# SWAPPING $2 \times 2$ BLOCKS IN THE SCHUR AND GENERALIZED SCHUR FORM* 

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#### Abstract

In this paper we describe how to swap two $2 \times 2$ blocks in a real Schur form and a generalized real Schur form. We pay special attention to the numerical stability of the method. We also illustrate the stability of our approach by a series of numerical tests.


1. Introduction. The real Schur form of a standard eigenvalue problem $\lambda I_{n}-A$ is an orthogonal matrix decomposition that plays a fundamental role in numerical linear algebra. It is an upper block triangular form $A_{S}$ that can be obtained under an orthogonal similarity transformation $U$ :

$$
A_{S}:=U^{T} A U=\left[\begin{array}{cccc}
A_{11} & A_{12} & \ldots & A_{1 k}  \tag{1.1}\\
0 & A_{22} & \ddots & A_{2 k} \\
\vdots & \ddots & \ddots & \vdots \\
0 & \ldots & 0 & A_{k k}
\end{array}\right]
$$

where the diagonal blocks $A_{i i}, i=1, \ldots, k$ are of dimension $1 \times 1$ or $2 \times 2$. This form is not only useful for computing the spectrum of $A$ but also for computing an invariant subspace of the matrix $A$ with a spectrum constrained to a particular region in the complex plane (see, e.g., $[1,3,7,8]$ ). For this, one needs to reorder the Schur form $A_{S}$ by updating the orthogonal transformation $U$ such that the diagonal blocks with their spectrum inside the considered region, appear first in the block triangular form. Such a reordering can always be performed by a series of swappings of two adjacent blocks in the real Schur form, as was shown in $[6,7]$.

There is a corresponding matrix decomposition known as the generalized real Schur form of a so-called regular pencil $\lambda B-A$, implying that $\operatorname{det}(\lambda B-A)$ is not identically zero. The equivalent generalized Schur form is upper block triangular and can be obtained under orthogonal equivalence transformations:

$$
\begin{aligned}
\lambda B_{S}-A_{S} & :=Z^{T}(\lambda B-A) Q \\
& =\lambda\left[\begin{array}{cccc}
B_{11} & B_{12} & \ldots & B_{1 k} \\
0 & B_{22} & \ddots & B_{2 k} \\
\vdots & \ddots & \ddots & \vdots \\
0 & \ldots & 0 & B_{k k}
\end{array}\right]-\left[\begin{array}{cccc}
A_{11} & A_{12} & \ldots & A_{1 k} \\
0 & A_{22} & \ddots & A_{2 k} \\
\vdots & \ddots & \ddots & \vdots \\
0 & \ldots & 0 & A_{k k}
\end{array}\right],
\end{aligned}
$$

where the subpencils $\left(\lambda B_{i i}-A_{i i}\right), i=1, \ldots, k$ are of dimension $1 \times 1$ or $2 \times 2$ and where $B_{S}$ is upper triangular. The concept corresponding to an invariant subspace

[^0]here is that of a deflating subspace, and again one needs to reorder the blocks in the generalized Schur form by updating the orthogonal transformations $Q$ and $Z$ such that the diagonal blocks $\left(\lambda B_{i i}-A_{i i}\right)$ with their spectrum inside the relevant region, appear first in the decomposition. The swapping problem for two consecutive blocks in the generalized Schur form was discussed in, e.g., $[5,8,9]$.

The swapping of a scalar block with another scalar block or with a $2 \times 2$ block, were addressed adequately in the papers [7-9], for both the standard and the generalized Schur form. But the problem of swapping two $2 \times 2$ blocks still remains a delicate problem, which can fail when restricting oneself to a direct method [1]. In this paper we revisit the problem of swapping two $2 \times 2$ blocks in both the standard and the generalized Schur form. This problem gained a renewed interest because of its use in the new rational $Q Z$ method [2,4] which makes extensive use of the swapping of $2 \times 2$ blocks that may be poorly conditioned. It is therefore needed to ensure that these swappings can be performed in a numerically reliable manner.

