

- [7] J. Dongarra, J. Di Croz, S. Hammarling, and R. Hanson, "An Extended Set of Fortran Basic Linear Algebra Subprograms," *ACM Trans. Math. Soft.*, 14, 1:1-17, March 1988.
- [8] J. Dongarra, J. Di Croz, S. Hammarling, and R. Hanson, "An Extended Set of Fortran Basic Linear Algebra Subprograms: Model Implementation and Test Programs," *ACM Trans. Math. Soft.*, 14, 1:18-32, March 1988.
- [9] C. L. Lawson, R. J. Hanson, D. R. Kincaid, and F. T. Krogh, "Basic Linear Algebra Subprograms for Fortran Usage," *ACM Trans. Math. Soft.*, 5, 3:308-323, September 1979.

## A.9 Send the Results to Tennessee

Congratulations! You have now finished installing and testing LAPACK. Your participation is greatly appreciated. If possible, results and comments should be sent by electronic mail to

sost@cs.utk.edu

Otherwise, results may be submitted either by sending the authors a hard copy of the output files or by returning the distribution tape with the output files stored on it.

We encourage you to make the LAPACK library available to your users and provide us with feedback from their experiences. You should make it clear that this software is still under development, and parts of it may be changed before the project is completed. The changes may affect the calling sequences of some routines, so the public release of LAPACK is not guaranteed to be compatible with this version.

If you would like to do more, please contact us so that we may coordinate your efforts with the development of the final test release of LAPACK. One option is to look at ways to improve the performance of LAPACK on your machine. If you do not have optimized BLAS, tuning the BLAS would likely have a dramatic effect on performance. Other suggestions on fine-tuning specific algorithms are also welcome. For example, one of our test sites noticed that the row interchanges in the LU factorization routine SGEHRF were degrading performance on the IBM3090 because of the non-unit stride in SSWP [2]. In response we added the auxiliary routine SLASWP to interchange a block of rows, so that users of the IBM3090 could easily replace this routine with one in which the row interchanges are applied to one column at a time.

## References

- [1] E. Anderson, Z. Bai, C. Bischof, J. Demmel, J. Dongarra, J. Du Croz, A. Greenbaum, S. Hammarling, A. McKenney, and D. Sorensen, *Preliminary LAPACK User's Guide*, University of Tennessee, July 1991.
- [2] E. Anderson and J. Dongarra, *LAPACK Working Note 16: Results from the Initial Release of LAPACK* University of Tennessee, CS-89-89, November 1989.
- [3] C. Bischof, J. Demmel, J. Dongarra, J. Du Croz, A. Greenbaum, S. Hammarling, and D. Sorensen, *LAPACK Working Note #5: Provisional Contents*, Argonne National Laboratory, AN-88-38, September 1988.
- [4] Z. Bai, J. Demmel, and A. McKenney, *LAPACK Working Note #3: On the Conditioning of the Nonsymmetric Eigenvalue Problem: Theory and Software*, University of Tennessee, CS-89-86, October 1989.
- [5] J. Dongarra, J. Du Croz, I. Duff, and S. Hammarling, "A Set of Level 3 Basic Linear Algebra Subprograms," *ACM Trans. Math. Soft.*, 16, 1:1-17, March 1990.
- [6] J. Dongarra, J. Du Croz, I. Duff, and S. Hammarling, "A Set of Level 3 Basic Linear Algebra Subprograms: Model Implementation and Test Program," *ACM Trans. Math. Soft.*, 16, 1:18-28, March 1990.

### A.8.3 Timing the Eigensystem Routines

Four input files are provided in each data type for timing the eigensystem routines, one for the nonsymmetric eigenvalue problem, one for the symmetric eigenvalue problem, one for the singular value decomposition, and one for the generalized nonsymmetric eigenvalue problem. For the REAL version, the small data sets are SCEPIMD, SNEPIMD, SSEPIMD, and SSVDIMD and the large data sets are SCFPIMD, SNFPIMD, SSFPIMD, and SSVDIMD. Each of the four input files reads a different set of parameters and the format of the input is indicated by a 3-character code on the first line.

The timing program for eigenvalue/singular value routines accumulates the operation count as the routines are executing using special instrumented versions of the LAPACK routines. The first step in compiling the timing program is therefore to make a library of the instrumented routines.

- a) Compile the files xEIGSRC and create an object library. If you have compiled either the S or C version, you must also compile and include the file SCISRC, and if you have compiled either the D or Z version, you must also compile and include the file DZISRC. If you did not compile the file ALBLASF and include it in your BLAS library as described in Section A.3, you must compile it now and include it in the instrumented LAPACK library.
- b) Compile the files xEIGIME with AEIGIME and link them to your test matrix generator library, the instrumented LAPACK library created in the previous step, your LAPACK library from Section A.5, and your BLAS library in that order (on some systems you may get unsatisfied external references if you specify the libraries in the wrong order). If you have compiled either the S or C version, you must also compile and include the file SCIGIME, and if you have compiled either the D or Z version, you must also compile and include the file DZIGIME.
- c) Make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value, or to restrict the number of tests if you are using a computer with performance in between that of a workstation and that of a supercomputer. Instead of decreasing the matrix dimensions to reduce the time, it would be better to reduce the number of matrix types to be timed, since the performance varies more with the matrix size than with the type. For example, for the nonsymmetric eigenvalue routines, you could use only one matrix of type 4 instead of four matrices of types 1, 3, 4, and 6. See Section 6 for further details.  
Associate the appropriate input file with Fortran unit number 5.
- d) The output file is written to Fortran unit number 6. Associate a suitably named file with this unit number (e.g., SCFPIMOUT, SNFPIMOUT, SSFPIMOUT, and SSVDIMOUT for the four runs of the REAL version).
- e) Run the programs in each data type you are using with the four data sets.
- f) Send the output files to the authors as directed in Section A.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

- c) The output file is written to Fortran unit number 6. Associate a suitably named file with this unit number (e.g., SLIIMOUT, SBLIIMOUT, and SRECIIMOUT for the REAL version).
- e) Run the timing programs in each data type you are using for each of the three input files.
- f) Send the output files to the authors as directed in section A.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

## A8.2 Timing the BLAS

The linear equation timing programs are also used to time the BLAS. Three input files are provided in each data type for timing the Level 2 and 3 BLAS. These input files time the BLAS using the matrix shapes encountered in the LAPACK routines, and we will use the results to analyze the performance of the LAPACK routines. For the REAL version, the small data sets are SBLIIMD, SBLIIMB, and SBLIIMC and the large data sets are SBLIIMAD, SBLIIMBD, and SBLIIMCD. There are three sets of inputs because there are three parameters in the Level 3 BLAS,  $M$ ,  $N$ , and  $K$ , and in most applications one of these parameters is small (on the order of the blocksize) while the other two are large (on the order of the matrix size). In SBLIIMD,  $M$  and  $N$  are large but  $K$  is small, while in SBLIIMB the small parameter is  $M$  and in SBLIIMC the small parameter is  $N$ . The Level 2 BLAS are timed only in the first data set, where  $K$  is also used as the bandwidth for the banded routines.

- a) Make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value. If you modified the values of  $N$  or  $NB$  in Section A.8.1, set  $M$ ,  $N$ , and  $K$  accordingly. The large parameters among  $M$ ,  $N$ , and  $K$  should be the same as the matrix sizes used in timing the linear equation routines, and the small parameter should be the same as the block sizes used in timing the linear equations routines. If necessary, the large data set can be simplified by using only one value of LDA.

Associate the appropriate input file with Fortran unit number 5.

- b) The output file is written to Fortran unit number 6. Associate a suitably named file with this unit number (e.g., SBLIIMACUT, SBLIIMBOUT, and SBLIIMCOUT for the three runs of the REAL version).
- c) Run the timing programs in each data type you are using for each of the three input files.
- d) Send the output files to the authors as directed in Section A.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

larger than those in the small data set, and the large data sets use additional values for parameters such as the block size `NB` and the leading array dimension `LDA`. The small input files end with the four characters `'TIMD` and the large input files end with the characters `'TMD` (except for the `BLAS` timing files, see Section A.8.2).

We encourage you to obtain timing results with the large data sets, as this allows us to compare different machines. If this would take too much time, suggestions for paring back the large data sets are given in the instructions below. We also encourage you to experiment with these timing programs and send us any interesting results, such as results for larger problems or for a wider range of block sizes. The main programs are dimensioned for the large data sets, so the parameters in the main program may have to be reduced in order to run the small data sets on a small machine, or increased to run experiments with larger problems.

The minimum time each subroutine will be timed is set to 0.0 in the large data files and to 0.05 in the small data files, and on many machines this value should be increased. If the timing interval is not long enough, the time for the subroutine after subtracting the overhead may be very small or zero, resulting in megaflop rates that are very large or zero. (To avoid division by zero, the megaflop rate is set to zero if the time is less than or equal to zero.) The minimum time that should be used depends on the machine and the resolution of the clock.

For more information on the timing programs and how to modify the input files, see Section 6.

### A8.1 Timing the Linear Equations Routines

Three input files are provided in each data type for timing the linear equation routines, one for square matrices, one for band matrices, and one for rectangular matrices. The small data sets are in `xLINIMD`, `xBNDIMD`, and `xRECTMD`, and the large data sets are in `xLINIMD`, `xBNDIMD`, and `xRECTMD`.

- a) Compile the files `xLINIME`, and link them to your LAPACK library and your BLAS library or libraries in that order (on some systems you may get unsatisfied external references if you specify the libraries in the wrong order). If you have compiled either the S or C version, you must also compile and include the file `SCNLSIE`, and if you have compiled either the D or Z version, you must also compile and include the file `DZLNLSIE`.
- b) Make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value, or to restrict the size of the tests if you are using a computer with performance in between that of a workstation and that of a supercomputer. The computational requirements can be cut in half by using only one value of `LDA`. If it is necessary to also reduce the matrix sizes or the values of the block size, corresponding changes should be made to the `BLAS` input files (see Section A.8.2).

