#### Abstract

We present a technique for deriving bounds on the relative change in the singular values of a real matrix (or the eigenvalues of a real symmetric matrix) due to a perturbation, as well as bounds on the angles between the unperturbed and perturbed singular vectors (or eigenvectors). In particular, we consider the class of perturbations  $\delta B$  for which  $B + \delta B = D_L B D_R$  for some non-singular matrices  $D_L$  and  $D_R$ . This class includes component-wise relative perturbations of the entries in a bidiagonal or biacyclic matrix, and perturbations that annihilate the off-diagonal block in a block triangular matrix. We show how many existing relative perturbation and deflation bounds can be derived from results for this general class of perturbations. We also present some new relative perturbation and deflation results for the singular values and vectors of biacyclic, triangular and shifted triangular matrices.

Relative Perturbation Techniques for Singular Value Problems<sup>†</sup>

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### 1. Introduction

We present a technique for deriving bounds on the *relative* change in the singular values of a real matrix (or the eigenvalues of a real symmetric matrix) due to a perturbation, as well as bounds on the angles between the unperturbed and perturbed singular vectors (or eigenvectors).

Perturbation and deflation criteria based on such relative perturbation results can be incorporated into QR-type algorithms for computing the singular values and singular vectors of dense or banded matrices, leading to highly accurate and efficient implementations. For instance, the implementation of the Golub-Kahan algorithm in [5, 6] computes the smallest singular values and associated singular vectors of a bidiagonal matrix to high relative accuracy; and the implementation of Rutishauser's QD algorithm in [11] computes all of the singular values of a bidiagonal matrix to high relative accuracy. In general deflation criteria enhance efficiency by breaking the problem up into smaller subproblems, which also makes them suitable for the design of accurate parallel divide-and-conquer algorithms [3].

In the context of the singular value decomposition of a real matrix B, we consider the class of perturbations  $\delta B$  where  $B + \delta B = D_L B D_R$  for some non-singular matrices  $D_L$  and  $D_R$  (cf. [17], [1, p. 770]). This is the class of all perturbations that do not change the rank of B. In particular it includes component-wise relative perturbations of a bidiagonal or biacyclic matrices and perturbations that annihilate the off-diagonal block in a block triangular matrix.

We show how many existing relative perturbation bounds for singular values and vectors of bidiagonal matrices can be derived from general results for this class of perturbations. We also present new relative perturbation results for the singular values and vectors of biacyclic, triangular and shifted triangular matrices. Surprisingly, all of these results can be derived from traditional absolute perturbation results.

The standard absolute perturbation result says that the eigenvalues of the real symmetric matrices M and  $M + \delta M$  satisfy

$$|\lambda_i[M + \delta M] - \lambda_i[M]| \le ||\delta M||, \tag{1.1}$$

where  $\lambda_i[X]$  denotes the  $i^{th}$  largest eigenvalue of X and  $\|\cdot\|$  denotes the two-norm (see [21, Fact 1-11]). Moreover, if w is an eigenvector of M corresponding to  $\lambda_i[M]$  and w' is an eigenvector of  $M + \delta M$  corresponding to some eigenvalue  $\lambda'$ , i.e.,

$$Mw = \lambda_i[M]w$$
 and  $(M + \delta M)w' = \lambda'w',$ 

then the  $\sin \theta$ -Theorem [4] implies that

$$|\sin \theta| \le \frac{\|\delta M\|}{\text{gap}_i},\tag{1.2}$$

where  $0 \le \theta \le \pi/2$  is the angle between the spaces spanned by w and by w' and

$$\operatorname{gap}_i \equiv \min_{j \neq i} |\lambda_j[M] - \lambda'|$$

is the absolute gap (see [21, Fact 1-12]).

These absolute perturbation results are used in §2 in the context of the eigenvalue decomposition to derive two relative perturbation results for perturbations of the form  $M + \delta M = D^T M D$ , where D is a nonsingular matrix. Since this is the class of congruence transformations of M, the inertia is preserved so that  $\lambda_i[M]$  and  $\lambda_i[M + \delta M]$  have the same sign (or are both zero). The eigenvalue result

$$|\lambda_i[M + \delta M] - \lambda_i[M]| \le |\lambda_i[M]| \|D^T D - I\|$$

is relative because it bounds the relative perturbation in the eigenvalue. The eigenvector result

$$\sin\theta \le \frac{\delta}{\rho_i[M] - \gamma} + \beta,$$

provided that  $\rho_i[M] > \gamma$ , where

$$\beta \equiv ||D - I||, \qquad \gamma \equiv ||D^T D - I||, \qquad \delta \equiv ||D^T D|| \, ||D^{-T} D^{-1} - I||,$$

 $\mathrm{and}^2$ 

$$\rho_i[M] \equiv \min_{j \neq i} \frac{|\lambda_j[M] - \lambda_i[M]|}{|\lambda_i[M]|},$$

is relative in the sense that it depends on the relative gap  $\rho_i[M]$ .

In §3 we derive analogous results for singular values and singular vectors by formulating the singular value problem as a symmetric eigenvalue problem. The remaining sections are devoted to applying these results to special classes of matrices and perturbations. In §4 we consider componentwise relative perturbations of symmetric tridiagonal matrices with zero diagonal elements, bidiagonal matrices, and biacyclic matrices. We obtain simple proofs of many of the relative perturbation results in [1, 5, 8, 6] and solve an open problem in [8]. In §5 we consider perturbations that introduce zeros into a block triangular or bidiagonal matrix. We give simple proofs of many of the deflation results in [2, 5, 6, 16, 20] and strengthen the deflation result in [11]. In §6 we discuss some further applications.

#### Notation

All matrices are real. The identity matrix is denoted by I, and its  $k^{th}$  column by  $e_k$ . The two-norm is denoted by  $\|\cdot\|$ , and the one-norm is denoted by  $\|\cdot\|_1$ .

<sup>&</sup>lt;sup>1</sup> Note that that  $\lambda_i[M]$  must be simple; otherwise  $\rho_i[M] = 0$  and this condition is not satisfied.

<sup>&</sup>lt;sup>2</sup> We adopt the convention that  $\rho_i[M] = +\infty$  when  $\lambda_i[M] = 0$ , in which case  $\delta/(\rho_i[M] - \gamma) = 0$ .