In Sections 2 and 3 we revisit the swapping problem for the standard real Schur form, and for the generalized real Schur form and we present a new method to perform the swapping. In both sections we perform a detailed error analysis to bound the relative norm of the off diagonal errors, after the swapping operation. In case these errors are not sufficiently small, we show how to update the swapping by an iterative process to further reduce these errors. In Section 4 we compare the different possible approaches for swapping $2 \times 2$ blocks and we show statistics for the accuracy and need for iterative refinement.
2. The standard eigenvalue problem. The problem we address in this section is the swapping of two $2 \times 2$ blocks in a matrix $A$ in real Schur form. We restrict ourselves to the case of a $4 \times 4$ matrix $A$

$$
A:=\left[\begin{array}{cc}
A_{11} & A_{12}  \tag{2.1}\\
0 & A_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & \times & \times
\end{array}\right]
$$

where the spectra of $A_{11}$ and $A_{22}$ are disjoint and not real

$$
\Lambda\left(A_{11}\right):=\left\{\alpha_{1} \pm \jmath \beta_{1}\right\} \neq \Lambda\left(A_{22}\right):=\left\{\alpha_{2} \pm \jmath \beta_{2}\right\}, \quad \beta_{i} \neq 0, i=1,2
$$

We are looking for a real orthogonal transformation $Q$ such that the transformed matrix $\tilde{A}:=Q^{T} A Q$ has the form

$$
\tilde{A}:=\left[\begin{array}{cc}
\tilde{A}_{11} & \tilde{A}_{12} \\
0 & \tilde{A}_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & \times & \times
\end{array}\right]
$$

with $\Lambda\left(\tilde{A}_{11}\right):=\Lambda\left(A_{22}\right), \quad \Lambda\left(\tilde{A}_{22}\right):=\Lambda\left(A_{11}\right)$.
The suggested construction is to find an orthonormal basis for the invariant subspace of $A$ corresponding to the spectrum $\Lambda\left(A_{22}\right)$, and to complete this to an orthonormal matrix $Q$. This can be obtained by the following method described in $[1,6]$.

Lemma 2.1. Let $A \in \mathbb{R}^{4 \times 4}$ be given as in (2.1), where the spectra of $A_{11}$ and $A_{22}$ are disjoint. Then the invariant subspace of $A$ corresponding to the spectrum of $A_{22}$ is spanned by the matrix

$$
\left[\begin{array}{r}
-X  \tag{2.2}\\
I_{2}
\end{array}\right], \quad \text { where } \quad A_{11} X-X A_{22}=A_{12}
$$

Moreover, the orthogonal similarity transformation $Q$ swaps the spectra of $A_{11}$ and $A_{22}$ if and only if

$$
\left[\begin{array}{r}
-X  \tag{2.3}\\
I_{2}
\end{array}\right]=Q\left[\begin{array}{c}
R \\
0
\end{array}\right]
$$

where $R$ is square and invertible.
The error analysis of Bai and Demmel [1] holds for the algorithmic implementation where $X$ is first solved from the Sylvester equation (2.2), and then $Q$ is constructed from the $Q R$ factorization (2.3).

Lemma 2.2 (Theorem 2 in [1]). Let $\hat{X}$ be the computed solution of the Sylvester equation (2.2), let $E:=-A_{12}-A_{11} \hat{X}+\hat{X} A_{22}$ be its residual and let $\hat{Q}$ be the computed factor of the $Q R$ factorization

$$
\left[\begin{array}{r}
-\hat{X} \\
I_{2}
\end{array}\right]=\hat{Q}\left[\begin{array}{c}
\hat{R} \\
0
\end{array}\right]
$$

Then the computed similarity transformation satisfies

$$
\left[\begin{array}{cc}
\tilde{A}_{11} & \tilde{A}_{12}  \tag{2.4}\\
\Delta & \tilde{A}_{22}
\end{array}\right]:=\hat{Q}^{T}\left[\begin{array}{cc}
A_{11} & A_{12} \\
0 & A_{22}
\end{array}\right] \hat{Q}
$$

where

$$
\|\Delta\|_{2} \leq\|E\|_{F} /\left[1+\sigma_{2}(X)_{2}^{2}\right]
$$

It is pointed out in [1] that their error analysis does not imply that $\|\Delta\|_{2} \leq \epsilon_{M}\|A\|_{2}$, where $\epsilon_{M}$ is the machine precision of the computer used. Nevertheless, the bound is often pessimistic and the observed errors often allow $\Delta$ to be safely dismissed.

In the rational QZ algorithm $[2,4] \mathcal{O}\left(n^{2}\right)$ swaps are executed, and failures occasionally happen due to inaccurate swaps of $2 \times 2$ blocks. In this paper, we try to address this problem when the above direct approach does not work properly.

For this we propose two adaptations of the so-called direct approach. The first one is the construction of the transformation $Q$ from (2.3), the second one is an iterative refinement step to get $\Delta$ from (2.4) as small as possible.