Associate the appropriate input file with Fortran unit number 5.

- b) The data files for the linear equation test program are called xLINISID. For each of the test programs, associate the appropriate data file with Fortran unit number 5.
- c) The output file is written to Fortran unit number 6. Associate a suitably named file (e.g., SLINIST.OUT) with this unit number.
- d) Run the test program.
- e) Send the output files to the authors as directed in Section A.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

### A7.2 Testing the Eigensystem Routines

- a) Compile the files xEIGSIF and link them to your matrix generator library, your LAPACK library, and your BLAS library or libraries in that order (on some systems you may get unsatisfied external references if you specify the libraries in the wrong order). If you have compiled either the S or C version, you must also compile and include the file SCIGSIF, and if you have compiled either the D or Z version, you must also compile and include the file ZIGSIF.
- b) There are ten sets of data files for the eigensystem test program xBAKISID, xBAE SID, xFCISID, xEDISID, xCGISID, xSBISID, xSGISID, NEPISID, SEPE SID, and SVDISID. Note that three of the input files (NEPISID, SEPE SID, and SVDISID) are used regardless of the data type of the test program. For each run of the test program, associate the appropriate data file with Fortran unit number 5.
- c) The output file is written to Fortran unit number 6. Associate suitably named files with this unit number (e.g., SNEPISID.OUT, SBAKISID.OUT, etc.).
- d) Run the test program.
- e) Send the output files to the authors as directed in Section A.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

## A.8 Run the LAPACK Timing Programs

There are two distinct timing programs for LAPACK routines in each data type, one for the linear equations routines and one for the eigensystem routines. The timing program for the linear equations routines is also used to time the BLAS. We encourage you to conduct these timing experiments in REAL and COMPLEX or in DOUBLE PRECISION and COMPLEX\*16; it is not necessary to send timing results in all four data types.

Two sets of input files are provided, a small set and a large set. The small data sets are appropriate for a standard workstation or other non-vector machine. The large data sets are appropriate for supercomputers, vector computers, and high-performance workstations. We are mainly interested in results from the large data sets, and it is not necessary to run both the large and small sets. The values of  $N$  in the large data sets are about five times

- c) The name of the output file is indicated on the first line of each input file and is currently defined to be `SHLAE2.SUMM` for the REAL Level 2 BLAS, with similar names for the other files. If necessary, edit the name of the output file to ensure that it is valid on your system.
- d) Run the Level 2 and 3 BLAS test programs.

If the tests using the supplied data files were completed successfully, consider whether the tests were sufficiently thorough. For example, on a machine with vector registers, at least one value of  $N$  greater than the length of the vector registers should be used; otherwise, important parts of the compiled code may not be exercised by the tests. If the tests were not successful, either because the program did not finish or the test ratios did not pass the threshold, you will probably have to find and correct the problem before continuing. If you have been testing a system-specific BLAS library, try using the Fortran BLAS for the routines that did not pass the tests. For more details on the BLAS test programs, see [8 and [6]. ]

## A. 5 Create the LAPACK Library

Compile the files `xLASRF` with `ALLANF` and create an object library. If you have compiled either the S or C version, you must also compile and include the files `SCLANF`, `SLAMCH`, and `SECNMF`, and if you have compiled either the D or Z version, you must also compile and include the files `DZLANF`, `DLAMCH`, and `DSECNF`. If you did not compile the file `ALLBLASF` and include it in your BLAS library as described in Section A.3, you must compile it now and include it in your LAPACK library.

## A. 6 Create the Test Matrix Generator Library

Compile the files `xMGENF` and create an object library. If you have compiled either the S or C version, you must also compile and include the file `SCAIGENF`, and if you have compiled either the D or Z version, you must also compile and include the file `DZAIGENF`.

## A. 7 Run the LAPACK Test Programs

There are two distinct test programs for LAPACK routines in each data type, one for the linear equations routines and one for the eigensystem routines. In each data type, there is one input file for testing the linear equation routines and ten input files for testing the eigenvalue routines. For more information on the test programs and how to modify the input files, see Section 5.

### A. 7.1 Testing the Linear Equation Routines

- a) Compile the files `xLINISIF` and link them to your matrix generator library, your LAPACK library, and your BLAS library or libraries in that order (on some systems you may get unsatisfied external references if you specify the libraries in the wrong order).

## A. 2. Installing SECOND and DSECOND

Both the timing routines and the test routines call `SECOND(DSECOND)`, a real function with no arguments that returns the time in seconds from some fixed starting time. Our version of this routine returns only “user time”, and not “user time + system time”. The version of `second` in `SECONF` calls `EIIME`, a Fortran library routine available on some computer systems. If `EIIME` is not available or a better local timing function exists, you will have to provide the correct interface to `SECOND` and `DSECOND` on your machine.

The test program in `TSECONF` performs a million operations using 5000 iterations of the `SANPY` operation  $y := y + \alpha x$  on a vector of length 100. The total time and `mgaflops` for this test is reported, then the operation is repeated including a call to `SECOND` on each of the 5000 iterations to determine the overhead due to calling `SECOND`. Compile `SECONF` and `TSECONF` and run the test program. There is no single right answer, but the times in seconds should be positive and the `mgaflop` ratios should be appropriate for your machine. Repeat this test for `DSECONF` and `DTSECONF` and save `SECOND` and `DSECOND` for inclusion in the `LAPACK` library in Section A.5.

## A. 3 Create the BLAS Library

Ideally, a highly optimized version of the `BLAS` library already exists on your machine. In this case you can go directly to Section A.4 to make the `BLAS` test programs. Otherwise, you must create a library using the files `xBLAS1F`, `xBLAS2F`, `xBLAS3F`, and `ALLBLASF`. You may already have a library containing some of the `BLAS`, but not all (Level 1 and 2, but not Level 3, for example). If so, you should use your local version of the `BLAS` wherever possible and, if necessary, delete the `BLAS` you already have from the provided files. The file `ALLBLASF` must be included if any part of `xBLAS2F` or `xBLAS3F` is used. Compile these files and create an object library.

## A. 4 Run the BLAS Test Programs

Test programs for the Level 2 and 3 `BLAS` are in the files `xBLA2F` and `xBLA3F`. A test program for the Level 1 `BLAS` is not included, in part because only a subset of the original set of Level 1 `BLAS` is actually used in `LAPACK` and the old test program was designed to test the full set of Level 1 `BLAS`. The original Level 1 `BLAS` test program is available from netlib as `TOMS` algorithm 539.

- a) Compile the files `xBLA2F` and `xBLA3F` and link them to your `BLAS` library or libraries. Note that each program includes a special version of the error-handling routine `XERRLA`, which tests the error-exits from the Level 2 and 3 `BLAS`. On most systems this will take precedence at link time over the standard version of `XERRLA` in the `BLAS` library. If this is not the case (the symptom will be that the program stops as soon as it tries to test an error-exit), you must temporarily delete `XERRLA` from `ALLBLASF` and recompile the `BLAS` library.
- b) Each `BLAS` test program has a corresponding data file `xBLA2D` or `xBLA3D`. Associate this file with Fortran unit number 5.



```

Epsilon                =      5.96046E-08
Safe minimum           =      1.17549E-38
Base                   =      2.00000
Precision              =      1.19209E-07
Number of digits in mantissa =      24.0000
Rounding mode         =      1.00000
Minimum exponent      =     -125.000
Underflow threshold   =      1.17549E-38
Largest exponent     =      128.000
Overflow threshold    =      3.40282E+38
Reciprocal of safe minimum =      8.50706E+37

```

On a Gray machine, the safe minimum underflows its output representation and the overflow threshold overflows its output representation, so the safe minimum is printed as 0.00000 and overflow is printed as R. This is normal. If you would prefer to print a representable number, you can modify the test program to print SEMN\*100. and RMAX/100. for the safe minimum and overflow thresholds.

Compile SLAMCH and TSLAMCH and run the test program. If the results from the test program are correct, save SLAMCH for inclusion in the LAPACK library. Repeat these steps with DLAMCH and TDLAMCH. If both tests were successful, go to Section A2.3.

If SLAMCH (or DLAMCH) returns an invalid value, you will have to create your own version of this function. The following options are used in LAPACK and must be set:

- 'B': Base of the machine
- 'E': Epsilon (relative machine precision)
- 'O': Overflow threshold
- 'P': Precision = Epsilon\*Base
- 'S': Safe minimum (often same as underflow threshold)
- 'U': Underflow threshold

Some people may be familiar with RUMCH (DUMCH), a primitive routine for setting machine parameters in which the user must comment out the appropriate assignment statements for the target machine. If a version of RUMCH is on hand, the assignments in SLAMCH can be made to refer to RUMCH using the correspondence

```

SLAMCH('U') = RUMCH(1)
SLAMCH('O') = RUMCH(2)
SLAMCH('E') = RUMCH(3)
SLAMCH('B') = RUMCH(5)

```

The safe minimum returned by SLAMCH('S') is initially set to the underflow value, but if  $1/(\text{overflow}) \geq (\text{underflow})$  it is recomputed as  $(1/(\text{overflow})) * (1 + \epsilon)$ , where  $\epsilon$  is the machine precision.

