

# 2052. Eigensensitivity of damped system with defective multiple eigenvalues

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**Abstract.** This paper considers the sensitivity of defective multiple eigenvalues of reducible matrix pencil, the average of eigenvalues is proved to be analytic, the derivatives of the average eigenvalues and the corresponding eigenvector matrices are obtained when the generalized eigenvalue is reducible. The sensitivity of defective multiple eigenvalues of a quadratic eigenvalue problem dependent on several parameters are also obtained by the result of generalized eigenvalue problem. The results are useful for investigating structural optimal design, model updating and structural damage detection.

**Keywords:** sensitivity analysis, generalized eigenvalue problem, defective eigenvalues, quadratic eigenvalue problem.

## 1. Introduction

Matrix eigensensitivity analysis has extensive applications in some engineering applications, for example, structural optimal design [1, 2], model updating [3, 4] and structural damage detection [5]. For example, in sensitivity-based finite element (FE) model updating, the eigensolutions of the analytical model serve to construct the objective function, and the eigensensitivities represent a linearized estimate of the change in the eigenvalues and eigenvectors due to perturbations of the elemental parameters of the FE model. As a result, the study of the variation of the eigenvalues and eigenvectors due to variations in the system parameters, or more precisely the sensitivity of eigensolutions, has emerged as an important area of research.

Most of the early theoretical works, such as [6-10], were concentrated on existence of derivatives of simple eigenvalues and their corresponding eigenvectors of a matrix or matrix pencil. However, in problems such as those of dynamics of symmetric structures, the corresponding matrices can have repeated eigenvalues. The sensitivity analysis of multiple eigenvalues and associated invariant subspaces of a matrix or matrix pencil have been investigated by many researchers [11-15]. More generally, Andrew et al. [16] discussed the existence of derivatives of eigenvalues and corresponding eigenvectors of an analytic matrix-valued function, which includes the quadratic eigenvalue problem considered here as a special case.

In practical numerical computation, a number of methods for the eigenpair derivatives have been developed [17-35], but most of the results are under the condition that the eigenvalues are simple or repeated eigenvalues with well-separated derivatives. The requirement that the repeated eigenvalues must have well-separated derivatives was relaxed in Andrew and Tan [36]. The sensitivity of multiple eigenpairs of the quadratic eigenvalue problem are considered in [37-41]. Li et al. [42] showed that the undamped viscously or nonviscously damped eigenproblems can be considered as a degenerated case of general nonlinear eigenproblems and developed an unified eigensensitivity method for both distinct and repeated eigenvalues. Recently Qian et al. [43] extended the methods of [36] for generalized eigenvalue problem to quadratic eigenvalue problem, and proposed numerical methods for computing first and higher order derivatives of multiple eigenpairs of quadratic eigenvalue problem, and obtained a particular solution to the governing equation of the derivatives of eigenvectors by using the QR decomposition or SVD. Some of the

existing works on sensitivity of eigensolutions of quadratic eigenvalue problem are reviewed by Adhikari in [44, 45].

Unfortunately, most of the works of sensitivity analysis of repeated eigenvalues and their corresponding eigenvectors are concentrated on non-defective matrices, although some results of a matrix are available in [14]. This paper aims at developing a method for sensitivity calculation of a defective repeated eigenpairs of generalized eigenvalue and quadratic eigenproblems. The remainder of this paper is arranged as follows. In Section 2, the sensitivity of defective multiple eigenvalues of generalized eigenvalue problem are derived. In Section 3, we focus on dealing with Sensitivity of defective multiple eigenvalues of quadratic eigenvalue problem, and derive the derivatives of the average eigenvalues and the corresponding eigenvector matrices. In Section 4, one example is performed for illustrative purpose. Finally, we make some concluding remarks in Section 5.

Throughout this paper we use the following notation.  $\mathbf{C}^{m \times n}$  denotes the set of complex matrices,  $\mathbf{C}^n = \mathbf{C}^{n \times 1}$ ,  $\mathbf{C} = \mathbf{C}^1$ .  $\mathbf{I}_n$  is the identity matrix of order  $n$ ,  $\text{diag}(a_1, \dots, a_n)$  stands for the diagonal matrix with diagonal elements  $a_1, \dots, a_n$ .  $\mathbf{A}^T$  denotes the transpose of a matrix  $\mathbf{A}$ . If  $\mathbf{A} = (a_{ij}) \in \mathbf{C}^{m \times n}$ ,  $\mathbf{B} = (b_{ij}) \in \mathbf{C}^{m \times n}$ , then  $\mathbf{A} \otimes \mathbf{B} = (a_{ij}\mathbf{B}) \in \mathbf{C}^{mn \times mn}$  denotes the Kronecker product of matrix  $\mathbf{A}$  and  $\mathbf{B}$ .

## 2. Sensitivity of defective eigenpairs of reducible generalized eigenvalue problem

Suppose that  $L$  is an open set of  $\mathbf{C}^N$ ,  $p = (p_1, \dots, p_N) \in L$ ,  $\mathbf{A}(p)$ ,  $\mathbf{B}(p) \in \mathbf{C}^{n \times n}$  are analytic matrix-valued functions in some neighborhood  $\mathfrak{N}(p^*)$  of the point  $p^* \in L$ . Without loss of generality we may assume that the point  $p^*$  is the origin of  $\mathbf{C}^N$ . In this section we study sensitivity analysis of the following generalized eigenvalue problem:

$$\mathbf{A}(p)\mathbf{x}(p) = \lambda(p)\mathbf{B}(p)\mathbf{x}(p), \quad \lambda(p) \in \mathbf{C}, \quad \mathbf{x}(p) \in \mathbf{C}^n, \quad p \in \mathfrak{N}(0). \tag{1}$$

Except where stated otherwise, we consider the case in which:

- (i)  $\{\mathbf{A}(p), \mathbf{B}(p)\}$  is a reducible matrix pencil for  $p \in \mathfrak{N}(0)$ , i.e.,  $\mathbf{B}(p)$  is invertible.
- (ii) Eq. (1) has an eigenvalue  $\lambda_1$  with multiplicity  $r > 1$  when  $p = 0$  and  $\lambda_1$  is a defective multiple eigenvalue of the matrix pencil  $\{\mathbf{A}(0), \mathbf{B}(0)\}$ .

Before giving the main result, we cite some related lemmas.