Let M be a  $n \times n$  symmetric matrix. The eigenvalues  $\{\lambda_i[M]\}_{i=1}^n$  of M are numbered in decreasing order so that  $\lambda_1[M] \geq \ldots \geq \lambda_n[M]$ . Let N be a  $m \times n$  matrix and let  $p = \min\{m, n\}$ . The singular values  $\{\sigma_i[N]\}_{i=1}^p$  of N are numbered in decreasing order so that  $\sigma_1[N] \geq \ldots \geq \sigma_p[N] \geq 0$ . Also  $\sigma_{max}[N] = \sigma_1[N]$  and  $\sigma_{min}[N] = \sigma_p[N]$ .

By convention, A and  $A + \delta A$  denote symmetric matrices;  $\lambda_i$  and  $\lambda'_i$  denote the  $i^{th}$  eigenvalues of A and  $A + \delta A$ , respectively;  $w_i$  and  $w'_i$  denote unit eigenvectors associated with  $\lambda_i$  and  $\lambda'_i$ , respectively;  $\theta_i$  denotes the angle  $(0 \le \theta_i \le \pi/2)$  between the subspaces spanned by  $w_i$  and by  $w'_i$ ; and<sup>3</sup>

$$\rho_i \equiv \min_{j \neq i} \frac{|\lambda_j - \lambda_i|}{|\lambda_i|}$$

denotes the relative gap for the  $i^{th}$  eigenvalue of A.

Similarly, B and  $B + \delta B$  denote general matrices;  $\sigma_i$  and  $\sigma_i'$  denote the  $i^{th}$  singular values of B and  $B + \delta B$ , respectively;  $u_i$  and  $u_i'$  denote unit left singular vectors associated with  $\lambda_i$  and  $\lambda_i'$ , respectively;  $v_i$  and  $v_i'$  denote the corresponding right singular vectors;  $\theta_i^u$  denotes the angle  $(0 \le \theta_i^u \le \pi/2)$  between the subspaces spanned by  $u_i$  and by  $u_i'$ ;  $\theta_i^v$  denotes the angle  $(0 \le \theta_i^v \le \pi/2)$  between the subspaces spanned by  $v_i$  and by  $v_i'$ ; and

$$\bar{\rho}_i \equiv \min \left\{ 2, \min_{j \neq i} \frac{|\sigma_j - \sigma_i|}{|\sigma_i|} \right\},$$

denotes the relative gap for the  $i^{th}$  singular value of B.

## 2. General Perturbation Results for the Eigenproblem

This section presents general relative perturbation results for the eigenvalues and eigenvectors of a symmetric matrix.

The first theorem is the basis for all of the eigenvalue and singular value results to come. It contains the bound on the relative perturbation in the eigenvalues from which we derive a bound on the relative perturbation in the singular values in §3.

THEOREM 2.1. Let  $A + \delta A = D^T A D$ , where D is a non-singular matrix. Then

$$\frac{|\lambda_i|}{\|(D^T D)^{-1}\|} \le |\lambda_i'| \le |\lambda_i| \|D^T D\|$$

and

$$|\lambda_i' - \lambda_i| \le |\lambda_i| \|D^T D - I\|.$$

<sup>&</sup>lt;sup>3</sup> Again we adopt the convention that  $\rho_i = +\infty$  if  $\lambda_i = 0$ .

<sup>&</sup>lt;sup>4</sup> We also adopt the convention that  $\bar{\rho}_i = +\infty$  if  $\sigma_i = 0$ .

**Proof:** These inequalities are an immediate consequence of Ostrowski's Theorem [14, Section 4.5.9]; we include the proofs for completeness.

Since 0 is the  $i^{th}$  eigenvalue of  $A - \lambda_i I$ , Sylvester's Inertia Theorem [21] implies that 0 is the  $i^{th}$  eigenvalue of

$$D^{T}(A - \lambda_{i}I)D = (A + \delta A) - \lambda_{i}D^{T}D.$$

Using (1.1) with  $M = A + \delta A$  and  $\delta M = -\lambda_i D^T D$  gives

$$|0 - \lambda_i'| \le ||-\lambda_i D^T D||$$

or

$$|\lambda_i'| \le |\lambda_i| \, ||D^T D||.$$

The derivation of the lower bound is similar. Furthermore,

$$D^{T}(A - \lambda_{i}I)D = (A + \delta A - \lambda_{i}I) + \lambda_{i} (I - D^{T}D).$$

Using (1.1) with  $M = A + \delta A - \lambda_i I$  and  $\delta M = \lambda_i (I - D^T D)$  yields

$$|0 - (\lambda_i' - \lambda_i)| \le ||\lambda_i (I - D^T D)||$$

or

$$|\lambda_i' - \lambda_i| \le |\lambda_i| \|D^T D - I\|.$$

These bounds are tight. For example, they give equality when D is a multiple of an orthogonal matrix.

The second theorem is the basis for all of the eigenvector and singular vector results to come. It bounds the angle between the corresponding eigenvector and a perturbed eigenvector in terms of the relative gap.

THEOREM 2.2. Let  $A + \delta A = D^T A D$ , where D is a non-singular matrix, and let

$$\beta \equiv \|D-I\|, \qquad \gamma \equiv \|D^TD-I\|, \qquad \text{and} \qquad \delta \equiv \|D^TD\| \, \|D^{-T}D^{-1}-I\|.$$

Then

$$\sin \theta_i \le \frac{\delta}{\rho_i - \gamma} + \beta,$$

provided that  $\rho_i > \gamma$ .

**Proof:** Let

$$M=A-\lambda_i'I,\quad \mu=\lambda_i-\lambda_i',\quad \delta M=\lambda_i'(I-D^{-T}D^{-1}),\quad \bar{\mu}=0,\quad ext{and}\quad \bar{w}=Dw_i',$$

so that

$$Mw_i = \mu w_i$$
 and  $(M + \delta M)\bar{w} = 0\bar{w}$ .

