# A Symmetric Structure-Preserving ΓQR Algorithm for Linear Response Eigenvalue Problems

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#### Abstract

In this paper, we present an efficient  $\Gamma QR$  algorithm for solving the linear response eigenvalue problem  $\mathscr{H}\boldsymbol{x} = \lambda \boldsymbol{x}$ , where  $\mathscr{H}$  is  $\boldsymbol{\Pi}^-$ -symmetric with respect to  $\Gamma_0 = \operatorname{diag}(I_n, -I_n)$ . Based on newly introduced  $\Gamma$ -orthogonal transformations, the  $\Gamma QR$  algorithm preserves the  $\boldsymbol{\Pi}^-$ -symmetric structure of  $\mathscr{H}$  throughout the whole process, which guarantees the computed eigenvalues to appear pairwise  $(\lambda, -\lambda)$  as they should. With the help of a newly established implicit  $\Gamma$ -orthogonality theorem, we incorporate the implicit multi-shift technique to accelerate the convergence of the  $\Gamma QR$  algorithm. Numerical experiments are given to show the effectiveness of the algorithm.

**Keywords.**  $II^{\pm}$ -matrix,  $\Gamma$ -orthogonality, structure preserving,  $\Gamma$ QR algorithm, linear response eigenvalue problem

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### 1 Introduction

In this paper, we consider the standard eigenvalue problem of the form

$$\mathcal{H}\boldsymbol{x} \equiv \begin{bmatrix} A & B \\ -B & -A \end{bmatrix} \begin{bmatrix} \boldsymbol{x}_1 \\ \boldsymbol{x}_2 \end{bmatrix} = \lambda \boldsymbol{x}, \tag{1.1}$$

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where A and B are  $n \times n$  real symmetric matrices. We refer to it a linear response eigenvalue problem (LREP). Any complex scalar  $\lambda$  and nonzero 2n-dimensional column vector  $\boldsymbol{x}$  that satisfy (1.1) are called an eigenvalue and its associated eigenvector, respectively, and correspondingly,  $(\lambda, \boldsymbol{x})$  is called an eigenpair.

Our consideration of this problem is motivated by Casida's eigenvalue equations in [10, 15, 19, 22]. In computational quantum chemistry and physics, the excitation states and response properties of molecules and clusters are predicted by the linear-response time-dependent density functional theory. The excitation energies and transition vectors (oscillator strengths) of molecular systems can be calculated by solving Casida's eigenvalue equations [10, 15, 19]. There has been a great deal of recent work on and interest in developing efficient numerical algorithms and simulation techniques for computing excitation responses of molecules and for material designs in energy science [2, 3, 12, 13, 17, 18, 20, 21].

Let

$$\Gamma_0 = \begin{bmatrix} I_n & 0 \\ 0 & -I_n \end{bmatrix}, \quad \Pi \equiv \Pi_{2n} = \begin{bmatrix} 0 & I_n \\ I_n & 0 \end{bmatrix}.$$
(1.2)

The matrix  $\mathcal{H}$  in (1.1) satisfies

$$\Gamma_0 \mathcal{H} = \begin{bmatrix} A & B \\ B & A \end{bmatrix} \text{ and } \mathcal{H} \Pi = -\Pi \mathcal{H}.$$
(1.3)

As a result of the second equation in (1.3), if  $(\lambda, \mathbf{x})$  is an eigenpair of  $\mathcal{H}$ , i.e.,  $\mathcal{H}\mathbf{x} = \lambda \mathbf{x}$ , then  $(-\lambda, \Pi \mathbf{x})$  is also an eigenpair of  $\mathcal{H}$ , and if also  $\lambda \notin \mathbb{R}$ , then  $(\bar{\lambda}, \bar{\mathbf{x}})$  and  $(-\bar{\lambda}, \Pi \bar{\mathbf{x}})$  are eigenpairs of  $\mathcal{H}$  as well, where  $\bar{\lambda}$  is the complex conjugate of  $\lambda$  and  $\bar{x}$  takes entrywise complex conjugation.

Previously in [2, 3, 23], LREP (1.1) was well studied under the condition that  $\Gamma_0 \mathcal{H}$  is positive definite. For the case, all eigenvalues of  $\mathcal{H}$  are real. Without the positive definite condition, the methods developed in [2, 3, 23] are not applicable.

Let  $\mathbb{J}_n$  be the set of all  $n \times n$  diagonal matrix with  $\pm 1$  on the diagonal and set

$$\Gamma_{2n} = \{ \operatorname{diag}(J, -J) : J \in \mathbb{J}_n \}.$$

Note that  $\Gamma_0 = \operatorname{diag}(I_n, -I_n) \in \Gamma_{2n}$ . In this paper, we will study an eigenvalue problem for which the condition that  $\Gamma_0 \mathcal{H}$  is positive definite is no longer assumed and it in fact includes LREP (1.1) as a special case. Specifically, we will consider the following eigenvalue problem

$$\mathcal{H}\boldsymbol{x} \equiv \begin{bmatrix} A & B \\ -B & -A \end{bmatrix} \boldsymbol{x} = \lambda \boldsymbol{x} \tag{1.4a}$$

with the structure property:

there is a 
$$\Gamma = \operatorname{diag}(J, -J) \in \Gamma_{2n}$$
 with  $J = \operatorname{diag}(\pm 1) \in \mathbb{J}_n$  such that  $\Gamma \mathcal{H} = \begin{bmatrix} JA & JB \\ JB & JA \end{bmatrix}$  with  $JA, JB \in \mathbb{R}^{n \times n}$  being symmetric. (1.4b)

There are two reasons for considering this more general eigenvalue problem (1.4). The first reason is that this includes (1.1), with/without  $\Gamma_0 \mathcal{H}$  being positive definite, as a special case, and the second one is that the intermediate eigenvalue problems in our later iterative QR-like algorithm for solving (1.1) are of this kind, i.e., with  $\Gamma \neq \Gamma_0$ .

It can be verified that the second equation in (1.3),  $\mathcal{H}\Pi = -\Pi\mathcal{H}$ , still holds in the case of (1.4b). Therefore the same results about the eigenvalue pattern we mentioned for (1.1) remain valid. Namely, if  $(\lambda, \boldsymbol{x})$  is an eigenpair, then  $(-\lambda, \Pi\boldsymbol{x})$  is also an eigenpair, and if also  $\lambda \notin \mathbb{R}$ , then  $(\bar{\lambda}, \bar{\boldsymbol{x}})$  and  $(-\bar{\lambda}, \Pi\bar{\boldsymbol{x}})$  are eigenpairs as well. Another interesting result is about the  $\Gamma$ -orthogonality among the eigenvectors of  $\mathcal{H}$ . Specifically, for two eigenpairs  $(\lambda, \boldsymbol{x})$  and  $(\mu, \boldsymbol{y})$  of  $\mathcal{H}$  if  $\lambda \neq \bar{\mu}$ , then it holds that  $\boldsymbol{y}^H \Gamma \boldsymbol{x} = 0$ , where  $\boldsymbol{y}^H$  is the conjugate transpose of  $\boldsymbol{y}$ . This is because using (1.4b), we have

$$\lambda \mathbf{y}^{\mathrm{H}} \Gamma \mathbf{x} = \mathbf{y}^{\mathrm{H}} \Gamma \mathcal{H} \mathbf{x} = \mathbf{y}^{\mathrm{H}} \mathcal{H}^{\mathrm{H}} \Gamma \mathbf{x} = \bar{\mu} \mathbf{y}^{\mathrm{H}} \Gamma \mathbf{x}$$

and thus  $(\lambda - \bar{\mu}) \mathbf{y}^{\mathrm{H}} \Gamma \mathbf{x} = 0$  which yields  $\mathbf{y}^{\mathrm{H}} \Gamma \mathbf{x} = 0$  when  $\lambda \neq \bar{\mu}$ .

The matrix  $\mathcal{H}$  in (1.4) has some nice block structures. In fact, the eigenvalue problem (1.4) can be written as a special Hamiltonian eigenvalue problem

$$\begin{bmatrix} 0 & JM \\ JK & 0 \end{bmatrix} \begin{bmatrix} \boldsymbol{y}_1 \\ \boldsymbol{y}_2 \end{bmatrix} = \lambda \begin{bmatrix} \boldsymbol{y}_1 \\ \boldsymbol{y}_2 \end{bmatrix} \text{ with } K = A - B, \quad M = A + B, \tag{1.5}$$

 $y_1 = x_1 - x_2$ , and  $y_2 = x_1 + x_2$ . There are several existing structure-preserving approaches [6, 8, 13, 16] can be applied to solve the eigenvalue problem (1.5).

- (a) A periodic QR (PQR) algorithm with orthogonal transformations [16] can be used to solve the product eigenvalue problem  $(JK)(JM)\mathbf{y}_2 = \lambda^2\mathbf{y}_2$  for (1.5). Here, the  $n \times n$  block structure in (1.4) is exploited and the symmetry of the spectrum can be preserved. However, the symmetric structures of JA and JB are destroyed during the iterations.
- (b) The KQZ algorithm [13] with orthogonal and  $\Pi$ -orthogonal transformations can be applied to solve  $\mathcal{H}$  in (1.1). The block structure  $\mathcal{H}\Pi = -\Pi\mathcal{H}$  is preserved during the KQZ iteration. The reduced matrices JA and JB in (1.4b) are no longer symmetric (tridiagonal), but only Hessenberg. It is mathematically different from the periodic QR algorithm, but they have the similar amount of computational costs.
- (c) An HR process proposed by Brebner and Grad (BG) [6] is used to reduce the product eigenvalue problem  $(JK)(JM)\mathbf{y}_2 = \lambda^2\mathbf{y}_2$  to a pseudosymmetric form  $C\mathbf{z} = \lambda^2 J'\mathbf{z}$ , where

C is symmetric tridiagonal and J' is the inertia sign-matrix of JK or JM. The tridiagonal pseudosymmetry in BG-algorithm are preserved during the HR iterations. The BG-algorithm has the similar amount of computational costs as our  $\Gamma$ QR algorithm. However, ill-conditioned K and M may cause the numerical instability of the BG-algorithm during constructing J'.

- (d) The symplectic QR-like algorithm with symplectic transformations [8] can also be applied to solve the special Hamiltonian eigenvalue problem (1.5). The Hamiltonian matrix is reduced to a condensed Hamiltonian form  $\begin{bmatrix} H_1 & H_3 \\ H^2 & -H_1 \end{bmatrix}$  with  $H_1$  and  $H_2$  being diagonal and  $H_3$  being symmetric triadiagonal, and then  $H_3$ -block will converge to a quasi-diagonal matrix. The symplectic QR-like algorithm does not really exploit the symmetry properties of JA and JB. Instability can occur when the symplectic Gaussian elimination matrices at some steps have larger condition numbers. This phenomenon can not be avoided because to maintain the Hamiltonian structure only the Gaussian elimination without pivot is allowed. Furthermore, the symplectic matrices Q in the symplectic QR-like algorithm satisfy  $Q^TJQ = J$ , where  $J = \Gamma_0 \Pi$ . The  $\Gamma$ -orthogonal matrices Q in our newly developed  $\Gamma QR$  algorithm (later) satisfy  $Q\Pi = \Pi Q$  and  $Q^T\Gamma Q = \Gamma'$  with  $\Gamma, \Gamma' \in \mathbf{\Gamma}_{2n}$ . The intersection of these two classes is the set of matrices Q that satisfy  $Q\Pi = \Pi Q$  and  $Q^T\Gamma Q = \Gamma'$  which is much smaller than two transformation sets.
- (e) The Hamiltonian QR-algorithm with symplectic orthogonal transformations proposed by [9] is only suitable for a very special Hamiltonian eigenvalue problem such as in  $(1.1)_{\text{one}}$  of rank(E), rank(E

The HR algorithm proposed in [7] is a pioneering work for solving the eigenvalue problem of an  $n \times n$  matrix  $\mathscr{A}$  having the property that there exists a so-called *pseudo-orthogonal* matrix H in the sense that  $H^{\top}JH = J'$  for some  $J = \operatorname{diag}(\pm 1)$  and  $J' = \operatorname{diag}(\pm 1)$  having the same inertia such that  $H^{-1}\mathscr{A}H = R$  is upper triangular. In light of the HR algorithm in [7], the main task of this paper is to develop iterative  $\Gamma QR$  algorithms for solving (1.4), while exploiting the inherent structures in  $\mathscr{H}$  and  $\Gamma$  for better numerical efficiency, based on  $(\Gamma, \Gamma')$ -orthogonal transformations with  $\Gamma, \Gamma' \in \Gamma_{2n}$  to be defined in Section 2. The transformations preserve the symmetry structures in  $\Gamma \mathscr{H}$  and the diagonal structure of  $\Gamma$ . Throughout this paper, we assume that  $\mathscr{H}$  is nonsingular, and thus 0 is not an eigenvalue of (1.1).