**Lemma 1.** [46] Let  $\mathbf{A}, \mathbf{B} \in \mathbf{C}^{n \times n}$  and suppose  $\{\mathbf{A}, \mathbf{B}\}$  is a regular matrix pencil, then there exist invertible matrices  $\mathbf{P}, \mathbf{Q} \in \mathbf{C}^{n \times n}$ , such that:

$$\mathbf{PAQ} = \begin{pmatrix} \mathbf{J} & \mathbf{0} \\ \mathbf{0} & \mathbf{I}_{n_2} \end{pmatrix}, \quad \mathbf{PBQ} = \begin{pmatrix} \mathbf{I}_{n_1} & \mathbf{0} \\ \mathbf{0} & \mathbf{N} \end{pmatrix},$$

where:

$$\mathbf{J} = \text{diag}(\mathbf{J}_1(\lambda_1), \dots, \mathbf{J}_r(\lambda_r)) \in \mathbf{C}^{n_1 \times n_1}, \quad \lambda_i \neq \lambda_j \quad (i \neq j), \quad 1 \leq i, j \leq r,$$

$$\mathbf{J}_i(\lambda_i) = \text{diag}(\mathbf{J}_i^{(1)}(\lambda_i), \dots, \mathbf{J}_i^{(k_i)}(\lambda_i)) \in \mathbf{C}^{n(\lambda_i) \times n(\lambda_i)},$$

$$\mathbf{J}_i^{(k)}(\lambda_i) = \begin{pmatrix} \lambda_i & 1 & & \\ & \ddots & \ddots & \\ & & \ddots & 1 \\ & & & \lambda_i \end{pmatrix} \in \mathbf{C}^{n_k(\lambda_i) \times n_k(\lambda_i)}, \quad 1 \leq k \leq k_i, \quad 1 \leq i \leq r,$$

$$\sum_{k=1}^{k_i} n_k(\lambda_i) = n(\lambda_i), \quad 1 \leq i \leq r, \quad \sum_{i=1}^r n(\lambda_i) = n_1,$$

$$\mathbf{N} = \text{diag}(\mathbf{N}^{(l_1)}, \dots, \mathbf{N}^{(l_s)}) \in \mathbf{C}^{n_2 \times n_2}, \quad \mathbf{N}^{(l_j)} = \begin{pmatrix} 0 & 1 & & \\ & \ddots & \ddots & \\ & & \ddots & 1 \\ & & & 0 \end{pmatrix} \in \mathbf{C}^{l_j \times l_j}, \quad 1 \leq j \leq s,$$

$$\sum_{j=1}^s l_j = n_2, \quad n_1 + n_2 = n.$$

**Lemma 2.** [14] Let  $\mathbf{A}(p) \in \mathbf{C}^{n \times n}$  be an analytic matrix-valued function in some neighborhood  $\mathfrak{N}(0)$  of the origin,  $\lambda_1$  is  $r$  multiple defective eigenvalue of  $\mathbf{A}(0)$ , i.e., there exist invertible matrices  $\mathbf{X} = (\mathbf{X}_1, \mathbf{X}_2) \in \mathbf{C}^{n \times n}$ ,  $\mathbf{Y} = (\mathbf{Y}_1, \mathbf{Y}_2) \in \mathbf{C}^{n \times n}$ ,  $\mathbf{X}_1, \mathbf{Y}_1 \in \mathbf{C}^{n \times r}$ , such that:

$$\mathbf{Y}^H \mathbf{A}(0) \mathbf{X} = \begin{pmatrix} \mathbf{A}_1 & 0 \\ 0 & \mathbf{A}_2 \end{pmatrix}, \quad \mathbf{Y}^H \mathbf{X} = \mathbf{I}_r, \quad \lambda(\mathbf{A}_1) = \lambda_1, \quad \lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset.$$

Then there exist a neighborhood of the origin  $\mathfrak{N}_1(0)$  and analytic matrix-valued functions  $\mathbf{X}_1(p), \mathbf{Y}_1(p) \in \mathbf{C}^{r \times r}$ ,  $\mathbf{A}_1(p) \in \mathbf{C}^{r \times r}$  on  $\mathfrak{N}_1(0)$  which satisfy:

- 1)  $\mathbf{A}(p) \mathbf{X}_1(p) = \mathbf{X}_1(p) \mathbf{A}_1(p)$ ,  $\mathbf{Y}_1^H(p) \mathbf{A}(p) = \mathbf{A}_1(p) \mathbf{Y}_1^H(p)$ ,  $\mathbf{Y}_1^H(p) \mathbf{X}(p) = \mathbf{I}_r$ .
- 2)  $\mathbf{A}_1(0) = \mathbf{A}_1$ ,  $\mathbf{X}_1(0) = \mathbf{X}_1$ ,  $\mathbf{Y}_1(0) = \mathbf{Y}_1$ .
- 3) Define  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$ , then  $\lambda_{aver}$  is analytic on  $\mathfrak{N}_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 \right).$$

$$4) \frac{\partial \mathbf{X}_1(0)}{\partial p_i} = \mathbf{X}_2 \Theta_1^{-1} \left( \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 \right), \quad \frac{\partial \mathbf{Y}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 \right) \mathbf{Y}_2^H, \text{ where:}$$

$$\Theta_1(\mathbf{V}) = \mathbf{V} \mathbf{A}_1 - \mathbf{A}_2 \mathbf{V}, \quad \Theta_2(\mathbf{V}) = \mathbf{A}_1 \mathbf{V} - \mathbf{V} \mathbf{A}_2,$$

when  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ ,  $\Theta_1(\mathbf{V}), \Theta_2(\mathbf{V})$  is invertible.

**Theorem 1.** Let  $p = (p_1, p_2, \dots, p_N) \in \mathbf{C}^N$ ,  $\mathfrak{N}(0)$  be a neighborhood of the origin of  $\mathbf{C}^N$ ,  $\mathbf{A}(p), \mathbf{B}(p) \in \mathbf{C}^{n \times n}$  be analytic on  $\mathfrak{N}(0)$ . Assume that  $\mathbf{B}(p)$  is invertible on  $\mathfrak{N}(0)$  and  $\lambda_1$  is a defective multiple eigenvalue of Eq. (1) at the origin with multiplicity  $r$  ( $r > 1$ ). By Lemma 1 there exist invertible matrices  $\mathbf{X} = (\mathbf{X}_1, \mathbf{X}_2) \in \mathbf{C}^{n \times n}$ ,  $\mathbf{Y} = (\mathbf{Y}_1, \mathbf{Y}_2) \in \mathbf{C}^{n \times n}$ ,  $\mathbf{X}_1, \mathbf{Y}_1 \in \mathbf{C}^{n \times r}$ , such that:

$$\mathbf{Y}^H \mathbf{A}(0) \mathbf{X} = \begin{pmatrix} \mathbf{A}_1 & 0 \\ 0 & \mathbf{A}_2 \end{pmatrix}, \quad \mathbf{Y}^H \mathbf{B}(0) \mathbf{X} = \mathbf{I}_n,$$

where  $\mathbf{A}_1 \in \mathbf{C}^{r \times r}$ ,  $\mathbf{A}_2 \in \mathbf{C}^{(n-r) \times (n-r)}$ ,  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ . Then there exist a neighborhood of the origin  $\mathfrak{N}_1(0)$  and analytic matrix-valued functions  $\mathbf{X}_1(p), \mathbf{Y}_1(p) \in \mathbf{C}^{r \times r}$ ,  $\mathbf{A}_1(p) \in \mathbf{C}^{r \times r}$  on  $\mathfrak{N}_1(0)$  which satisfy:

- 1)  $\mathbf{A}(p) \mathbf{X}_1(p) = \mathbf{B}(p) \mathbf{X}_1(p) \mathbf{A}_1(p)$ ,  $\mathbf{Y}_1^H(p) \mathbf{A}(p) = \mathbf{A}_1(p) \mathbf{Y}_1^H(p) \mathbf{B}(p)$ ;
- 2)  $\mathbf{A}_1(0) = \mathbf{A}_1$ ,  $\mathbf{X}_1(0) = \mathbf{X}_1$ ,  $\mathbf{Y}_1(0) = \mathbf{Y}_1$ ;
- 3) Define  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$ , then  $\lambda_{aver}$  is analytic on  $\mathfrak{N}_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right);$$

$$4) \frac{\partial \mathbf{X}_1(0)}{\partial p_i} = \mathbf{X}_2 \Theta_1^{-1} \left( \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_2^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right),$$

$$\frac{\partial \mathbf{Y}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_2 \right) \mathbf{Y}_2^H - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0),$$

where  $\Theta_1(\mathbf{V}) = \mathbf{V}\mathbf{A}_1 - \mathbf{A}_2\mathbf{V}$ ,  $\Theta_2(\mathbf{V}) = \mathbf{A}_1\mathbf{V} - \mathbf{V}\mathbf{A}_2$ , (when  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ ,  $\Theta_1(\mathbf{V})$ ,  $\Theta_2(\mathbf{V})$  is invertible).

**Proof.** Let  $\mathbf{C}(p) = \mathbf{B}^{-1}(p)\mathbf{A}(p)$ , then Eq. (1) is equivalent to the following equation:

$$\mathbf{C}(p)\mathbf{x}(p) = \lambda(p)\mathbf{x}(p), \quad \lambda(p) \in \mathcal{C}, \quad \mathbf{x}(p) \in \mathbf{C}^n, \quad p \in \aleph(0). \tag{2}$$

Let  $\tilde{\mathbf{Y}}_1^H = \mathbf{Y}_1^H \mathbf{B}(0)$ ,  $\tilde{\mathbf{Y}}_2^H = \mathbf{Y}_2^H \mathbf{B}(0)$ ,  $\tilde{\mathbf{Y}}^H = (\tilde{\mathbf{Y}}_1^H, \tilde{\mathbf{Y}}_2^H)$ , then  $\tilde{\mathbf{Y}}_1^H$  satisfies:

$$\tilde{\mathbf{Y}}_1^H \mathbf{C}(0)\mathbf{X} = \begin{pmatrix} \mathbf{A}_1 & 0 \\ 0 & \mathbf{A}_2 \end{pmatrix}, \quad \tilde{\mathbf{Y}}_1^H \mathbf{X} = \mathbf{I}_n.$$

According to Lemma 2, there exist a neighborhood of the origin  $\aleph_1(0)$  and analytic matrix-valued functions  $\mathbf{X}_1(p)$ ,  $\mathbf{Y}_1(p) \in \mathbf{C}^{n \times r}$ ,  $\mathbf{A}_1(p) \in \mathbf{C}^{r \times r}$  on  $\aleph_1(0)$  which satisfy Eqs. (3)-(6):

$$\mathbf{C}(p)\mathbf{X}_1(p) = \mathbf{X}_1(p)\mathbf{A}_1(p), \quad \tilde{\mathbf{Y}}_1^H(p)\mathbf{C}(p) = \mathbf{A}_1(p)\tilde{\mathbf{Y}}_1^H(p), \quad \tilde{\mathbf{Y}}_1^H(p)\mathbf{X}(p) = \mathbf{I}_r, \tag{3}$$

$$\mathbf{A}_1(0) = \mathbf{A}_1, \quad \mathbf{X}_1(0) = \mathbf{X}_1, \quad \mathbf{Y}_1(0) = \mathbf{Y}_1. \tag{4}$$

$\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$  is analytic on  $\aleph_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left( \tilde{\mathbf{Y}}_1^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_1 \right), \tag{5}$$

$$\frac{\partial \mathbf{X}_1(0)}{\partial p_i} = \mathbf{X}_2 \Theta_1^{-1} \left( \tilde{\mathbf{Y}}_2^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_1 \right), \quad \frac{\partial \mathbf{Y}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left( \tilde{\mathbf{Y}}_1^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_2 \right) \tilde{\mathbf{Y}}_2^H, \tag{6}$$

where  $\Theta_1(\mathbf{V}) = \mathbf{V}\mathbf{A}_1 - \mathbf{A}_2\mathbf{V}$ ,  $\Theta_2(\mathbf{V}) = \mathbf{A}_1\mathbf{V} - \mathbf{V}\mathbf{A}_2$ , (when  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ ,  $\Theta_1(\mathbf{V})$ ,  $\Theta_2(\mathbf{V})$  is invertible).

Let  $\mathbf{Y}_1^H(p) = \tilde{\mathbf{Y}}_1^H(p)\mathbf{B}^{-1}(p)$ , from Eq. (3), we have:

$$\mathbf{A}(p)\mathbf{X}_1(p) = \mathbf{B}(p)\mathbf{X}_1(p)\mathbf{A}_1(p), \quad \mathbf{Y}_1^H(p)\mathbf{A}(p) = \mathbf{A}_1(p)\mathbf{Y}_1^H(p)\mathbf{B}(p),$$

and  $\mathbf{A}_1(0) = \mathbf{A}_1$ ,  $\mathbf{X}_1(0) = \mathbf{X}_1$ ,  $\mathbf{Y}_1(0) = \mathbf{Y}_1$ , then we get Eqs. (1), (2) of the Theorem 1.