If we perform a singular value decomposition of $X$ then we obtain such a factorization from

$$
X=U_{X} \Sigma_{X} V_{X}^{T} \quad \Longrightarrow \quad Q=\left[\begin{array}{cc}
U_{X} & 0 \\
0 & V_{X}
\end{array}\right]\left[\begin{array}{cc}
C_{X} & S_{X} \\
-S_{X} & C_{X}
\end{array}\right], \quad R=-S_{X}^{-1} V_{X}^{T}
$$

where $S_{X}:=\left(I+\Sigma_{X}^{2}\right)^{-\frac{1}{2}}$ and $C_{X}:=\Sigma_{X} S_{X}$ are diagonal matrices that can be computed elementwise from $\Sigma_{X}$. The recommended procedure for computing the diagonal elements $c_{i}, s_{i}(i=1,2)$ of $C_{X}$ and $S_{X}$ is

$$
\begin{cases}s_{i}:=1 / \sqrt{1+\sigma_{i}^{2}}, & c_{i}:=s_{i} \cdot \sigma_{i}, \\ \text { if } \sigma_{i} \geq 1 \\ c_{i}:=1 / \sqrt{1+1 / \sigma_{i}^{2}}, & s_{i}:=c_{i} / \sigma_{i}, \\ \text { if } \sigma_{i}<1\end{cases}
$$

because this computes the smallest quantities to a higher relative accuracy. In the numerical examples given below, we show that this alternative construction for the transformation $Q$ often gives better results than the use of a simple $Q R$ factorization.

But even for this approach, there is no guarantee that the similarity transformation (2.4) yields a negligible off-diagonal block $\Delta$. We recommend then to perform an updating similarity to further reduce the norm of the block $\Delta$ in $\tilde{A}$. This can be done by applying a similarity transformation:

$$
Q_{u p}^{T} \tilde{A} Q_{u p}=\left[\begin{array}{cc}
\hat{A}_{11} & \hat{A}_{12} \\
0 & \hat{A}_{22}
\end{array}\right], \quad Q_{u p}:=\left[\begin{array}{cc}
I & Y^{T} \\
-Y & I
\end{array}\right]\left[\begin{array}{cc}
R_{Y} & 0 \\
0 & R_{Y^{T}}
\end{array}\right]
$$

where $R_{Y}$ and $R_{Y^{T}}$ are normalizations to make $Q_{u p}$ orthonormal, and where $Y$ is computed from the quadratic matrix equation

$$
Y \tilde{A}_{12} Y+\tilde{A}_{22} Y-Y \tilde{A}_{11}-\Delta=0
$$

Since the updating transformation $Q_{u p}$ is assumed to be close to the identity, $\|Y\|_{2}$ is very small and can then be well approximated by the solution of the linear equation

$$
\tilde{A}_{22} Y-Y \tilde{A}_{11}=\Delta
$$

The solution $Y$ of this linear system can again be computed using Kronecker products, and the implementation of the rotation is best performed via the singular value decomposition of $Y$ since $Y:=U_{Y} \Sigma_{Y} V_{Y}^{T}$ implies

$$
Q_{u p}=\left[\begin{array}{cc}
V_{Y} & 0 \\
0 & U_{Y}
\end{array}\right]\left[\begin{array}{cc}
C_{Y} & S_{Y} \\
-S_{Y} & C_{Y}
\end{array}\right], \quad R_{Y}=C_{Y}^{-1} V_{Y}^{T}, \quad R_{Y^{T}}=C_{Y}^{-1} U_{Y}^{T}
$$

where $C_{Y}:=\left(I+\Sigma_{Y}^{2}\right)^{-\frac{1}{2}}$ and $S_{Y}:=\Sigma_{Y} C_{Y}$ are diagonal matrices that can be computed elementwise from $\Sigma_{Y}$ using the procedure described earlier. The updating transformation will not yield a new off-diagonal black that is exactly zero, but its norm can be expected to be of the order of $\|\Delta\|_{2}^{2}$, since this is the Riccati equation approach for computing invariant subspaces. This would imply that in most cases, one updating transformation $Q_{u p}$ is enough to be able to dismiss the new off-diagonal block. The numerical experiments given below, support this claim.
3. The generalized eigenvalue problem. In the corresponding problem for pencils of matrices we need to swap two $2 \times 2$ pencils in a pencil $\lambda B-A$ in real Schur form, with $B$ and $A$ like

$$
\left[\begin{array}{cc}
B_{11} & B_{12}  \tag{3.1}\\
0 & B_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
0 & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & 0 & \times
\end{array}\right],\left[\begin{array}{cc}
A_{11} & A_{12} \\
0 & A_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & \times & \times
\end{array}\right] .
$$

