Short manual for the JDQZ-package

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1 Introduction

The JDQZ algorithm computes a number of eigenpairs near a target value of the generalized eigen problem

$$\beta \mathbf{A} \mathbf{x} = \alpha \mathbf{B} \mathbf{x}.\tag{1}$$

The algorithm is described in the paper "Jacobi-Davidson style QR and QZ algorithms for the reduction of matrix pencils" by D.R. Fokkema, G.L.G. Sleijpen, and H.A. van der Vorst [?]. The paper can be obtained from the homepage of Gerard Sleijpen (http://www.math.ruu.nl/people/sleijpen/) or from the homepage of Henk van der Vorst (http://www.math.ruu.nl/people/vorst/). This package is a Fortran 77 implementation of the algorithm. The code has been developed by Diederik Fokkema and has been modified and is currently maintained by Martin van Gijzen.

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2 Obtaining the code

To obtain the code you can send an e-mail to: vangyzen@math.ruu.nl. In return you will receive a uuencoded, gziped and tared file. Save this file as jdqz.tar.gz.uue. You can install the code by typing the following commands:

```
uudecode jdqz.tar.gz.uue
gunzip jdqz.tar.gz
tar xvf jdqz.tar
```

This has created a directory jdqz, with two subdirectories jdlib and jdtest. You can find this manual in the directory jdqz, the code is in the subdirectories. Makefiles are supplied, so if the necessary libraries are present, in particular LAPACK, typing make should make the jdqz-library.

3 Usage of the package

The package is written in Fortran 77 and uses the double complex data-type. It calls routines from the LAPACK and BLAS libraries. These libraries are installed on many machines. LAPACK and BLAS routines can also be obtained from Netlib (http://www.netlib.org/index.html). The directory jdlib contains the actual JDQZ-code, the directory jdtest contains some examples that may be of help in using the code.

3.1 User supplied subroutines

The user has to supply three problem dependent routines: one for the multiplication of a vector with the operator \mathbf{A} , one for multiplication with \mathbf{B} , and one for performing the preconditioning operation. The subroutine to multiply with \mathbf{A} must be called AMUL and must have the following header:

```
subroutine AMUL( n, q, r )

c.... Subroutine to compute r = Aq

c.... integer n
    double complex q(n), r(n)
```

 ${\tt q}$ is the input vector, ${\tt r}$ the output vector. ${\tt n}$ is the dimension of the problem. The subroutine to multiply with ${\tt B}$ must be called BMUL and must have the following header:

Finally, the routine to perform the preconditioning operation must be called PRECON and must have the header

```
subroutine PRECON( n, q )
c....
c... Subroutine to compute q = K^-1 q
c....
integer n
double complex q(n)
```

The preconditioning matrix should be an approximation of the matrix $\mathbf{A} - \tau \mathbf{B}$, with τ the prechosen target value. Preconditioning within the JDQZ algorithm is described in section 3.4 of [?]. Preconditioning is not essential for the correct behavior of the algorithm. It should improve the rate of convergence, but leaving the vector \mathbf{q} untouched should have no influence on the correctness of the results.

Note that data for the matrices, like the nonzero coeffcients and their indices, should be passed via common blocks, not via the parameter lists.

3.2 Calling JDQZ

JDQZ itself can be called with the statement

```
call JDQZ( alpha, beta, eivec, wanted, n, target, eps,

$ kmax, jmax, jmin, method, m, 1, mxmv, maxstep,

$ lock, order, testspace, zwork, lwork )
```