Because Eq. (1) is equivalent to the Eq. (2), the eigenvalues of Eq. (1) is same with Eq. (2), then the average of eigenvalues of Eq. (1) equals to the average of eigenvalues of Eq. (2). From Eq. (5),  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$  is analytic on  $\aleph_1(0)$  and:

$$\begin{aligned} \frac{\partial \lambda_{aver}(0)}{\partial p_i} &= \frac{1}{r} \text{tr} \left( \tilde{\mathbf{Y}}_1^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_1 \right) = \frac{1}{r} \text{tr} \left\{ \mathbf{Y}_1^H \mathbf{B}(0) \left[ \mathbf{B}^{-1}(0) \left( \frac{\partial \mathbf{A}(0)}{\partial p_i} - \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{C}(0) \right) \right] \mathbf{X}_1 \right\} \\ &= \frac{1}{r} \text{tr} \left[ \mathbf{Y}_1^H \left( \frac{\partial \mathbf{A}(0)}{\partial p_i} - \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0)\mathbf{A}(0) \right) \mathbf{X}_1 \right] = \frac{1}{r} \text{tr} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right). \end{aligned}$$

From Eq. (6), we get the derivatives of right generalized eigenvectors matrices of Eq. (1):

$$\begin{aligned} \frac{\partial \mathbf{X}_1(0)}{\partial p_i} &= \mathbf{X}_2 \Theta_1^{-1} \left( \tilde{\mathbf{Y}}_2^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_1 \right) = \mathbf{X}_2 \Theta_1^{-1} \left\{ \mathbf{Y}_2^H \mathbf{B}(0) \left[ \mathbf{B}^{-1}(0) \left( \frac{\partial \mathbf{A}(0)}{\partial p_i} - \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{C}(0) \right) \right] \mathbf{X}_1 \right\} \\ &= \mathbf{X}_2 \Theta_1^{-1} \left[ \mathbf{Y}_2^H \left( \frac{\partial \mathbf{A}(0)}{\partial p_i} - \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0)\mathbf{A}(0) \right) \mathbf{X}_1 \right] \\ &= \mathbf{X}_2 \Theta_1^{-1} \left( \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_2^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right). \end{aligned}$$

According to:

$$\begin{aligned} \frac{\partial \tilde{\mathbf{Y}}_1^H(0)}{\partial p_i} &= \Theta_2^{-1} \left( \tilde{\mathbf{Y}}_1^H \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{X}_2 \right) \tilde{\mathbf{Y}}_2^H \\ &= \Theta_2^{-1} \left\{ \mathbf{Y}_1^H \mathbf{B}(0) \left[ \mathbf{B}^{-1}(0) \left( \frac{\partial \mathbf{A}(0)}{\partial p_i} - \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{C}(0) \right) \right] \mathbf{X}_2 \right\} \mathbf{Y}_2^H \mathbf{B}(0) \\ &= \Theta_2^{-1} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_2 \mathbf{A}_2 \right) \mathbf{Y}_2^H \mathbf{B}(0), \end{aligned}$$

and  $\tilde{\mathbf{Y}}_1^H(p) = \mathbf{Y}_1^H(p) \mathbf{B}(P)$ , we get the derivatives of left generalized eigenvectors matrices of Eq. (1):

$$\begin{aligned} \frac{\partial \mathbf{Y}_1^H(0)}{\partial p_i} &= \frac{\tilde{\mathbf{Y}}_1^H(0)}{\partial p_i} \mathbf{B}^{-1}(0) - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0) \\ &= \Theta_2^{-1} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_2 \mathbf{A}_2 \right) \mathbf{Y}_2^H - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0). \end{aligned}$$

So the proof of the Theorem 1 is completed.

### 3. Sensitivity of defective eigenpairs of quadratic eigenvalue problem

In this section we study sensitivity of the following quadratic eigenvalue problem:

$$(\lambda^2(p)\mathbf{M}(p) + \lambda(p)\mathbf{C}(p) + \mathbf{K}(p))\mathbf{u}(p) = 0, \quad \lambda(p) \in \mathbf{C}, \quad \mathbf{x}(p) \in \mathbf{C}^n, \quad p \in \aleph(0), \quad (7)$$

where  $p = (p_1, p_2, \dots, p_N) \in \mathbf{C}^N$ ,  $\mathbf{M}(p)$ ,  $\mathbf{K}(p)$ ,  $\mathbf{C}(p) \in \mathbf{C}^{n \times n}$  are analytic matrix-valued functions in some neighborhood  $\aleph(p^*)$  of the point  $p^*$  and  $\mathbf{M}(p)$  is invertible. Without loss of generality we may assume that the point  $p^*$  is the origin of  $\mathbf{C}^N$ .

**Lemma 3** [38]. Suppose  $\mathbf{M}$ ,  $\mathbf{C}$ ,  $\mathbf{K} \in \mathbf{C}^{n \times n}$  and  $\mathbf{M}$  is invertible, let:

$$\mathbf{A} = \begin{pmatrix} -\mathbf{K} & \mathbf{0} \\ \mathbf{0} & \mathbf{M} \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} \mathbf{C} & \mathbf{M} \\ \mathbf{M} & \mathbf{0} \end{pmatrix}.$$

Then, we have:

1) Quadratic eigenvalue problem Eq. (7) is equivalent to generalized eigenvalue problem  $\mathbf{A}\mathbf{x} = \lambda\mathbf{B}\mathbf{x}$ , where  $\mathbf{x}^H = [\mathbf{u}^H, \lambda\mathbf{u}^H]$ .

2) Suppose  $\lambda$  is the eigenvalue of Eq. (7), the columns of  $\mathbf{U}$ ,  $\mathbf{V} \in \mathbf{C}^{n \times r}$  are the corresponding right eigenvectors and left eigenvectors of  $\lambda$ , let  $\mathbf{X}^H = [\mathbf{U}^H, \lambda\mathbf{U}^H]$ ,  $\mathbf{Y}^H = [\mathbf{V}^H, \lambda\mathbf{V}^H]$ , then:

$$\mathbf{A}\mathbf{X} = \lambda\mathbf{B}\mathbf{X}, \quad \mathbf{Y}^H \mathbf{A} = \lambda\mathbf{Y}^H \mathbf{B}, \quad \mathbf{Y}^H \mathbf{B}\mathbf{X} = \mathbf{V}^H (2\lambda\mathbf{M} + \mathbf{C})\mathbf{U}.$$

3) If  $\mathbf{U}_1 \in \mathbf{C}^{r \times r}$ ,  $\mathbf{D}_1 \in \mathbf{C}^{r \times r}$  satisfy  $\mathbf{M}\mathbf{U}_1\mathbf{D}_1^2 + \mathbf{C}\mathbf{U}_1\mathbf{D}_1 + \mathbf{K}\mathbf{U}_1 = \mathbf{0}$ , then  $\lambda(\mathbf{D}_1) \subset \lambda(\mathbf{M}, \mathbf{C}, \mathbf{K})$ .