The rest of this paper is organized as follows. In Section 2, we introduce some basic definitions and state their immediately implied properties. In Section 3, we first give two kinds of  $\Gamma$ -orthogonal transformations, and then prove existence and uniqueness of the  $\Gamma QR$  factorization and propose an algorithm to compute the factorization for a given matrix G with  $G\Pi = -\Pi G$ . In Section 4, we present the  $\Pi^-$ -upper Hessenberg reduction/tridiagonalization and prove the implicit  $\Gamma$ -orthogonality theorem of a  $\Pi^-$ -matrix G. In Section 5, we develop  $\Gamma QR$  algorithms for computing all eigenpairs of  $\mathcal{H}$  and

analyze their convergence with the goal of an efficient implicit multi-shift  $\Gamma QR$  algorithm. Numerical results of the  $\Gamma QR$  algorithm compared to the other existing algorithm are shown in Section 6. Finally, some conclusions are drawn in Section 7.

**Notation**.  $\mathbb{C}^{n\times m}$  is the set of all  $n\times m$  complex matrices with real entries,  $\mathbb{R}^n=\mathbb{R}^{n\times 1}$ , and  $\mathbb{R}=\mathbb{R}^1$ . We denote by  $I_n$  and  $\mathbf{0}_{m\times n}(\mathbf{0}_m)$  the  $n\times n$  identity matrix and  $m\times n$   $(m\times m)$  zero matrix, respectively, and their subscripts may be dropped if their sizes can be read from the context.  $I_0$  and  $I_{2n}$  are reserved as given by (1.2), and often the subscript to  $I_{2n}$  is dropped, too, when no confusion is possible. The jth column of the identity matrix I is  $e_j$  whose size will be determined by the context. We shall also adopt MATLAB-like convention to access the entries of vectors and matrices. Let i:j be the set of integers from i to j inclusive. For a vector  $\mathbf{u}$  and a matrix X,  $\mathbf{u}_{(j)}$  is the jth entry of  $\mathbf{u}$  and  $X_{(i,j)}$  is the (i,j)th entry of X;  $X_{(\mathbb{I}_1,\mathbb{I}_2)}$  is the submatrix of X consist of intersections of all rows  $i \in \mathbb{I}_1$  and all columns  $j \in \mathbb{I}_2$ ;  $X^{\top}$  is the transpose of X. We denote by eig(A) the spectrum of matrix A, and diag(X, Y) is the 2-by-2 block-diagonal matrix with diagonal blocks X and Y.

### 2 Definitions and Preliminaries

In this section, we introduce several kinds of matrix classes and their essential properties. Recall  $\Pi_{2n}$  defined in (1.2).

**Definition 2.1.** Let  $G \in \mathbb{R}^{2n \times 2m}$  with  $m \leq n$ . G is called a  $\mathbf{\Pi}^{\pm}$ -matrix if i.e.,

$$G\Pi_{2m} = \pm \Pi_{2n}G$$
, i.e.,  $G = \begin{bmatrix} G_1 & G_2 \\ \pm G_2 & \pm G_1 \end{bmatrix}$  with  $G_1, G_2 \in \mathbb{R}^{n \times m}$ . (2.1)

Denote by  $\mathbf{\Pi}_{2n\times 2m}^{\pm}$  the set of all  $2n\times 2m$   $\mathbf{\Pi}^{\pm}$ -matrices, and  $\mathbf{\Pi}_{2n}^{\pm}:=\mathbf{\Pi}_{2n\times 2n}^{\pm}$  for short.

In this definition and those below, to save space and avoid repetitions, we often pack two statements into one: one for  $\Pi^+$ -matrix and the other for  $\Pi^-$ -matrix. In the same spirit, in definitions/statements in the rest of this paper, any of them with phrases in parentheses is understood that the phrases can be used to replace the phrases immediately before for another definition/statement.

We say a matrix X, possibly nonsquare, is upper Hessenberg if  $X_{(i,j)} = 0$  for i > j+1, upper triangular if  $X_{(i,j)} = 0$  for i > j, and diagonal if  $X_{(i,j)} = 0$  for  $i \neq j$ . This is consistent with the standard definitions of upper Hessenberg, upper triangular, and diagonal matrices which are usually for square matrices.

A quasi-upper triangular matrix means that it is a block upper triangular matrix with diagonal blocks being  $1 \times 1$  or  $2 \times 2$ . Similarly, a quasi-diagonal matrix means that it is a block diagonal matrix with diagonal blocks being  $1 \times 1$  or  $2 \times 2$ .

**Definition 2.2.** Let  $G \in \Pi^-_{2n \times 2m}$  as in (2.1).

- 1. G is  $\Pi^-$ -upper Hessenberg if  $G_1$  is upper Hessenberg and  $G_2$  is upper triangular. G is unreduced  $\Pi^-$ -upper Hessenberg if  $G_1$  is unreduced upper Hessenberg.
- 2. G is  $\Pi^-$ -upper ( $\Pi^-$ -quasi-upper) triangular if  $G_1$  is upper (quasi-upper) triangular and  $G_2$  is strictly upper triangular.

Denote by  $\mathbb{U}_{2n\times 2m}^-$  ( $q\mathbb{U}_{2n\times 2m}^-$ ) the set of all  $2n\times 2m$   $\mathbf{\Pi}^-$ -upper ( $\mathbf{\Pi}^-$ -quasi-upper) triangular matrices, and write, for short,  $\mathbb{U}_{2n}^- := \mathbb{U}_{2n\times 2n}^-$  and  $q\mathbb{U}_{2n}^- := q\mathbb{U}_{2n\times 2n}^-$ .

3. G is  $\Pi^-$ -diagonal ( $\Pi^-$ -quasi-diagonal) if  $G_1$  is diagonal (quasi-diagonal) and  $G_2$  is diagonal.

Denote by  $\mathbb{D}_{2n\times 2m}^-$  ( $q\mathbb{D}_{2n\times 2m}^-$ ) the set of all  $2n\times 2n$   $\mathbf{\Pi}^-$ -diagonal ( $\mathbf{\Pi}^-$ -quasidiagonal) matrices, and write, for short,  $\mathbb{D}_{2n}^- := \mathbb{D}_{2n\times 2n}^-$  and  $q\mathbb{D}_{2n}^- := q\mathbb{D}_{2n\times 2n}^-$ .

**Definition 2.3.** 1. Let  $G \in \Pi_{2n}^+$  as in (2.1). G is  $\Pi^+$ -symmetric ( $\Pi^+$ -sym-tridiagonal), if  $G_1$  is symmetric (symmetric tridiagonal) and  $G_2$  is symmetric (diagonal).

- 2. Let  $G \in \Pi_{2n}^-$  as in (2.1). G is  $\Pi^-$ -symmetric ( $\Pi^-$ -sym-tridiagonal) with respect to  $\Gamma \in \Gamma_{2n}$  if  $\Gamma G$  is  $\Pi^+$ -symmetric ( $\Pi^+$ -sym-tridiagonal).
- 3. G is unreduced  $\Pi^{\pm}$ -sym-tridiagonal if it is  $\Pi^{\pm}$ -sym-tridiagonal and  $G_1$  is unreduced.

The following propositions are direct consequences of Definitions 2.1 - 2.3 and are rather straightforward to verify.

**Proposition 2.1.** (i)  $G \in \Pi_{2n \times 2m}^{\pm}$  if and only if  $\Gamma G \in \Pi_{2n \times 2m}^{\mp}$  and  $G\Gamma' \in \Pi_{2n \times 2m}^{\mp}$  for any  $\Gamma \in \Gamma_{2n}$  and  $\Gamma' \in \Gamma_{2m}$ .

- (ii) The inverse of a  $\mathbf{H}^{\pm}$ -(upper triangular) matrix is still a  $\mathbf{H}^{\pm}$ -(upper triangular) matrix.
- (iii) The product  $\widetilde{G}G$  of two  $2n \times 2n$  matrices  $\widetilde{G}$  and G in their respective categories belongs to the one as listed in the following table.

don't need the other table? keep it anyway?

$\widetilde{G}$	$\Pi^+$	$\Pi^-$
$I\!\!I^+$	$II^+$	$\Pi^-$
$II^-$	$\Pi^-$	$\Pi^+$

**Proposition 2.2.** Let  $G \in \Pi_{2n}^-$ . Then G has 2n eigenvalues, appearing in pairs  $(\lambda, -\lambda)$  for real or purely imaginary eigenvalues  $\lambda$  and in quadruples  $(\pm \lambda, \pm \bar{\lambda})$  for complex eigenvalues  $\lambda$ .

*Proof.* By Definition 2.1, it holds that

$$\det(G - \lambda I) = \det(\Pi G - \lambda \Pi) = \det(G\Pi + \lambda \Pi) = \det(G + \lambda I) = 0.$$

The assertion follows immediately.

**Definition 2.4.**  $Q \in \Pi_{2n \times 2m}^+$  is  $\Gamma$ -orthogonal with respect to  $\Gamma \in \Gamma_{2n}$  if  $\Gamma' := Q^\top \Gamma Q \in \Gamma_{2m}$ . Denote by  $\mathbb{O}_{2n \times 2m}^{\Gamma}$  the set of all  $2n \times 2m$   $\Gamma$ -orthogonal matrices, and  $\mathbb{O}_{2n}^{\Gamma} := \mathbb{O}_{2n \times 2n}^{\Gamma}$  for short.

Often, for short, we may say  $Q \in \mathbf{\Pi}_{2n \times 2m}^+$  is  $\Gamma$ -orthogonal, by which we mean there is  $\Gamma \in \mathbf{\Gamma}_{2n}$  that has the requirement of the definition satisfied. Similarly, we may simply say Q is  $\Gamma$ -orthogonal. The same understanding applies to the expression  $Q \in \mathbb{O}_{2n \times 2m}^{\mathbf{\Gamma}}$ .

**Proposition 2.3.** Let  $Q_i \in \mathbb{O}_{2n}^{\Gamma}$  with respect to  $\Gamma_i \in \Gamma_{2n}$  for i = 1, 2, and suppose  $\Gamma_2 = Q_1^{\top} \Gamma_1 Q_1$ .

- (i)  $Q_1Q_2 \in \mathbb{O}_{2n}^{\mathbf{\Gamma}}$  with respect to  $\Gamma_1$ .
- (ii)  $Q_i$  is nonsingular and  $Q_i^{-1} = \Gamma_{i+1}Q_i^{\top}\Gamma_i$ , where  $\Gamma_3 := Q_2^{\top}\Gamma_2Q_2$ .
- (iii) If also  $Q_i \in \mathbb{U}_{2n}^+$ , then  $Q_i = J_i \oplus J_i$  for some  $J_i \in \mathbb{J}_n$ .

*Proof.* Items (i) and (ii) follow from Definition 2.4 directly. For item (iii),  $Q_i^{-1} \in \mathbb{U}_{2n}^+$  by Proposition 2.1(ii). On the other hand, by item (ii),  $Q_i^{-1} = \Gamma_{i+1}Q_i^{\top}\Gamma_i$  which is  $\mathbf{\Pi}^+$ -lower triangular. This implies that  $Q_i = J_i \oplus J_i$  for some  $J_i \in \mathbb{J}_n$ , completing the proof of item (iii).

# **3** ΓQR Factorization

**Definition 3.1.** G = QR is called a  $\Gamma QR$  factorization of  $G \in \mathbf{\Pi}_{2n \times 2m}^-$  with respect to  $\Gamma \in \mathbf{\Gamma}_{2n}$  if  $R \in \mathbb{U}_{2n \times 2m}^-$  and  $Q \in \mathbb{O}_{2n}^{\mathbf{\Gamma}}$  with respect to  $\Gamma$  or if  $R \in \mathbb{U}_{2m}^-$  and  $Q \in \mathbb{O}_{2n \times 2m}^{\mathbf{\Gamma}}$  with respect to  $\Gamma$ .

