# Computing the Singular Values of 2-by-2 Complex Matrices 

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May 2002


#### Abstract

This paper describes an algorithm for the singular value decomposition of a 2-by-2 complex matrix. It computes accurate singular values.


Keywords Singular value decomposition (SVD), Jacobi method.

## 1 Introduction

In this note we describe an accurate algorithm for singular values of a 2 -by- 2 complex matrix

$$
\begin{equation*}
B=U \Sigma V^{\mathrm{H}} . \tag{1}
\end{equation*}
$$

The singular value decomposition (SVD) of a 2 -by- 2 complex matrix does not occur as frequently as that of a 2-by-2 real upper triangular matrix. The commonly used QR method for SVD consists of two stages. In the first stage, a matrix is reduced to bidiagonal form using, say, Householder transformations on both sides. In the second stage, the bidiagonal matrix resulted from the first stage is diagonalized using QR iterations, where we deal with the SVDs of 2 -by- 2 blocks. Since the first stage, we can assume the 2-by-2 blocks real and upper triangular. However, in Jacobi methods, which are suitable for parallel computing, we have to deal with the SVD of 2-by- 2 complex matrix if the original matrix is complex.

Our algorithm consists of two stages. The first stage reduces $B$ to real and upper triangular:

$$
B=U_{1} R V_{1}^{\mathrm{H}} \quad \text { where } \quad R=\left[\begin{array}{ll}
f & g  \tag{2}\\
0 & h
\end{array}\right]
$$

and $U_{1}$ and $V_{1}$ are unitary. The second stage computes the SVD

$$
\begin{equation*}
R=U_{2} \Sigma V_{2}^{\mathrm{T}}, \tag{3}
\end{equation*}
$$

where $U_{2}$ and $V_{2}$ are orthogonal, using the algorithm by Demmel and Kahan [2]. Then, from (2) and (3), the SVD (1) can be obtained by $U=U_{1} U_{2}$ and $V=V_{1} V_{2}$.

Demmel and Kahan's algorithm computes singular values of a real and upper triangular 2-by-2 matrix accurate to almost machine precision and is efficient, however, no details were given in [2]. In [1], Bai and Demmel briefly described the algorithm and listed its FORTRAN code SLASV2. In this note, we explain SLASV2 and give a C code.

## 2 Triangularization

To triangularize $B=\left[b_{i j}\right]$, we first find a unitary rotation

$$
Q_{1}=\left[\begin{array}{rr}
c & -\bar{s} \\
s & \bar{c}
\end{array}\right] \quad \text { where } \quad|c|^{2}+|s|^{2}=1
$$

such that

$$
Q_{1}^{\mathrm{H}} B=\left[\begin{array}{cc}
f & z_{1} \\
0 & z_{2}
\end{array}\right]
$$

We may assume that $f$ is real. Next, we make the two entries $z_{1}$ and $z_{2}$ real:

$$
Q_{2}^{\mathrm{H}}\left[\begin{array}{ll}
f & z_{1} \\
0 & z_{2}
\end{array}\right] V_{1}=\left[\begin{array}{ll}
f & g \\
0 & h
\end{array}\right],
$$

where

$$
Q_{2}=\left[\begin{array}{cc}
1 & 0 \\
0 & \bar{z}_{1} z_{2} /\left|z_{1} z_{2}\right|
\end{array}\right] \quad \text { and } \quad V_{1}=\left[\begin{array}{cc}
1 & 0 \\
0 & \bar{z}_{1} /\left|z_{1}\right|
\end{array}\right] .
$$

We can see that $g=\left|z_{1}\right|$ and $h=\left|z_{2}\right|$ are real.

## 3 Demmel and Kahan's Algorithm

Let

$$
A \equiv\left[\begin{array}{ll}
a_{11} & a_{12}  \tag{4}\\
a_{12} & a_{22}
\end{array}\right]=R R^{\mathrm{T}}=\left[\begin{array}{cc}
f^{2}+g^{2} & g h \\
g h & h^{2}
\end{array}\right] .
$$

Since (3), $A=U_{2} \Sigma^{2} U_{2}^{\mathrm{T}}$ or $A U_{2}=U_{2} \Sigma^{2}$. Let $U_{2}=\left[\begin{array}{ll}u_{1} & u_{2}\end{array}\right]$ and $\Sigma=$ $\operatorname{diag}\left(\sigma_{1}, \sigma_{2}\right)$, then

$$
\left[\begin{array}{ll}
a_{11} & a_{12} \\
a_{12} & a_{22}
\end{array}\right] u_{i}=\sigma_{i}^{2} u_{i}, \quad i=1,2 .
$$