Let  $0 \le \bar{\theta} \le \pi/2$  be the angle between the spaces spanned by  $w_i$  and by  $\bar{w}$ . By (1.2),

$$\sin \bar{\theta} \le |\lambda_i'| \|I - D^{-T}D^{-1}\| \left( \min_{j \ne i} |\lambda_j - \lambda_i'| \right)^{-1}$$

By Theorem 2.1,

$$|\lambda_i'| \le |\lambda_i| \, ||D^T D||$$

and

$$|\lambda_j - \lambda_i'| \ge |\lambda_j - \lambda_i| - |\lambda_i' - \lambda_i| \ge |\lambda_j - \lambda_i| - |\lambda_i| \|D^T D - I\|,$$

so that

$$\sin \bar{\theta} \le |\lambda_i| \|D^T D\| \|I - D^{-T} D^{-1}\| \left( \min_{j \ne i} |\lambda_j - \lambda_i| - |\lambda_i| \|D^T D - I\| \right)^{-1} = \frac{\delta}{\rho_i - \gamma}.$$

To bound  $\sin \theta_i$ , we use the triangle inequality

$$\sin \theta_i \le \sin \bar{\theta} + \sin \theta',$$

where  $0 \le \theta' \le \pi/2$  is the angle between the spaces spanned by  $\bar{w}$  and by  $w'_i$ . The second summand is bounded above by

$$\sin \theta' \le \|\bar{w} - w_i'\| = \|(D - I)w_i'\| \le \beta.$$

### 3. General Perturbation Results for the Singular Value Problem

This section presents general relative perturbation results for singular values and singular vectors. The results are derived by transforming the singular value problem to an eigenvalue problem. This can be done in two ways. First, the positive square-roots of the eigenvalues of  $B^TB$  or  $BB^T$  are the singular values of B, i.e.,  $\sigma_i = \sqrt{\lambda_i[B^TB]} = \sqrt{\lambda_i[BB^T]}$ . This transformation is used in Theorem 3.1 to derive bounds on the relative perturbation in the singular values. Second, the eigenvalues of the Jordan-Wielandt matrix

$$A = \begin{pmatrix} 0 & B^T \\ B & 0 \end{pmatrix}$$

are  $\pm \sigma_i$  (and 0 if  $m \neq n$ ), and

$$w_i = \frac{1}{\sqrt{2}} \left( \begin{array}{c} v_i \\ u_i \end{array} \right)$$

is a unit eigenvector of A associated with  $\lambda_i = \sigma_i$  (see [10], [22, Theorem I.4.2]). This transformation is used in Theorem 3.3 to derive the result for the singular vectors.

Theorem 3.1. Let  $B + \delta B = D_L B D_R$ , where  $D_L$  and  $D_R$  are nonsingular matrices. Then

$$\frac{\sigma_i}{\|D_L^{-1}\| \|D_R^{-1}\|} \le \sigma_i' \le \sigma_i \|D_L\| \|D_R\|.$$

**Proof:** Let  $C = D_L B$ . Apply Theorem 2.1 with  $A = C^T C$  and  $D = D_R$  to get

$$\frac{\lambda_i[C^T C]}{\|D_R^{-1}\|^2} \le \lambda_i[D_R^T (C^T C) D_R] = {\sigma'_i}^2 \le \lambda_i[C^T C] \|D_R\|^2;$$

and again with  $A = BB^T$  and  $D = D_L^T$  to get

$$\frac{\sigma_i^2}{\|D_L^{-1}\|^2} = \frac{\lambda_i[BB^T]}{\|D_L^{-1}\|^2} \le \lambda_i[D_L(BB^T)D_L^T] = \lambda_i[CC^T] \le \lambda_i[BB^T] \|D_L\|^2 = \sigma_i^2 \|D_L\|^2.$$

But  $\lambda_i[C^TC] = \lambda_i[CC^T]$  for  $1 \le i \le \min\{m, n\}$ .

This result could also be proved by means of the inequalities

$$\sigma_{i+j-1}[DB] \le \sigma_j[D] \ \sigma_i[B]$$
 and  $\sigma_{i+j-1}[BD] \le \sigma_i[B] \ \sigma_j[D]$ 

[15, Theorem 3.3.16] with j = 1.

COROLLARY 3.2 (BARLOW AND DEMMEL [1, p. 771]). Let  $B + \delta B = D_L B D_R$ , where  $D_L$  and  $D_R$  are nonsingular diagonal matrices. Then

$$\sigma_i \cdot \min_j |(D_L)_{jj}| \cdot \min_k |(D_R)_{kk}| \leq \sigma_i' \leq \sigma_i \cdot \max_j |(D_L)_{jj}| \cdot \max_k |(D_R)_{kk}|.$$

THEOREM 3.3. Let  $B + \delta B = D_L B D_R$ , where  $D_L$  and  $D_R$  are nonsingular matrices, and let

$$\bar{\beta} \equiv \max\{\|D_L^T - I\|, \|D_R - I\|\}, \qquad \bar{\gamma} \equiv \max\{\|D_L D_L^T - I\|, \|D_R^T D_R - I\|\},$$

and

$$\bar{\delta} \equiv \max\{\|D_L D_L^T\|, \|D_R^T D_R\|\} \cdot \max\{\|D_L^{-1} D_L^{-T} - I\|, \|D_R^{-T} D_R^{-1} - I\|\}.$$

Then

$$|\sigma_i' - \sigma_i| \leq \bar{\gamma} \ \sigma_i;$$

and, provided that  $\bar{\rho}_i > \bar{\gamma}$ ,

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \left(\frac{\bar{\delta}}{\bar{\rho}_i - \bar{\gamma}} + \bar{\beta}\right).$$

Proof: Let

$$A = \begin{pmatrix} 0 & B^T \\ B & 0 \end{pmatrix}, \quad A + \delta A = \begin{pmatrix} 0 & (B + \delta B)^T \\ B + \delta B & 0 \end{pmatrix}, \quad ext{and} \quad D = \begin{pmatrix} D_L^T & \\ & D_R \end{pmatrix},$$

so that

$$w_i = rac{1}{\sqrt{2}} \left(egin{array}{c} v_i \ u_i \end{array}
ight), \qquad w_i' = rac{1}{\sqrt{2}} \left(egin{array}{c} v_i' \ u_i' \end{array}
ight), \qquad ext{and} \qquad A + \delta A = D^T A D.$$