LSAME	LOGICAL	Test if two characters are the same regardless of case
SLAMCH	REAL	Determine machine-dependent parameters
DLAMCH	DOUBLE PRECISION	Determine machine-dependent parameters
SECON	REAL	Return time in seconds from a fixed starting time
DSECON	DOUBLE PRECISION	Return time in seconds from a fixed starting time

If you are working only in single precision, you do not need to install DLAMCH and DSECON and if you are working only in double precision, you do not need to install SLAMCH and SECON. These five subroutines and their test programs are provided in the files LSAMEF and TLSAMEF, SLAMCHF and TSLAMCHF, etc.

## A. 2. Installing LSAME

LSAME is a logical function with two character parameters, A and B. It returns .TRUE. if A and B are the same regardless of case, or .FALSE. if they are different. For example, the expression

```
LSAME( UPL0, 'U' )
```

is equivalent to

```
( UPL0.EQ.'U' ).OR.( UPL0.EQ.'u' )
```

The supplied version works correctly on all systems that use the ASCII code for internal representations of characters. For systems that use the EBCDIC code, one constant must be changed. For CDC systems with 6-12 bit representation, alternative code is provided in the comments. The test program in TLSAMEF tests all combinations of the same character in upper and lower case for A and B and two cases where A and B are different characters.

Compile LSAMEF and TLSAMEF and run the test program. If LSAME works correctly, the only message you should see is

```
ASCII character set
Tests completed
```

The working version of LSAME should be appended to the file ALIBLASE. This file, which also contains the error handler XERBLA, will be compiled with either the BLAS library in Section A3 or the LAPACK library in Section A5.

## A. 2. Installing SLAMCH and DLAMCH

SLAMCH and DLAMCH are real functions with a single character parameter that indicates the machine parameter to be returned. The test program in TSLAMCHF simply prints out the different values computed by SLAMCH so you need to know something about what the values should be. For example, the output of the test program for SLAMCH on a Sun SPARC station is

- 194. DNEPMD
- 195. DSEPMMD
- 196. DSDMMD
  
- 197. ZCEPMD
- 198. ZNEPMD
- 199. ZSEPMMD
- 200. ZSDMMD

## A Installing LAPACK on a non-Unix System

Installing and testing the non-Unix version of LAPACK involves the following steps:

1. Read the tape.
2. Test and install the machine-dependent routines.
3. Create the BLAS library, if necessary.
4. Run the Level 2 and 3 BLAS test programs.
5. Create the LAPACK library.
6. Create the library of test matrix generators.
7. Run the LAPACK test programs.
8. Run the LAPACK timing programs.
9. Send the results from steps 7 and 8 to the authors at the University of Tennessee.

### A. 1 Read the Tape

Read the tape and assign names to the files, preferably as indicated in the beginning of this appendix. The first file (named `README`) is a list of the files in the order specified in the beginning of this appendix. You will need about 28 megabytes to read in the complete tape. On a Sun SPARC station, the libraries used 14 MB and the LAPACK executable files used 20 MB. In addition, the object files used 18 MB, but the object files can be deleted after creating the libraries and executable files. Your actual space requirements will be less if you do not use all four data types. The total space requirements including the object files is approximately 70 MB for all four data types.

### A. 2 Test and Install the Machine-Dependent Routines.

There are five machine-dependent functions in the test and timing package, at least three of which must be installed. They are

- 159. SEIGIME Timing program for the eigensystem routines
- 160. CEIGIME
- 161. DEIGIME
- 162. ZEIGIME
  
- 163. SEISRCF Instrumented LAPACK routines
- 164. CEISRCF
- 165. DEISRCF
- 166. ZEISRCF
  
- 167. SCISRCF Instrumented auxiliary routines used in S and C versions
- 168. DCISRCF Instrumented auxiliary routines used in D and Z versions
  
- 169. SCEPIMD Data file 1 for timing Generalized Nonsymmetric Eigenvalue Problem
- 170. SNEPIMD Data file 1 for timing Nonsymmetric Eigenvalue Problem
- 171. SSEPIMD Data file 1 for timing Symmetric Eigenvalue Problem
- 172. SVDIMD Data file 1 for timing Singular Value Decomposition
  
- 173. CEPIMD
- 174. CNEPIMD
- 175. SEPIMD
- 176. SVDIMD
  
- 177. DEPIMD
- 178. DNEPIMD
- 179. DSEPIMD
- 180. DSDIMD
  
- 181. ZEPIMD
- 182. ZNEPIMD
- 183. ZSEPIMD
- 184. ZSDIMD
  
- 185. SCEPIMD Data file 2 for timing Generalized Nonsymmetric Eigenvalue Problem
- 186. SNEPIMD Data file 2 for timing Nonsymmetric Eigenvalue Problem
- 187. SSEPIMD Data file 2 for timing Symmetric Eigenvalue Problem
- 188. SVDIMD Data file 2 for timing Singular Value Decomposition
  
- 189. CEPIMD
- 190. CNEPIMD
- 191. SEPIMD
- 192. SVDIMD
  
- 193. DEPIMD

- 123. ZBEIMD
  
- 124. SBEIMD Data file 1-b for timing the BLAS
- 125. DBEIMD
- 126. CBEIMD
- 127. ZBEIMD
  
- 128. SBEIMD Data file 1-c for timing the BLAS
- 129. DBEIMD
- 130. CBEIMD
- 131. ZBEIMD
  
- 132. SLINMD Data file 2 for timing dense square linear equations
- 133. DLINMD
- 134. CLINMD
- 135. ZLINMD
  
- 136. SRECMD Data file 2 for timing dense rectangular linear equations
- 137. DRECMD
- 138. CRECMD
- 139. ZRECMD
  
- 140. SBNDMD Data file 2 for timing banded linear equations
- 141. DBNDMD
- 142. CBNDMD
- 143. ZBNDMD
  
- 144. SBEIMD Data file 2-a for timing the BLAS
- 145. DBEIMD
- 146. CBEIMD
- 147. ZBEIMD
  
- 148. SBEIMD Data file 2-b for timing the BLAS
- 149. DBEIMD
- 150. CBEIMD
- 151. ZBEIMD
  
- 152. SBEIMD Data file 2-c for timing the BLAS
- 153. DBEIMD
- 154. CBEIMD
- 155. ZBEIMD
  
- 156. AEIGMF Auxiliary routines for the eigensystem timing program
- 157. SCIGMF
- 158. DIIGMF

- 87. **SSGSID**     Data file for testing symmetric generalized eigenvalue routines
- 88. **BSGSID**
- 89. **CGSID**
- 90. **ZGSID**
  
- 91. **AEIGSIF**     Auxiliary routines for the eigensystemtest program
- 92. **SEIGSIF**
- 93. **DEIGSIF**
  
- 94. **SEIGSIF**     Test program for eigensystem routines
- 95. **CEIGSIF**
- 96. **DEIGSIF**
- 97. **ZEIGSIF**
  
- 98. **NEPSID**     Data file for testing Nonsymmetric Eigenvalue Problem
- 99. **SEPSID**     Data file for testing Symmetric Eigenvalue Problem
- 100. **SVDSID**     Data file for testing Singular Value Decomposition
  
- 101. **ALINIMF**     Auxiliary routines for the linear system timing program
- 102. **SLINIMF**
- 103. **DLINIMF**
  
- 104. **SLINIMF**     Timing program for linear equations
- 105. **CLINIMF**
- 106. **DLINIMF**
- 107. **ZLINIMF**
  
- 108. **SLINI1D**     Data file 1 for timing dense square linear equations
- 109. **DLINI1D**
- 110. **CLINI1D**
- 111. **ZLINI1D**
  
- 112. **SRECI1D**     Data file 1 for timing dense rectangular linear equations
- 113. **DRECI1D**
- 114. **CRECI1D**
- 115. **ZRECI1D**
  
- 116. **SBNDI1D**     Data file 1 for timing banded linear equations
- 117. **DBNDI1D**
- 118. **CBNDI1D**
- 119. **ZBNDI1D**
  
- 120. **SBEI1MD**     Data file 1-a for timing the BLAS
- 121. **DBEI1MD**
- 122. **CEI1MD**

- 52. ~~DMGENF~~
- 53. ~~ZMGENF~~
  
- 54. ~~AMNISIF~~ Auxiliary routines for the linear equation test program
  
- 55. ~~SLINISIF~~ Test program for linear equation routines
- 56. ~~CLINISIF~~
- 57. ~~DLINISIF~~
- 58. ~~ZLINISIF~~
  
- 59. ~~SLINISID~~ Data file 1 for linear equation test program
- 60. ~~DLINISID~~
- 61. ~~CLINISID~~
- 62. ~~ZLINISID~~
  
- 63. ~~SBAKISID~~ Data file for testing ~~SCBAK~~
- 64. ~~DBAKISID~~ Data file for testing ~~DCBAK~~
- 65. ~~CBAKISID~~ Data file for testing ~~CCBAK~~
- 66. ~~ZBAKISID~~ Data file for testing ~~ZCBAK~~
  
- 67. ~~SBALISID~~ Data file for testing ~~SCBAL~~
- 68. ~~DBALISID~~ Data file for testing ~~DCBAL~~
- 69. ~~CBALISID~~ Data file for testing ~~CCBAL~~
- 70. ~~ZBALISID~~ Data file for testing ~~ZCBAL~~
  
- 71. ~~SECISID~~ Data file for testing eigencondition routines
- 72. ~~DECISID~~
- 73. ~~CECISID~~
- 74. ~~ZECISID~~
  
- 75. ~~SEDISID~~ Data file for testing nonsymmetric eigenvalue driver routines
- 76. ~~DEDISID~~
- 77. ~~CEDISID~~
- 78. ~~ZEDISID~~
  
- 79. ~~SCGISID~~ Data file for testing nonsymmetric generalized eigenvalue routines
- 80. ~~DCGISID~~
- 81. ~~CCGISID~~
- 82. ~~ZCGISID~~
  
