nag_real_general_eigensystem (f02bjc)

1. Purpose

nag_real_general_eigensystem (f02bjc) calculates all the eigenvalues and, if required, all the eigenvectors of the generalized eigenproblem $Ax = \lambda Bx$ where A and B are real, square matrices, using the QZ algorithm.

2. Specification

```
#include <nag.h>
#include <nagf02.h>
```

3. Description

All the eigenvalues and, if required, all the eigenvectors of the generalized eigenproblem $Ax = \lambda Bx$ where A and B are real, square matrices, are determined using the QZ algorithm. The QZ algorithm consists of four stages:

- (a) A is reduced to upper Hessenberg form and at the same time B is reduced to upper triangular form.
- (b) A is further reduced to quasi-triangular form while the triangular form of B is maintained.
- (c) The quasi-triangular form of A is reduced to triangular form and the eigenvalues extracted.

This function does not actually produce the eigenvalues $\lambda_j,$ but instead returns α_j and β_j such that

$$\lambda_j = \alpha_j / \beta_j, \qquad j = 1, 2, \dots, n.$$

The division by β_j becomes the responsibility of the user's program, since β_j may be zero indicating an infinite eigenvalue. Pairs of complex eigenvalues occur with α_j/β_j and α_{j+1}/β_{j+1} complex conjugates, even though α_i and α_{j+1} are not conjugate.

(d) If the eigenvectors are required (wanty = \mathbf{TRUE}), they are obtained from the triangular matrices and then transformed back into the original co-ordinate system.

4. Parameters

\mathbf{n}

Input: n, the order of the matrices A and B. Constraint: $\mathbf{n} \geq 1$.

a[n][tda]

Input: the n by n matrix A. Output: the array is overwritten.

tda

Input: the second dimension of the array \mathbf{a} as declared in the function from which nag_real_general_eigensystem is called. Constraint: $\mathbf{tda} \geq \mathbf{n}$.

Constraint: to

b[n][tdb]

Input: the n by n matrix B. Output: the array is overwritten.

\mathbf{tdb}

Input: the second dimension of the array **b** as declared in the function from which nag_real_general_eigensystem is called. Constraint: $tdb \ge n$.

tol

Input: the tolerance used to determine negligible elements. If tol > 0.0, an element will be considered negligible if it is less than tol times the norm of its matrix. If $tol \le 0.0$, machine precision is used in place of tol. A value of tol greater than machine precision may result in faster execution but less accurate results.

alfa[n]

Output: α_i , for $j = 1, 2, \ldots, n$.

beta[n]

Output: β_j , for $j = 1, 2, \ldots, n$.

wantv

Input: wantv must be set to **TRUE** if the eigenvectors are required. If wantv is set to **FALSE** then the array v is not referenced.

v[n][tdv]

Output: if wantv = TRUE, then

(i) if the *j*th eigenvalue is real, the *j*th column of **v** contains its eigenvector;

(ii) if the *j*th and (j + 1)th eigenvalues form a complex pair, the *j*th and (j + 1)th columns of **v** contain the real and imaginary parts of the eigenvector associated with the first eigenvalue of the pair. The conjugate of this vector is the eigenvector for the conjugate eigenvalue.

Each eigenvector is normalised so that the component of largest modulus is real and the sum of squares of the moduli equal one.

If wanty = FALSE, v is not referenced and may be set to the null pointer, i.e., (double *)0.

 $\mathbf{t}\mathbf{d}\mathbf{v}$

Input: the second dimension of the array ${\bf v}$ as declared in the function from which nag_real_general_eigensystem is called.

Constraint: $\mathbf{tdv} \geq \mathbf{n}$ if $\mathbf{wantv} = \mathbf{TRUE}$.

iter[n]

Output: iter[j - 1] contains the number of iterations needed to obtain the *j*th eigenvalue. Note that the eigenvalues are obtained in reverse order, starting with the *n*th.

fail

The NAG error parameter, see the Essential Introduction to the NAG C Library.