The parameters must be of the following data types:

```
alpha, beta double complex array, size jmax
```

eivec two dimensional double complex array, size $n \times kmax$

wanted logical, scalar n integer, scalar

target double complex, scalar eps double precision, scalar

kmax integer, scalar jmax integer, scalar integer, scalar jmin integer, scalar method m integer, scalar 1 integer, scalar mxmvinteger, scalar integer, scalar maxstep

lock double precision, scalar

order integer, scalar testspace integer, scalar

zwork two dimensional double complex array, size n × lwork

lwork integer, scalar

The meaning of most of the parameters can be found in [?] and we will refer to the appropriate section if possible.

```
alpha, beta
                 Obvious from equation (??)
wanted
                  Compute the converged eigenvectors (if wanted =
                  .true.)
eivec
                  Converged eigenvectors if wanted = .true., else con-
                  verged Schur vectors
                  The size of the problem
                  The value near which the eigenvalues are sought
target
                  Tolerance of the eigensolutions, \|\beta \mathbf{A} \mathbf{x} - \alpha \mathbf{B} \mathbf{x}\|/|\alpha/\beta| < \epsilon
eps
kmax
                  Number of wanted eigensolutions, on output: number of
                  converged eigenpairs
                  Maximum size of the search space
jmax
jmin
                  Minimum size of the search space
                  Linear equation solver:
method
                  GMRES_m, [?]
    1:
    2:
                  BiCGstab(\ell), [?]
                  Maximum dimension of searchspace of GMRES_m
m
1
                  Degree of GMRES-polynomial in Bi-CGstab(\ell)
                  Maximum number of matrix-vector multiplications in
mxmv
                  GMRES_m or BiCGstab(\ell)
                  Maximum number of Jacobi-Davidson iterations
maxstep
lock
                  Tracking parameter (section 2.5.1)
order
                  Selection criterion for Ritz values:
    0:
                  nearest to target
    -1:
                  smallest real part
    1:
                 largest real part
    -2:
                  smallest imaginary part
    2:
                  largest imaginary part
                  Determines how to expand the testspace W
testspace
                  \mathbf{w} = \text{"Standard Petrov"} \times \mathbf{v} \text{ (Section 3.1.1)}
    1:
    2:
                  \mathbf{w} = \text{"Standard 'variable' Petrov"} \times \mathbf{v} \text{ (Section 3.1.2)}
    3:
                  \mathbf{w} = \text{"Harmonic Petrov"} \times \mathbf{v} \text{ (Section 3.5.1)}
                  Workspace
zwork
lwork
                  Size of workspace, >= 4+m+5 \text{ jmax}+3 \text{kmax} if GMRES<sub>m</sub>
                  is used, >= 10 + 6\ell + 5 \text{ jmax} + 3 \text{kmax} if Bi-CGstab(\ell) is
                  used.
```

4 Guidelines for chosing the parameters.

In this section we will try to give some sound guidelines how to choose the parameters. Optimal parameters are of course problem and system dependent, but we will give "on average" reasonable values, based on practical experience.

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eps	10^{-9} , don't take it too large. A relatively large value for eps may cause problems when computing many eigen
	solutions. Moreover, convergence from moderate to high accuracy is fast.
1	·
kmax	It may be wise to take kmax a bit larger than the actual
	number of eigensolutions you want, to avoid missing one
jmax	very problem and memory dependent, reasonable value:
	$3 \times kmax$ with a minimum of 20
jmin	$2 imes \mathtt{kmax}$
method	$2 (BiCGstab(\ell))$
m	30
1	2
mxmv	Very problem dependent, any value in the range 5-100
	is reasonable. Suggestion: 100
maxstep	1000
lock	Take it small to avoid missing eigensolutions, eg. lock =
	10^{-9} .
testspace	3 if a reasonable value for target is known, else take 2

5 Example

In this section we will illustrate the usage of the code by a very simple example. The example can be found in the directory jdtest. For $\bf A$ and $\bf B$ we take diagonal matrices. The action of $\bf A$ and $\bf B$ is described by the following piece of Fortran code:

```
r(i) = i*q(i)
    end do
С
    end
    subroutine BMUL( n, q, r )
C...........
      Subroutine to compute r = Bq
C...............
    integer
               n, i
    double complex q(n), r(n)
С
    do i = 1, n
      r(i) = q(i)/i
    end do
С
    end
```

Obviously, we have for the eigenvalues α/β of (??)