That is

$$
\left[\begin{array}{cc}
a_{11}-\sigma_{i}^{2} & a_{12} \\
a_{12} & a_{22}-\sigma_{i}^{2}
\end{array}\right] u_{i}=0
$$

It then follows that

$$
\operatorname{det}\left[\begin{array}{cc}
a_{11}-\sigma_{i}^{2} & a_{12} \\
a_{12} & a_{22}-\sigma_{i}^{2}
\end{array}\right]=0
$$

Equivalently, $\sigma_{i}^{2}$ are the zeros of the quadratic polynomial

$$
\lambda^{2}-\left(a_{11}+a_{22}\right) \lambda+a_{11} a_{22}-a_{12}^{2}
$$

Thus $\sigma_{i}$ are the square roots of

$$
\begin{equation*}
\frac{\left(a_{11}+a_{22}\right) \pm \sqrt{\left(a_{11}+a_{22}\right)^{2}-4 a_{11} a_{22}+4 a_{12}^{2}}}{2} \tag{5}
\end{equation*}
$$

Direct use of the above expression may cause unnecessary under/overflow and catastrophic cancellation. It is reorganized as follows to avoid those numerical problems. Expression (5) equals

$$
\frac{4\left(a_{11} a_{22}-a_{12}^{2}\right)}{2\left(\left(a_{11}+a_{22}\right) \mp \sqrt{\left(a_{11}-a_{22}\right)^{2}+4 a_{12}^{2}}\right)}
$$

From (4), substituting $a_{11}, a_{12}$, and $a_{22}$ with $f^{2}+g^{2}, g h$, and $h^{2}$ respectively, we get

$$
\frac{4\left(\left(f^{2}+g^{2}\right) h^{2}-(g h)^{2}\right)}{2\left(f^{2}+g^{2}+h^{2} \mp \sqrt{\left(f^{2}+g^{2}-h^{2}\right)^{2}+4(g h)^{2}}\right)} .
$$

The numerator can be reduced to $4(f h)^{2}$, which is a perfect square. For the denominator, it can be verified that

$$
2\left(f^{2}+g^{2}+h^{2}\right)=(f+h)^{2}+g^{2}+(f-h)^{2}+g^{2}
$$

and

$$
\begin{aligned}
& \left(f^{2}+g^{2}-h^{2}\right)^{2}+4(g h)^{2} \\
= & \left(f^{2}-h^{2}\right)^{2}+g^{4}+2\left(f^{2}-h^{2}\right) g^{2}+4(g h)^{2} \\
= & \left(f^{2}-h^{2}\right)^{2}+g^{4}+2\left(f^{2}+h^{2}\right) g^{2} \\
= & (f+h)^{2}(f-h)^{2}+g^{4}+\left((f+h)^{2}+(f-h)^{2}\right) g^{2} \\
= & \left((f+h)^{2}+g^{2}\right)\left((f-h)^{2}+g^{2}\right)
\end{aligned}
$$

So, $\left.2\left(f^{2}+g^{2}+h^{2}\right) \mp \sqrt{\left(f^{2}+g^{2}-h^{2}\right)^{2}+4(g h)^{2}}\right)$ is also a perfect square $\left(\sqrt{(f+h)^{2}+g^{2}} \mp \sqrt{(f-h)^{2}+g^{2}}\right)^{2}$. Thus the singular values

$$
\sigma_{i}=\frac{2|f h|}{\left|\sqrt{(f+h)^{2}+g^{2}} \mp \sqrt{(f-h)^{2}+g^{2}}\right|} .
$$

The smaller singular value is given by

$$
\begin{equation*}
\sigma_{2}=\frac{2|f h|}{\sqrt{(f+h)^{2}+g^{2}}+\sqrt{(f-h)^{2}+g^{2}}}, \tag{6}
\end{equation*}
$$

and the larger one is

$$
\sigma_{1}=|f h| / \sigma_{2} .
$$

Now, we describe the computation of $\sigma_{2}$. For simplicity, we assume $f, g, h \geq 0$.

