ -S_k \end{pmatrix} \quad (3.5)$$

$$\tilde{A}_{21} = \frac{1}{2} \begin{pmatrix} I_k & -S_k \end{pmatrix} (A^* - A_0) \begin{pmatrix} I_k \\ S_k \end{pmatrix} \quad (3.6)$$

$$\tilde{A}_{22} = \frac{1}{2} \begin{pmatrix} I_k & -S_k \end{pmatrix} (A^* - A_0) \begin{pmatrix} I_k \\ -S_k \end{pmatrix} \quad (3.7)$$

Because

$$\begin{aligned} \|A^* - A\|^2 &= \left\| A^* - A_0 - D \begin{pmatrix} U_2G_1U_2^T & 0 \\ 0 & P_2G_2P_2^T \end{pmatrix} D^T \right\|^2 \\ &= \left\| D^T(A^* - A_0)D - \begin{pmatrix} U_2G_1U_2^T & 0 \\ 0 & P_2G_2P_2^T \end{pmatrix} \right\|^2 \\ &= \|\tilde{A}_{11} - U_2G_1U_2^T\|^2 + \|\tilde{A}_{12}\|^2 + \|\tilde{A}_{21}\|^2 + \|\tilde{A}_{22} - P_2G_2P_2^T\|^2. \end{aligned}$$

Hence  $\|A^* - A\| = \inf_{A \in ABSR^{n \times n}}$  is equivalent to

$$\|\tilde{A}_{11} - U_2 G_1 U_2^T\| = \min_{G_1 \in ASR^{k \times k}} \tag{3.8}$$

$$\|\tilde{A}_{22} - P_2 G_2 P_2^T\| = \min_{G_2 \in ASR^{k \times k}} \tag{3.9}$$

But

$$\begin{aligned} \|\tilde{A}_{11} - U_2 G_1 U_2^T\|^2 &= \|U^T \tilde{A}_{11} U - U^T U_2 G_1 U_2^T U\|^2 \\ &= \|U_1^T \tilde{A}_{11} U_1\|^2 + \|U_1^T \tilde{A}_{11} U_2\|^2 + \|U_2^T \tilde{A}_{11} U_1\|^2 + \|U_2^T \tilde{A}_{11} U_2 - G_1\|^2 \\ &= \|U_1^T \tilde{A}_{11} U_1\|^2 + \|U_1^T \tilde{A}_{11} U_2\|^2 + \|U_2^T \tilde{A}_{11} U_1\|^2 \\ &\quad + \|U_2^T \frac{\tilde{A}_{11} - \tilde{A}_{11}^T}{2} U_2 - G_1\|^2 + \|U_2^T \frac{\tilde{A}_{11} + \tilde{A}_{11}^T}{2} U_2\|^2 \end{aligned}$$

Therefore (3.8) holds if and only if

$$G_1 = U_2^T \frac{\tilde{A}_{11} - \tilde{A}_{11}^T}{2} U_2 \tag{3.10}$$

By the similar method (3.9) holds if and only if

$$G_2 = P_2^T \frac{\tilde{A}_{22} - \tilde{A}_{22}^T}{2} P_2. \tag{3.11}$$

Substituting (3.10), (3.11) to (2.17) we obtain the solution of Problem II is

$$\begin{aligned} \hat{A} &= A_0 + D \begin{pmatrix} U_2 U_2^T \frac{\tilde{A}_{11} - \tilde{A}_{11}^T}{2} U_2 U_2^T & 0 \\ 0 & P_2 P_2^T \frac{\tilde{A}_{22} - \tilde{A}_{22}^T}{2} P_2 P_2^T \end{pmatrix} D^T \\ &= A_0 + D \begin{pmatrix} (I - X_1 X_1^+) \frac{\tilde{A}_{11} - \tilde{A}_{11}^T}{2} (I - X_1 X_1^+) & 0 \\ 0 & (I - Y_1 Y_1^+) \frac{\tilde{A}_{22} - \tilde{A}_{22}^T}{2} (I - Y_1 Y_1^+) \end{pmatrix} D^T. \end{aligned}$$

According to similar discussion in Theorem 4 when  $n$  is odd number we have

**Theorem 5.** *Given  $A^* \in R^{n \times n}$ ,  $X \in R^{n \times m}$  ( $n = 2k + 1$ ) and the notation of  $X, \Lambda$  and conditions are the same as Theorem 3. Then there is a unique soluiton  $\hat{A} \in S_E$  and  $\hat{A}$  can be represented as*

$$\hat{A} = A'_0 + D \begin{pmatrix} \frac{(I - X' X'^+)(\bar{A}_{11} - \bar{A}_{11}^T)(I - X' X'^+)}{2} & 0 \\ 0 & \frac{(I - Y_1 Y_1^+)(\bar{A}_{22} - \bar{A}_{22}^T)(I - Y_1 Y_1^+)}{2} \end{pmatrix} D^T$$

where  $A'_0$  is the same as (2.32) and  $D$  is the same as (2.3)

$$\begin{aligned} \bar{A}_{11} &= \frac{1}{2} \begin{pmatrix} I_k & 0 & S_k \\ 0 & \sqrt{2} & 0 \end{pmatrix} (A^* - A'_0) \begin{pmatrix} I_k & 0 \\ 0 & \sqrt{2} \\ S_k & 0 \end{pmatrix}, \\ \bar{A}_{22} &= \frac{1}{2} \begin{pmatrix} I_k & 0 & -S_k \end{pmatrix} (A^* - A'_0) \begin{pmatrix} I_k \\ 0 \\ -S_k \end{pmatrix}. \end{aligned}$$

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