Since we are looking for two pairs of disjoint complex conjugate eigenvalues, the matrix $B$ is invertible and the spectrum of the above $4 \times 4$ pencil is the union of $\Lambda\left(B_{11}^{-1} A_{11}\right):=\left\{\alpha_{1} \pm \jmath \beta_{1}\right\}, \beta_{1} \neq 0$ and $\Lambda\left(B_{22}^{-1} A_{22}\right):=\left\{\alpha_{2} \pm \jmath \beta_{2}\right\}, \beta_{2} \neq 0$.

We are looking for two real orthogonal transformations $Q$ and $Z$ such that the transformed pencil $\lambda \tilde{B}-\tilde{A}:=Z^{T}(\lambda B-A) Q$ has $\tilde{B}$ and $\tilde{A}$ of the form

$$
\left[\begin{array}{cc}
\tilde{B}_{11} & \tilde{B}_{12} \\
0 & \tilde{B}_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
0 & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & 0 & \times
\end{array}\right],\left[\begin{array}{cc}
\tilde{A}_{11} & \tilde{A}_{12} \\
0 & \tilde{A}_{22}
\end{array}\right]=\left[\begin{array}{cc|cc}
\times & \times & \times & \times \\
\times & \times & \times & \times \\
\hline 0 & 0 & \times & \times \\
0 & 0 & \times & \times
\end{array}\right]
$$

with $\Lambda\left(\tilde{B}_{11}^{-1} \tilde{A}_{11}\right):=\Lambda\left(B_{22}^{-1} A_{22}\right), \quad \Lambda\left(\tilde{B}_{22}^{-1} \tilde{A}_{22}\right):=\Lambda\left(B_{11}^{-1} A_{11}\right)$. The suggested construction is to find a pair of orthonormal bases for the right and left deflating subspaces of $\lambda B-A$ corresponding to the spectrum of $B_{22}^{-1} A_{22}$ and $B_{11}^{-1} A_{11}$, respectively, and to complete these bases to orthonormal transformations $Q$ and $Z$, respectively. This can be obtained by the following direct method, described in [5].

Lemma 3.1. Let the pencil $\lambda B-A$ be given as in (3.1), where the spectra of $B_{11}^{-1} A_{11}$ and $B_{22}^{-1} A_{22}$ are disjoint. Then the right and left deflating subspaces of $\lambda B-A$ corresponding to the spectrum of $B_{22}^{-1} A_{22}$ and $B_{11}^{-1} A_{11}$, respectively, are spanned by the columns and rows of the matrices

$$
\left[\begin{array}{r}
-X  \tag{3.2}\\
I_{2}
\end{array}\right], \quad\left[\begin{array}{ll}
I_{2} & Y
\end{array}\right], \quad \text { where } \quad\left\{\begin{array}{l}
A_{11} X-Y A_{22}=A_{12} \\
B_{11} X-Y B_{22}=B_{12}
\end{array}\right.
$$

Moreover, the orthogonal equivalence transformations $Q$ and $Z$ swap the spectra of the diagonal blocks in $Z^{T}(\lambda B-A) Q$ if and only if

$$
\left[\begin{array}{r}
-X  \tag{3.3}\\
I_{2}
\end{array}\right]=Q\left[\begin{array}{c}
R_{X} \\
0
\end{array}\right], \quad \text { and } \quad\left[\begin{array}{ll}
I_{2} & Y
\end{array}\right]=\left[\begin{array}{ll}
0 & R_{Y}
\end{array}\right] Z^{T}
$$

where $R_{X}$ and $R_{Y}$ are square and invertible.
The error analysis of Kågström's algorithm [5] holds again for a particular algorithmic implementation where $X$ and $Y$ are first solved from the generalized Sylvester equations (3.2), and then $Q$ and $Z$ are constructed from the $Q R$ factorizations (3.3).