First, we make $f \geq h$. In the trivial case when $g=0, \sigma_{1}=f$ and $\sigma_{2}=h$. When $g \neq 0$ and $f / g<\epsilon_{M}$, where $\epsilon_{M}$ is the machine precision, $g$ is very large comparing with $f$ and $h$, then $\sigma_{1}=g$. To avoid unnecessary overflow, $\sigma_{2}$ is computed by

$$
\sigma_{2}= \begin{cases}f /(g / h) & h>1.0 \\ (f / g) h & h \leq 1.0\end{cases}
$$

Then, we consider the normal case when $g$ is not very large. Since $f \geq h$, we scale (6) and get

$$
\sigma_{2}=\frac{h}{\frac{1}{2}\left(\sqrt{s^{2}+q^{2}}+\sqrt{d^{2}+q^{2}}\right)},
$$

where we define the quotient $q=g / f$, the difference $d=(f-h) / f$, which is computed by

$$
d= \begin{cases}1.0 & \text { if } f-h=f \\ (f-h) / f & \text { otherwise }\end{cases}
$$

to deal with the case when $f=\infty$, and the sum $s=(f+h) / f=2-d$. Let

$$
a=\frac{1}{2}\left(\sqrt{d^{2}+q^{2}}+\sqrt{s^{2}+q^{2}}\right)
$$

then

$$
\sigma_{2}=h / a \quad \text { and } \quad \sigma_{1}=f a
$$

Now we consider the right singular vectors $v_{i}(i=1,2)$ in $V_{2}=\left[\begin{array}{ll}v_{1} & v_{2}\end{array}\right]$. From (3),

$$
R^{\mathrm{T}} R=\left[\begin{array}{ll}
v_{1} & v_{2}
\end{array}\right]\left[\begin{array}{cc}
\sigma_{1}^{2} & 0 \\
0 & \sigma_{2}^{2}
\end{array}\right]\left[\begin{array}{ll}
v_{1} & v_{2}
\end{array}\right]^{\mathrm{T}},
$$

we have

$$
\left[\begin{array}{cc}
f^{2}-\sigma_{1}^{2} & f g \\
f g & g^{2}+h^{2}-\sigma_{1}^{2}
\end{array}\right] v_{1}=0
$$

This shows that the vector

$$
x=\left[\begin{array}{c}
f^{2}-\sigma_{1}^{2} \\
f g
\end{array}\right]
$$

is orthogonal to $v_{1}$. It then follows that $v_{2}$ is parallel to $x$. Let

$$
V_{2}=\left[\begin{array}{rr}
c_{v} & -s_{v} \\
s_{v} & c_{v}
\end{array}\right] \quad c_{v}^{2}+s_{v}^{2}=1
$$

then $\left[s_{v} c_{v}\right]$ is parallel to $\left[-f^{2}+\sigma_{1}^{2} f g\right]$, which is parallel to $[2(a-1)(a+$ 1)/ $q$ 2]. To avoid cancellation, we derive

$$
\begin{aligned}
\frac{2 a-2}{q} & =\frac{\sqrt{s^{2}+q^{2}}+\sqrt{d^{2}+q^{2}}-s-d}{q} \\
& =\frac{\sqrt{s^{2}+q^{2}}-s}{q}+\frac{\sqrt{d^{2}+q^{2}}-d}{q} \\
& =\frac{q}{\sqrt{s^{2}+q^{2}}+s}+\frac{q}{\sqrt{d^{2}+q^{2}}+d}
\end{aligned}
$$

Let

$$
t=\left(\frac{q}{\sqrt{s^{2}+q^{2}}+s}+\frac{q}{\sqrt{d^{2}+q^{2}}+d}\right)(1+a) .
$$

When $q^{2}$ underflows, $\sqrt{s^{2}+q^{2}}=s, \sqrt{d^{2}+q^{2}}=d$, and $a=(s+d) / 2=1$, thus

$$
\begin{aligned}
t & =\left(\frac{q}{2 s}+\frac{q}{2 d}\right)(1+a) \\
& =\frac{q}{s}+\frac{q}{d} \\
& =\frac{g}{f-h}+\frac{q}{s} .
\end{aligned}
$$

In the above, since $q=g / f$ is small, possibly underflows, but $d=(f-h) / f$ may also be small. So, $q / d$ is computed by $q / d=g /(f-h)$. Since $s=$ $(f+h) / f \geq 1$, we have no problem with the computation of $q / s$.