The bound for the singular values follows directly from Theorem 2.1. By Theorem 2.2,

$$\sin \theta_i \le \frac{\delta}{\rho_i - \gamma} + \beta,$$

where

$$\beta = ||D - I||, \qquad \gamma = ||D^T D - I||, \qquad \delta = ||D^T D|| \, ||D^{-T} D^{-1} - I||,$$

and

$$\rho_i = \min_{\ell \neq i} \frac{|\lambda_\ell - \sigma_i|}{\sigma_i},$$

provided that  $\rho_i > \gamma$ . But  $\beta = \bar{\beta}$ ,  $\gamma = \bar{\gamma}$ ,  $\delta = \bar{\delta}$ , and  $\rho_i = \bar{\rho}_i$  (recall that  $\lambda_{\ell} = \pm \sigma_j$  for some j), so that

$$\sin \theta_i \le \frac{\bar{\delta}}{\bar{\rho}_i - \bar{\gamma}} + \bar{\beta},$$

provided that  $\bar{\rho}_i > \bar{\gamma}$ . Now

$$2\sin\frac{\theta_i^u}{2} = \|u_i - u_i'\| \le \sqrt{2} \|w_i - w_i'\| = 2\sqrt{2}\sin\frac{\theta_i}{2},$$

and, since  $0 \le \theta_i, \theta_i^u \le \pi/2$ , this implies that

$$\sin \theta_i^u \le \sqrt{2} \sin \theta_i.$$

The derivation of the bound for  $\sin \theta_i^v$  is similar.

In the remaining sections, we apply the general perturbation results in this section and the previous one to particular classes of matrices and perturbations.

# 4. Componentwise Relative Perturbations

This section deals with perturbations that cause componentwise relative changes in the non-zero elements of tridiagonal matrices with zero diagonal elements, bidiagonal matrices, and biacyclic matrices. By applying the results in §2 and §3 we get simple proofs of many of the relative perturbation results in [1, 5, 8, 6] and solve an open problem in [8].

### 4.1. Eigenvalues and Singular Values

Symmetric tridiagonal matrices with zero diagonal elements are symmetric permutations of matrices of the form

$$\begin{pmatrix} 0 & B^T \\ B & 0 \end{pmatrix},$$

where B is bidiagonal. They result from formulating the singular value problem as a symmetric eigenvalue problem (see  $\S 3$ ). We begin by bounding the effect of a relative change in a single pair of nonzero elements.

COROLLARY 4.1 (KAHAN [18], DEMMEL AND KAHAN [6, THEOREM 2]). Let A be a symmetric tridiagonal matrix with zero diagonal elements, and let  $A + \delta A$  equal A except for the off-diagonal elements  $a_{k,k+1} = a_{k+1,k}$  that are perturbed to  $\alpha a_{k,k+1} = \alpha a_{k+1,k}$  for some  $\alpha \neq 0$ . Then

$$\frac{1}{1+\eta} |\lambda_i| \le |\lambda_i'| \le (1+\eta) |\lambda_i|,$$

where  $\eta \equiv \max\{|\alpha|, 1/|\alpha|\} - 1$ .

Proof: Let

$$D = \operatorname{diag} \left( \cdots \sqrt{\alpha} \quad 1/\sqrt{\alpha} \quad \sqrt{\alpha} \quad \sqrt{\alpha} \quad 1/\sqrt{\alpha} \quad \sqrt{\alpha} \quad \cdots \right),$$

so that  $A + \delta A = D^T A D$  (see [1, p. 770]). Now apply Theorem 2.1 and note that  $||D^T D|| = ||(D^T D)^{-1}|| = 1 + \eta$ .

By repeated application of this result we can bound the effect of relative changes in every pair of nonzero elements (see [6, Corollary 1]). The result is analogous to the following result for bidiagonal matrices.

COROLLARY 4.2 (BARLOW AND DEMMEL [1, THEOREM 1], DEIFT ET AL. [5, THEOREM 2.12], DEMMEL AND KAHAN [6, COROLLARY 2]). Let

$$B = \begin{pmatrix} b_{11} & b_{12} & & & & \\ & b_{22} & \ddots & & & \\ & & \ddots & b_{n-1,n} \\ & & & b_{nn} \end{pmatrix} \quad and \quad B + \delta B = \begin{pmatrix} \alpha_1 b_{11} & \alpha_2 b_{12} & & & \\ & \alpha_3 b_{22} & \ddots & & \\ & & & \ddots & \alpha_{2n-2} b_{n-1,n} \\ & & & & \alpha_{2n-1} b_{nn} \end{pmatrix},$$

where  $\alpha_j \neq 0$ . Then

$$\frac{1}{1+\eta} \, \sigma_i \le \sigma_i' \le (1+\eta) \, \sigma_i,$$

where  $\eta \equiv \prod_{j=1}^{2n-1} \max\{|\alpha_j|, 1/|\alpha_j|\} - 1$ .

Proof: Let

$$D_L = \operatorname{diag}\left(\alpha_1, \frac{\alpha_1 \alpha_3}{\alpha_2}, \frac{\alpha_1 \alpha_3 \alpha_5}{\alpha_2 \alpha_4}, \frac{\alpha_1 \alpha_3 \alpha_5 \alpha_7}{\alpha_2 \alpha_4 \alpha_6}, \ldots\right)$$

and

$$D_R = \operatorname{diag}\left(1, \frac{\alpha_2}{\alpha_1}, \frac{\alpha_2\alpha_4}{\alpha_1\alpha_3}, \frac{\alpha_2\alpha_4\alpha_6}{\alpha_1\alpha_3\alpha_5}, \ldots\right),\,$$

so that  $B + \delta B = D_L B D_R$  (see [1, p. 770]). Now apply Theorem 3.1 and note that  $||D_L|| ||D_R|| \le 1 + \eta$  and  $||D_L^{-1}|| ||D_R^{-1}|| \le 1 + \eta$ .

Biacyclic matrices are a generalization of bidiagonal matrices in which the underlying bipartite graph can be any acyclic bipartite graph [8]. Only biacyclic matrices have the property that relative perturbations of the matrix elements cause relative perturbations in the singular values that are independent of the values of the elements [8].

COROLLARY 4.3 (DEMMEL AND GRAGG [8, THEOREM 1]). Let B be biacyclic, and let  $B + \delta B$  be a componentwise relative perturbation of B, i.e.,  $[B + \delta B]_{k,\ell} = \alpha_{k,\ell} b_{k,\ell}$  for all k and  $\ell$ , where  $\alpha_{k,\ell} \neq 0$ . Then

$$\frac{1}{1+\eta} \, \sigma_i \le \sigma_i' \le (1+\eta) \, \sigma_i,$$

where  $\eta \equiv \prod_{b_{k,\ell} \neq 0} \max\{|\alpha_{k,\ell}|, 1/|\alpha_{k,\ell}|\} - 1$ .