The case when  $R \in \mathbb{U}_{2m}^-$  and  $Q \in \mathbb{O}_{2n \times 2m}^{\mathbf{\Gamma}}$  with respect to  $\Gamma$  in this definition corresponds to the so-called *skinny* QR factorization in numerical linear algebra.

**Definition 3.2.** Let 
$$M = \begin{bmatrix} M_1 & M_2 \\ -M_2 & -M_1 \end{bmatrix} \in \boldsymbol{I}_{2n}^-$$
, and set

$$M_{1i} = (M_1)_{(1:i,1:i)}, \quad M_{2i} = (M_2)_{(1:i,1:i)}.$$

 $\begin{bmatrix} M_{1i} & M_{2i} \\ -M_{2i} & -M_{1i} \end{bmatrix}$ is called the *i*th  $\Pi^-$ -leading principal submatrix of M and its determinant is called the *i*th  $\Pi^-$ -leading principal minor of M.

The next theorem shows that almost every  $\Pi^-$ -matrix  $G \in \Pi^-_{2n \times 2m}$  has a  $\Gamma QR$  factorization with respect to a given  $\Gamma \in \Gamma_{2n}$  and the factorization is unique if it is required that the top-left quarter of the R-factor has positive diagonal entries.

**Theorem 3.1.** Suppose that  $G \in \Pi^-_{2n \times 2m}(m \le n)$  has full column rank and  $\Gamma \in \Gamma_{2n}$ .

(i) If  $G = QR = \widetilde{Q}\widetilde{R}$  (with Q,  $\widetilde{Q} \in \mathbb{O}_{2n \times 2m}^{\Gamma}$  and R,  $\widetilde{R} \in \mathbb{U}_{2m}^{-}$ ) are two  $\Gamma QR$  factorizations of G with respect to  $\Gamma$ , then

$$\widetilde{Q}^{\top} \Gamma \widetilde{Q} = Q^{\top} \Gamma Q \in \mathbf{\Gamma}_{2m}$$

and there is a  $\Pi^+$ -diagonal matrix  $D = J \oplus J$  with  $J \in \mathbb{J}_m$  such that  $\widetilde{Q} = QD$  and  $\widetilde{R} = DR$ . In particular, if the top-left quarters of R and  $\widetilde{R}$  have positive diagonal entries, then  $D = I_{2m}$ ,  $Q = \widetilde{Q}$ , and  $R = \widetilde{R}$ .

(ii) G has a  $\Gamma QR$  factorization with respect to  $\Gamma$  if and only if no  $\mathbf{\Pi}^-$ -leading principal minor of  $G^{\top}\Gamma G$  vanishes.

*Proof.* We first prove item (i). Let  $\Gamma' = Q^{\top} \Gamma Q$  and  $\widetilde{\Gamma}' = \widetilde{Q}^{\top} \Gamma \widetilde{Q}$ . From the assumption we have

$$\Gamma' R = Q^\top \Gamma Q R = Q^\top \Gamma \widetilde{Q} \widetilde{R} \quad \Rightarrow \quad Q^\top \Gamma \widetilde{Q} = \Gamma' R \widetilde{R}^{-1}.$$

Similarly,  $\widetilde{Q}^{\top} \Gamma Q = \widetilde{\Gamma}' \widetilde{R} R^{-1}$ . Therefore

$$\widetilde{\varGamma}'\widetilde{R}R^{-1} = \widetilde{Q}^{\top}\varGamma Q = (Q^{\top}\varGamma\widetilde{Q})^{\top} = (\varGamma'R\widetilde{R}^{-1})^{\top} = \widetilde{R}^{-\top}R^{\top}\varGamma'. \tag{3.1}$$

Because  $\widetilde{\Gamma}'\widetilde{R}R^{-1} \in \mathbb{U}_{2m}^-$  and at the same time  $\widetilde{R}^{-\top}R^{\top}\Gamma'$  is  $\mathbf{\Pi}^-$ -lower triangular, we conclude that  $\widetilde{R}R^{-1}$  and  $\widetilde{R}^{-\top}R^{\top}$  must be diagonal. Set

$$D = \widetilde{R}R^{-1} \in \mathbf{\Pi}_{2m}^+ \tag{3.2}$$

which implies  $\widetilde{R}^{-\top}R^{\top} = (\widetilde{R}R^{-1})^{-\top} = D^{-1}$ . Thus, from (3.1) and (3.2), we have  $\widetilde{\Gamma}'D = \widetilde{Q}^{\top}\Gamma Q$  and  $\Gamma'D^{-1} = Q^{\top}\Gamma\widetilde{Q}$ . This implies  $\Gamma' = D\widetilde{\Gamma}'D$ , and thus

$$D^2 = I_{2m}, \quad \Gamma' = \widetilde{\Gamma}' \in \Gamma_{2m}.$$

So  $D = \operatorname{diag}(J, -J)$  for some  $J \in \mathbb{J}_m$  and  $\widetilde{R} = DR$ . Furthermore, since  $G = QR = \widetilde{Q}\widetilde{R}$  has full column rank, it follows that  $\widetilde{Q} = QD$ .

Now if also the top-left quarters of R and R have positive diagonal entries, then  $\widetilde{R} = DR$  implies  $D = I_{2m}$ , as expected.

Next we prove item (ii).

**Necessity.** Let P be the permutation matrix

$$P = [\boldsymbol{e}_1, \boldsymbol{e}_3, \cdots, \boldsymbol{e}_{2m-1} | \boldsymbol{e}_2, \boldsymbol{e}_4, \cdots, \boldsymbol{e}_{2m}] \in \mathbb{R}^{2m \times 2m}.$$
(3.3)

Suppose that G = QR is a  $\Gamma QR$  factorization with respect to  $\Gamma$ , and let  $\Gamma' = Q^{\top} \Gamma Q$ . Then

$$P^{\top}G^{\top}\Gamma GP = (P^{\top}R^{\top}P)(P^{\top}\Gamma'P)(P^{\top}RP) =: R_{\mathbf{p}}^{\top}\Gamma'_{\mathbf{p}}R_{\mathbf{p}},$$

where  $R_{\mathbf{p}} = P^{\top}RP$  is upper triangular and  $\Gamma'_{\mathbf{p}} = P^{\top}\Gamma'P$  is diagonal, as in

$$R_{\mathbf{p}} = \begin{bmatrix} R_{11} & \cdots & R_{1m} \\ \vdots & \ddots & \vdots \\ 0 & \cdots & R_{mm} \end{bmatrix}, \quad \Gamma_{\mathbf{p}}' = \begin{bmatrix} \Gamma_{11}' & 0 \\ & \ddots & \\ 0 & \Gamma_{mm}' \end{bmatrix}$$
(3.4)

with  $R_{ij} \in \mathbf{\Pi}_2^-$ ,  $R_{ii} = \begin{bmatrix} d_i & 0 \\ 0 & -d_i \end{bmatrix}$ , and  $\Gamma'_{ii} \in \mathbf{\Gamma}_2$  for  $i, j = 1, \dots, m$ . Since G has full column rank, it follows that  $\det(R_{ii}) \neq 0$  for  $i = 1, \dots, m$ . Therefore, there is no leading principal minor of  $P^{\mathsf{T}}G^{\mathsf{T}}\Gamma GP$  of even order vanishes, i.e., no  $\mathbf{\Pi}^-$ -leading principal minor of  $G^{\mathsf{T}}\Gamma G$  vanishes.

**Sufficiency.** Suppose that  $G \in \mathbf{\Pi}_{2n \times 2m}^-$  and no  $\mathbf{\Pi}^-$ -leading principal minor of  $M := G^\top \Gamma G$  vanishes. Then there is an LU factorization of  $M_{\mathbf{p}} := P^\top M P$ :  $M_{\mathbf{p}} = L_{\mathbf{p}} \widehat{R}_{\mathbf{p}}$  with nonsingular

$$L_{\mathbf{p}} = \begin{bmatrix} I_{2} & & & 0 \\ L_{21} & I_{2} & & \\ \vdots & \ddots & \ddots & \\ L_{m1} & \cdots & L_{m,m-1} & I_{2} \end{bmatrix}, \quad \widehat{R}_{\mathbf{p}} = \begin{bmatrix} \widehat{R}_{11} & \cdots & \widehat{R}_{1m} \\ & \ddots & \vdots \\ 0 & & \widehat{R}_{mm} \end{bmatrix}, \tag{3.5}$$

where  $L_{ij} \in \mathbf{\Pi}_2^+$  and  $\widehat{R}_{ij} \in \mathbf{\Pi}_2^-$ . Decompose  $\widehat{R}_{\mathbf{p}}$  as

$$\widehat{R}_{\mathbf{p}} = \begin{bmatrix} \widehat{R}_{11} & 0 \\ & \ddots \\ 0 & \widehat{R}_{mm} \end{bmatrix} \begin{bmatrix} I_{2} & R_{12} & \cdots & R_{1m} \\ & \ddots & \ddots & \vdots \\ & & \ddots & R_{m-1,m} \\ & & & I_{2} \end{bmatrix} =: \widehat{D}_{\mathbf{p}} R_{\mathbf{p}}$$
(3.6)

with  $R_{ij} = \hat{R}_{ii}^{-1} \hat{R}_{ij} \in \mathbf{II}_2^+$ . Then we have

$$M_{\mathbf{p}} = L_{\mathbf{p}} \widehat{R}_{\mathbf{p}} = L_{\mathbf{p}} \widehat{D}_{\mathbf{p}} R_{\mathbf{p}} = R_{\mathbf{p}}^{\top} \widehat{D}_{\mathbf{p}}^{\top} L_{\mathbf{p}}^{\top} = M_{\mathbf{p}}^{\top}.$$
(3.7)

The uniqueness of the LU factorization implies that  $L_{\mathbf{p}}^{\top} = R_{\mathbf{p}}$ . Since  $M_{\mathbf{p}}$  is symmetric, it follows that  $\widehat{D}_{\mathbf{p}} = \operatorname{diag}(\{\widehat{R}_{ii}\}_{i=1}^m)$  is symmetric. Because  $\widehat{R}_{ii} \in \mathbf{\Pi}_2^-$ ,  $\widehat{R}_{ii}$  must be of the form  $\widehat{R}_{ii} = \begin{bmatrix} d_i & 0 \\ 0 & -d_i \end{bmatrix}$  for  $i = 1, \dots, m$ . Write

$$\widehat{R}_{ii} = \begin{bmatrix} \sqrt{|d_i|} & 0\\ 0 & \sqrt{|d_i|} \end{bmatrix} \begin{bmatrix} \operatorname{sgn}(d_i) & 0\\ 0 & -\operatorname{sgn}(d_i) \end{bmatrix} \begin{bmatrix} \sqrt{|d_i|} & 0\\ 0 & \sqrt{|d_i|} \end{bmatrix}$$

and denote

$$D_{\mathbf{p}}^{1/2} = \operatorname{diag}\left(\left\{\begin{bmatrix} \sqrt{|d_i|} & 0\\ 0 & \sqrt{|d_i|} \end{bmatrix}\right\}_{i=1}^m\right), \quad \Gamma_{\mathbf{p}}' = \operatorname{diag}\left(\left\{\begin{bmatrix} \operatorname{sgn}(d_i) & 0\\ 0 & -\operatorname{sgn}(d_i) \end{bmatrix}\right\}_{i=1}^m\right). \quad (3.8)$$

From (3.7) and (3.8) we have

$$G^{\top} \Gamma G = P M_{\mathbf{p}} P^{\top} = (P L_{\mathbf{p}} D_{\mathbf{p}}^{1/2} P^{\top}) (P \Gamma_{\mathbf{p}}^{\prime} P^{\top}) (P D_{\mathbf{p}}^{1/2} L_{\mathbf{p}}^{\top} P^{\top}) =: R^{\top} \Gamma^{\prime} R, \tag{3.9}$$

where  $R=PD_{\mathbf{p}}^{1/2}L_{\mathbf{p}}^{\top}P^{\top}\in\mathbb{U}_{2m}^{+}$  and  $\Gamma'=P\Gamma'_{\mathbf{p}}P^{\top}$ . Let

$$Q_{-} := GR^{-1}\Gamma' \in \mathbf{\Pi}_{2n \times 2m}^{+}. \tag{3.10}$$

With the help of (3.9), it can be verified that

$$Q_-^\top \Gamma Q_- = (\Gamma' R^{-\top} G^\top) \Gamma(G R^{-1} \Gamma') = \Gamma' R^{-\top} (R^\top \Gamma' R) R^{-1} \Gamma' = \Gamma'$$

which says  $Q_{-} \in \mathbb{O}_{2n \times 2m}^{\mathbf{\Gamma}}$ . Therefore

$$G = (GR^{-1}\Gamma')(\Gamma'R) =: Q_{-}R_{-} \text{ with } R_{-} \in \mathbb{U}_{2m}^{-}$$
 (3.11)

to give G = QR, a  $\Gamma QR$  factorization.