**Theorem 2.** Suppose  $\mathbf{M}$ ,  $\mathbf{C}$ ,  $\mathbf{K} \in \mathbf{C}^{n \times n}$  and  $\mathbf{M}$  is invertible. If  $\lambda_1$  is a defective multiple eigenvalues of Eq. (7) with multiplicity  $r$ . Then there exist:

$$\begin{aligned} \mathbf{U} &= (\mathbf{U}_1, \mathbf{U}_2) \in \mathbf{C}^{n \times 2n}, \quad \mathbf{V} = (\mathbf{V}_1, \mathbf{V}_2) \in \mathbf{C}^{n \times 2n}, \quad \mathbf{U}_1, \mathbf{V}_1 \in \mathbf{C}_r^{n \times r}, \\ \mathbf{A}_1 &\in \mathbf{C}^{r \times r}, \quad \mathbf{A}_2 \in \mathbf{C}^{(2n-r) \times (2n-r)}, \end{aligned}$$

such that  $\mathbf{U}_1, \mathbf{V}_1$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_1$  respectively,  $\mathbf{U}_2, \mathbf{V}_2$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_2$  respectively. Moreover:

$$\begin{aligned} \mathbf{V}_1^H \mathbf{C} \mathbf{U}_1 + \mathbf{A}_1 \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 + \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 \mathbf{A}_1 &= \mathbf{I}_r, \\ \mathbf{V}_2^H \mathbf{C} \mathbf{U}_2 + \mathbf{A}_2 \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 + \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 \mathbf{A}_2 &= \mathbf{I}_{2n-r}, \\ \lambda(\mathbf{A}_1) &= \lambda_1, \quad \lambda_1 \notin \lambda(\mathbf{A}_2). \end{aligned}$$

**Proof.** Since  $\lambda_1$  is a defective multiple eigenvalues of Eq. (7) with multiplicity  $r$ , from Lemma 3,  $\lambda_1$  is also a defective multiple eigenvalues of  $\mathbf{A}\mathbf{x} = \lambda\mathbf{B}\mathbf{x}$ , where:

$$\mathbf{A} = \begin{pmatrix} -\mathbf{K} & \mathbf{0} \\ \mathbf{0} & \mathbf{M} \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} \mathbf{C} & \mathbf{M} \\ \mathbf{M} & \mathbf{0} \end{pmatrix}.$$

By Lemma 1, there exists  $\mathbf{X} = (\mathbf{X}_1, \mathbf{X}_2)$ ,  $\mathbf{Y} = (\mathbf{Y}_1, \mathbf{Y}_2)$ ,  $\mathbf{X}_1, \mathbf{Y}_1 \in \mathbb{C}^{2n \times r}$  such that:

$$\mathbf{Y}^H \mathbf{A} \mathbf{X} = \begin{pmatrix} \mathbf{A}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{A}_2 \end{pmatrix}, \quad \mathbf{Y}^H \mathbf{B} \mathbf{X} = \mathbf{I}_{2n}, \quad \lambda_1 \notin \lambda(\mathbf{A}_2). \tag{8}$$

From Eq. (8) we get:

$$\mathbf{A} \mathbf{X}_1 = \mathbf{B} \mathbf{X}_1 \mathbf{A}_1, \quad \mathbf{Y}_1^H \mathbf{A} = \mathbf{A}_1 \mathbf{Y}_1^H \mathbf{B}, \tag{9}$$

$$\mathbf{A} \mathbf{X}_2 = \mathbf{B} \mathbf{X}_2 \mathbf{A}_2, \quad \mathbf{Y}_2^H \mathbf{A} = \mathbf{A}_2 \mathbf{Y}_2^H \mathbf{B}, \tag{10}$$

$$\mathbf{Y}_1^H \mathbf{B} \mathbf{X}_1 = \mathbf{I}_r, \quad \mathbf{Y}_2^H \mathbf{B} \mathbf{X}_2 = \mathbf{I}_{2n-r}. \tag{11}$$

Let  $\mathbf{U}_i = [\mathbf{I}_n, \mathbf{0}] \mathbf{X}_i$ ,  $\mathbf{V}_i = [\mathbf{I}_n, \mathbf{0}] \mathbf{Y}_i$  ( $i = 1, 2$ ), from Eq. (9) we have:

$$\mathbf{X}_1^H = [\mathbf{U}_1^H, \mathbf{A}_1^H \mathbf{U}_1^H], \quad \mathbf{Y}_1^H = [\mathbf{V}_1^H, \mathbf{A}_1 \mathbf{V}_1^H], \tag{12}$$

$$\mathbf{M} \mathbf{U}_1 \mathbf{A}_1^2 + \mathbf{C} \mathbf{U}_1 \mathbf{A}_1 + \mathbf{K} \mathbf{U}_1 = \mathbf{0}, \quad \mathbf{A}_1^2 \mathbf{V}_1^H \mathbf{M} + \mathbf{A}_1 \mathbf{V}_1^H \mathbf{C} + \mathbf{V}_1^H \mathbf{K} = \mathbf{0}. \tag{13}$$

From Eq. (10) we get:

$$\mathbf{X}_2^H = [\mathbf{U}_2^H, \mathbf{A}_2^H \mathbf{U}_2^H], \quad \mathbf{Y}_2^H = [\mathbf{V}_2^H, \mathbf{A}_2 \mathbf{V}_2^H], \tag{14}$$

$$\mathbf{M} \mathbf{U}_2 \mathbf{A}_2^2 + \mathbf{C} \mathbf{U}_2 \mathbf{A}_2 + \mathbf{K} \mathbf{U}_2 = \mathbf{0}, \quad \mathbf{A}_2^2 \mathbf{V}_2^H \mathbf{M} + \mathbf{A}_2 \mathbf{V}_2^H \mathbf{C} + \mathbf{V}_2^H \mathbf{K} = \mathbf{0}. \tag{15}$$

Thus  $\mathbf{U}_1$  and  $\mathbf{V}_1$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_1$  respectively,  $\mathbf{U}_2$  and  $\mathbf{V}_2$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_2$  respectively. From Eqs. (11), (12) and (14) we get:

$$\mathbf{V}_1^H \mathbf{C} \mathbf{U}_1 + \mathbf{A}_1 \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 + \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 \mathbf{A}_1 = \mathbf{I}_r, \tag{16}$$

$$\mathbf{V}_2^H \mathbf{C} \mathbf{U}_2 + \mathbf{A}_2 \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 + \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 \mathbf{A}_2 = \mathbf{I}_{2n-r}. \tag{17}$$

So the proof of the theorem is completed.