**Proof:** By Lemma 1 in [8] there exist diagonal matrices  $D_L$  and  $D_R$  such that  $B + \delta B = D_L B D_R$ , where the diagonal entries of  $D_L$  and  $D_R$  are quotients of products of distinct  $\alpha_{k,\ell}$  and each  $\alpha_{k,\ell}$  can only appear either in the numerators of  $D_L$  and the denominators of  $D_R$  or in the numerators of  $D_R$  and the denominators of  $D_L$ . Now apply Theorem 3.1 and verify that  $||D_L|| ||D_R|| \le 1 + \eta$  and  $||D_L^{-1}|| ||D_R^{-1}|| \le 1 + \eta$ .

## 4.2. Eigenvectors and Singular Vectors

The corresponding results for eigenvectors and singular vectors again follow directly from the results in §2 and §3.

COROLLARY 4.4. Let A be a symmetric tridiagonal matrix with zero diagonal elements, and let  $A + \delta A$  equal A except for the off-diagonal elements  $a_{k,k+1} = a_{k+1,k}$  that change to  $\alpha a_{k,k+1} = \alpha a_{k+1,k}$  for some  $\alpha > 0$ . Then

$$\sin \theta_i \le \frac{\eta(1+\eta)}{\rho_i - \eta} + \frac{\eta}{2},$$

where  $\eta \equiv \max \{\alpha, 1/\alpha\} - 1$ , provided that  $\rho_i > \eta$ .

**Proof:** Define D as in the proof of Corollary 4.1 so that  $A + \delta A = D^T A D$ . Now apply Theorem 2.2 and note that

$$\beta = \max\{|\sqrt{\alpha} - 1|, |1/\sqrt{\alpha} - 1|\} \le \frac{\eta}{2},$$

$$\gamma = \max\{|\alpha - 1|, |1/\alpha - 1|\} = \eta,$$

and

$$\delta = \max\{\alpha, 1/\alpha\} \cdot \max\{|\alpha - 1|, |1/\alpha - 1|\} \le (1 + \eta)\eta.$$

Deift et al. [5, p. 1472] prove that

$$\sin \theta_i \le \frac{\eta(1+\eta')}{\mathrm{relgap}_i - \eta'},$$

where

$$\eta' \equiv e^{\eta} - 1$$
 and  $\operatorname{relgap}_i \equiv \min_{j \neq i} \frac{|\lambda_j - \lambda_i|}{|\lambda_j| + |\lambda_i|}$ ,

provided that  $\operatorname{relgap}_i > \eta'$ . If  $\rho_i$  is small, then  $\operatorname{relgap}_i \approx \rho_i/2$ . Hence, if  $\eta$  is also small, then this bound is slightly weaker than the bound in Corollary 4.4.

COROLLARY 4.5. Let

$$B = \begin{pmatrix} b_{11} & b_{12} & & & \\ & b_{22} & \ddots & & \\ & & \ddots & b_{n-1,n} \\ & & & b_{nn} \end{pmatrix} \quad and \quad B + \delta B = \begin{pmatrix} \alpha_1 b_{11} & \alpha_2 b_{12} & & & \\ & \alpha_3 b_{22} & \ddots & & \\ & & & \ddots & \alpha_{2n-2} b_{n-1,n} \\ & & & & \alpha_{2n-1} b_{nn} \end{pmatrix},$$

where  $\alpha_j > 0$ . Then

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta(1+\eta)}{\bar{\rho}_i - \eta} + \frac{\eta}{2} \right),\,$$

where  $\eta \equiv \prod_{j=1}^{2n-1} \max \{\alpha_j, 1/\alpha_j\} - 1$ , provided that  $\bar{\rho}_i > \eta$ .

Proof: Let

$$D_L = \frac{1}{\alpha_0} \operatorname{diag} \left( \alpha_1, \frac{\alpha_1 \alpha_3}{\alpha_2}, \frac{\alpha_1 \alpha_3 \alpha_5}{\alpha_2 \alpha_4}, \frac{\alpha_1 \alpha_3 \alpha_5 \alpha_7}{\alpha_2 \alpha_4 \alpha_6}, \ldots \right)$$

and

$$D_R = \alpha_0 \operatorname{diag} \left( 1, \frac{\alpha_2}{\alpha_1}, \frac{\alpha_2 \alpha_4}{\alpha_1 \alpha_3}, \frac{\alpha_2 \alpha_4 \alpha_6}{\alpha_1 \alpha_3 \alpha_5}, \dots \right),$$

so that  $B + \delta B = D_L B D_R$  (see [1, p. 770]). Choose  $\alpha_0$  to minimize

$$\max \{(D_L)_{ii}, (D_R)_{ii}, (D_L^{-1})_{ii}, (D_R^{-1})_{ii}\}$$

so that

$$\frac{1}{\sqrt{1+\eta}} \le (D_L)_{ii}, (D_R)_{ii}, (D_L^{-1})_{ii}, (D_R^{-1})_{ii} \le \sqrt{1+\eta}.$$

Now apply Theorem 3.3 and note that  $\bar{\beta} \leq \frac{\eta}{2}$ ,  $\bar{\gamma} \leq \eta$ , and  $\bar{\delta} \leq \eta(1+\eta)$  (cf. the proof of Corollary 4.4).

Deift et al. [5, Theorem 2.12] prove that

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \frac{\eta'(1+\eta')}{\text{relgap}_i - \eta'},$$

where

$$\eta' \equiv e^{\eta} - 1$$
 and  $\operatorname{relgap}_i \equiv \min_{j \neq i} \frac{|\sigma_j - \sigma_i|}{\sigma_j + \sigma_i},$ 

provided that  $\operatorname{relgap}_i > \eta'$ . Again, when  $\bar{\rho}_i$  and  $\eta$  are small, this bound is slightly weaker than the bound in Corollary 4.5.

The following result solves an open problem in [8]; the proof is similar to the proof of Corollary 4.5.

COROLLARY 4.6. Let B be biacyclic, and let  $B + \delta B$  be a componentwise relative perturbation of B, i.e.,  $[B + \delta B]_{k,\ell} = \alpha_{k,\ell} b_{k,\ell}$  for all k and  $\ell$ , where  $\alpha_{k,\ell} > 0$ . Then

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta(1+\eta)}{\bar{\rho}_i - \eta} + \frac{\eta}{2} \right),$$

where  $\eta \equiv \prod_{b_{k,\ell} \neq 0} \max\{\alpha_{k,\ell}, 1/\alpha_{k,\ell}\} - 1$ , provided that  $\bar{\rho}_i > \eta$ .