5. Error Indications and Warnings

NE_INT_ARG_LT

On entry, **n** must not be less than 1: $\mathbf{n} = \langle value \rangle$.

NE_2_INT_ARG_LT

On entry $\mathbf{tda} = \langle value \rangle$ while $\mathbf{n} = \langle value \rangle$. These parameters must satisfy $\mathbf{tda} \geq \mathbf{n}$. On entry $\mathbf{tdb} = \langle value \rangle$ while $\mathbf{n} = \langle value \rangle$. These parameters must satisfy $\mathbf{tdb} \geq \mathbf{n}$. On entry $\mathbf{tdv} = \langle value \rangle$ while $\mathbf{n} = \langle value \rangle$. These parameters must satisfy $\mathbf{tdv} \geq \mathbf{n}$.

NE_ITERATIONS_QZ

More than $\mathbf{n} \times 30$ iterations are required to determine all the diagonal 1 by 1 or 2 by 2 blocks of the quasi-triangular form in the second step of the QZ algorithm. This failure occurs at the *i*th eigenvalue, $i = \langle value \rangle$. α_j and β_j are correct for j = i + 1, i + 2, ..., n but \mathbf{v} does not contain any correct eigenvectors.

The value of i will be returned in member error of the NAG error structure provided **NAGERR_DEFAULT** is not used as the error parameter.

6. Further Comments

The time taken by the function is approximately proportional to n^3 and also depends on the value chosen for parameter **tol**.

6.1. Accuracy

The computed eigenvalues are always exact for a problem $(A + E)x = \lambda(B + F)x$ where ||E||/||A||and ||F||/||B|| are both of the order of max(tol, ε), tol being defined as in Section 4 and ε being the **machine precision**.

Note: interpretation of results obtained with the QZ algorithm often requires a clear understanding of the effects of small changes in the original data. These effects are reviewed in Wilkinson (1979), in relation to the significance of small values of α_j and β_j . It should be noted that if α_j and β_j are **both** small for any j, it may be that no reliance can be placed on **any** of the computed eigenvalues $\lambda_i = \alpha_i/\beta_i$. The user is recommended to study Wilkinson (1979) and, if in difficulty, to seek expert advice on determining the sensitivity of the eigenvalues to perturbations in the data.

6.2. References

Moler C B and Stewart G W (1973) An Algorithm for Generalized Matrix Eigenproblems SIAM J. Numer. Anal. 10 241–256.

Ward R C (1975) The Combination Shift QZ Algorithm SIAM J. Numer. Anal. 12 835–853. Wilkinson I H (1979) Kronecker's Canonical Form and the QZ Algorithm Lingar Algebra and Ar

Wilkinson J H (1979) Kronecker's Canonical Form and the QZ Algorithm Linear Algebra and Appl. 28 285–303.

7. See Also

None

8. Example

To find all the eigenvalues and eigenvectors of $Ax = \lambda Bx$ where

A	=	/ 3.9	9 4.3	4.3	4.4
		12.5	5 21.5	21.5	26.0
		-34.5	5 - 47.5	-43.5	-46.0
		\ -0.5	5 7.5	3.5	6.0/

and

$$B = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 2 & 3 & 3 & 3 \\ -3 & -5 & -4 & -4 \\ 1 & 4 & 3 & 4 \end{pmatrix}.$$