**Theorem 3.** Let  $p = (p_1, p_2, \dots, p_N) \in \mathbb{C}^N$ ,  $\mathfrak{N}(0)$  be a neighborhood of the origin of  $\mathbb{C}^N$ ,  $\mathbf{M}(p)$ ,  $\mathbf{C}(p)$ ,  $\mathbf{K}(p) \in \mathbb{C}^{n \times n}$  be analytic on  $\mathfrak{N}(0)$ . Assume that  $\mathbf{M}(p)$  is invertible on  $\mathfrak{N}(0)$  and  $\lambda_1$  is a defective multiple eigenvalue of Eq. (7) at the origin with multiplicity  $r$  ( $r > 1$ ), i.e., there exist invertible matrices  $\mathbf{U} = (\mathbf{U}_1, \mathbf{U}_2) \in \mathbb{C}^{n \times 2n}$ ,  $\mathbf{V} = (\mathbf{V}_1, \mathbf{V}_2) \in \mathbb{C}^{n \times 2n}$ ,  $\mathbf{U}_1, \mathbf{V}_1 \in \mathbb{C}^{r \times r}$ ,  $\mathbf{A}_1 \in \mathbb{C}^{r \times r}$ ,  $\mathbf{A}_2 \in \mathbb{C}^{(2n-r) \times (2n-r)}$ , such that  $\mathbf{U}_1, \mathbf{V}_1$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_1$  respectively,  $\mathbf{U}_2, \mathbf{V}_2$  are right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_2$  respectively. Moreover:

$$\mathbf{V}_1^H \mathbf{C} \mathbf{U}_1 + \mathbf{A}_1 \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 + \mathbf{V}_1^H \mathbf{M} \mathbf{U}_1 \mathbf{A}_1 = \mathbf{I}_r,$$

$$\mathbf{V}_2^H \mathbf{C} \mathbf{U}_2 + \mathbf{A}_2 \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 + \mathbf{V}_2^H \mathbf{M} \mathbf{U}_2 \mathbf{A}_2 = \mathbf{I}_{2n-r},$$

$$\lambda(\mathbf{A}_1) = \lambda_1, \quad \lambda_1 \notin \lambda(\mathbf{A}_2).$$

Then there exist a neighborhood of the origin  $\mathfrak{N}_1(0)$  and analytic matrix-valued functions  $\mathbf{U}_1(p), \mathbf{V}_1(p) \in \mathbb{C}^{r \times r}$ ,  $\mathbf{A}_1(p) \in \mathbb{C}^{r \times r}$  on  $\mathfrak{N}_1(0)$  which satisfy:

- 1)  $\mathbf{M}(p)\mathbf{Y}_1(p)\mathbf{A}_1^2(p) + \mathbf{C}(p)\mathbf{U}_1(p)\mathbf{A}_1(p) + \mathbf{K}(p)\mathbf{U}_1(p) = \mathbf{0}$ ,  
 $\mathbf{A}_1^2(p)\mathbf{V}_1^H(p)\mathbf{M}(p) + \mathbf{A}_1(p)\mathbf{V}_1^H(p)\mathbf{C}(p) + \mathbf{V}_1^H(p)\mathbf{K}(p) = \mathbf{0}$ .
- 2)  $\mathbf{A}_1(0) = \mathbf{A}_1$ ,  $\mathbf{U}_1(0) = \mathbf{U}_1$ ,  $\mathbf{V}_1(0) = \mathbf{V}_1$ .
- 3) The average of the eigenvalues  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$  is analytic on  $\aleph_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left[ -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right) \right].$$

$$4) \frac{\partial \mathbf{U}_1(0)}{\partial p_i} = \mathbf{U}_2 \Theta_1^{-1} \left[ -\mathbf{V}_2^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right) \right],$$

$$\frac{\partial \mathbf{V}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left[ -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_2 \right) \right] \mathbf{V}_2^H - \mathbf{V}_1^H \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{M}^{-1}(0),$$

where  $\Theta_1(\mathbf{V}) = \mathbf{V}\mathbf{A}_1 - \mathbf{A}_2\mathbf{V}$ ,  $\Theta_2(\mathbf{V}) = \mathbf{A}_1\mathbf{V} - \mathbf{V}\mathbf{A}_2$ , (when  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ ,  $\Theta_1$ ,  $\Theta_2$  is invertible).

**Proof.** Let:

$$\mathbf{A}(p) = \begin{pmatrix} -\mathbf{K}(p) & \mathbf{0} \\ \mathbf{0} & \mathbf{M}(p) \end{pmatrix}, \quad \mathbf{B}(p) = \begin{pmatrix} \mathbf{C}(p) & \mathbf{M}(p) \\ \mathbf{M}(p) & \mathbf{0} \end{pmatrix}, \quad \mathbf{x}(p) = \begin{pmatrix} \mathbf{u}(p) \\ \lambda(p)\mathbf{u}(p) \end{pmatrix},$$

$$\mathbf{X} = (\mathbf{X}_1, \mathbf{X}_2), \quad \mathbf{X}_1^H = [\mathbf{U}_1^H, \mathbf{A}_1^H \mathbf{U}_1^H], \quad \mathbf{X}_2^H = [\mathbf{U}_2^H, \mathbf{A}_2^H \mathbf{U}_2^H],$$

$$\mathbf{Y} = (\mathbf{Y}_1, \mathbf{Y}_2), \quad \mathbf{Y}_1^H = [\mathbf{V}_1^H, \mathbf{A}_1 \mathbf{V}_1^H], \quad \mathbf{Y}_2^H = [\mathbf{V}_2^H, \mathbf{A}_2 \mathbf{V}_2^H].$$

From Lemma 3 and the proof of Theorem 2, Eq. (7) is equivalent to generalized eigenvalue problem  $\mathbf{A}(p)\mathbf{x}(p) = \lambda(p)\mathbf{B}(p)\mathbf{x}(p)$ ,  $\mathbf{B}(p)$  is invertible,  $\lambda_1$  is a defective multiple eigenvalue of  $\{\mathbf{A}(0), \mathbf{B}(0)\}$  at the origin with multiplicity  $r$  and:

$$\mathbf{Y}^H \mathbf{A}(0) \mathbf{X} = \begin{pmatrix} \mathbf{A}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{A}_2 \end{pmatrix}, \quad \mathbf{Y}^H \mathbf{B} \mathbf{X} = \mathbf{I}_{2n}, \quad \lambda_1 \notin \lambda(\mathbf{A}_2).$$