$$\alpha/\beta = i^2 \quad (i = 1, \dots, n) \tag{2}$$

We want to compute the five eigenvalues closest to 31, so target = 31, and kmax = 5. The preconditioning operation should mimic the action of the matrix $(\mathbf{A} - \tau \mathbf{B})^{-1}$. Since we are dealing with diagonal matrices we can take the exact inverse. The preconditioning operation is then described by

We take for the problem size n = 100. We are searching near a target, hence order = 0. For the rest of the parameters we take the values suggested in the previous section:

```
10^{-9}
eps
             20
jmax
jmin
             10
method
             2
             30 (does not matter, is not used)
1
             100
mxmv
             1000
maxstep
             10^{-9}.
lock
testspace
```

At the end we want to compute to compute the norms of the residuals

$$\|\mathbf{r}_i\| = \|\beta_i \mathbf{A} \mathbf{x}_i - \alpha_i \mathbf{B} \mathbf{x}_i\| \tag{3}$$

and therefore need the eigenvectors. Hence wanted = .true.. The parameter lwork is set to 137, the minimum allowed for the choice of parameters. The following piece of code

```
call JDQZ(alpha, beta, eivec, wanted, n, target, eps,
     $
           kmax, jmax, jmin,
           method, m, 1, maxnmv, maxstep,
     $
           lock, order, testspace, zwork, lwork)
С
      elapse = etime( tarray )
С
с...
         Compute the norms of the residuals:
         tmp is a double complex array of size n
с...
      do j = 1, kmax
         call amul ( n, eivec(1,j), residu )
         call zscal ( n, beta(j), residu, 1)
         call bmul (n, eivec(1,j), tmp)
         call zaxpy( n, -alpha(j), tmp, 1, residu, 1 )
         print '("lambda(",i2,"): (",1p,e11.4,",",e11.4,
              " )")', j,alpha(j)/beta(j)
     $
         print '(a30,d13.6)', '||beta Ax - alpha Bx||:',
     $
                dznrm2( n, residu, 1 )
      end do
      write(*,10) tarray(1), elapse
```

```
10 format(1x,'END JDQZ AFTER ',f6.2,' SEC. CPU-TIME AND ', f6.2,
            ' SEC. ELAPSED TIME')
gives as output
lambda(1): (3.6000E+01, 0.0000E+00)
       ||beta Ax - alpha Bx||: 0.109282D-10
lambda(2): (2.5000E+01, 0.0000E+00)
       ||beta Ax - alpha Bx||: 0.200064D-09
lambda(3): (1.6000E+01, 0.0000E+00)
       ||beta Ax - alpha Bx||: 0.207683D-10
lambda(4): (4.9000E+01, 0.0000E+00)
       ||beta Ax - alpha Bx||: 0.373422D-09
lambda(5): (9.0000E+00, 0.0000E+00)
       ||beta Ax - alpha Bx||: 0.526224D-10
END JDQZ AFTER
                0.84 SEC. CPU-TIME AND
                                         0.96 SEC. ELAPSED TIME
```

The preconditioning reduces the computing time considerably. JDQZ without preconditioning gives the same eigenvalues, but for the timing it gives

```
END JDQZ AFTER 3.97 SEC. CPU-TIME AND 4.13 SEC. ELAPSED TIME
```

hence, preconditioning gives a reduction of the computing time of about a factor four for this example.

References

- [1] D.R. Fokkema, G.L.G. Sleijpen, and H.A. van der Vorst, *Jacobi-Davidson style QR and QZ algorithms for the partial reduction of matrix Pencils*. Preprint 941, Department of Mathematics, Utrecht University, January, 1996.
- [2] Y. Saad and M.H. Schultz. GMRES: A generalized minimum residual algorithm for solving nonsymmetric linear systems. SIAM J. Sci. Statist. Comput., 7:856–869, 1986.
- [3] G.L.G. Sleijpen and D.R. Fokkema. BiCGstab(ℓ) for linear equations involving matrices with complex spectrum. ETNA~1:11-32,~1994