Our goal in this paper is to develop a structure-preserving QR-like algorithm to compute all eigenvalues of  $\mathcal{H} \in \mathbf{\Pi}_{2n}^-$ . The basic idea is to calculate a sequence of  $\Gamma$ -orthogonal matrices  $\{Q_i\}$ , based on a  $\Gamma$ QR factorization, such that

$$\mathscr{H}_{i+1} = Q_i^{-1} \mathscr{H}_i Q_i, \quad Q_i^{\top} \Gamma_i Q_i = \Gamma_{i+1} \quad \text{for } i = 1, 2, \dots,$$

where initially  $\mathcal{H}_0 = \mathcal{H}$ . For this purpose, at first, we introduce two elementary  $\Gamma$ orthogonal transformations which will be used to zero out a specific entry of a vector.
Specifically, given  $\Gamma \in \Gamma_{2n}$  and  $\mathbf{u} \in \mathbb{R}^{2n}$ , we seek  $Q \in \mathbb{Q}_{2n}^{\Gamma}$  to zero out some portion of  $\mathbf{u}$ . Two different kinds of matrices Q will be used to deal with all possible scenarios that will occur in computing the  $\Gamma QR$  factorizations in Algorithm 3.1 later.

Let  $\mathbf{a} \in \mathbb{R}^k$   $(1 \le k \le n)$ ,  $J \equiv \operatorname{diag}(j_1, ..., j_k) \in \mathbb{J}_k$ . Assume that  $\mathbf{a}^\top J \mathbf{a} \ne 0$ . Let  $P_a$  be a permutation which is chosen by interchanging row 1 and row r  $(2 \le r \le k)$  of J such that  $\hat{j}_1 \mathbf{a}^\top J \mathbf{a} = \hat{j}_1 \hat{\mathbf{a}}^\top \widehat{J} \hat{\mathbf{a}} > 0$ , where  $\hat{\mathbf{a}} = P_a \mathbf{a}$  and  $\widehat{J} = P_a J P_a = \operatorname{diag}(\hat{j}_1, ..., \hat{j}_k)$ . A Householder-like transformation is proposed by [7] to zero out the elements of  $\hat{\mathbf{a}}_{(2:k)}$  as follows. Let

$$H(\boldsymbol{a})^{-1} = I - \frac{\hat{\jmath}_1}{\beta} (\hat{\boldsymbol{a}} - \alpha e_1) (\hat{\boldsymbol{a}} - \alpha e_1)^{\top} \widehat{J}, \quad \widehat{H}(\boldsymbol{a})^{-1} = H(\boldsymbol{a})^{-1} P_a,$$
(3.12a)

where  $\alpha = -\text{sign}(\hat{\boldsymbol{a}}_{(1)})\sqrt{\hat{\jmath}_1\hat{\boldsymbol{a}}^{\top}\hat{\boldsymbol{J}}\hat{\boldsymbol{a}}}$  and  $\beta = \alpha[\alpha - \hat{\boldsymbol{a}}_{(1)}]$ . Then it can be verified that

$$\widehat{H}(\boldsymbol{a})^{-1}\boldsymbol{a} = [H(\boldsymbol{a})^{-1}P_a]\boldsymbol{a} = H(\boldsymbol{a})^{-1}\widehat{\boldsymbol{a}} = \alpha e_1, \quad \widehat{H}(\boldsymbol{a})^{\top}J\widehat{H}(\boldsymbol{a}) = \widehat{J}.$$
(3.12b)

Hyp\_Householder (hyperbolic Householder) transformation: Suppose  $1 \le \ell < m \le n$ ,  $\mathbf{u} \in \mathbb{R}^{2n}$  and  $\Gamma = \operatorname{diag}(\gamma_1, \dots, \gamma_n, -\gamma_1, \dots, -\gamma_n) \in \mathbf{\Gamma}_{2n}$ . There are two cases:

Case 1.  $\boldsymbol{a} \leftarrow \boldsymbol{u}_{(\ell':m')}$  with  $\ell' = n + \ell$  and m' = n + m,  $J = \operatorname{diag}(\gamma_{\ell}, \dots, \gamma_{m})$ ;

Case 2.  $\boldsymbol{a} \leftarrow \boldsymbol{u}_{(\ell:m)}, J = \operatorname{diag}(\gamma_{\ell}, \dots, \gamma_{m}).$ 

Using (3.12) we construct a hyperbolic Householder  $\Gamma$ -orthogonal transformation Q with respect to  $\Gamma$  through its inverse by  $Q_h(\ell':m';\boldsymbol{u}) \text{ not defined yet}$ 

$$Q^{-1} = \begin{cases} Q_h(\ell' : m'; \mathbf{u}), & \text{case 1;} \\ Q_h(\ell : m; \mathbf{u}), & \text{case 2,} \end{cases}$$
$$= \operatorname{diag}(I_{\ell-1}, \hat{H}(a)^{-1}, I_{n-m}, I_{\ell-1}, \hat{H}(a)^{-1}, I_{n-m}). \tag{3.13}$$

Then it holds that

$$Q^{-1}\mathbf{u} = \hat{\mathbf{u}} \text{ with } \begin{cases} \hat{\mathbf{u}}_{(\ell'+1:m')} = 0, & \text{case 1;} \\ \hat{\mathbf{u}}_{(\ell+1:m)} = 0, & \text{case 2,} \end{cases}$$
(3.14)

and  $Q^{\top} \Gamma Q = \Gamma'$ , where  $\Gamma' = (\gamma'_1, \dots, \gamma'_n, -\gamma'_1, \dots, -\gamma'_n)$  is given by

$$\begin{cases} \gamma'_{s} = \gamma_{s}, & s = 1, \dots, \ell - 1 \text{ and } m + 1, \dots, n, \\ \gamma'_{s+\ell-1} = \widehat{j}_{s}, & s = 1, \dots, m - \ell + 1. \end{cases}$$
 (3.15)

Hyp\_Givens (hyperbolic Givens) transformation: Suppose  $1 \le \ell \le n$ ,  $\mathbf{u} \in \mathbb{R}^{2n}$  and  $\Gamma = \operatorname{diag}(\gamma_1, \dots, \gamma_n, -\gamma_1, \dots, -\gamma_n) \in \Gamma_{2n}$ . Let  $\alpha \leftarrow \mathbf{u}_{(\ell)}$ ,  $\beta \leftarrow \mathbf{u}_{(n+\ell)}$ . Define

$$\begin{cases}
c = \frac{1}{\sqrt{1 - r^2}}, s = \frac{r}{\sqrt{1 - r^2}} & \text{with } r = \frac{\beta}{\alpha}, \text{ if } |\alpha| > |\beta|, \\
c = \frac{r}{\sqrt{1 - r^2}}, s = \frac{1}{\sqrt{1 - r^2}} & \text{with } r = \frac{\alpha}{\beta}, \text{ if } |\alpha| < |\beta|.
\end{cases}$$
(3.16)

We construct a hyperbolic Givens  $\Gamma$ -orthogonal transformation with respect to  $\Gamma$  through its inverse by

$$Q^{-1} = \mathcal{Q}_g(\ell; \alpha; \beta) = \begin{bmatrix} C & S \\ S & C \end{bmatrix} \in \Pi_{2n}^+, \tag{3.17}$$

where C is obtained from  $I_n$  by resetting  $C_{(\ell,\ell)}=c$  and S from  $O_{n\times n}$  by resetting  $S_{(\ell,\ell)}=-s$ . Then we have

$$Q^{-1}\boldsymbol{u} = \hat{\boldsymbol{u}}$$
 with  $\hat{\boldsymbol{u}}_{(n+\ell)} = 0$ ,

#### **Algorithm 3.1** ΓQR factorization

Input:  $G \in \mathbf{\Pi}_{2n \times 2m}^-$ ,  $\Gamma = \operatorname{diag}(J, -J) \in \mathbf{\Gamma}_{2n}$  with  $J = \operatorname{diag}(\gamma_1, \dots, \gamma_n)$ ,  $n' \leftarrow 2n$ ; **Output:**  $Q \in \mathbb{O}_{2n}^{\Gamma}$  with respect to  $\Gamma$ ,  $\Gamma' = Q^{\top} \Gamma Q \in \mathbb{J}_n$ , and  $R \in \mathbb{U}_{2m}^{-}$  such that G = QR;

- 1:  $Q \leftarrow I_{2n}, \ \Gamma_{\text{sav}} \leftarrow \Gamma;$
- 2: **for**  $\ell = 1 : m \$ **do**
- $\ell' \leftarrow n + \ell$ ,  $\boldsymbol{u} \leftarrow G_{(:,\ell)}$ ;
- compute Hyp\_Householder  $\Gamma$ -orthogonal transformation:  $\widetilde{Q}^{-1} = \mathcal{Q}_h(\ell':n';\boldsymbol{u})$  with respect to  $\Gamma$  (by (3.13));
- $Q \leftarrow Q\widetilde{Q}$ ,  $G \leftarrow \widetilde{Q}^{-1}G$ ,  $\Gamma \leftarrow -\Gamma'$  (by (3.15)); 5:
- $\alpha \leftarrow G_{(\ell,\ell)}, \beta \leftarrow G_{(\ell',\ell)};$ 6:
- compute Hyp\_Givens  $\Gamma$ -orthogonal transformation:  $\widetilde{Q}^{-1} = \mathcal{Q}_q(\ell; \alpha, \beta)$  with respect to  $\Gamma$  (by (3.17));
- $Q \leftarrow Q\widetilde{Q}$ ,  $G \leftarrow \widetilde{Q}^{-1}G$ ,  $\Gamma \leftarrow \Gamma'$  (by (3.18));
- 9:
- compute Hyp\_Householder  $\Gamma$ -orthogonal transformation:  $\widetilde{Q}^{-1} = \mathcal{Q}_h(\ell:n;\boldsymbol{u})$  with 10: respect to  $\Gamma$  (by (3.13));
- $Q \leftarrow Q\widetilde{Q}$ ,  $G \leftarrow \widetilde{Q}^{-1}G$ ,  $\Gamma \leftarrow \Gamma'$  (by (3.18));
- 12: end for

12: end for
13: return 
$$Q \leftarrow Q_{(:,[1:m,n+1:n+m])}$$
,  $\Gamma' \leftarrow \Gamma$ ,  $\Gamma \leftarrow \Gamma_{\text{sav}}$ , and  $R = \begin{bmatrix} G_{(1:m,:)} \\ G_{(n+1:n+m,:)} \end{bmatrix}$ .

and  $Q^{\top} \Gamma Q = \Gamma'$ , where

$$\begin{cases}
\gamma'_{\ell} = \delta \gamma_{\ell}, & \delta = c^2 - s^2 = \pm 1, \\
\gamma'_{j} = \gamma_{j}, & j \neq \ell.
\end{cases}$$
(3.18)

**Remark 3.1.** (i) Utilizing the special structure of  $\mathbf{\Pi}^-$ -matrix  $G \in \mathbf{\Pi}^-_{2n \times 2m}$ ,  $\widetilde{Q}^{-1}$  at lines 4, 7 and 10 of Algorithm 3.1 eliminates the  $(n + \ell + 1 : 2n, \ell)$  and  $(\ell + 1 : n, \ell)$ or the  $(n+\ell,\ell)$ th and  $(\ell,n+\ell)$ th or the  $(\ell+1:n,\ell)$  and  $(n+\ell+1:2n,\ell)$  entries of G simultaneously, for  $\ell = 1, \ldots, m$ . (ii) Upon exit, Algorithm 3.1 computes G = QR, where  $Q \in \mathbb{O}_{2n}^{\Gamma}$  is a  $\Gamma$ -orthogonal matrix with respect to  $\Gamma$  and  $R \in \mathbb{U}_{2n \times 2m}^-$ . It is worth noting that  $\Gamma'$  is unknown before G = QR is computed but it is unique, according to Theorem 3.1(i).