According to Theorem 1, Then there exist a neighborhood of the origin  $\aleph_1(0)$  and analytic matrix-valued functions  $\mathbf{X}_1(p)$ ,  $\mathbf{Y}_1(p) \in \mathbf{C}_r^{2n \times r}$ ,  $\mathbf{A}_1(p) \in \mathbf{C}^{r \times r}$  on  $\aleph_1(0)$  which satisfy Eqs. (18-22):

$$\mathbf{A}(p)\mathbf{X}_1(p) = \mathbf{B}(p)\mathbf{X}_1(p)\mathbf{A}_1(p), \quad \mathbf{Y}_1^H(p)\mathbf{A}(p) = \mathbf{A}_1(p)\mathbf{Y}_1^H(p)\mathbf{B}(p), \quad (18)$$

$$\mathbf{A}_1(0) = \mathbf{A}_1, \quad \mathbf{X}_1(0) = \mathbf{X}_1, \quad \mathbf{Y}_1(0) = \mathbf{Y}_1. \quad (19)$$

The average of the eigenvalues  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$  is analytic on  $\aleph_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right), \quad (20)$$

$$\frac{\partial \mathbf{X}_1(0)}{\partial p_i} = \mathbf{X}_2 \Theta_1^{-1} \left( \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_2^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right), \quad (21)$$

$$\frac{\partial \mathbf{Y}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_2 \right) \mathbf{Y}_2^H - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{B}^{-1}(0), \quad (22)$$

where  $\Theta_1(\mathbf{V}) = \mathbf{V}\mathbf{A}_1 - \mathbf{A}_2\mathbf{V}$ ,  $\Theta_2(\mathbf{V}) = \mathbf{A}_1\mathbf{V} - \mathbf{V}\mathbf{A}_2$ , (when  $\lambda(\mathbf{A}_1) \cap \lambda(\mathbf{A}_2) = \emptyset$ ,  $\Theta_1(\mathbf{V})$ ,  $\Theta_2(\mathbf{V})$  is invertible).

Let  $\mathbf{U}_i(p) = [\mathbf{I}_n, \mathbf{0}]\mathbf{X}_i(p)$ ,  $\mathbf{V}_i(p) = [\mathbf{I}_n, \mathbf{0}]\mathbf{Y}_i(p)$ , ( $i = 1, 2$ ), then  $\mathbf{U}_1(p)$ ,  $\mathbf{V}_1(p)$  is analytic on  $\aleph_1(0)$ . From Eq. (18) we get:

$$\mathbf{M}(p)\mathbf{U}_1(p)\mathbf{A}_1^2(p) + \mathbf{C}(p)\mathbf{U}_1(p)\mathbf{A}_1(p) + \mathbf{K}(p)\mathbf{U}_1(p) = \mathbf{0},$$

$$\mathbf{A}_1^2(p)\mathbf{V}_1^H(p)\mathbf{M}(p) + \mathbf{A}_1(p)\mathbf{V}_1^H(p)\mathbf{C}(p) + \mathbf{V}_1^H(p)\mathbf{K}(p) = \mathbf{0}.$$

Thus  $\mathbf{U}_1(p)$ ,  $\mathbf{V}_1(p)$  are the right and left eigenvector matrices corresponding to eigenvalues matrix  $\mathbf{A}_1(p)$  of the Eq. (7) respectively. By Eq. (19),  $\mathbf{U}_1(0) = \mathbf{U}_1$ ,  $\mathbf{V}_1(0) = \mathbf{V}_1$ , so we get Eqs. (1), (2) of this theorem. By Eq. (20), the average of the eigenvalues  $\lambda_{aver} = \text{tr}(\mathbf{A}_1(p))/r$  is analytic on  $\mathfrak{K}_1(0)$  and:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left( \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right).$$

Since:

$$\begin{aligned} \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 &= (\mathbf{V}_1, \mathbf{A}_1 \mathbf{V}_1^H) \begin{pmatrix} -\frac{\partial \mathbf{K}(0)}{\partial p_i} & \\ & \frac{\partial \mathbf{M}(0)}{\partial p_i} \end{pmatrix} \begin{pmatrix} \mathbf{U}_1 \\ \mathbf{U}_1 \mathbf{A}_1 \end{pmatrix} \\ &- (\mathbf{V}_1, \mathbf{A}_1 \mathbf{V}_1^H) \begin{pmatrix} \frac{\partial \mathbf{C}(0)}{\partial p_i} & \frac{\partial \mathbf{M}(0)}{\partial p_i} \\ \frac{\partial \mathbf{M}(0)}{\partial p_i} & \mathbf{0} \end{pmatrix} \begin{pmatrix} \mathbf{U}_1 \\ \mathbf{U}_1 \mathbf{A}_1 \end{pmatrix} \mathbf{A}_1 \\ &= -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right). \end{aligned}$$

Then:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_i} = \frac{1}{r} \text{tr} \left[ -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right) \right],$$

so we get Eq. (3) of the theorem. According to Eq. (21):

$$\frac{\partial \mathbf{U}_1(0)}{\partial p_i} = \mathbf{U}_2 \Theta_1^{-1} \left( \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_2^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 \right).$$

Since:

$$\begin{aligned} \mathbf{Y}_2^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_1 - \mathbf{Y}_2^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_1 \mathbf{A}_1 &= (\mathbf{V}_2^H, \mathbf{A}_2 \mathbf{V}_2^H) \begin{pmatrix} -\frac{\partial \mathbf{K}(0)}{\partial p_i} & \\ & \frac{\partial \mathbf{M}(0)}{\partial p_i} \end{pmatrix} \begin{pmatrix} \mathbf{U}_1 \\ \mathbf{U}_1 \mathbf{A}_1 \end{pmatrix} \\ &- (\mathbf{V}_2^H, \mathbf{A}_2 \mathbf{V}_2^H) \begin{pmatrix} \frac{\partial \mathbf{C}(0)}{\partial p_i} & \frac{\partial \mathbf{M}(0)}{\partial p_i} \\ \frac{\partial \mathbf{M}(0)}{\partial p_i} & \mathbf{0} \end{pmatrix} \begin{pmatrix} \mathbf{U}_1 \\ \mathbf{U}_1 \mathbf{A}_1 \end{pmatrix} \mathbf{A}_1 \\ &= -\mathbf{V}_2^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right). \end{aligned}$$