When $d=(f-h) / f$ underflows, i.e., $f$ and $h$ are very close, we have $\sqrt{d^{2}+q^{2}}=q$ and $\sqrt{s^{2}+q^{2}}=s=2-d=2$. Consequently, $a=(2+q) / 2=$ $1+q / 2$. Since $q^{2}$ underflows, $q<\epsilon_{M}$, thus $a=1.0$. Therefore

$$
\begin{aligned}
t & =\frac{f}{g}\left(\sqrt{s^{2}+q^{2}}+\sqrt{d^{2}+q^{2}}-2\right)(1+a) \\
& =\frac{f}{g} q(a+1) \\
& =2.0
\end{aligned}
$$

In summary,
$t= \begin{cases}2 & \text { if } q^{2}=0 \text { and } d=0 \\ g /(f-h)+q / s & \text { if } q^{2}=0 \text { and } d \neq 0 \\ \left(q /\left(\sqrt{s^{2}+q^{2}}+s\right)+q /\left(\sqrt{d^{2}+q^{2}}+d\right)\right)(1+a) & \text { if } q^{2} \neq 0 .\end{cases}$
Finally, $c_{v}=2 / \sqrt{t^{2}+4}$ and $s_{v}=t / \sqrt{t^{2}+4}$.
To compute $c_{u}$ and $s_{u}$ from $c_{v}$ and $s_{v}$, we let

$$
U_{2}=\left[\begin{array}{rr}
c_{u} & -s_{u} \\
s_{u} & c_{u}
\end{array}\right] \quad c_{u}^{2}+s_{u}^{2}=1
$$

then from (3) we have

$$
\left[\begin{array}{ll}
f & g \\
0 & h
\end{array}\right]\left[\begin{array}{l}
c_{v} \\
s_{v}
\end{array}\right]=\sigma_{1}\left[\begin{array}{l}
c_{u} \\
s_{u}
\end{array}\right]
$$

which leads to

$$
c_{u}=\left(f c_{v}+g s_{v}\right) / \sigma_{1}=\left(c_{v}+q s_{v}\right) / a
$$

and

$$
s_{u}=h s_{v} / \sigma_{1}=(h / f)\left(s_{v} / a\right),
$$

which guarantees $\left|s_{u}\right| \leq 1$ since $h / f \leq 1$ and $a \geq 1$.
The model used for estimating accuracy is the IEEE floating-point standard: the computed result is the same as if the operation is carried out exactly and then round the result to a floating-point number. For example, floating-point addition of two positive numbers:

$$
f l(a+b)=(a+b)(1+\delta), \quad|\delta| \leq u
$$

where $u$ is the unit of round off. Introducing errors to the operands, we have

$$
\begin{aligned}
& f l\left(a\left(1+\delta_{1}\right)+b\left(1+\delta_{2}\right)\right) \\
= & \left(a\left(1+\delta_{1}\right)+b\left(1+\delta_{2}\right)\right)\left(1+\delta_{3}\right) \\
\approx & a+b+a \delta_{1}+b \delta_{2}+(a+b) \delta_{3} \\
\leq & (a+b)\left(1+\max \left(\delta_{1}, \delta_{2}\right)+\delta_{3}\right)
\end{aligned}
$$

If $\left|\delta_{1}\right| \leq c_{1} u$ and $\left|\delta_{2}\right| \leq c_{2} u$ for some positive constants $c_{1}$ and $c_{2}$, then

$$
f l\left(a\left(1+\delta_{1}\right)+b\left(1+\delta_{2}\right)\right) \leq(a+b)(1+(c+1) \delta), \quad|\delta| \leq u
$$

where $c=\max \left(c_{1}, c_{2}\right)$. Thus,

- addition of positive quantities, where the relative error of the result is at most 1 ulp larger than the maximum of the relative errors of the inputs.

Similarly,

- multiplication and division, where the relative error of the result is at most 1 ulp larger than the sum of the relative errors of the inputs;
- square root, where the relative error of the result is at most 1 ulp more than half the relative error of the input.