# 5. Perturbations that Change the Sparsity Pattern

This section presents bounds on the relative perturbation in the singular values and singular vectors of a block triangular or bidiagonal matrix when an off-diagonal block is changed. This change is called *deflation* if the off-diagonal block is set to zero. By applying the results in §2 and §3 we get simple proofs of many of the deflation results in [2, 5, 6, 16, 20] and strengthen the deflation result in [11].

## 5.1. Block-Triangular Matrices

First we derive bounds on the relative perturbation in the singular values and vectors of block-triangular matrices. Then we derive bounds on the singular values of the diagonal blocks.

LEMMA 5.1. Let  $B + \delta B = D_L B$  or  $B + \delta B = B D_R$ , where

$$D_L = egin{array}{ccc} k & m-k & & & k & n-k \ I & X & & & & \\ m-k & & I & & & & \\ & & I & & & & \\ \end{array} \qquad or \qquad D_R = egin{array}{ccc} k & n-k & & & \\ I & X & & & \\ n-k & & & I \end{array} 
ight).$$

Then

$$\frac{1}{1+\eta} \, \sigma_i \le \sigma_i' \le (1+\eta) \, \sigma_i$$

and

$$|\sigma_i' - \sigma_i| \leq \eta \ \sigma_i,$$

where  $\eta \equiv ||X||$ . Moreover,

$$\max\{\sin\theta_i^u,\sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta'(1+\eta')}{\bar{\rho}_i - \eta'} + \frac{\eta}{2} \right),$$

where  $\eta' \equiv e^{\eta} - 1$ , provided that  $\bar{\rho}_i > \eta'$ .

**Proof:** We treat the case  $B + \delta B = D_L B$ ; the other cases are similar. Since

$$D_L^{-1} = \begin{pmatrix} I & -X \\ & I \end{pmatrix}$$

we have  $||D_L|| = ||D_L^{-1}|| \le 1 + \eta$ . To get the first pair of inequalities, apply Theorem 3.1 with  $D_R = I$ . The second inequality follows immediately. The final inequality follows from Theorem 3.3 after noting that

$$\bar{\beta} = \left\| \begin{pmatrix} 0 & X \\ 0 & 0 \end{pmatrix} \right\| = \eta, \qquad \bar{\gamma} = \left\| \begin{pmatrix} 0 & X \\ X^T & X^T X \end{pmatrix} \right\| \leq \eta',$$

and

$$\bar{\delta} = \left\| \begin{pmatrix} I & X \\ X^T & I + X^T X \end{pmatrix} \right\| \left\| \begin{pmatrix} 0 & -X \\ -X^T & X^T X \end{pmatrix} \right\| \leq \eta' (1 + \eta'),$$

where we have used the fact that

$$\left\| \begin{pmatrix} 0 & \alpha \\ \alpha & \alpha^2 \end{pmatrix} \right\| \le e^{\alpha} - 1$$

when  $\alpha \geq 0$ .

When B is block triangular, we can bound the effect of setting the off-diagonal block to zero.

THEOREM 5.2. Let

$$B=egin{array}{ccc} k & n-k \ B= k & \left(egin{array}{ccc} B_{11} & B_{12} \ & B_{22} \end{array}
ight) & and & B+\delta B= \left(egin{array}{ccc} B_{11} & \ & B_{22} \end{array}
ight),$$

where  $B_{11}$  or  $B_{22}$  is non-singular. Then

$$|\sigma_i' - \sigma_i| \le \sigma_i \, \eta,$$

where

$$\eta \equiv \frac{\|B_{12}\|}{\max\{\sigma_{min}[B_{11}], \sigma_{min}[B_{22}]\}}.$$

Moreover,

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta'(1+\eta')}{\bar{\rho}_i - \eta'} + \frac{\eta}{2} \right),$$

where  $\eta' \equiv e^{\eta} - 1$ , provided that  $\bar{\rho}_i > \eta'$ . Finally, if  $\sigma_{min}[B_{11}] > \sigma_{max}[B_{22}]$ , then

$$0 \le \sigma_i - \sigma_i[B_{11}] \le \sigma_i \frac{\|B_{12}\|}{\sigma_{min}[B_{11}]}, \qquad 1 \le i \le k$$

and

$$0 \le \sigma_i[B_{22}] - \sigma_{k+i} \le \sigma_{k+i} \frac{\|B_{12}\|}{\sigma_{min}[B_{11}]}, \qquad 1 \le i \le n-k.$$

**Proof:** If  $B_{11}$  is non-singular, then

$$\begin{pmatrix} B_{11} & B_{12} \\ & B_{22} \end{pmatrix} \begin{pmatrix} I & -B_{11}^{-1}B_{12} \\ & I \end{pmatrix} = \begin{pmatrix} B_{11} \\ & B_{22} \end{pmatrix},$$

and by Lemma 5.1,

$$|\sigma_i' - \sigma_i| \le \sigma_i \|B_{11}^{-1} B_{12}\| \le \sigma_i \frac{\|B_{12}\|}{\sigma_{min}[B_{11}]}.$$

If  $B_{22}$  is non-singular, then

$$\begin{pmatrix} I & -B_{12}B_{22}^{-1} \\ & I \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ & B_{22} \end{pmatrix} = \begin{pmatrix} B_{11} \\ & B_{22} \end{pmatrix},$$

and by Lemma 5.1,

$$|\sigma_i' - \sigma_i| \le \sigma_i \|B_{12}B_{22}^{-1}\| \le \sigma_i \frac{\|B_{12}\|}{\sigma_{min}[B_{22}]}.$$

This proves the first inequality. The second inequality follows from Lemma 5.1.

If  $\sigma_{min}[B_{11}] > \sigma_{max}[B_{22}]$ , then

$$\sigma'_{i} = \sigma_{i}[B_{11}], \qquad 1 \leq i \leq k,$$
  
$$\sigma'_{k+i} = \sigma_{i}[B_{22}], \qquad 1 \leq i \leq n-k,$$

and

$$\max\{\sigma_{min}[B_{11}], \sigma_{min}[B_{22}]\} = \sigma_{min}[B_{11}].$$

The lower bounds in the last two sets of inequalities follow from the Interlacing Theorem for singular values [13, Corollary 8.3.3].