- 83. ~~SSBISID~~ Data file for testing ~~SSBIRD~~
- 84. ~~DSBISID~~ Data file for testing ~~DSBIRD~~
- 85. ~~CSBISID~~ Data file for testing ~~CSBIRD~~
- 86. ~~ZSBISID~~ Data file for testing ~~ZSBIRD~~

- 17. DSECNF DSECNF function to return time in seconds
- 18. TIDSECNF Test program for DSECNF
  
- 19. AILBLASF Auxiliary routines for the BLAS (and LAPACK)
  
- 20. SBLAS1F Level 1 BLAS
- 21. CBLAS1F
- 22. DBLAS1F
- 23. ZBLAS1F
  
- 24. SBLAS2F Level 2 BLAS
- 25. CBLAS2F
- 26. DBLAS2F
- 27. ZBLAS2F
  
- 28. SBLAS3F Level 3 BLAS
- 29. CBLAS3F
- 30. DBLAS3F
- 31. ZBLAS3F
  
- 32. SBLA2F Test program for Level 2 BLAS
- 33. CBLA2F
- 34. DBLA2F
- 35. ZBLA2F
  
- 36. SBLA2D Data file for testing Level 2 BLAS
- 37. CBLA2D
- 38. DBLA2D
- 39. ZBLA2D
  
- 40. SBLA3F Test program for Level 3 BLAS
- 41. CBLA3F
- 42. DBLA3F
- 43. ZBLA3F
  
- 44. SBLA3D Data file for testing Level 3 BLAS
- 45. CBLA3D
- 46. DBLA3D
- 47. ZBLA3D
  
- 48. SCAIGEN Auxiliary routines for the test matrix generators
- 49. DZAI GEN
  
- 50. SMIGEN Test matrix generators
- 51. CMIGEN



## Appendix F: Implementation Guide for Non-Unix Systems

In the non-Unix version, the software is distributed on an unlabeled ASCII tape containing 200 files. All files consist of 80-character fixed-length records, with a maximum block size of 8000.

In the installation instructions, each file will be identified by the name given below and we recommend that you assign these names to the files when the tape is read. Files with names ending in 'F' contain Fortran source code; those with names ending in 'D' contain data for input to the test and timing programs. There are two sets of data for each timing run; data file 1 for small, non-vector computers, such as workstations, and data file 2 for large computers, particularly Cray-class supercomputers. All file names have at most eight characters.

The leading one or two characters of the file name generally indicates which of the different versions of the library or test program will use it:

A all four data types  
SC REAL and COMPLEX  
DE DOUBLE PRECISION and COMPLEX\*16  
S: REAL  
D DOUBLE PRECISION  
C COMPLEX  
Z COMPLEX\*16

Many of the files occur in groups of four, corresponding to the four different Fortran floating-point data types, and we will frequently refer to these files generically, using 'x' in place of the first letter (for example, xLASRF).

1. **README** List of files as in this section
2. **ALANF** LAPACK auxiliary routines used in all versions
3. **SCLANF** LAPACK auxiliary routines used in S and C versions
4. **DZLANF** LAPACK auxiliary routines used in D and Z versions
5. **SLASRF** LAPACK routines and auxiliary routines
6. **CLASRF**
7. **DLASRF**
8. **ZLASRF**
9. **ISAMEF** ISAME function to compare two characters
10. **TISAMEF** Test program for ISAME
11. **SLAMCH** SLAMCH function to determine machine parameters
12. **TSLAMCH** Test program for SLAMCH
13. **DLAMCH** DLAMCH function to determine machine parameters
14. **TDLAMCH** Test program for DLAMCH
15. **SECONF** SECONF function to return time in seconds
16. **TSECONF** Test program for SECONF

Test/timing run	Data set	S	C
Linear eqn testing	_test.in	44	101
Eigensystemtesting	rep.in	7	10
	sep.in	30	
	svd.in	41	56
	_ec.in	73	12
	_ed.in	42	
	_gg.in	18	24
	_sg.in	9	13
	_sb.in	1	1
	_bal.in	< 1	< 1
	_bak.in	< 1	< 1
Linear eqn timing	_TIME.in	475	3171
	_TIME2.in	374	1661
	_BAND.in	44	293
BLAS timing	_BLAS.in1	246	1445
	_BLAS.in2	41	283
	_BLAS.in3	45	319
Eigensystemtiming	_NEP.M.in	230	836
	_SEP.M.in	60	757
	_SVD.M.in	73	155
	_CEP.M.in	390	1001

Table 15: Gray **MP**– 1 processor, execution times (in seconds)

## Appendix E: Estimated Time

In this appendix we list the execution times (in seconds) for the test and timing runs on a Sun SPARCstation and on one processor of a Gray VM. For timing, the small data sets were used for the SPARCstation and the large data sets for the Gray VM. The minimum time was set to 0.05 seconds for the Sun and 0.0 seconds for the Gray. The Fortran BLAS were used on the Sun and the Gray BLAS were used on the Gray, except for the complex tests and the timings with the input files CBAND.in and CIIME.in, for which the Fortran BLAS were used on the Gray as well. These times (particularly for the Gray) were obtained on a loaded machine and should be considered rough approximations.

On the Sun, the Fortran files were compiled with `f77 -O`. Tests were done using two versions of the Sun-4 compiler, with an older version, `ctrft2.f`, `chetrs.f`, `cchol.f`, `zchol.f`, and `dstein.f` (from the LAPACK libraries) had to be compiled without optimization. On the Gray, the Fortran files were compiled with `cf77 -Zp` using `cf77 5.0` and `UNICOS 7.0`.

Test/timing run	Data set	S	C	D	Z
Linear eqn testing	_test.in	439	2046	516	1921
Eigensystemtesting	rep.in	43	282	69	314
	sep.in	147	647	267	726
	svd.in	210	1224	347	1341
	_ec.in	340	40	434	44
	_ed.in	154	509	238	621
	_gg.in	91	555	152	610
	_sg.in	55	355	87	340
	_sb.in	5	36	6	34
	_bal.in	<1	<1	<1	<1
_bak.in	<1	<1	<1	<1	
Linear eqn timing	_tim.in	116	941	151	786
	_tim2.in	136	1180	188	983
	_band.in	38	322	56	266
BLAS timing	_blas.in1	120	925	139	779
	_blas.in2	35	182	36	158
	_blas.in3	35	178	36	158
Eigensystemtiming	_neptim.in	64	544	113	596
	_septim.in	41	887	80	907
	_svdtim.in	42	259	60	261
	_geptim.in	163	4239	322	1784

Table 14: Sun SPARCstation execution times (in seconds)

```

        SMALL = SQRT( SMALL )
        LARGE = SQRT( LARGE )
    END IF

```

Users of other machines with similar restrictions on the effective range of usable numbers may have to modify this test so that the square roots are done on their machine as well. In the Unix version, SLABAD is found in LAPACK/SRC and in the non-Unix version it is in SCLAUXF.

In the eigensystem timing program calls are made to the IINPACK and EISPACK equivalents of the LAPACK routines to allow a direct comparison of performance measures. In some cases we have increased the minimum number of iterations in the IINPACK and EISPACK routines to allow them to converge for our test problems, but even this may not be enough. One goal of the LAPACK project is to improve the convergence properties of these routines, so error messages in the output file indicating that a IINPACK or EISPACK routine did not converge should not be regarded with alarm.

In the eigensystem timing program we have equivalenced some work arrays and then passed them to a subroutine, where both arrays are modified. This is a violation of the Fortran 77 standard, which says "if a subprogram reference causes a dummy argument in the referenced subprogram to become associated with another dummy argument in the referenced subprogram neither dummy argument may become defined during execution of the subprogram"<sup>2</sup> If this causes any difficulties, the equivalence can be commented out as explained in the comments for the main eigensystem timing program.

We have added a lot of new software since the second release of LAPACK. Expect a few bugs. We will try to correct them before the public release.

---

<sup>2</sup>ANSI X3.9-1978, sec. 15.9.3.6

## Appendix D: Caveats

In this appendix we list the machine-specific difficulties we have encountered in our own experience with LAPACK. We assume the user has installed the machine-specific routines correctly and that the Level 2 and 3 BLAS test programs have run successfully, so we do not list any warnings associated with those routines.

LAPACK is written in Fortran 77. Prospective users with only a Fortran 66 compiler will not be able to use this package.

Some IBM compilers do not recognize DBLE as a generic function as used in LAPACK. The software tools we use to convert from single precision to double precision convert REAL(C) and AIMG(C), where C is COMPLEX to DBL(Z) and DIMG(Z), where Z is COMPLEX\*16, but IBM compilers use DREAL(Z) and DIMG(Z) to take the real and imaginary parts of a double complex number. IBM users can fix this problem by changing DBLE to DREAL when the argument of DBLE is COMPLEX\*16.

IBM compilers do not permit the data type COMPLEX\*16 in a FUNCTION subprogram definition. The data type on the first line of the function subprogram must be changed from COMPLEX\*16 to DOUBLE COMPLEX for the following functions:

ZBEG from the Level 2 BLAS test program  
ZBEG from the Level 3 BLAS test program  
ZLAIV from the LAPACK library  
ZLRND from the test matrix generator library  
ZLAM from the test matrix generator library  
ZLAM from the test matrix generator library

The functions ZDZIC and ZDZIU from the Level 1 BLAS are already declared DOUBLE COMPLEX. If that doesn't work, try the declaration COMPLEX FUNCTION\*16.

If compiling on a SUN you may run out of space in /tmp (especially when compiling in the LAPACK/SRC directory). Thus, you will need to have your system administrator increase the size of your tmp partition.