In the following, we use a small example with n=3 and m=2 by Wilkinson's

diagram to illustrate the elimination process in computing a  $\Gamma QR$  factorization of G.

In general, after m steps we have computed 3m  $\Gamma$ -orthogonal matrices  $\widetilde{Q}_1^{-1}, \ldots, \widetilde{Q}_{3m}^{-1}$  such what  $(\widehat{Q}_{3m}^{-1}, \ldots, \widehat{Q}_1^{-1})G = R$  is a  $\Pi$ -upper triangular.

Remark 3.2. Theorem 3.1(ii) reveals that almost all matrices in  $\mathbf{\Pi}_{2n\times 2m}^-(m\leq n)$  have  $\Gamma$ QR factorizations. In practice, for a given  $2n\times 2m$   $\mathbf{\Pi}^-$ -matrix G, one way to construct its  $\Gamma$ QR factorization with respect to given  $\Gamma\in\Gamma_{2n}$  is through reducing G to a  $2n\times 2m$   $\mathbf{\Pi}^-$ -upper triangular matrix by a sequence of  $\Gamma$ -orthogonal transformations: Hyp\_Householder  $\Gamma$ -orthogonal transformations and Hyp\_Givens  $\Gamma$ -orthogonal transformations. The Hyp\_Householder transformation in (3.12a) may not exist if  $\hat{\mathbf{a}}^{\mathrm{T}}J\hat{\mathbf{a}}=0$ . Similarly, the Hyp\_Givens transformation in (3.17) may not exist if  $|\alpha|=|\beta|$ . In [1], it is said that these cases can occur when the matrix is artificially designed. There is clearly a numerical stability issue if  $\hat{\mathbf{a}}^{\mathrm{T}}J\hat{\mathbf{a}}\approx 0$  or  $|\alpha|\approx |\beta|$ . The danger of severe cancellation can occurs and is discussed in [4, 5]. If a dangerous cancellation occurs at some  $\ell$ th step of Algorithm 3.1, it is recommended pre-multiply the current G by a randomly generated  $\Gamma$ -orthogonal  $\widetilde{Q}^{-1}$  with  $\widetilde{Q}^{\mathrm{T}}\Gamma\widetilde{Q}^{\mathrm{T}}=\Gamma'$ . Then we set  $G\leftarrow\widetilde{Q}^{-1}G$ ,  $\Gamma\leftarrow\Gamma'$ , and continue performing Algorithm 3.1 from the  $\ell$ th step. It usually can successfully circumvent the cancellation by this [4, 5].

# 4 Implicit $\Gamma$ -orthogonality Theorem

Here and hereafter we suppose that  $\mathscr{H} \in \Pi_{2n}^-$  is  $\Pi^-$ -symmetric with respect to  $\Gamma \in \Gamma_{2n}$ .

**Definition 4.1.** Given  $\mathbf{q}_1 \in \mathbb{R}^{2n}$  with  $\mathbf{q}_1^{\top} \Gamma \mathbf{q}_1 = \pm 1$ . For  $1 \leq m \leq n$ , the *m*th order  $\mathbf{\Pi}^+$ -Krylov matrix of  $\mathcal{H}$  on  $\mathbf{q}_1$  is defined as

$$K_{2m} \equiv K_{2m}(\mathcal{H}, \boldsymbol{q}_1)$$

$$:= \left[ \boldsymbol{q}_1, \mathcal{H} \boldsymbol{q}_1, \cdots, \mathcal{H}^{m-1} \boldsymbol{q}_1 \mid \Pi \boldsymbol{q}_1, \Pi(\mathcal{H} \boldsymbol{q}_1), \cdots, \Pi(\mathcal{H}^{m-1} \boldsymbol{q}_1) \right] \in \mathbb{R}^{2n \times 2m}. \quad (4.1)$$

Let  $\mathcal{K}_{2m} \equiv \mathcal{K}_{2m}(\mathcal{H}, \mathbf{q}_1)$  be the subspace spanned by the columns of  $K_{2m}$ .

**Theorem 4.1.** Let  $K_{2m} \equiv K_{2m}(\mathcal{H}, \mathbf{q}_1)$  be the  $\mathbf{\Pi}^+$ -Krylov matrix (4.1), where  $m \leq n$ . Suppose  $\operatorname{rank}(K_{2m}) = 2m$ , and let  $K_{2m} = Q_{2m}R_{2m}$  (with  $Q_{2m} \in \mathbb{O}_{2n \times 2m}^{\mathbf{\Gamma}}$  and  $R_{2m} \in \mathbb{U}_{2m}^{\mathbf{\Gamma}}$ ) be a  $\Gamma \operatorname{QR}$  factorization with respect to  $\Gamma$  and set  $\Gamma' = Q_{2m}^{\mathbf{\Gamma}} \Gamma Q_{2m} \in \mathbf{\Gamma}_{2m}$ . Then

$$\mathcal{H}Q_{2m} = Q_{2m}T_{2m} + \boldsymbol{z}_{m}\boldsymbol{e}_{m}^{\top} - \boldsymbol{\Pi}\boldsymbol{z}_{m}\boldsymbol{e}_{2m}^{\top}, \tag{4.2a}$$

$$Q_{2m}^{\top} \Gamma \boldsymbol{z}_m = Q_{2m}^{\top} \Gamma(\Pi \boldsymbol{z}_m) = 0, \tag{4.2b}$$

$$T_{2m} = (\Gamma' Q_{2m}^{\top} \Gamma) \mathcal{H} Q_{2m}, \tag{4.2c}$$

for some  $\mathbf{z}_m \in \mathbb{R}^{2n}$ , and  $\Gamma'T_{2m}$  is unreduced  $\mathbf{\Pi}^+$ -sym-tridiagonal, i.e.,  $T_{2m}$  is unreduced  $\mathbf{\Pi}^-$ -sym-tridiagonal with respect to  $\Gamma'$ .

*Proof.* Since  $K_{2m} = Q_{2m}R_{2m}$  has full column rank and is the  $2n \times 2m$   $\Pi^+$ -Krylov matrix by assumption,  $R_{2m}$  is  $\Pi^+$ -upper triangular and nonsingular and so is  $R_{2m}^{-1}$ . Using  $\mathcal{H}\Pi = -\Pi\mathcal{H}$ , we have

$$\mathcal{H}K_{2m} = \mathcal{H}\left[\boldsymbol{q}_{1}, \mathcal{H}\boldsymbol{q}_{1}, \cdots, \mathcal{H}^{m-1}\boldsymbol{q}_{1}, \Pi\boldsymbol{q}_{1}, \Pi(\mathcal{H}\boldsymbol{q}_{1}), \cdots, \Pi(\mathcal{H}^{m-1}\boldsymbol{q}_{1})\right]$$

$$= K_{2m}C_{2m} + \mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{m}^{\top} - \Pi(\mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{2m}^{\top}), \tag{4.3}$$

where  $C_{2m} = \operatorname{diag}(C_1, -C_1)$  with  $C_1 = \begin{bmatrix} \mathbf{0}_{m-1}^\top & 0 \\ I_{m-1} & \mathbf{0}_{m-1} \end{bmatrix}$ . Substituting  $K_{2m}$  by  $Q_{2m}R_{2m}$  into (4.3), we get

$$\mathcal{H}Q_{2m} = Q_{2m} \left[ \underbrace{R_{2m}C_{2m}R_{2m}^{-1} + \Gamma'Q_{2m}^{\top}\Gamma(\mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{m}^{\top} - \Pi\mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{2m}^{\top})R_{2m}^{-1}}_{=:T_{2m}} \right]$$

$$+ (I - Q_{2m}\Gamma'Q_{2m}^{\top}\Gamma)(\mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{m}^{\top} - \Pi\mathcal{H}^{m}\boldsymbol{q}_{1}\boldsymbol{e}_{2m}^{\top})R_{2m}^{-1}.$$

$$(4.4)$$

Set  $\gamma_{mm} = \boldsymbol{e}_m^{\top} R_{2m}^{-1} \boldsymbol{e}_m = \boldsymbol{e}_{2m}^{\top} R_{2m}^{-1} \boldsymbol{e}_{2m}$  and  $\boldsymbol{z}_m = \gamma_{mm} (I - Q_{2m} \Gamma' Q_{2m}^{\top} \Gamma) \mathcal{H}^m \boldsymbol{q}_1$ . From (4.4), we have

$$\mathscr{H}Q_{2m} = Q_{2m}T_{2m} + \boldsymbol{z}_{m}\boldsymbol{e}_{m}^{\top} - \boldsymbol{\Pi}\boldsymbol{z}_{m}\boldsymbol{e}_{2m}^{\top}. \tag{4.5}$$

From the fact that  $Q_{2m}^{\top} \Gamma Q_{2m} = \Gamma'$  it follows that

$$Q_{2m}^{\mathsf{T}} \Gamma \boldsymbol{z}_m = Q_{2m}^{\mathsf{T}} \Gamma(\Pi \boldsymbol{z}_m) = 0. \tag{4.6}$$

Therefore  $(\Gamma'Q_{2m}^{\top}\Gamma)\mathcal{H}Q_{2m} = T_{2m}$  by (4.5) and (4.6). Because  $C_{2m}$  in (4.4) is unreduced  $\Pi^-$ -upper Hessenberg, by Proposition 2.1 we know that  $T_{2m}$  is unreduced  $\Pi^+$ -upper Hessenberg. Furthermore, since  $\mathcal{H}$  is  $\Pi^-$ -symmetric with respect to  $\Gamma$ ,  $Q_{2m}^{\top}\Gamma\mathcal{H}Q_{2m}$  is  $\Pi^+$ -sym-tridiagonal and thus  $T_{2m}$  is  $\Pi^-$ -sym-tridiagonal with respect to  $\Gamma'$ . This completes the proof.

**Theorem 4.2.** Let  $Q_{2m} \in \mathbb{R}^{2n \times 2m}$   $(m \leq n)$  be a  $\Gamma$ -orthogonal with respect to  $\Gamma$  such that  $Q_{2m} \boldsymbol{e}_1 = \boldsymbol{q}_1$ . Let  $\Gamma' = Q_{2m}^{\top} \Gamma Q_{2m}$ . If  $Q_{2m}$  satisfies (4.2a) for some unreduced  $\Pi^+$ -sym-tridiagonal,  $\Gamma' T_{2m}$  and  $\boldsymbol{z}_m \in \mathbb{R}^{2n}$ , then

$$K_{2m}(\mathcal{H}, \mathbf{q}_1) = Q_{2m} \left[ \mathbf{e}_1, T_{2m} \mathbf{e}_1, \cdots, T_{2m}^{m-1} \mathbf{e}_1, \Pi \mathbf{e}_1, \Pi(T_{2m} \mathbf{e}_1), \cdots, \Pi(T_{2m}^{m-1} \mathbf{e}_1) \right]$$

$$=: Q_{2m} R_{2m}$$

$$(4.7)$$

is a  $\Gamma QR$  factorization of  $K_{2m}(\mathcal{H}, \mathbf{q}_1)$  and  $\operatorname{rank}(K_{2m}(\mathcal{H}, \mathbf{q}_1)) = 2m$ .