8.1. Program Text

/* nag_real_general_eigensystem(f02bjc) Example Program

```
* Copyright 1991 Numerical Algorithms Group.
 * Mark 2, 1991.
 */
#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <math.h>
#include <nagf02.h>
#include <nagx02.h>
#define NMAX 8
#define TDA NMAX
#define TDB NMAX
#define TDZ NMAX
main()
{
  Integer i, j, k, n, ip, iter[NMAX];
  Complex alfa[NMAX];
```

```
double beta[NMAX], eps1;
double a[NMAX][TDA], b[NMAX][TDB], z[NMAX][TDZ];
  Boolean matz;
  Vprintf("f02bjc Example Program Results\n");
  Vscanf("%*[^\n]"); /* Skip heading in data file */
Vscanf("%ld", &n);
  if (n>0 && n<=NMAX)
    {
      for (i=0; i<n; ++i)</pre>
         for (j=0; j<n; ++j)
Vscanf("%lf", &a[i][j]);</pre>
       for (i=0; i<n; ++i)</pre>
         for (j=0; j<n; ++j)
Vscanf("%lf",&b[i][j]);</pre>
      matz = TRUE;
       eps1 = X02AJC;
      f02bjc(n, (double *)a, (Integer)TDA, (double *)b, (Integer)TDB, eps1,
               alfa, beta, matz, (double *)z, (Integer)TDZ, iter,
               NAGERR_DEFAULT);
      ip = 0;
       for (i=0; i<n; ++i)</pre>
         {
           Vprintf("Eigensolution %4ld\n",i+1);
           Vprintf("alfa[%ld].re %7.3f",i,alfa[i].re);
Vprintf(" alfa[%ld].im %7.3f",i,alfa[i].im);
           Vprintf(" beta[%ld] %7.3f\n",i,beta[i]);
           if (beta[i] == 0.0)
             Vprintf("lambda is infinite");
           else
              if (alfa[i].im == 0.0)
                {
                  Vprintf("lambda
                                         %7.3f\n",alfa[i].re/beta[i]);
                  Vprintf("Eigenvector\n");
                  for (j=0; j<n; ++j)
    Vprintf("%7.3f\n", z[j][i]);</pre>
                }
              else
                {
                  Vprintf("lambda
                                        %7.3f
                                                  %7.3f\n",
                            alfa[i].re/beta[i], alfa[i].im/beta[i] );
                  Vprintf("Eigenvector\n");
                  k = (Integer)pow((double)-1, (double)(ip+2));
                  for (j=0; j<n; ++j)
                     ſ
                       Vprintf("%7.3f",z[j][i-ip]);
                       Vprintf("%7.3f\n",k*z[j][i-ip+1]);
                     }
                  ip = 1-ip;
                }
         }
      Vprintf("Number of iterations (machine-dependent)\n");
       for (i=0; i<n; ++i)</pre>
         Vprintf("%2ld",iter[i]);
       Vprintf("\n");
       exit(EXIT_SUCCESS);
    }
  else
    ſ
       Vfprintf(stderr,"n is out of range: n = %4ld,\n",n);
       exit(EXIT_FAILURE);
    }
}
```

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8.2. Program Data

f02bjc Example Program Data 4

3.9 4 3	12.5	-34.5	-0.5
4.0	21.0	41.0	7.5
4.3	21.5	-43.5	3.5
4.4	26.0	-46.0	6.0
1.0	2.0	-3.0	1.0
1.0	3.0	-5.0	4.0
1.0	3.0	-4.0	3.0
1.0	3.0	-4.0	4.0

8.3. Program Results

```
f02bjc Example Program Results
Eigensolution
                 1
alfa[0].re 3.801 alfa[0].im 0.000 beta[0] 1.900
            2.000
lambda
Eigenvector
  0.996
 0.006
 0.063
 0.063
                 2
Eigensolution
alfa[1].re 1.563 alfa[1].im 2.084 beta[1] 0.521
         3.000
lambda
                      4.000
Eigenvector
 0.945 0.000
0.189 0.000
 0.113 -0.151
 0.113 -0.151
Eigensolution
                  3
alfa[2].re 3.030 alfa[2].im -4.040 beta[2] 1.010
lambda 3.000 -4.000
Eigenvector
 0.945 0.000
0.189 0.000
0.113 0.151
  0.113 0.151
Eigensolution
                  4
alfa[3].re 4.000 alfa[3].im 0.000 beta[3] 1.000
           4.000
lambda
Eigenvector
  0.988
 0.011
 -0.033
 0.154
Number of iterations (machine-dependent)
 0 0 5 0
```