Lemma 3.2 (Theorem 3.1 in [5]). Let $\hat{X}$ and $\hat{Y}$ be the computed solutions of the generalized Sylvester equation (3.2), let $E:=-A_{12}-A_{11} \hat{X}+\hat{Y} A_{22}$ and $F:=$ $-B_{12}-B_{11} \hat{X}+\hat{Y} B_{22}$ be their residuals and let $\hat{Q}$ and $\hat{Z}$ be the computed factors of the $Q R$ factorizations

$$
\left[\begin{array}{c}
-\hat{X} \\
I_{2}
\end{array}\right]=\hat{Q}\left[\begin{array}{c}
\hat{R}_{X} \\
0
\end{array}\right], \quad\left[\begin{array}{c}
I_{2} \\
\hat{Y}^{T}
\end{array}\right]=\hat{Z}\left[\begin{array}{c}
0 \\
\hat{R}_{Y}^{T}
\end{array}\right] .
$$

Then the computed equivalence transformation satisfies

$$
\left[\begin{array}{cc}
\lambda \tilde{B}_{11}-\tilde{A}_{11} & \lambda \tilde{B}_{11}-\tilde{A}_{12} \\
\lambda \Delta_{B}-\Delta_{A} & \lambda \tilde{B}_{22}-\tilde{A}_{22}
\end{array}\right]:=\hat{Z}^{T}\left[\begin{array}{cc}
\lambda B_{11}-A_{11} & \lambda B_{12}-A_{12} \\
0 & \lambda B_{22}-A_{22}
\end{array}\right] \hat{Q}
$$

where

$$
\begin{aligned}
& \left\|\Delta_{A}\right\|_{2} \leq\|E\|_{F} / \sqrt{\left(1+\sigma_{2}(X)_{2}^{2}\right)\left(1+\sigma_{2}(Y)_{2}^{2}\right)} \\
& \left\|\Delta_{B}\right\|_{2} \leq\|F\|_{F} / \sqrt{\left(1+\sigma_{2}(X)_{2}^{2}\right)\left(1+\sigma_{2}(Y)_{2}^{2}\right)}
\end{aligned}
$$

It is again pointed out in [5] that this error analysis does not imply that the offdiagonal block $\lambda \Delta_{B}-\Delta_{A}$ can be dismissed. We therefore recommend to apply the same techniques here as for the standard eigenvalue problem: an alternative manner of computing the transformation and an iterative refinement step.

For the constructions of $Q$ and $Z$ from the matrices $X$ and $Y$ we thus recommend to first compute the singular value decompositions

$$
X=U_{X} \Sigma_{X} V_{X}^{T}, \quad Y^{T}=U_{Y} \Sigma_{Y} V_{Y}^{T}
$$

from which we can construct

$$
Q=\left[\begin{array}{cc}
U_{X} & 0 \\
0 & V_{X}
\end{array}\right]\left[\begin{array}{cc}
C_{X} & S_{X} \\
-S_{X} & C_{X}
\end{array}\right], \quad R_{X}=-S_{X}^{-1} V_{X}^{T}
$$

where $S_{X}:=\left(I+\Sigma_{X}^{2}\right)^{-\frac{1}{2}}$ and $C_{X}:=\Sigma_{X} S_{X}$, and

$$
Z=\left[\begin{array}{cc}
V_{Y} & 0 \\
0 & U_{Y}
\end{array}\right]\left[\begin{array}{cc}
C_{Y} & -S_{Y} \\
S_{Y} & C_{Y}
\end{array}\right], \quad R_{Y}^{T}=C_{Y}^{-1} V_{Y}^{T}
$$

where $C_{Y}:=\left(I+\Sigma_{Y}^{2}\right)^{-\frac{1}{2}}$ and $S_{Y}:=\Sigma_{Y} C_{Y}$.
The updating of the equivalence transformation to further reduce the norm of the block $\lambda \Delta_{B}-\Delta_{A}$ in $\lambda \tilde{B}-\tilde{A}$ can be done by applying an equivalence transformation

$$
\begin{gathered}
{\left[\begin{array}{cc}
\lambda \hat{B}_{11}-\hat{A}_{11} & \lambda \hat{B}_{12}-\hat{A}_{12} \\
0 & \lambda \hat{B}_{22}-\hat{A}_{22}
\end{array}\right]:=Z_{u p}^{T}\left[\begin{array}{cc}
\lambda \tilde{B}_{11}-\tilde{A}_{11} & \lambda \tilde{B}_{11}-\tilde{A}_{12} \\
\lambda \Delta_{B}-\Delta_{A} & \lambda \tilde{B}_{22}-\tilde{A}_{22}
\end{array}\right] Q_{u p},} \\
Q_{u p}:=\left[\begin{array}{cc}
I & X^{T} \\
-X & I
\end{array}\right]\left[\begin{array}{cc}
R_{X} & 0 \\
0 & R_{X^{T}}
\end{array}\right], \quad Z_{u p}:=\left[\begin{array}{cc}
I & Y^{T} \\
-Y & I
\end{array}\right]\left[\begin{array}{cc}
R_{Y} & 0 \\
0 & R_{Y^{T}}
\end{array}\right],
\end{gathered}
$$