*Proof.* It holds that

$$\mathcal{H}\boldsymbol{q}_1 = \mathcal{H}Q_{2m}\boldsymbol{e}_1 = (Q_{2m}T_{2m} + \boldsymbol{z}_m\boldsymbol{e}_m^\top - \boldsymbol{\Pi}\boldsymbol{z}_m\boldsymbol{e}_{2m}^\top)\boldsymbol{e}_1 = Q_{2m}T_{2m}\boldsymbol{e}_1. \tag{4.8}$$

Assume that  $\mathscr{H}^{i-1}\boldsymbol{q}_1 = \mathscr{H}^{i-1}Q_{2m}\boldsymbol{e}_1 = Q_{2m}T_{2m}^{i-1}\boldsymbol{e}_1$  holds for  $i=2,\cdots,m-1$ . Then

$$egin{aligned} \mathscr{H}^i oldsymbol{q}_1 &= \mathscr{H} Q_{2m} T_{2m}^{i-1} oldsymbol{e}_1 \ &= (Q_{2m} T_{2m} + oldsymbol{z}_m oldsymbol{e}_m^ op - \Pi oldsymbol{z}_m oldsymbol{e}_{2m}^ op) T_{2m}^{i-1} oldsymbol{e}_1 \ &= Q_{2m} T_{2m}^i oldsymbol{e}_1 + oldsymbol{z}_m oldsymbol{e}_m^ op T_{2m}^{i-1} oldsymbol{e}_1 - \Pi oldsymbol{z}_m oldsymbol{e}_{2m}^ op T_{2m}^{i-1} oldsymbol{e}_1. \end{aligned}$$

It can be verified that  $\mathbf{e}_m^{\top} T_{2m}^{i-1} \mathbf{e}_1 = \mathbf{e}_{2m}^{\top} T_{2m}^{i-1} \mathbf{e}_1 = 0$ . Therefore, we have  $\mathscr{H}^i \mathbf{q}_1 = Q_{2m} T_{2m}^i \mathbf{e}_1$  and thus (4.7) holds. Furthermore,  $R_{2m}$  in (4.7) is nonsingular and  $\mathbf{\Pi}^+$ -upper triangular. Hence, rank $(K_{2m}(\mathscr{H}, \mathbf{q}_1)) = 2m$ .

**Theorem 4.3** (Implicit  $\Gamma$ -orthogonality theorem). Let  $\mathscr{H} \in \Pi_{2n}^-$  be  $\Pi^-$ -symmetric with respect to  $\Gamma$  and Q,  $\widetilde{Q} \in \mathbb{R}^{2n \times 2n}$  be two  $\Gamma$ -orthogonal  $\Pi^+$ -matrices with respect to  $\Gamma \in \Gamma_{2n}$ . Assume  $Q\mathbf{e}_1 = \widetilde{Q}\mathbf{e}_1$  and let

$$\Gamma' = Q^{\top} \Gamma Q, \quad \widetilde{\Gamma}' = \widetilde{Q}^{\top} \Gamma \widetilde{Q}.$$

If  $\mathscr{H}Q = QT_{2n}$  and  $\mathscr{H}\widetilde{Q} = \widetilde{Q}\widetilde{T}_{2n}$ , where  $\Gamma'T_{2n}$  and  $\widetilde{\Gamma}'\widetilde{T}_{2n}$  are unreduced  $\Pi^+$ -symtridiagonal, i.e.,  $T_{2n}$  and  $\widetilde{T}_{2n}$  are unreduced  $\Pi^-$ -sym-tridiagonal with respect to  $\Gamma'$  and  $\widetilde{\Gamma}'$ , respectively, then  $Q = \widetilde{Q}D$ ,  $\Gamma' = \widetilde{\Gamma}'$ , and  $T_{2n} = D\widetilde{T}_{2n}D$  for some  $D = \operatorname{diag}(J,J)$  with  $J \in \mathbb{J}_n$ .

*Proof.* By Theorem 4.2, it holds that

$$K_{2n}(\mathcal{H}, Q\mathbf{e}_1) = QR = \widetilde{Q}\widetilde{R} = K_{2n}(\mathcal{H}, \widetilde{Q}\mathbf{e}_1),$$

where R and  $\widetilde{R}$  are nonsingular and  $\Pi^+$ -upper triangular. From Theorem 3.1(i), we know that the  $\Gamma QR$  factorization of the nonsingular  $K_{2n}(\mathcal{H}, Q\mathbf{e}_1)$  is unique modulo some  $\Pi^+$ -diagonal  $D = \operatorname{diag}(J, J)$  with  $J \in \mathbb{J}_n$ . Thus,  $\widetilde{Q} = QD$ ,  $\widetilde{R} = DR$ , and consequently

$$\widetilde{\Gamma}' = \widetilde{Q}^{\top} \Gamma \widetilde{Q} = D Q^{\top} \Gamma Q D = D \Gamma' D = \Gamma',$$

$$\widetilde{T}_{2n} = \widetilde{\Gamma}' \widetilde{Q}^{\top} \Gamma \mathscr{H} \widetilde{Q} = \Gamma' D Q^{\top} \Gamma \mathscr{H} Q D = D (\Gamma' Q^{\top} \Gamma) \mathscr{H} Q D = D T_{2n} D,$$

as expected.

# 5 $\Gamma QR$ Algorithms

Based on  $\Gamma$ QR factorizations of  $\mathbf{H}^-$ -matrices and inspired by the usual QR algorithm for the standard eigenvalue problem [11], in this section, we will develop structure-preserving  $\Gamma$ QR algorithms for solving the LREP (1.4) of a  $\mathbf{H}^-$ -symmetric matrix  $\mathscr{H}$  with respect to  $\Gamma \in \Gamma_{2n}$ . A straightforward extension of the usual QR algorithm is outlined in Algorithm 5.1.

#### **Algorithm 5.1** The simple $\Gamma QR$ algorithm

Input:  $\mathcal{H} \in \Pi_{2n}^-$  a  $\Pi^-$ -symmetric matrix with respect to  $\Gamma \in \Gamma_{2n}$ ; Output:  $\Gamma'\mathcal{H} \in q\mathbb{D}_{2n}^+$  ( $\Pi^+$ -quasi-diagonal).

- 1:  $\mathcal{H}_1 \leftarrow \mathcal{H}, \Gamma_1 \leftarrow \Gamma, i = 1$ ;
- 2: repeat
- 3: compute the  $\Gamma QR$  factorization with respect to  $\Gamma_i$ :  $\mathcal{H}_i = Q_i R_i$ ;
- 4:  $\Gamma_{i+1} \leftarrow Q_i^{\top} \Gamma_i Q_i \in \boldsymbol{\Gamma}_{2n}, \, \mathscr{H}_{i+1} \leftarrow R_i Q_i;$
- 5:  $i \leftarrow i + 1$ ;
- 6: **until** convergence  $\mathcal{H} \leftarrow \mathcal{H}_i$ .

**Proposition 5.1.** The following statements holds for Algorithm 5.1.

(i) 
$$Q_i^{-1} = \Gamma_{i+1} Q_i^{\top} \Gamma_i$$
,  $\mathcal{H}_{i+1} = Q_i^{-1} \mathcal{H}_i Q_i = R_i \mathcal{H}_i R_i^{-1}$ ,  $\Gamma_{i+1} \mathcal{H}_{i+1} = Q_i^{\top} (\Gamma_i \mathcal{H}_i) Q_i$ .

(ii) 
$$\mathcal{H}_{i+1} = (Q_1 \cdots Q_i)^{-1} \mathcal{H}(Q_1 \cdots Q_i), \ \Gamma_{i+1} \mathcal{H}_{i+1} = (Q_1 \cdots Q_i)^{\top} (\Gamma \mathcal{H})(Q_1 \cdots Q_i).$$
  
Furthermore,  $\mathcal{H}_i$  is  $\Pi^-$ -symmetric with respect to  $\Gamma_i$ .

(iii) 
$$\mathcal{H}_{i+1} = (R_i \cdots R_1) \mathcal{H}(R_i \cdots R_1)^{-1}, \ \Gamma_{i+1} = (Q_1 \cdots Q_i)^{\top} \Gamma(Q_1 \cdots Q_i).$$

(iv) 
$$\mathcal{H}^i = (Q_1 \cdots Q_i)(R_i \cdots R_1).$$

Let the spectral decomposition of  $\mathcal{H}$  be

$$\mathcal{H}X = X\Lambda \text{ with } \Lambda \in q\mathbb{D}_{2n}^-, X \in \mathbf{\Pi}_{2n}^+.$$
 (5.1a)

For P as in (3.3), we have

$$P^{\top} \Lambda P = \operatorname{diag}(\Lambda_1, \dots, \Lambda_{\ell}) \text{ with } \operatorname{eig}(\Lambda_i) = \{\pm \lambda_i\}, \text{ or } \operatorname{eig}(\Lambda_i) = \{\pm \lambda_i, \pm \bar{\lambda}_i\}.$$
 (5.1b)

In light of the convergence proof of the HR algorithm in [7], we can also prove the convergence of Algorithm 5.1.

**Theorem 5.1.** Given  $\Gamma \in \Gamma_{2n}$ , let  $\mathscr{H} \in \Pi_{2n}^-$  be  $\Pi^-$ -symmetric with respect to  $\Gamma$  having the spectral decomposition (5.1). Suppose  $|\lambda_1| > \cdots > |\lambda_\ell| > 0$  and Algorithm 5.1 is executable for  $\mathscr{H}$  in the sense that all  $\Gamma QR$  factorizations at line 3 exist. If the  $\Pi^+$ -LU factorization of  $X^{-1} = L_{\mathbf{x}}U_{\mathbf{x}}$  exists, where  $L_{\mathbf{x}}^{\top}$ ,  $U_{\mathbf{x}} \in q\mathbb{U}_{2n}^+$  with  $L_{\mathbf{x}}$  having  $1 \times 1$  and/or  $2 \times 2$  unit diagonal blocks, conforming to the block structure of  $\Lambda$  and if the  $\Gamma QR$  factorization of  $X = Q_{\mathbf{x}}R_{\mathbf{x}}$  with respect to  $\Gamma$  exists, then  $\mathscr{H}_i$  in Algorithm 5.1 converges to a  $\Pi^-$ -quasi-diagonal matrix with its eigenvalues  $\lambda_i$  emerging in the order of  $\lambda_1, \lambda_2, \ldots, \lambda_\ell$ , as  $i \to \infty$ .

Proof. From the assumption, it follows that  $\Lambda^i L_{\mathbf{x}} \Lambda^{-i} = I + E_i$  with  $E_i \to 0$  as  $i \to \infty$ . Let  $(I + R_{\mathbf{x}} E_i R_{\mathbf{x}}^{-1}) = \widetilde{Q}_i \widetilde{R}_i$  be the  $\Gamma QR$  factorization with respect to  $\Gamma' := Q_{\mathbf{x}}^{\top} \Gamma Q_{\mathbf{x}}$ , and set  $\widetilde{\Gamma}_{i+1} := \widetilde{Q}_i^{\top} \Gamma' \widetilde{Q}_i \in \mathbf{\Gamma}_{2n}$ . It holds that  $\widetilde{Q}_i \to I_{2n}$  because  $E_i \to 0$ .

For i sufficiently large, we have

$$\begin{split} \mathscr{H}^i &= X \varLambda^i X^{-1} = X \varLambda^i L_{\mathbf{x}} \varLambda^{-i} \varLambda^i U_{\mathbf{x}} \\ &= X (I + E_i) \varLambda^i U_{\mathbf{x}} \\ &= Q_{\mathbf{x}} (I + R_{\mathbf{x}} E_i R_{\mathbf{x}}^{-1}) R_{\mathbf{x}} \varLambda^i U_{\mathbf{x}} \\ &= (Q_{\mathbf{x}} \widetilde{Q}_i) \widetilde{R}_i R_{\mathbf{x}} \varLambda^i U_{\mathbf{x}} \end{split}$$

which gives a  $\Gamma QR$  factorization of  $\mathcal{H}^i$ . On the other hand,  $\mathcal{H}^i = (Q_1 \cdots Q_i)(R_i \cdots R_1)$  by Proposition 5.1(ii). Now apply the uniqueness of the  $\Gamma QR$  factorization as stated in Theorem 3.1(i) to conclude that there is  $D_i = \operatorname{diag}(J_i, J_i)$  with  $J_i \in \mathbb{J}_n$  such that

$$(Q_{\mathbf{x}}\widetilde{Q}_i)D_i = Q_1 \cdots Q_i, \quad D_i(\widetilde{R}_i R_{\mathbf{x}} \Lambda^i U_{\mathbf{x}}) = R_i \cdots R_1,$$

and  $\widetilde{\Gamma}_{i+1} = \Gamma_{i+1}$ . By Proposition 5.1(ii), we have

$$\mathscr{H}_{i+1} = D_i \widetilde{Q}_i^{-1} Q_{\mathbf{x}}^{-1} \mathscr{H} Q_{\mathbf{x}} \widetilde{Q}_i D_i = D_i \widetilde{Q}_i^{-1} R_{\mathbf{x}} \Lambda R_{\mathbf{x}}^{-1} \widetilde{Q}_i D_i$$

which converges to a  $\Pi^-$ -quasi-diagonal matrix with its eigenvalues emerging in the order of  $\lambda_1, \lambda_2, \dots, \lambda_\ell$ , as  $i \to \infty$ .