We have not included test programs for the Level 1 BLAS. Users should therefore be aware of a common problem in machine-specific implementations of xNRM, the function to compute the 2-norm of a vector. The Fortran version of xNRM avoids underflow or overflow by scaling intermediate results, but some library versions of xNRM are not so careful about scaling. If xNRM is implemented without scaling intermediate results, some of the LAPACK test ratios may be unusually high, or a floating point exception may occur in the problems scaled near underflow or overflow. The solution to these problems is to link the Fortran version of xNRM with the test program.

Some of our test matrices are scaled near overflow or underflow but on the Crays, problems with the arithmetic near overflow and underflow forced us to scale by only the square root of overflow and underflow. The LAPACK auxiliary routine SLABAD (or DLABAD) is called to take the square root of underflow and overflow in cases where it could cause difficulties. We assume we are on a Cray if  $\log_{10}(\text{overflow})$  is greater than 2000 and take the square root of underflow and overflow in this case. The test in SLABAD is as follows:

```
IF( LOG10( LARGE ) .GT. 2000. ) THEN
```

SCRLQ or SCRLQ

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 2mnk - (m+n)k^2 + 2/3k^3 + mk + nk - k^2 - 2/3k \\
 \text{addi ti ons:} \quad 2mnk - (m+n)k^2 + 2/3k^3 + mk - nk + 1/3k \\
 \hline
 \text{total flops:} \quad 4mnk - 2(m+n)k^2 + 4/3k^3 + 2mk - k^2 - 1/3k
 \end{array}$$

SEQRS

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad \mathbf{NHS} [2mn - 1/2n^2 + 5/2n] \\
 \text{addi ti ons:} \quad \mathbf{NHS} [2mn - 1/2n^2 + 1/2n] \\
 \hline
 \text{total flops:} \quad \mathbf{NHS} [4mn - n^2 + 3n]
 \end{array}$$

SCMR, SCMLQ, SCML or SCMLQ (SIDE='L')

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 2nmk - nk^2 + 2nk \\
 \text{addi ti ons:} \quad 2nmk - nk^2 + nk \\
 \hline
 \text{total flops:} \quad 4nmk - 2nk^2 + 3nk
 \end{array}$$

SCMR, SCMLQ, SCML or SCMLQ (SIDE='R')

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 2nmk - mk^2 + mk + nk - 1/2k^2 + 1/2k \\
 \text{addi ti ons:} \quad 2nmk - mk^2 + mk \\
 \hline
 \text{total flops:} \quad 4nmk - 2mk^2 + 2mk + nk - 1/2k^2 + 1/2k
 \end{array}$$

SRIR

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 1/6n^3 + 1/2n^2 + 1/3n \\
 \text{addi ti ons:} \quad 1/6n^3 - 1/2n^2 + 1/3n \\
 \hline
 \text{total flops:} \quad 1/3n^3 + 2/3n
 \end{array}$$

SEHRD

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 5/3n^3 + 1/2n^2 - 7/6n - 13 \\
 \text{addi ti ons:} \quad 5/3n^3 - n^2 - 2/3n - 8 \\
 \hline
 \text{total flops:} \quad 10/3n^3 - 1/2n^2 - 11/6n - 21
 \end{array}$$

SSYRD

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 2/3n^3 + 5/2n^2 - 1/6n - 15 \\
 \text{addi ti ons:} \quad 2/3n^3 + n^2 - 8/3n - 4 \\
 \hline
 \text{total flops:} \quad 4/3n^3 + 3n^2 - 17/6n - 19
 \end{array}$$

SEBRD( $m \geq n$ )

$$\begin{array}{l}
 \text{mul ti pli cati ons:} \quad 2mn^2 - 2/3n^3 + 2n^2 + 20/3n \\
 \text{addi ti ons:} \quad 2mn^2 - 2/3n^3 + n^2 - mn + 5/3n \\
 \hline
 \text{total flops:} \quad 4mn^2 - 4/3n^3 + 3n^2 - mn + 25/3n
 \end{array}$$

SEBRD( $m < n$ )

exchange  $m$  and  $n$  in above

$$\begin{array}{l}
\text{SSYIRF} \quad \text{mul ti pli cati ons:} \quad 1/6n^3 + 1/2n^2 + 10/3n \\
\text{addi ti ons:} \quad 1/6n^3 - 1/6n \\
\hline
\text{total flops:} \quad 1/3n^3 + 1/2n^2 + 19/6n
\end{array}$$

$$\begin{array}{l}
\text{SSYIR} \quad \text{mul ti pli cati ons:} \quad 1/3n^3 + 2/3n \\
\text{addi ti ons:} \quad 1/3n^3 - 1/3n \\
\hline
\text{total flops:} \quad 2/3n^3 + 1/3n
\end{array}$$

$$\begin{array}{l}
\text{SSYIR} \quad \text{mul ti pli cati ons:} \quad \mathbb{NBS} [n^2 + n] \\
\text{addi ti ons:} \quad \mathbb{NBS} [n^2 - n] \\
\hline
\text{total flops:} \quad \mathbb{NBS} [2n^2]
\end{array}$$

$$\begin{array}{l}
\text{SCQRF or SCQLF} (m \geq n) \\
\text{mul ti pli cati ons:} \quad mn^2 - 1/3n^3 + mn + 1/2n^2 + 23/6n \\
\text{addi ti ons:} \quad mn^2 - 1/3n^3 + 1/2n^2 + 5/6n \\
\hline
\text{total flops:} \quad 2mn^2 - 2/3n^3 + mn + n^2 + 14/3n
\end{array}$$

$$\begin{array}{l}
\text{SCQRF or SCQLF} (m \leq n) \\
\text{mul ti pli cati ons:} \quad nm^2 - 1/3m^3 + 2nm - 1/2m^2 + 23/6m \\
\text{addi ti ons:} \quad nm^2 - 1/3m^3 + nm - 1/2m^2 + 5/6m \\
\hline
\text{total flops:} \quad 2nm^2 - 2/3m^3 + 3nm - m^2 + 14/3n
\end{array}$$

$$\begin{array}{l}
\text{SCRF or SCQLF} (m \geq n) \\
\text{mul ti pli cati ons:} \quad mn^2 - 1/3n^3 + mn + 1/2n^2 + 29/6n \\
\text{addi ti ons:} \quad mn^2 - 1/3n^3 + mn - 1/2n^2 + 5/6n \\
\hline
\text{total flops:} \quad 2mn^2 - 2/3n^3 + 2mn + 17/3n
\end{array}$$

$$\begin{array}{l}
\text{SCRF or SCQLF} (m \leq n) \\
\text{mul ti pli cati ons:} \quad nm^2 - 1/3m^3 + 2nm - 1/2m^2 + 29/6m \\
\text{addi ti ons:} \quad nm^2 - 1/3m^3 + 1/2m^2 + 5/6m \\
\hline
\text{total flops:} \quad 2nm^2 - 2/3m^3 + 2nm + 17/3n
\end{array}$$

$$\begin{array}{l}
\text{SRCQR or SRCQL} \\
\text{mul ti pli cati ons:} \quad 2mnk - (m+n)k^2 + 2/3k^3 + 2nk - k^2 - 5/3k \\
\text{addi ti ons:} \quad 2mnk - (m+n)k^2 + 2/3k^3 + nk - mk + 1/3k \\
\hline
\text{total flops:} \quad 4mnk - 2(m+n)k^2 + 4/3k^3 + 3nk - mk - k^2 - 4/3k
\end{array}$$

LAPACK routines:

SCEIRF	mul ti pli cati ons:	$1/2mn^2 - 1/6n^3 + 1/2mn - 1/2n^2 + 2/3n$
	addi ti ons:	$1/2mn^2 - 1/6n^3 - 1/2mn + 1/6n$
	total flops:	$mn^2 - 1/3n^3 - 1/2n^2 + 5/6n$

SCEIR	mul ti pli cati ons:	$2/3n^3 + 1/2n^2 + 5/6n$
	addi ti ons:	$2/3n^3 - 3/2n^2 + 5/6n$
	total flops:	$4/3n^3 - n^2 + 5/3n$

SCEIRS	mul ti pli cati ons:	$\mathbf{NBS} [n^2]$
	addi ti ons:	$\mathbf{NBS} [n^2 - n]$
	total flops:	$\mathbf{NBS} [2n^2 - n]$

SFOIRF	mul ti pli cati ons:	$1/6n^3 + 1/2n^2 + 1/3n$
	addi ti ons:	$1/6n^3 - 1/6n$
	total flops:	$1/3n^3 + 1/2n^2 + 1/6n$

SFOIR	mul ti pli cati ons:	$1/3n^3 + n^2 + 2/3n$
	addi ti ons:	$1/3n^3 - 1/2n^2 + 1/6n$
	total flops:	$2/3n^3 + 1/2n^2 + 5/6n$

SFOIRS	mul ti pli cati ons:	$\mathbf{NBS} [n^2 + n]$
	addi ti ons:	$\mathbf{NBS} [n^2 - n]$
	total flops:	$\mathbf{NBS} [2n^2]$

SFBIRF	mul ti pli cati ons:	$n(1/2k^2 + 3/2k + 1) - 1/3k^3 - k^2 - 2/3k$
	addi ti ons:	$n(1/2k^2 + 1/2k) - 1/3k^3 - 1/2k^2 - 1/6k$
	total flops:	$n(k^2 + 2k + 1) - 2/3k^3 - 3/2k^2 - 5/6k$

SFBIRS	mul ti pli cati ons:	$\mathbf{NBS} [2nk + 2n - k^2 - k]$
	addi ti ons:	$\mathbf{NBS} [2nk - k^2 - k]$
	total flops:	$\mathbf{NBS} [4nk + 2n - 2k^2 - 2k]$