Weaker bounds on the singular values appear in [2].

COROLLARY 5.3 (CHANDRASEKARAN AND IPSEN [2, THEOREM 5.2.1]). With the notation of Theorem 5.2, assume that  $\sigma_{min}[B_{11}] > \sigma_{max}[B_{22}]$ . Then

$$|\sigma_i[B_{22}] - \sigma_{k+i}| \le \sigma_{max}[B_{22}] \frac{||B_{12}||}{\sigma_{min}[B_{11}] - \sigma_{max}[B_{22}]}, \qquad 1 \le i \le n-k.$$

**Proof:** By Theorem 5.2,

$$0 \le \sigma_i[B_{22}] - \sigma_{k+i} \le \sigma_{k+i} \frac{\|B_{12}\|}{\sigma_{min}[B_{11}]}.$$

Thus

$$\sigma_{k+i} \le \sigma_i[B_{22}] \le \sigma_{max}[B_{22}],$$

and the result follows immediately.

By applying Theorem 3.1 directly, we get the following result.

COROLLARY 5.4. With the notation of Theorem 5.2, assume that  $B_{11}$  is non-singular and  $||B_{22}|| < \sigma_{min}[B_{11}]$ . Then

$$\sigma_i[B_{11}] \le \sigma_i \le \left(1 + \frac{\|B_{12}\|^2}{\sigma_{min}[B_{11}]^2 - \|B_{22}\|^2}\right)^{1/2} \sigma_i[B_{11}], \qquad 1 \le i \le k$$

and

$$\left(1 + \frac{\|B_{12}\|^2}{\sigma_{min}[B_{11}]^2 - \|B_{22}\|^2}\right)^{-1/2} \sigma_i[B_{22}] \le \sigma_{k+i} \le \sigma_i[B_{22}], \qquad 1 \le i \le n-k.$$

**Proof:** By Theorem 5.2, we have  $\sigma_i[B_{11}] \leq \sigma_i$ . Let  $\rho \equiv ||B_{22}||/\sigma_{min}[B_{11}] < 1$ . Since  $B_{11}$  is nonsingular,

$$B \equiv \begin{pmatrix} B_{11} & B_{12} \\ & B_{22} \end{pmatrix} = \begin{pmatrix} B_{11} & & \\ & \frac{1}{2}B_{22} \end{pmatrix} \begin{pmatrix} I & B_{11}^{-1}B_{12} \\ & \rho I \end{pmatrix} \equiv \hat{B}\hat{D}.$$

By Theorem 3.1 with  $D_L = I$ ,  $B = \hat{B}$ , and  $D_R = \hat{D}$ , we have  $\sigma_i \leq \sigma_i[\hat{B}] \|\hat{D}\|$ . Since  $\sigma_{max}[\frac{1}{\rho}B_{22}] = \sigma_{min}[B_{11}]$ ,

$$\sigma_i[\hat{B}] = \sigma_i[B_{11}], \qquad 1 \le i \le k.$$

Now

$$\|\hat{D}\| = \max_{j} \left\| \begin{pmatrix} 1 & \sigma_{j}[B_{11}^{-1}B_{12}] \\ \rho \end{pmatrix} \right\| \le \max_{j} \left( 1 + \frac{\sigma_{j}[B_{11}^{-1}B_{12}]^{2}}{1 - \rho^{2}} \right)^{1/2}$$

$$= \left( 1 + \frac{\|B_{11}^{-1}B_{12}\|^{2}}{1 - \rho^{2}} \right)^{1/2} \le \left( 1 + \frac{\|B_{12}\|^{2}}{\sigma_{min}[B_{11}]^{2} - \|B_{22}\|^{2}} \right)^{1/2}.$$

The second inequality follows in a similar fashion from the identity

$$B\hat{D} \equiv \begin{pmatrix} B_{11} & B_{12} \\ & B_{22} \end{pmatrix} \begin{pmatrix} \rho I & -B_{11}^{-1}B_{12} \\ & I \end{pmatrix} = \begin{pmatrix} \rho B_{11} & \\ & B_{22} \end{pmatrix} \equiv \hat{B}.$$

Mathias and Stewart [20, Theorem 3.1] prove the slightly weaker results

$$\sigma_{i}[B_{11}] \leq \sigma_{i} \leq \left(1 - \frac{\|B_{12}\|^{2}}{\sigma_{min}[B_{11}]^{2} - \|B_{22}\|^{2}}\right)^{-1/2} \sigma_{i}[B_{11}], \qquad 1 \leq i \leq k$$

and

$$\left(1 - \frac{\|B_{12}\|^2}{\sigma_{min}[B_{11}]^2 - \|B_{22}\|^2}\right)^{1/2} \sigma_i[B_{22}] \le \sigma_{k+i} \le \sigma_i[B_{22}], \qquad 1 \le i \le n-k$$

that are used in their analysis of the URV decomposition. For the special case where B is bidiagonal and k = n - 1, Demmel and Kahan [6, Theorem 5] prove the slightly weaker result that

$$|\sigma_i' - \sigma_i| \le 2 \sigma_i' \frac{\|B_{12}\|^2}{\sigma_{min}[B_{11}]^2 - \|B_{22}\|^2}$$

### 5.2. Bidiagonal Matrices

This section applies the results of §5.1 to bidiagonal matrices. In particular, we derive the deflation criteria used in [5, 6] to split a matrix into independent submatrices.

THEOREM 5.5. Let

$$B = \frac{k}{n-k} \begin{pmatrix} B_{11} & b_{k,k+1}e_ke_1^T \\ & B_{22} \end{pmatrix} \quad and \quad B + \delta B = \begin{pmatrix} B_{11} & \\ & B_{22} \end{pmatrix}$$

be non-singular bidiagonal matrices. Then

$$\frac{\sigma_i}{1+\eta} \le \sigma_i' \le (1+\eta) \, \sigma_i,$$

where

$$\eta \equiv |b_{k,k+1}| \cdot \min\{||B_{11}^{-1}e_k||, ||B_{22}^{-T}e_1||\}.$$

Moreover,

$$\max\{\sin\theta_i^u,\sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta'(1+\eta')}{\bar{\rho}_i - \eta'} + \frac{\eta}{2} \right),$$

where  $\eta' \equiv e^{\eta} - 1$ , provided that  $\bar{\rho}_i > \eta'$ .