This completes our explanation.
Finally, we list our C code translated from SLASV2 in [1].

```
#include <stdlib.h>
#include <math.h>
#define EPS 2.1e-16 // double precsion
#define isign(i) ((i)<0 ? (-1) : (+1)) // int sign function
#define sign(x) ((x)<0.0 ? (-1) : (+1)) // float sign function
//-----------------------------------------------------------------------------------
// Lasv2
// SVD of a 2x2 real upper triangular matrix
// [lf g] = [cu -su]*[smax 0 ] [ [ [ [ cv sv]
// [0 h] [su cu] [ 0 smin] [-sv cv]
// smax is the larger singular value and smin is the smaller
```

```
//
// "smin" and "smax" are singular values
// "cv" and "sv" are the entries in the right singular vector matrix
// "cu" and "su" are the entries in the left singular vector matrix
// "f", "g", and "h" are the entries in the upper triangular matrix
//
// This code is translated from the FORTRAN code SLASV2 listed in
// Z.Bai and J.Demmel,
// "Computing the Generalized Singular Value Decomposition",
// SIAM J. Sci. Comput., Vol. 14, No. 6, pp. 1464-1486, November 1993
//-------------------------------------------------------------------------------
void
Lasv2(double *smin, double *smax, double *sv, double *cv,
            double *su, double *cu, double f, double g, double h)
{
    double svt, cvt, sut, cut; // temporary sv, cv, su, and cu
    double ft = f, gt = g, ht = h; // temporary f, g, h
    double fa = fabs(f), ga = fabs(g), ha = fabs(h);
                                    // |f|, |g|, and |h|
    int pmax = 1,
        swap = 0, // is swapped
        glarge = 0, // is g very large
        tsign;
    // tmp sign
    double fmh,
            d,
            dd,
            q,
            qq,
            s,
            ss,
            spq,
            dpq,
            a;
    double tmp,
            tt;
        // make fa>=ha
        if (fa<ha) {
        pmax = 3;
        tmp = ft; ft = ht; ht = tmp; // swap ft and ht
```

```
    tmp = fa; fa = ha; ha = tmp; // swap fa and ha
    swap = 1;
} // if fa<ha
if (ga==0.0) { // diagonal
    *smin = ha;
    *smax = fa;
    cut = 1.0; sut = 0.0; // identity
    cvt = 1.0; svt = 0.0;
} else { // not diagonal
    if (ga>fa) { // g is the largest entry
        pmax = 2;
        if ((fa/ga)<EPS) { // g is very large
                glarge = 1;
                *smax = ga; // 1 ulp
                if (ha>1.0)
                    *smin = fa/(ga/ha); // 2 ulps
                else
                    *smin = (fa/ga)*ha; // 2 ulps
                cut = 1.0; sut = ht/gt;
                cvt = 1.0; svt = ft/gt;
        } // if g large
    } // if ga>fa
    if (glarge==0) { // normal case
        fmh = fa - ha; // 1 ulp
        if (fmh==fa) // cope with infinite f or h
                d = 1.0;
        else
            d = fmh/fa; // note 0<=d<==1.0, 2 ulps
        q = gt/ft; // note |q|<1/EPS, 1 ulp
        s = 2.0 - d; // note s>=1.0, 3 ulps
        qq = q*q; ss = s*s;
        spq = sqrt(ss + qq); // note 1<=spq<=1+1/EPS, 5 ulps
        if (d==0.0)
                dpq = fabs(q); // 0 ulp
        else
            dpq = sqrt(d*d + qq); // note 0<=dpq<=1+1/EPS, 3.5 ulps
        a = 0.5*(spq + dpq); // note 1<=a<=1 + |q|, 6 ulps
        *smin = ha/a; // 7 ulps
        *smax = fa*a; // 7 ulps
```

```
        if (qq==0.0) { // qq underflow
        if (d==0.0)
            tmp = sign(ft)*2*sign(gt);
                                    // 0 ulp
        else
            tmp = gt/(sign(ft)*fmh) + q/s;
                                    // 6 ulps
            } else {
                tmp = (q/(spq + s) + q/(dpq + d))*(1.0 + a);
                                    // 17 ulps
            } // if qq
            tt = sqrt(tmp*tmp + 4.0);
                // 18.5 ulps
            cvt = 2.0/tt; // 19.5 ulps
            svt = tmp/tt; // 36.5 ulps
            cut = (cvt + svt*q)/a; // 46.5 ulps
            sut = (ht/ft)*svt/a; // 45.5 ulps
        } // if g not large
    } // if ga
    if (swap==1) {
        *cu = svt; *su = cvt;
        *cv = sut; *sv = cut;
    } else {
        *cu = cut; *su = sut;
        *cv = cvt; *sv = svt;
    } // if swap
    // correct the signs of smax and smin
    if (pmax==1) tsign = sign(*cv)*sign(*cu)*sign(f);
    if (pmax==2) tsign = sign(*sv)*sign(*cu)*sign(g);
    if (pmax==3) tsign = sign(*sv)*sign(*su)*sign(h);
    *smax = isign(tsign)*(*smax);
    *smin = isign(tsign*sign(f)*sign(h))*(*smin);
} // Lasv2
```


## References

[1] Zhaojun Bai and James W. Demmel, Computing the generalized singular value decomposition, SIAM J. Sci. Comput., Vol. 14, No. 6, pp. 14641486, November 1993.
[2] James Demmel and W. Kahan, Accurate singular values of bidiagonal matrices, SIAM J. Sci. Stat. Comput., Vol. 11, No. 5, pp. 873-912, September 1990.