Level 2 BLAS	multipl ications	additi ons	total flops
SGEMV 1,2	$mn$	$mn$	$2mn$
SSYMV 3,4	$n^2$	$n^2$	$2n^2$
SSEMV 3,4	$n(2k+1) - k(k+1)$	$n(2k+1) - k(k+1)$	$n(4k+2) - 2k(k+1)$
SIRMV 3,4,5	$n(n+1)/2$	$(n-1)n/2$	$n^2$
SIBMV 3,4,5	$n(k+1) - k(k+1)/2$	$nk - k(k+1)/2$	$n(2k+1) - k(k+1)$
SIRSV 5	$n(n+1)/2$	$(n-1)n/2$	$n^2$
SIBSV 5	$n(k+1) - k(k+1)/2$	$nk - k(k+1)/2$	$n(2k+1) - k(k+1)$
SGER 1	$mn$	$mn$	$2mn$
SSYR 3	$n(n+1)/2$	$n(n+1)/2$	$n(n+1)$
SSYR2 3	$n(n+1)$	$n^2$	$2n^2 + n$

- 1 - Uses  $m$  multiplies if  $\alpha \neq \pm 1$   
2 - Uses  $m$  multiplies if  $\beta \neq \pm 1$  or 0  
3 - Uses  $n$  multiplies if  $\alpha \neq \pm 1$   
4 - Uses  $n$  multiplies if  $\beta \neq \pm 1$  or 0  
5 - Less  $n$  multiplies if matrix is unit triangular

Table 12: Operation counts for the Level 2 BLAS

Level 3 BLAS	multipl ications	additi ons	total flops
SGEMM	$mkn$	$mkn$	$2mkn$
SSYMM(SIDE='L')	$m^2n$	$m^2n$	$2m^2n$
SSYMM(SIDE='R')	$mn^2$	$mn^2$	$2mn^2$
SSYRK	$kn(n+1)/2$	$kn(n+1)/2$	$kn(n+1)$
SSYRK	$kn^2$	$kn^2 + n$	$2kn^2 + n$
SIRMM(SIDE='L')	$nm(m+1)/2$	$nm(m-1)/2$	$nm^2$
SIRMM(SIDE='R')	$mn(n+1)/2$	$mn(n-1)/2$	$mn^2$
SIRSM(SIDE='L')	$nm(m+1)/2$	$nm(m-1)/2$	$nm^2$
SIRSM(SIDE='R')	$mn(n+1)/2$	$mn(n-1)/2$	$mn^2$

Table 13: Operation counts for the Level 3 BLAS

## Appendix C: Operation Counts for the BLAS and LAPACK

In this appendix we reproduce in tabular form the formulas we have used to compute operation counts for the BLAS and LAPACK routines. In single precision, the functions `SCHE2`, `SCHE3`, `SCPAK` and `SCHA` return the operation counts for the Level 2 BLAS, Level 3 BLAS, LAPACK auxiliary routines, and LAPACK routines, respectively. All four functions are found in the directory `LAPACK/TIMING/LIN` in the Unix version and in `SCINTSTF` in the non-Unix version.

In the tables below we give operation counts for the single precision real dense and banded routines (the counts for the symmetric packed routines are the same as for the dense routines). Separate counts are given for multiplies (including divisions) and additions, and the total is the sum of these expressions. For the complex analogues of these routines, each multiplication would count as 6 operations and each addition as 2 operations, so the total would be different. For the double precision routines, we use the same operation counts as for the single precision real or complex routines.

### Operation Counts for the Level 2 BLAS

The four parameters used in counting operations for the Level 2 BLAS are the matrix dimensions  $m$  and  $n$  and the upper and lower bandwidths  $k_u$  and  $k_l$  for the band routines ( $k$  if symmetric or triangular). An exact count also depends slightly on the values of the scaling factors  $\alpha$  and  $\beta$ , since some common special cases (such as  $\alpha = 1$  and  $\beta = 0$ ) can be treated separately.

The count for `SGMV` from the Level 2 BLAS is as follows:

<code>SGMV</code>	multiplies:	$mn - (m - k_l - 1)(m - k_l)/2 - (n - k_u - 1)(n - k_u)/2$
	additions:	$mn - (m - k_l - 1)(m - k_l)/2 - (n - k_u - 1)(n - k_u)/2$
	total flops:	$2mn - (m - k_l - 1)(m - k_l) - (n - k_u - 1)(n - k_u)$

plus  $m$  multiplies if  $\alpha \neq \pm 1$  and another  $m$  multiplies if  $\beta \neq \pm 1$  or 0. The other Level 2 BLAS operation counts are shown in Table 12.

### Operation Counts for the Level 3 BLAS

Three parameters are used to count operations for the Level 3 BLAS: the matrix dimensions  $m$ ,  $n$ , and  $k$ . In some cases we also must know whether the matrix is multiplied on the left or right. An exact count depends slightly on the values of the scaling factors  $\alpha$  and  $\beta$ , but in Table 13 we assume these parameters are always  $\pm 1$  or 0, since that is how they are used in the LAPACK routines.

### Operation Counts for the LAPACK Routines

The parameters used in counting operations for the LAPACK routines are the matrix dimensions  $m$  and  $n$ , the upper and lower bandwidths  $k_u$  and  $k_l$  for the band routines ( $k$  if symmetric or triangular), and `NRHS`, the number of right hand sides in the solution phase. The operation counts for the LAPACK routines not listed here are not computed by a formula. In particular, the operation counts for the eigenvalue routines are problem dependent and are computed during execution of the timing program.

\_LAQSY equilibrate a symmetric matrix  
 \_LAQTR solve a real or complex quasi-triangular system  
 \_LAR2V apply real plane rotations from both sides to a sequence  
     of 2 by 2 real symmetric matrices  
 \_LARF apply (multiply by) an elementary reflector  
 \_LARFB apply (multiply by) a block reflector  
 \_LARFG generate an elementary reflector  
 \_LARFT form the triangular factor of a block reflector  
 \_LAREX unrolled version of xLARF  
 \_LARCV generate a vector of plane rotations  
 \_LARIG generate a plane rotation  
 \_LARIV apply a vector of plane rotations to a pair of vectors  
 \_LAS2 (real) Compute singular values of a 2 x 2 triangular matrix  
 \_LASCL scale a matrix by C/CFROM  
 \_LASET initializes a matrix to BETA on the diagonal and ALPHA on  
     the off-diagonals  
 \_LASR Apply a sequence of plane rotations to a rectangular matrix  
 \_LASSQ Compute a scaled sum of squares of the elements of a vector  
 \_LASV2 (real) Compute singular values and singular vectors of a 2 x 2 triangular  
     matrix  
 \_LASW Perform a series of row interchanges  
 \_LASY2 solve for a matrix X that satisfies the equation  
      $TL * X + ISGN * X * TR = SCALE * B$   
 \_LASYF compute part of the diagonal pivoting factorization of a symmetric matrix  
 \_LAIBS solve a triangular band system with scaling to prevent overflow  
 \_LAIPS solve a packed triangular system with scaling to prevent overflow  
 \_LAIRD reduce NB rows and columns of a real symmetric or complex Hermitian  
     matrix to tridiagonal form  
 \_LAIBS solve a triangular system with scaling to prevent overflow  
 \_LAIZM apply a Householder matrix generated by xTRQF to a matrix  
 \_LAU2 Unblocked version of \_LAUM  
 \_LAUM Compute the product U\*U or L\*L (blocked version)  
 \_LAPY Add a multiple of a matrix to another matrix  
 \_LARO Initialize a rectangular matrix (usually to zero)

`_UNM2` (complex) multiply by the unitary matrix from `CHQF`  
`_UMR2` (complex) multiply by the unitary matrix from `CHQF`

Other LAPACK auxiliary routines:

`_LABD` (real) returns square root of underflow and overflow if exponent range is large  
`_LABD` reduce `NB` rows or columns of a matrix to upper or lower bidiagonal form  
`_LACN` estimate the norm of a matrix for use in condition estimation  
`_LAPY` copy a matrix to another matrix  
`_LAIV` perform complex division in real arithmetic  
`_LAE2` compute eigenvalues of a 2 x 2 real symmetric or complex Hermitian matrix  
`_LAEB` chases a bulge down an upper Hessenberg block  
`_LAEBZ` compute and use the count of eigenvalues of a symmetric tridiagonal matrix  
`_LAEIN` Use inverse iteration to find a specified right and/or left eigenvector of an upper Hessenberg matrix  
`_LAEQU` perform an orthogonal similarity transformation to standardize a 2 by 2 diagonal block of a quasi-triangular matrix  
`_LAEQZ` unblocked single-/double-shift version of QZ method  
`_LAESY` (complex) Compute eigenvalues and eigenvectors of a complex symmetric 2 x 2 matrix  
`_LAEZ` Compute eigenvalues and eigenvectors of a 2 x 2 real symmetric or complex Hermitian matrix  
`_LAEX` swap adjacent diagonal blocks in a quasi-upper triangular matrix  
`_LAG2` compute the eigenvalues of a 2 by 2 generalized eigenvalue problem with scaling to avoid over-/underflow  
`_LAGC` bulge-chasing for the multishift QZ method  
`_LAGIF` factorizes the matrix  $(T - \lambda I)$   
`_LAGIM` matrix-vector product where the matrix is tridiagonal  
`_LAGIS` solves a system of equations  $(T - \lambda I)x = y$  where  $T$  is a tridiagonal matrix  
`_LAHEF` (complex) compute part of the diagonal pivoting factorization of a Hermitian matrix  
`_LAQRR` Find the Schur factorization of a Hessenberg matrix (modified version of HQR from EISPACK)  
`_LAHD` reduce `NB` columns of a general matrix to Hessenberg form  
`_LAI1` apply one step of incremental condition estimation  
`_LAN2` (real) Solve a 1 x 1 or 2 x 2 linear system  
`_LARD` sort the elements of a vector in increasing or decreasing order  
`_LAPIM` multiply a matrix by a symmetric tridiagonal matrix  
`_LAP2` Compute square root of  $X^{*2} + Y^{*2}$   
`_LAP3` (real) Compute square root of  $X^{*2} + Y^{*2} + Z^{*2}$   
`_LAQCB` equilibrate a general band matrix  
`_LAQCE` equilibrate a general matrix  
`_LAQSB` equilibrate a symmetric band matrix  
`_LAQSP` equilibrate a symmetric packed matrix