**Proof:** As in the proof of Theorem 5.2, we apply Lemma 5.1 with  $B + \delta B = BD_R$  and  $X = -b_{k,k+1}B_{11}^{-1}e_ke_1^T$  so that

$$||X|| \le |b_{k,k+1}| ||B_{11}^{-1}e_k||;$$

or with  $B + \delta B = D_L B$  and  $X = -b_{k,k+1} e_k e_1^T B_{22}^{-1}$  so that

$$||X|| \le |b_{k,k+1}| ||B_{22}^{-T}e_1||.$$

The result follows immediately.

Now we show that the above theorem justifies Convergence Criterion 1 from [5, 6], which is used in the zero-shift Golub-Kahan algorithm for computing the singular values of bidiagonal matrices to high relative accuracy.

COROLLARY 5.6. Assume that

$$B = \begin{pmatrix} b_{11} & b_{12} & & & \\ & b_{22} & \ddots & & \\ & & \ddots & b_{n-1,n} \\ & & & b_{nn} \end{pmatrix},$$

is nonsingular, and let  $B + \delta B$  equal B except that the (k, k + 1) entry is set to zero. Let

$$\eta \equiv |b_{k,k+1}| \cdot \min \left\{ ||S^{-1}e_{2k}||, ||S^{-1}e_{2k+1}|| \right\},\,$$

where

$$S = \begin{pmatrix} 0 & b_{11} \\ b_{11} & 0 & b_{12} \\ & b_{12} & 0 & \ddots \\ & & \ddots & \ddots & b_{n-1,n} \\ & & & b_{n1,n} & 0 & b_{nn} \\ & & & & b_{nn} & 0 \end{pmatrix},$$

and let  $\eta' \equiv e^{\eta} - 1$ . Then

$$\frac{\sigma_i}{1+\eta} \le \sigma_i' \le (1+\eta) \, \sigma_i.$$

Moreover,

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \sqrt{2} \left( \frac{\eta'(1+\eta')}{\bar{\rho}_i - \eta'} + \frac{\eta}{2} \right),$$

provided that  $\bar{\rho}_i > \eta'$ . Finally, these results hold when  $\eta$  and  $\eta'$  are replaced by

$$\eta_1 \equiv |b_{k,k+1}| \cdot \min \left\{ \|S^{-1}e_{2k}\|_1, \|S^{-1}e_{2k+1}\|_1 \right\} \quad and \quad \eta_1' \equiv e^{\eta_1} - 1.$$

**Proof:** Let

$$B = \frac{k}{n-k} \begin{pmatrix} B_{11} & B_{12} \\ & B_{22} \end{pmatrix}.$$

From Theorem 5.5 it suffices to show that

$$||B_{11}^{-1}e_k|| = ||S^{-1}e_{2k}||$$
 and  $||B_{22}^{-T}e_1|| = ||S^{-1}e_{2k+1}||$ .

But

$$S = P \begin{pmatrix} B^T \\ B \end{pmatrix} P^T,$$

where the permutation matrix P represents the mapping

$$P: (1 \ 2 \ 3 \ 4 \ \dots \ 2n-1 \ 2n) \mapsto (1 \ n+1 \ 2 \ n+2 \ \dots \ n \ 2n).$$

Hence

$$S^{-1}e_{2k} = P \begin{pmatrix} B^{-1} \\ B^{-T} \end{pmatrix} P^{T}e_{2k} = P \begin{pmatrix} B^{-1}e_{k} \\ 0_{n} \end{pmatrix} = P \begin{pmatrix} B_{11}^{-1}e_{k} \\ 0_{2n-k} \end{pmatrix},$$

where  $0_j$  represents a  $j \times 1$  zero vector, and

$$S^{-1}e_{2k+1} = P \begin{pmatrix} & B^{-1} \\ B^{-T} & \end{pmatrix} P^T e_{2k+1} = P \begin{pmatrix} 0_n \\ B^{-T}e_{k+1} \end{pmatrix} = P \begin{pmatrix} 0_{n+k} \\ B_{22}^{-T}e_1 \end{pmatrix}.$$

The final result follows from the vector-norm inequality  $\|\cdot\| \leq \|\cdot\|_1$ .

Demmel and Kahan [6, Theorem 4] prove that

$$e^{-\eta/\sqrt{2}}\sigma_i \leq \sigma_i' \leq e^{\eta/\sqrt{2}}\sigma_i$$

which is slightly stronger than the bound in Corollary 5.6 when  $\eta$  is small.<sup>5</sup> Deift et al. [5, Theorem 4.7] prove that

$$\frac{\sigma_i}{1+\eta'} \le \sigma_i' \le (1+\eta') \, \sigma_i,$$

which is slightly weaker.

Deift et al. [5, Theorem 4.7] also prove that

$$\max\{\sin\theta_i^u, \sin\theta_i^v\} \le \frac{\eta'(1+\eta')}{\operatorname{relgap}_i - \eta'} + \eta',$$

and

$$\max\{\sin\theta_i^u,\sin\theta_i^v\} \leq \sqrt{\frac{2n+5}{4}} \frac{\eta_1'(1+\eta_1')}{\mathrm{relgap}_i - \eta_1'} + \frac{1}{\sqrt{2}} \left(1 + \sqrt{n-\frac{1}{2}}\right) \eta_1',$$

where

$$\operatorname{relgap}_i \equiv \min_{j \neq i} \frac{|\sigma_j - \sigma_i|}{\sigma_j + \sigma_i}.$$

If  $\bar{\rho}_i$  and  $\eta$  are small, then relgap<sub>i</sub>  $\approx \bar{\rho}_i/2$  and  $\eta' \approx \eta$  so that these bounds are slightly weaker than the bounds in Corollary 5.6.

#### 5.3. Shifted Matrices

Faddeev, Kublanovskaya, and Faddeeva [9] and Chandrasekaran and Ipsen [2, 16] present an unshifted QR algorithm for computing the singular values and vectors of a triangular matrix B. The algorithm is a generalization of the Golub-Kahan algorithm [12] for bidiagonal matrices because it does not form  $B^TB$  or  $BB^T$ . All of the deflation criteria presented in §5.1

<sup>&</sup>lt;sup>5</sup> This bound can be obtained by sharpening the bound on  $||D_L||$  or  $||D_L||$  in in Lemma 5.1 to  $e^{\eta/\sqrt{2}}$ .

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