# Algorithm 5.2 $\Pi^-$ -sym-tridiagonalization

```
Input: \Pi^--symmetric matrix \mathscr{H} with respect to \Gamma = \operatorname{diag}(J, -J) with J = \operatorname{diag}(\gamma_1, \dots, \gamma_n) \in \mathbb{J}_n;
```

**Output:**  $Q \in \mathbb{O}_{2n}^{\Gamma}$  with respect to  $\Gamma$ ,  $\Gamma' = Q^{\top} \Gamma Q \in \Gamma_{2n}$ , and  $\mathscr{H}' = Q^{-1} \mathscr{H} Q$  is a  $\Pi^-$ -sym-tridiagonal matrix with respect to  $\Gamma'$ .

```
1: Q \leftarrow I_{2n}, \ \Gamma_{\text{sav}} \leftarrow \Gamma, \ \mathscr{H}_{\text{sav}} \leftarrow \mathscr{H}, \ n' \leftarrow 2n;
  2: for \ell = 1 : n - 1 do
             \ell' \leftarrow n + \ell + 1, \, \boldsymbol{u} \leftarrow G_{(:,\ell)};
             compute Hyp_Householder \Gamma-orthogonal transformation \widetilde{Q}^{-1} = (\mathcal{Q}_h)_{(\ell':n',n)} with
             respect to \Gamma (by (3.13)); Q \leftarrow Q\widetilde{Q}, \mathcal{H} \leftarrow \widetilde{Q}^{-1}\mathcal{H}\widetilde{Q}, \Gamma \leftarrow -\Gamma' (by (3.15));
             \alpha \leftarrow \mathscr{H}_{(\ell+1,\ell)}, \ \beta \leftarrow \mathscr{H}_{(\ell',\ell)};
  6:
             compute Hyp_Givens \Gamma-orthogonal transformation \widetilde{Q}^{-1} = (\mathcal{Q}_q)_{(\ell+1:\alpha,\beta)} with re-
             spect to \Gamma (by (3.17));
             \widetilde{Q} \leftarrow Q\widetilde{Q}, \mathscr{H} \leftarrow \widetilde{Q}^{-1}\mathscr{H}\widetilde{Q}, \Gamma \leftarrow \Gamma' \text{ (by (3.18))};
  8:
             \boldsymbol{u} \leftarrow \mathscr{H}_{(:,\ell)};
  9:
             Compute Hyp_Householder \Gamma-orthogonal transformation \widetilde{Q}^{-1} = (\mathcal{Q}_h)_{(\ell+1:n:n)} with
10:
             respect to \Gamma (by (3.13));
             Q \leftarrow Q\widetilde{Q}, \mathcal{H} \leftarrow \widetilde{Q}^{-1}\mathcal{H}\widetilde{Q}, \Gamma \leftarrow \Gamma' \text{ (by (3.18))};
12: end for
13: return \Gamma' \leftarrow \Gamma, \Gamma \leftarrow \Gamma_{\text{sav}}, \mathcal{H}' \leftarrow \mathcal{H}, \mathcal{H} \leftarrow \mathcal{H}_{\text{sav}};
Note that we set (Q_h)_{(\ell:m;\boldsymbol{u})} = I if \ell = m.
```

check red text

In what follows, using Algorithm 5.1 as the basis, we focus on developing an efficient structure-preserving  $\Gamma$ QR algorithm to solve the eigenvalue problem (1.4) for the  $\Pi^-$ -symmetric matrix  $\mathscr{H}$  with respect to a given  $\Gamma$  such as  $\Gamma_0$  in (1.2). To do so, we first reduce  $\mathscr{H}$  to its  $\Pi^-$ -sym-tridiagonal form with respect to  $\Gamma$  and then use the two special  $\Gamma$ -orthogonal transformations described in section 3 to implicitly carry out lines 3 and 4 in Algorithm 5.1. The first phase, the  $\Pi^-$ -sym-tridiagonalization, is given in Algorithm 5.2. To illustrate the elimination process in the  $\Pi^-$ -sym-tridiagonalization, we trace actions on a small example with n=4 in Table 5.1.

Table 5.1: The  $\Pi^-$ -sym-tridiagonalization for n=4

In general, after n-1 step in  $\mathbf{\Pi}^-$ -sym-tridiagonalization, we have computed 3n-2  $\Gamma$ -orthogonal matrices  $\widetilde{Q}_1,\cdots,\widetilde{Q}_{3n-2}$  such that

$$(\widetilde{Q}_1 \cdots \widetilde{Q}_{3n-2})^{-1} \mathcal{H}(\widetilde{Q}_1 \cdots \widetilde{Q}_{3n-2}) = \mathcal{H}'$$

is  $\Pi^-$ -sym-tridiagonal with respect to  $\Gamma'$ .

As in the usual QR algorithm, the shift technique should be incorporated to accelerate the convergence of the simple  $\Gamma$ QR algorithm – Algorithm 5.1. By Proposition 2.2, we choose the filtering polynomials p(x) as

$$\begin{cases} p(x) = (x - \lambda)(x + \lambda) & \text{for real or imaginary } \lambda, \\ p(x) = (x - \lambda)(x + \lambda)(x - \bar{\lambda})(x + \bar{\lambda}) & \text{for complex } \lambda \end{cases}$$
 (5.2)

to ensure  $p(\mathscr{H}) \in \mathcal{I}_{2n}^+$ . On the other hand, from Theorem 4.3, because of the uniqueness of the  $\mathcal{I}_{2n}^-$ -sym-tridiagonalization of  $\mathscr{H}$ , the  $\Gamma QR$  factorization can be performed without explicitly computing the  $\Gamma QR$  factorization of  $p(\mathscr{H})$ . It only needs to construct a  $\Gamma$ -orthogonal transformation Q to reduce the first column of  $p(\mathscr{H})$  to a vector parallel to  $e_1$ . We outline the implicit multi-shift  $\Gamma QR$  algorithm in Algorithm 5.3.

**Remark 5.1.** (i) In Algorithm 5.3, lines 11–13 can be executed in two substeps with  $p_1(x)$  and  $p_2(x)$ , respectively, where

$$p_1(x) = (x - \lambda_1)(x + \lambda_1), \quad p_2(x) = (x - \lambda_2)(x + \lambda_2)$$

for real or purely imaginary  $\lambda_1$  and  $\lambda_2$ , and

$$p_1(x) = (x - \lambda_1)(x - \bar{\lambda}_1), \quad p_2(x) = (x + \lambda_1)(x + \bar{\lambda}_1)$$

for complex  $\lambda_1$ . Doing so enables that all computations are done in the real arithmetics.

(ii) There are many other structure-preserving approaches as discussed in Section 1. Based on the characteristic analysis of these algorithms, we will compare the performance of the  $\Gamma$ QR algorithm with that of the PQR algorithm [16] in our numerical studies. The flop counts of the implicit multi-shift  $\Gamma$ QR algorithm and the PQR algorithm for a  $\mathbf{H}^-$ -symmetric matrix  $\mathscr{H}$  with respect to  $\Gamma$  are summarized in Table 5.2. In each phase, the  $\Gamma$ QR algorithm consumes less than the PQR algorithm, especially in each iterative step in Phase ii. This is because the PQR algorithm cannot take advantage of the symmetric structures in JA, JB of (1.4b) but has to treat them like an n-by-n nonsymmetric matrix.

#### Algorithm 5.3 Implicit Multi-shift ΓQR Algorithm

```
Input: \Pi^--symmetric matrix \mathcal{H} with respect to \Gamma = \operatorname{diag}(J, -J) with J =
        \operatorname{diag}(\gamma_1, \cdots, \gamma_n) \in \mathbb{J}_n, and tolerance \epsilon;
Output: Q \in \mathbb{O}_{2n}^{\mathbf{\Gamma}} with respect to \Gamma, \Gamma' = Q^{\top} \Gamma Q \in \mathbf{\Gamma}_{2n}, \Lambda = Q^{-1} \mathcal{H} Q \in q \mathbb{D}_{2n}^{-};
  1: \Gamma_{\text{sav}} \leftarrow \Gamma, \mathscr{H}_{\text{sav}} \leftarrow \mathscr{H}, Q \leftarrow I_{2n}, \Lambda_i \leftarrow \emptyset (i = 1, 2);
  2: while n > 2 do
             Use Algorithm 5.2 to perform \Pi^--sym-tridiagonalization: \mathscr{H} \leftarrow \widetilde{Q}^{-1}\mathscr{H}\widetilde{Q}, Q \leftarrow
              QQ, \Gamma \leftarrow \Gamma' \ (\Gamma' \text{ is an output of Algorithm 5.2});
             if |\mathscr{H}_{(n,n-1)}| < \epsilon(|\mathscr{H}_{(n-1,n-1)}| + |\mathscr{H}_{(n,n)}|) then
                  \Lambda_1 \leftarrow \operatorname{diag}(\mathscr{H}_{(n,n)}, \Lambda_1), \ \Lambda_2 \leftarrow \operatorname{diag}(\mathscr{H}_{(2n,n)}, \Lambda_2),
                  \mathbb{I} = [1, \cdots, n-1, n+1, \cdots, 2n-1], \ \mathcal{H} \leftarrow \mathcal{H}_{(\mathbb{I}, \mathbb{I})}, \ \Gamma \leftarrow \Gamma_{(\mathbb{I}, \mathbb{I})}, \ n \leftarrow n-1;
  6:
                  if |\mathcal{H}_{(n-1,n-2)}| < \epsilon(|\mathcal{H}_{(n-2,n-2)}| + |\mathcal{H}_{(n-1,n-1)}|) then
  7:
                        \begin{split} & \Lambda_1 \leftarrow \operatorname{diag}(\mathscr{H}_{([n-1,n],[n-1,n])},\Lambda_1), \ \Lambda_2 \leftarrow \operatorname{diag}(\mathscr{H}_{([2n-1,2n],[n-1,n])},\Lambda_2), \\ & \mathbb{I} = [1,\cdots,n-2,n+1,\cdots,2n-2], \ \mathscr{H} \leftarrow \mathscr{H}_{(\mathbb{I},\mathbb{I})}, \ \Gamma \leftarrow \Gamma_{(\mathbb{I},\mathbb{I})}, \ n \leftarrow n-2; \end{split}
  8:
                  else
  9:
                        \mathbb{I} = [n-1, n, 2n-1, 2n], H_4 = \mathcal{H}_{(\mathbb{I},\mathbb{I})}, \text{ and compute } eig(H_4) = \{\pm l_1, \pm l_2\};
 10:
                        \mathbf{h} = p(\mathcal{H})\mathbf{e}_1, where p(x) is as given in (5.2) (see also Remark 5.1(i)).
11:
                        construct the \Gamma-orthogonal transformation Q_1 such that Q_1^{-1}\mathbf{h} = h\mathbf{e}_1;
                        \mathcal{H} \leftarrow Q_1^{-1} \mathcal{H} Q_1, \ Q \leftarrow Q Q_1, \ \Gamma \leftarrow \Gamma';
13:
14:
             end if
15:
16: end while
17: \Lambda := \begin{bmatrix} \Lambda_1 & \Lambda_2 \\ -\Lambda_2 & -\Lambda_1 \end{bmatrix} \in q\mathbb{D}_{2n}^-, \ \Gamma' \leftarrow \Gamma, \ \Gamma \leftarrow \Gamma_{\text{sav}}, \ \mathscr{H} \leftarrow \mathscr{H}_{\text{sav}}.
```

Methods	Phase		Flop	
ΓQR	$\Pi^-$ -sym-tridiagonalization on $\mathscr H$ by Algorithm 8			
1 011	ii	one step of implicit double-shift ΓQR iteration (Al-	120n	
		gorithm 5.3)		
DOD	i	Hessenberg-triangular reduction by Householder	$11n^3$	
PQR		transformation on $JKJM$ in $(1.5)$		
	ii	ii one step of implicit double-shift PQR iteration [14]		

Table 5.2: The flop counts of the  $\Gamma QR$  algorithm and the PQR algorithm

# 6 Numerical Experiments

To test the efficiency of  $\Gamma QR$  algorithm, we borrow K and M in the numerical example of [2] for the sodium dimer Na<sub>2</sub> with order n=1862. They are symmetric positive definite. We then recover the Casida's eigenvalue problem as in (1.1) by  $A=\frac{1}{2}(K+M)-4.88I_n$  and  $B=\frac{1}{2}(K-M)$  with an excitation energy 4.88, and reset K=A-B, M=A+B. In Table 6.1, we list the CPU time by the  $\Gamma QR$  algorithm and PQR algorithm for the computation of eigenvalues of  $\mathscr{H}=\begin{bmatrix}A&B\\-B&-A\end{bmatrix}$  and KM, respectively.