Then:



$$\frac{\partial \mathbf{U}_1(0)}{\partial p_i} = \mathbf{U}_2 \Theta_1^{-1} \left[ -\mathbf{V}_2^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_1 \mathbf{A}_1 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_1 \right) \right].$$

Utilizing Eq. (22) and the following equation:

$$\begin{aligned} \mathbf{Y}_1^H \frac{\partial \mathbf{A}(0)}{\partial p_i} \mathbf{X}_2 - \mathbf{Y}_1^H \frac{\partial \mathbf{B}(0)}{\partial p_i} \mathbf{X}_2 &= (\mathbf{V}_1^H, \mathbf{A}_1 \mathbf{V}_1^H) \begin{pmatrix} -\frac{\partial \mathbf{K}(0)}{\partial p_i} & \\ & \frac{\partial \mathbf{M}(0)}{\partial p_i} \end{pmatrix} \begin{pmatrix} \mathbf{U}_2 \\ \mathbf{U}_2 \mathbf{A}_2 \end{pmatrix} \\ &- (\mathbf{V}_1^H, \mathbf{A}_1 \mathbf{V}_1^H) \begin{pmatrix} \frac{\partial \mathbf{C}(0)}{\partial p_i} & \frac{\partial \mathbf{M}(0)}{\partial p_i} \\ \frac{\partial \mathbf{M}(0)}{\partial p_i} & \mathbf{0} \end{pmatrix} \begin{pmatrix} \mathbf{U}_2 \\ \mathbf{U}_2 \mathbf{A}_2 \end{pmatrix} \mathbf{A}_2 \\ &= -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_2 \right), \end{aligned}$$

we have:

$$\frac{\partial \mathbf{V}_1^H(0)}{\partial p_i} = \Theta_2^{-1} \left[ -\mathbf{V}_1^H \left( \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2^2 + \frac{\partial \mathbf{C}(0)}{\partial p_i} \mathbf{U}_2 \mathbf{A}_2 + \frac{\partial \mathbf{K}(0)}{\partial p_i} \mathbf{U}_2 \right) \right] \mathbf{V}_2^H - \mathbf{V}_1^H \frac{\partial \mathbf{M}(0)}{\partial p_i} \mathbf{M}^{-1}(0).$$

So the proof of the theorem is completed.

#### 4. An example

Let  $p = (p_1, p_2)$ , consider the matrix, consider the matrix:

$$\mathbf{A}(p) = \begin{pmatrix} 2p_1 + p_2 + 1 & p_1 + p_2 + 1 & p_1 \\ 2p_2 & p_1 + 1 & p_2 \\ p_1 & p_2 & 0 \end{pmatrix}, \quad \mathbf{B}(p) = \begin{pmatrix} 1 & p_1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

Then:

$$\mathbf{A}(0) = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad \mathbf{B}(p) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

The eigenvalue of  $\{\mathbf{A}(0), \mathbf{B}(0)\}$  is  $\lambda = 1$  with multiplicity  $r = 2$ . The derivatives of the matrices are:

$$\begin{aligned} \frac{\partial \mathbf{A}(0)}{\partial p_1} &= \begin{pmatrix} 2 & 1 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}, & \frac{\partial \mathbf{B}(0)}{\partial p_1} &= \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \\ \frac{\partial \mathbf{A}(0)}{\partial p_2} &= \begin{pmatrix} 1 & 1 & 0 \\ 2 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}, & \frac{\partial \mathbf{B}(0)}{\partial p_2} &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}. \end{aligned}$$

Let  $\mathbf{X} = \mathbf{I}_3$  and  $\mathbf{Y} = \mathbf{I}_3$ , then:

$$\mathbf{Y}^H \mathbf{A}(0) \mathbf{X} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \mathbf{Y}^H \mathbf{B}(0) \mathbf{X} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

From Theorem 1 we have:

$$\frac{\partial \lambda_{aver}(0)}{\partial p_1} = \frac{1}{2} \operatorname{tr} \left[ \begin{pmatrix} 2 & 1 \\ 0 & 1 \end{pmatrix} - \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \right] = \frac{1}{2} \operatorname{tr} \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} = \frac{3}{2},$$

$$\frac{\partial \lambda_{aver}(0)}{\partial p_2} = \frac{1}{2} \operatorname{tr} \left[ \begin{pmatrix} 1 & 1 \\ 2 & 0 \end{pmatrix} - \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \right] = \frac{1}{2} \operatorname{tr} \begin{pmatrix} 1 & 1 \\ 2 & 0 \end{pmatrix} = \frac{1}{2},$$

$$\frac{\partial \mathbf{X}_1(0)}{\partial p_1} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \Theta_1^{-1} \begin{pmatrix} 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 1 & -1 \end{pmatrix},$$

$$\frac{\partial \mathbf{X}_1(0)}{\partial p_2} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \Theta_1^{-1} \begin{pmatrix} 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 1 \end{pmatrix},$$

$$\frac{\partial \mathbf{X}_2(0)}{\partial p_1} = \Theta_2^{-1} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \end{pmatrix} - \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -1 & 1 \\ 0 & 0 & 0 \end{pmatrix},$$

$$\frac{\partial \mathbf{X}_2(0)}{\partial p_2} = \Theta_2^{-1} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & -1 \\ 0 & 0 & 1 \end{pmatrix}.$$

## 5. Conclusion

The derivatives of eigenvalues and eigenvectors, which characterize the tendency of variation for frequencies and mode shapes with respect to design parameters, are widely used in many applications. For example, one of the most commonly used methods in damage detection of bridge is vibration test analysis method, whose idea is to discrete bridge by finite element method or finite difference method and convert the dynamic model of the bridge structure to computing the eigensensitivity of a damped system. This paper proposes the theoretical analysis on sensitivity of defective multiple eigenvalues. For generalized eigenvalue problem, the derivatives of the average eigenvalues and the corresponding eigenvector matrices (Theorem 1) are both obtained by the equivalence of generalized eigenvalue problem and standard eigenvalue problem when mass matrix is invertible. For quadratic eigenvalue problem, we can translate quadratic eigenvalue problem into generalized eigenvalue problem, then by the result of generalized eigenvalue problem, the average of eigenvalues is proved to be analytic, the derivatives of the average eigenvalues and the corresponding eigenvector matrices (Theorem 3) are obtained when the mass matrix is invertible.

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