where $R_{X}, R_{X^{T}} R_{Y}$ and $R_{Y^{T}}$ are normalization factors to make $Q_{u p}$ and $Z_{u p}$ orthonormal. The $(X, Y)$ are computed from the system of quadratic matrix equations

$$
\tilde{\Delta}_{A}-\tilde{A}_{22} X+Y \tilde{A}_{11}-Y \tilde{A}_{12} X=0, \quad \tilde{\Delta}_{B}-\tilde{B}_{22} X+Y \tilde{B}_{11}-Y \tilde{B}_{12} X=0
$$

These can be approximated by the system of linear equations

$$
\tilde{\Delta}_{A}=\tilde{A}_{22} X-Y \tilde{A}_{11}, \quad \tilde{\Delta}_{B}=\tilde{B}_{22} X-Y \tilde{B}_{11}
$$

since $\|X\|_{2}$ and $\|Y\|_{2}$ are very small. The solution $(X, Y)$ of this linear system can be computed using Kronecker products.
4. Numerical results. In the numerical experiments, we limit ourselves to the standard eigenvalue problem because most perturbation phenomena can already be observed there and the numerical examples are easy to generate. We have randomly generated $4 \times 4$ block triangular matrices with two pairs of complex conjugate eigenvalues. The diagonal blocks are generated as:

$$
A_{11}=\left[\begin{array}{cc}
a & b k  \tag{4.1}\\
-b / k & a
\end{array}\right], \quad A_{22}=\left[\begin{array}{cc}
a+r_{1} g & \left(b+r_{2} g\right) k \\
-\left(b+r_{2} g\right) / k & a+r_{1} g
\end{array}\right]
$$

with $a, b, r_{1}$ and $r_{2}$ random values drawn from the standard normal distribution. We abuse notation and denote by $\lambda_{1}$ the complex pair of eigenvalues $a \pm \jmath b$ and by $\lambda_{2}$ the pair $\left(a+r_{1} g\right) \pm \jmath\left(b+r_{2} g\right)$; with $\left|\lambda_{1}-\lambda_{2}\right|$ we denote the maximum of the difference of the corresponding eigenvalues. We use the parameters $g$ and $k$ to change respectively the gap $\left|\lambda_{1}-\lambda_{2}\right|$ and the condition number of the eigenvector. Both $\left|\lambda_{1}-\lambda_{2}\right|$ and $k$ are varied from $10^{-12}$ up to $10^{12}$ in 30 logarithmically spaced increments. For every combination, 20 random matrices have been generated for which the eigenvalues are swapped. Figure 4.1 shows the percentage of times the swap needed iterative refinement when $Q$ was either computed from the SVD of $X$ or from the QR factorization of $\left[\begin{array}{c}-X \\ I\end{array}\right]$. Figure 4.1 indicates that the first method


Fig. 4.1. Percentage of times the swap required iterative refinement.


Fig. 4.2. Relative backward error of the swap.
outperforms the second. Figure 4.2 shows the relative backward error for both cases. This figure indicates backward stability in both cases, but again the first method outperforms the second.

Finally, Figures 4.3 and 4.4 show the relative forward errors on the eigenvalues after the swap for respectively the SVD and $Q R$ update.

The right part of Figure 4.4 shows that there are two regions in the parameter domain where the relative forward error on $\hat{\lambda}_{2}$, computed with the QR update, is of the order of $10^{-12}$ which is significantly higher than the error with the SVD update which is of order $10^{-15}$.
5. Concluding remarks. In this paper we have revisited the problem of swapping two $2 \times 2$ blocks on the diagonal of a standard or generalized Schur form. We have introduced two simple modifications of the direct swapping techniques introduced and analyzed in $[1,5,6]$. The modifications are based on the use of the generalized signular value decomposition of an orthogonal matrix, and the iterative refinement of invariant and deflating subspaces, based on Riccati equations and their linear approximations. These modifications reveal substantial improvements.


Fig. 4.3. Relative forward error on the eigenvalues after the swap computed with the SVD update.


FIG. 4.4. Relative forward error on the eigenvalues after the swap computed with the $Q R$ update.
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