All numerical computations are carried out by MATLAB Version 2014b, on a Mac-Book Pro with a 2.8GHz Intel Core i7 processor and 8GB RAM, with the unit machine roundoff  $\mathfrak{u}=2^{-53}=1.11\times10^{-16}$ .

Methods		Phase	Time (secs.)
ΓQR	i	$\Pi^-$ -sym-tridiagonalization of ${\mathscr H}$ with $Q$ accumulated	326.05
1 QII	ii	Implicit ΓQR algorithm (Algorithm 5.3)	399.23
$PQR^{1}$	i	Hessenberg-triangular reduction on $KM$	643.02
1 611	ii	Implicit PQR algorithm	2078.09

Table 6.1: The CPU time by  $\Gamma QR$  and PQR.

Table 6.1 shows that the PQR algorithm takes about 3.6 times as much the CPU time as  $\Gamma$ QR algorithm does. The  $\Gamma$ QR algorithm is much cheaper than the PQR algorithm, as expected from Table 5.2.

To demonstrate accuracies in computed approximations, we calculate the relative errors of eigenvalues and the normalized residual norms for the jth approximate eigenpair  $(\mu_j, \mathbf{z}_j)$ :  $(\mu_j, \mathbf{z}_j):$   $(\mu_j, \mathbf{z}_j):$  (

$$\epsilon(\mu_j) = \frac{|\mu_j - \lambda_j|}{|\lambda_j|} \quad \text{and} \quad r(\mu_j) = \frac{||\mathcal{H} \mathbf{z}_j - \mu_j \mathbf{z}_j||_1}{(||\mathcal{H}||_1 + |\mu_j|)||\mathbf{z}_j||_1},$$

where  $\lambda_i$  denotes the jth "exact" eigenvalue of  $\mathcal{H}$  obtained by MATLAB's function eig.

The approximate eigenpair  $(\mu_j, \mathbf{z}_j)$  is computed as follows: (1) apply the inverse iteration with the computed eigenvalue  $\mu_j^0$  (by Algorithm 5.3) as the shift to the  $\mathbf{\Pi}^-$ -symtridiagonal matrix (an output of Algorithm 5.2) to compute an approximate eigenpair  $(\tilde{\mu}_j, \tilde{\mathbf{z}}_j)$  of the  $\mathbf{\Pi}^-$ -sym-tridiagonal matrix, and (2) apply the inverse iteration again on  $(\tilde{\mu}_j, \tilde{Q}\tilde{\mathbf{z}}_j)$  to the original  $\mathbf{\Pi}^-$ -symmetric  $\mathscr{H}(\tilde{Q})$  is another output of Algorithm 5.2) to getwhy does the corresponding approximate eigenpair  $(\mu_j, \mathbf{z}_j)$  of the original  $\mathscr{H}$ . On the other hand, a different way from the approximate eigenpair of the PQR algorithm is computed by applying the inverse

<sup>&</sup>lt;sup>1</sup>jupiter.math.nctu.edu.tw/~wwlin/code/PQZ.zip

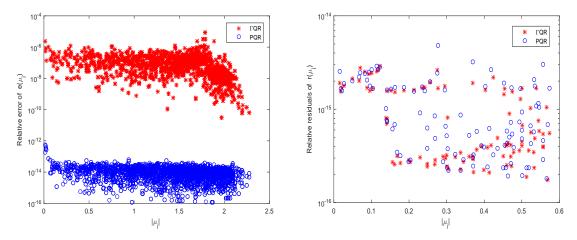


Figure 6.1: Left: relative errors of eigenvalues; Right: relative residual norm of eigenpairs.

iteration with the computed eigenvalue as the shift directly to the original  $\Pi^-$ -symmetric matrix  $\mathcal{H}$ .

Therefore, the dominant step for the computation of eigenpairs by the  $\Gamma QR$  algorithm or the PQR algorithm is the LU factorization of the matrix  $\mathcal{H} - \mu_j I$  for the inverse iteration. As before, the  $\Gamma QR$  algorithm is slightly more expensive than the PQR algorithm because an extra linear system of the  $II^-$ -sym-tridiagonal matrix needs to be solved with O(n) flop counts.

In Figure 6.1 (left), we plot the relative errors of  $\epsilon(\mu_j)$  for  $j=1,2,\ldots,1862$ . Unfortunately, we see that the  $\Gamma QR$  algorithm achieves only half of the accuracy that the PQR algorithm can. This is because the Hyp\_Householder and Hyp\_Givens transformations are not orthogonal matrices which lead to loss of accuracy during the reduction/iteration process in the  $\Gamma QR$  algorithm. However, the Hyp\_Householder and Hyp\_Givens transformations are  $\Gamma$ -orthogonal, which are strongly structure-preserving for  $\mathbf{\Pi}^-$ -symmetric matrix with respect to  $\Gamma$ , and mutually contain so that the entry sizes of the reduced<sub>This</sub> reason is  $\mathbf{\Pi}^-$ -symmetric matrix always achieve a balanced state. This is the reason why the  $\Gamma QR$  algorithm still keep the half accuracy.

In Figure 6.1 (right), we plot the normalized residual norms  $r(\mu_j)$ , j = 1, ..., 100 for the first 100 smallest positive eigenvalues of  $\mathcal{H}$ . It is clear that if all eigenpairs are obtained in this way, then the cost will dominant those of Algorithm 5.2 and 5.3. But computing different eigenpairs by the inverse iteration are highly independent and thus highly parallelizable.

#### 7 Conclusions

In this paper, we have developed an efficient implicit multi-shift  $\Gamma QR$  algorithm for solving the linear response eigenvalue problem (LREP) in (1.1) structurally. This algorithm relies on two basic  $\Gamma$ -orthogonal transformations, which preserve  $\mathbf{H}^-$ -symmetric structure of  $\mathcal{H}$  with respect to  $\Gamma$  throughout the whole computation. Thus the computed eigenvalues and eigenvectors are guaranteed to appear pairwise as in  $\{(\lambda, \mathbf{x}), (-\lambda, \Pi \mathbf{x})\}$ for a real or purely imaginary eigenvalue  $\lambda$  and in  $\{(\lambda, x), (-\lambda, \Pi x), (\lambda, \bar{x}), (-\lambda, \Pi \bar{x})\}$  for a true complex eigenvalue  $\lambda$ . Note that, these structures will be lost if the  $\Gamma$ -orthogonality is not preserved owing to roundoff errors, as in the HR algorithm [7] and the usual QR algorithm. We accelerate the convergence of the  $\Gamma QR$  algorithm by using the doubleshift technique based on the implicit  $\Gamma$ -orthogonality theorem, the final algorithm can be found in Algorithm 5.3. To compare with the block structure-preserving periodic QR algorithm, our algorithm costs much less than the PQR algorithm, and furthermore, the numerical experiment shows that the ΓQR algorithm can compute eigenpairs overall as This contradicts Fig. accurate as the PQR algorithm. In summary, the ΓQR algorithm is an efficient structure-gest we drop the plots preserving algorithm for solving the  $\Pi^-$ -symmetric eigenvalue problem compared with the for relative errors per preserving algorithm for solving the  $\Pi^-$ -symmetric eigenvalue problem compared with the formula  $\Pi^-$ -symmetric eigenvalue problem compared with the formula  $\Pi^-$ -symmetric eigenvalue problem compared with the formula  $\Pi^-$ -symmetric eigenvalue problem compared with  $\Pi^-$ -symmetric eigenvalue problem compared w the other existing algorithms. ues" by eig are truly exact. PQR matches well with QR in com-

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puted eigenvalues because both are QR-

based alg.

## References

- [1] S. T. Alexander, C.-T. Pan, and R. J. Plemmons. Analysis of a recursive least squares hyperbolic rotation algorithm for signal processing. *Linear Algebra Appl.*, 98:3–40, 1988.
- [2] Z. Bai and R.-C. Li. Minimization principle for linear response eigenvalue problem, I: Theory. SIAM J. Matrix Anal. Appl., 33(4):1075–1100, 2012.
- [3] Z. Bai and R.-C. Li. Minimization principles for the linear response eigenvalue problem II: Computation. SIAM J. Matrix Anal. Appl., 34(2):392–416, 2013.
- [4] M. A. Brebner and J. Grad. Similarity transformation for pseudosymmetric matrices with particular reference to the HR method. Research Paper 245, Department of Mathematics, Statistics and Computing Science, University of Calgary, 1974.
- [5] M. A. Brebner and J. Grad. The general form and certain properties of similarity transformation for pseudosymmetric matrices. Research Paper 281, Department of Mathematics, Statistics and Computing Science, University of Calgary, 1975.

- [6] M.A. Brebner and J.Grad. Eigenvalues of  $Ax = \lambda Bx$  for real symmetric matrices A and B computed by reduction to a pseudo symmetric form and the HR process. Linear Algebra Appl., 43:99–118, 1982.
- [7] A. Bunse-Gerstner. An analysis of the HR algorithm for computing the eigenvalues of a matrix. *Linear Algebra Appl.*, 35:155–173, 1981.
- [8] A. Bunse-Gerstner and V. Mehrmann. A symplectic QR-like algorithm for the solution of the real algebraic Riccati equation. *IEEE Trans. Automat. Control*, 31:1104–1113, 1986.
- [9] R. Byers. A Hamiltonian QR algorithm. SIAM J. Sci. Statist. Comput., 7:212–229, 1986.
- [10] M. E. Casida. Time-dependent density-functional response theory for molecules. In D. P. Chong, editor, Recent advances in Density Functional Methods, pages 155–189, World Scientific, Singapore, 1995.
- [11] J. Demmel. Applied Numerical Linear Algebra. SIAM, Philadelphia, PA, 1997.
- [12] W. R. Ferng, K.-Y. Lin, and W.-W. Lin. A novel nonsymmetric k-lanczos algorithm for the generalized nonsymmetric k-eigenvalue problems. *Linear Algebra Appl.*, 252:81–105, 1997.
- [13] U. Flaschka, W.-W. Lin, and J.-L. Wu. A KQZ algorithm for solving linear-response eigenvalue equations. *Linear Algebra Appl.*, 165:93–123, 1992.
- [14] G. H. Golub and C. F. Van Loan. Matrix Computations. Johns Hopkins University Press, Baltimore, Maryland, 3rd edition, 1996.
- [15] A. Ipatov, F. Cordova, L. J. Doriol, and M. E. Casida. Excited-state spin-contamination in time-dependent density-functional theory for molecules with open-shell ground states. J. Molecular Struct.: THEOCHEM, 914(13):60-73, 2009.
- [16] D. Kressner. The periodic QR algorithm is a disguised QR algorithm. Linear Algebra Appl., 417:423–433, 2006.
- [17] M. T. Lusk and A. E. Mattsson. High-performance computing for materials design to advance energy science. *MRS Bulletin*, 36:169–174, 2011.
- [18] J. Olsen and P. Jorgensen. Linear and nonlinear response functions for an exact state and for an MCSCF state. J. Chem. Phys., 82(7):3235–3264, 1985.
- [19] P. Papakonstantinou. Reduction of the RPA eigenvalue problem and a generalized Cholesky decomposition for real-symmetric matrices. *Europhysics Letters*, 78(1):12001, 2007.
- [20] D. Rocca, D. Lu, and G. Galli. *Ab initio* calculations of optical absorpation spectra: solution of the Bethe-Salpeter equation within density matrix perturbation theory. *J. Chem. Phys.*, 133(16):164109, 2010.
- [21] Y. Saad, J. R. Chelikowsky, and S. M. Shontz. Numerical methods for electronic structure calculations of materials. *SIAM Rev.*, 52:3–54, 2010.
- [22] R. E. Stratmann, G. E. Scuseria, and M. J. Frisch. An efficient implementation of timedependent density-functional theory for the calculation of excitation of large molecules. J. Chem. Phys., 109:8218–8824, 1998.

[23] Z. Teng and R.-C. Li. Convergence analysis of Lanczos-type methods for the linear response eigenvalue problem. *J. Comput. Appl. Math.*, 247:17–33, 2013.