\_SBMV (complex) Symmetric band matrix times vector  
 \_SYR (complex) Symmetric rank-1 update  
 \_SPR (complex) Symmetric rank-1 update of a packed matrix  
 ICMAI Find the index of element whose real part has max. abs. value  
 IZMAI Find the index of element whose real part has max. abs. value  
 SCSUM Sum absolute values of a complex vector  
 DCSUM Double precision version of SCSUM  
 \_RSCL (real) Scale a vector by the reciprocal of a constant  
 CSRCL Scale a complex vector by the reciprocal of a real constant  
 ZRSCL Double precision version of CSRCL

Level 2 BLAS versions of the block routines:

\_GBF2 compute the LU factorization of a general band matrix  
 \_GBD2 reduce a general matrix to bidiagonal form  
 \_GHD2 reduce a square matrix to upper Hessenberg form  
 \_GEQ2 compute an LQ factorization without pivoting  
 \_GEQ2 compute a QL factorization without pivoting  
 \_GEQ2 compute a QR factorization without pivoting  
 \_GEQ2 compute an RQ factorization without pivoting  
 \_GEF2 compute the LU factorization of a general matrix  
 \_HGS2 (complex) reduce a Hermitian-definite generalized eigenvalue problem to standard form  
 \_HEID2 (complex) reduce a Hermitian matrix to real tridiagonal form  
 \_HEIF2 (complex) compute diagonal pivoting factorization of a Hermitian matrix  
 \_OR2L (real) generate the orthogonal matrix from xGEQLF  
 \_OR2R (real) generate the orthogonal matrix from xGEQRF  
 \_OR2L (real) generate the orthogonal matrix from xGEQLF  
 \_OR2R (real) generate the orthogonal matrix from xGEQRF  
 \_ORM2L (real) multiply by the orthogonal matrix from xGEQLF  
 \_ORM2R (real) multiply by the orthogonal matrix from xGEQRF  
 \_ORM2L (real) multiply by the orthogonal matrix from xGEQLF  
 \_ORM2R (real) multiply by the orthogonal matrix from xGEQRF  
 \_PBIF2 compute the Cholesky factorization of a positive definite band matrix  
 \_POIF2 compute the Cholesky factorization of a positive definite matrix  
 \_SYGS2 (real) reduce a symmetric-definite generalized eigenvalue problem to standard form  
 \_SYID2 (real) reduce a symmetric matrix to tridiagonal form  
 \_SYIF2 compute the diagonal pivoting factorization of a symmetric matrix  
 \_TRIF2 compute the inverse of a triangular matrix  
 \_UNG2L (complex) generate the unitary matrix from xGEQLF  
 \_UNG2R (complex) generate the unitary matrix from xGEQRF  
 \_UNG2L (complex) generate the unitary matrix from xGEQLF  
 \_UNG2R (complex) generate the unitary matrix from xGEQRF  
 \_UNM2L (complex) multiply by the unitary matrix from xGEQLF  
 \_UNM2R (complex) multiply by the unitary matrix from xGEQRF

## Appendix B: LAPACK Auxiliary Routines

This appendix lists all of the auxiliary routines (except for the BLAS) that are called from the LAPACK routines. These routines are found in the directory LAPACK/SRC in the Unix version and in the files xxLAPACK and xLAPACK in the non-Unix version. Routines specified with an underscore as the first character are available in all four data types (S, D, C, and Z), except those marked (real), for which the first character may be 'S' or 'D', and those marked (complex), for which the first character may be 'C' or 'Z'.

Special subroutines:

XERBLA Error handler for the BLAS and LAPACK routines

Special functions:

ILAEN	INTEGER	Return block size and other parameters
LSAME	LOGICAL	Return .TRUE. if two characters are the same regardless of case
LSAMN	LOGICAL	Return .TRUE. if two character strings are the same regardless of case
SLAMH	REAL	Return single precision machine parameters
DLAMH	DOUBLE PRECISION	Return double precision machine parameters

Functions for computing norms:

_LANCB	General band matrix
_LANGE	General matrix
_LANGT	General tridiagonal matrix
_LANHB	(complex) Hermitian band matrix
_LANHE	(complex) Hermitian matrix
_LANHP	(complex) Hermitian packed matrix
_LANHS	Upper Hessenberg matrix
_LANSB	Symmetric band matrix
_LANSP	Symmetric packed matrix
_LANST	Symmetric tridiagonal matrix
_LANSY	Symmetric matrix
_LANIB	Triangular band matrix
_LANIP	Triangular packed matrix
_LANIR	Trapezoidal matrix

Extensions to the Level 1 and 2 BLAS:

CROF	Apply a plane rotation to a pair of complex vectors, where the cos is real and the sin is complex
CSROF	Apply a real plane rotation to a pair of complex vectors
ZDROF	Double precision version of CSROF
_SSYM	(complex) Symmetric matrix times vector
_SSPM	(complex) Symmetric packed matrix times vector

Orthogonal/unitary transformation routines have also been provided for the reductions that use elementary transformations.

	UN	UP
	QR	QP
GR	×	
GR	×	×
GR	×	
MR	×	
MR	×	×
MR	×	

In addition, a number of driver routines are provided with this release. The naming convention for the driver routines is the same as for the LAPACK routines, but the last 3 characters **YYY** have the following meanings (note an 'X' in the last character position indicates a more expert driver):

- SV factor the matrix and solve a system of equations
- SVX equilibrate, factor, solve, compute error bounds and do iterative refinement, and estimate the condition number
- IS solve over- or underdetermined linear system using orthogonal factorizations
- ISX compute a minimum norm solution using a complete orthogonal factorization (using QR with column pivoting)
- ISS solve least squares problem using the SVD
- EV compute all eigenvalues and/or eigenvectors
- EXX compute selected eigenvalues and eigenvectors
- ES compute all eigenvalues, Schur form and/or Schur vectors
- ESX compute all eigenvalues, Schur form and/or Schur vectors and the conditioning of selected eigenvalues or eigenvectors
- GV compute generalized eigenvalues and/or generalized eigenvectors
- GS compute generalized eigenvalues, Schur form and/or Schur vectors
- SVD compute the SVD and/or singular vectors

The driver routines provided in LAPACK are indicated by the following table:

	GE	GB	GT	FO	FP	FB	PT	HE	HP	HB	ST
								SY	SP	SB	
SV	×	×	×	×	×	×	×	×	×		
SVX	×	×	×	×	×	×	×	×	×		
IS	×										
ISX	×										
ISS	×										
EV	×							×	×	×	×
EXX	×							×	×	×	×
ES	×										
ESX	×										
GV	×							×	×		
GS	×										
SVD	×										

- SYL solve the Sylvester matrix equation
- TRD reduce a symmetric matrix to real symmetric tridiagonal form
- TRF compute a triangular factorization (LU, Cholesky, etc.)
- TRI compute inverse (based on triangular factorization)
- TRS solve systems of linear equations (based on triangular factorization)

Given these definitions, the following table indicates the LAPACK subroutines for the solution of systems of linear equations:

	GE	GB	GF	PO	PP	PB	PT	HE SY	HP SP	TR	TP	TB	UN CR
TRF	×	×	×	×	×	×	×	×	×				
TRS	×	×	×	×	×	×	×	×	×	×	×	×	
RES	×	×	×	×	×	×	×	×	×	×	×	×	
TRI	×			×	×			×	×	×	×		
CON	×	×	×	×	×	×	×	×	×	×	×	×	
EQU	×	×		×	×	×							
QPF	×												
QRF †	×												
QRS †	×												
QQR †													×
MQR †													×

†—also RQ, QL, and IQ

The following table indicates the LAPACK subroutines for finding eigenvalues and eigenvectors or singular values and singular vectors:

	GE	GG	HB	HG	TR	TG	HE SY	HP SP	HB SB	ST	PT	BD
HD	×	×										
TRD							×	×	×			
HD	×											
EQR			×							×	×	
EQZ				×								
EN			×							×		
ENC					×	×						
EBZ										×		
EBF										×		
SQR												×
SEN					×							
SNA					×							
SYL					×							
EXC					×							
BAL	×	×										
BAK	×	×										
GST							×	×				



The last three characters, *YYY*, indicate the computation done by a particular subroutine. Included in this release are subroutines to perform the following computations:

**BAK** back transformation of eigenvectors after balancing  
**BAL** permute and/or balance to isolate eigenvalues  
**BD** reduce to bidiagonal form by orthogonal transformations  
**CON** estimate condition number  
**EBZ** compute selected eigenvalues by bisection  
**FIN** compute selected eigenvectors by inverse iteration  
**EQR** compute eigenvalues and/or the Schur form using the QR algorithm  
**EQU** equilibrate a matrix to reduce its condition number  
**EQZ** compute generalized eigenvalues and/or generalized Schur form by QZ method  
**ERF** compute eigenvectors using the Pal-Walker-Kahan variant of the QL or QR algorithm  
**EXC** compute eigenvectors from Schur factorization  
**EXC** swap adjacent diagonal blocks in a quasi-upper triangular matrix  
**GR** generate the orthogonal/unitary matrix from **GBD**  
**GR** generate the orthogonal/unitary matrix from **GBD**  
**GQ** generate the orthogonal/unitary matrix from **GLQF**  
**QL** generate the orthogonal/unitary matrix from **GLQF**  
**QR** generate the orthogonal/unitary matrix from **GLQF**  
**GQ** generate the orthogonal/unitary matrix from **GLQF**  
**GST** reduce a symmetric-definite generalized eigenvalue problem to standard form  
**GIR** generate the orthogonal/unitary matrix from **xxxTRD**  
**HD** reduce to upper Hessenberg form by orthogonal transformations  
**IQF** compute an IQ factorization without pivoting  
**IQS** compute a minimum norm solution using the IQ factorization  
**MR** multiply by the orthogonal/unitary matrix from **GBD**  
**MR** multiply by the orthogonal/unitary matrix from **GBD**  
**MQ** multiply by the orthogonal/unitary matrix from **GLQF**  
**ML** multiply by the orthogonal/unitary matrix from **GLQF**  
**MR** multiply by the orthogonal/unitary matrix from **GLQF**  
**MQ** multiply by the orthogonal/unitary matrix from **GLQF**  
**MR** multiply by the orthogonal/unitary matrix from **xxxTRD**  
**QIF** compute a QL factorization without pivoting  
**QIS** solve a least squares problem using the QL factorization  
**QIF** compute a QR factorization with column pivoting  
**QIF** compute a QR factorization without pivoting  
**QIS** solve a least squares problem using the QR factorization  
**RES** refine initial solution returned by **TRS** routines  
**RQF** compute an RQ factorization without pivoting  
**RQS** compute a minimum norm solution using the RQ factorization  
**SEN** compute a basis and/or reciprocal condition number (sensitivity) of an invariant subspace  
**SNA** estimate reciprocal condition numbers of eigenvalue/-vector pairs  
**SQR** compute singular values and/or singular vectors using the QR algorithm

## Appendix A: LAPACK Routines

In this appendix, we review the subroutine naming scheme for LAPACK as proposed in [3] and indicate by means of a table which subroutines are included in this release. We also list the driver routines, which are new since release 2.

Each subroutine name in LAPACK is a coded specification of the computation done by the subroutine. All names consist of six characters in the form TXXYY. The first letter, T, indicates the matrix data type as follows:

S	REAL
D	DOUBLE PRECISION
C	COMPLEX
Z	COMPLEX*16 (if available)

The next two letters, XX, indicate the type of matrix. Most of these two-letter codes apply to both real and complex routines; a few apply specifically to one or the other, as indicated below

BD	bi diagonal
GB	general band
GE	general (i.e. unsymmetric, in some cases rectangular)
GG	general matrices, generalized problem (i.e. a pair of general matrices)
GT	general tri diagonal
HB	(complex) Hermitian band
HE	(complex) Hermitian
HG	upper Hessenberg matrix, generalized problem (i.e., a Hessenberg and a triangular matrix)
HP	(complex) Hermitian, packed storage
HS	upper Hessenberg
OR	(real) orthogonal
OP	(real) orthogonal, packed storage
PB	symmetric or Hermitian positive definite band
PO	symmetric or Hermitian positive definite
PP	symmetric or Hermitian positive definite, packed storage
PT	symmetric or Hermitian positive definite tri diagonal
SB	(real) symmetric band
SP	symmetric, packed storage
ST	symmetric tri diagonal
SY	symmetric
TB	triangular band
TG	triangular matrices, generalized problem (i.e., a pair of triangular matrices)
TP	triangular, packed storage
TR	triangular (or in some cases quasi-triangular)
TZ	trapezoidal
UN	(complex) unitary
UP	(complex) unitary, packed storage

- line 2: The number of values of N
- line 3: The values of N the matrix dimension
- line 4: Number of values of the parameters
- line 5: The values for NB, the blocksize
- line 6: The values for NS, the number of shifts
- line 7: The values for MNB, the multishift crossover point
- line 8: The values for MNB, determines minimum blocksize
- line 9: The values for MNBK, also determines minimum blocksize
- line 10: The values for the leading dimension LDA
- line 11: The minimum time (in seconds) that a subroutine will be timed. If TIMMIN is zero, each routine should be timed only once.
- line 12: NYPES, the number of matrix types to be used

If NYPES  $\geq 4$ , all the types are used. If  $0 < \text{NYPES} < 4$ , then line 13 specifies NYPES integer values, which are the numbers of the matrix types to be used. The remaining lines specify a path name and the specific routines to be timed. For the generalized nonsymmetric eigenvalue problem the path names for the four data types are SHG, CHG, DHG, and ZHG. A line to request all the routines in the REAL path has the form

```
SHG  T T T T T T T T T T T T T T T T T
```

where the first 3 characters specify the path name, and up to MAXYP nonblank characters may appear in columns 4-80. If the  $k^{\text{th}}$  such character is 'T' or 't', the  $k^{\text{th}}$  routine will be timed. If at least one but fewer than 18 nonblank characters are specified, the remaining routines will not be timed. If columns 4-80 are blank, all the routines will be timed, so the input line

```
SHG
```

is equivalent to the line above.

The output is in the form of a table which shows the absolute times in seconds, floating point operation counts, and megaflop rates for each routine over all relevant input parameters. For the SHCQZ routine, the table has one line for each different combination of NB, NS, MNB, MNBK and MNBK.

## Acknowledgments

Jim Demmel of the University of California-Berkeley, Sven Hammarling of NAG Ltd., and Alan McKenney of the Courant Institute of Mathematical Sciences, New York University, also contributed to this report.

Four different matrix types are provided for timing the generalized nonsymmetric eigenvalue routines. A variety of matrix types is allowed because the number of iterations to compute the eigenvalues, and hence the timing, can depend on the type of matrix whose eigendecomposition is desired. The matrices used for timing have at least one zero, one infinite, and one singular ( $\alpha = \beta = 0$ ) generalized eigenvalue. The remaining eigenvalues are sometimes real and sometimes complex, distributed in magnitude as follows:

- “clustered” entries  $1, \varepsilon, \dots, \varepsilon$  with random signs;
- evenly spaced entries from 1 down to  $\varepsilon$  with random signs;
- geometrically spaced entries from 1 down to  $\varepsilon$  with random signs;
- eigenvalues randomly chosen from the interval  $(\varepsilon, 1)$ .

## 6.6. Input File for Timing the Generalized Nonsymmetric Eigenproblem

An annotated example of an input file for timing the REAL generalized nonsymmetric eigenproblem routines is shown below

```
GEP:  Data file for timing Generalized Nonsymmetric Eigenvalue Problem
4                                     Number of values of N
50 100 150 200                       Values of N (dimension)
4                                     Number of parameter values
10  10  10  10                       Values of NB (blocksize)
2   2   4   4                       Values of NS (no. of shifts)
200 2   4   4                       Values of MAXB (multishift crossover pt)
200 200 200 10                      Values of MINNB (minimum blocksize)
200 200 200 10                      Values of MINBLK (minimum blocksize)
201 201 201 201                     Values of LDA (leading dimension)
0.0                                  Minimum time in seconds
5                                     Number of matrix types
SHG  T T T T T T T T T T T T T T T T
```

The first line of the input file must contain the characters GEP in columns 1-3. Lines 2-12 are read using list-directed input and specify the following values:

2. `SGHRD(Q)` (LAPACK reduction to generalized upper Hessenberg form, computing  $U$  but not  $V$ .)
3. `SGHRD(Z)` (LAPACK reduction to generalized upper Hessenberg form, computing  $V$  but not  $U$ .)
4. `SGHRD(Q, Z)` (LAPACK reduction to generalized upper Hessenberg form, computing  $U$  and  $V$ .)
5. `SHHQZ(F)` (LAPACK computation of generalized eigenvalues only of a pair of matrices in generalized Hessenberg form)
6. `SHHQZ(S)` (LAPACK computation of generalized Schur form of a pair of matrices in generalized Hessenberg form)
7. `SHHQZ(Q)` (LAPACK computation of generalized Schur form of a pair of matrices in generalized Hessenberg form and  $Q$ )
8. `SHHQZ(Z)` (LAPACK computation of generalized Schur form of a pair of matrices in generalized Hessenberg form and  $Z$ )
9. `SHHQZ(Q, Z)` (LAPACK computation of generalized Schur form of a pair of matrices in generalized Hessenberg form and  $Q$  and  $Z$ )
10. `STGEV(A, L)` (LAPACK computation of the the left generalized eigenvectors of a matrix pair in generalized Schur form)
11. `STGEV(B, L)` (LAPACK computation of the the left generalized eigenvectors of a matrix pair in generalized Schur form back transformed by  $Q$ )
12. `STGEV(A, R)` (LAPACK computation of the the right generalized eigenvectors of a matrix pair in generalized Schur form)
13. `STGEV(B, R)` (LAPACK computation of the the right generalized eigenvectors of a matrix pair in generalized Schur form back transformed by  $Z$ )
14. `QZHES(F)` (ELSPACK reduction to generalized upper Hessenberg form with `MAIZ = FALSE`, so  $V$  is not computed.)
15. `QZHES(T)` (ELSPACK reduction to generalized upper Hessenberg form with `MAIZ = TRUE`, so  $V$  is computed.)
16. `QZIT(F)` (QZIT followed by `QZAL` with `MAIZ = FALSE`: ELSPACK computation of generalized eigenvalues only of a pair of matrices in generalized Hessenberg form)
17. `QZIT(T)` (QZIT followed by `QZAL` with `MAIZ = TRUE`: ELSPACK computation of generalized Schur form of a pair of matrices in generalized Hessenberg form and  $Z$ )
18. `QZMEC` (ELSPACK computation of the the right generalized eigenvectors of a matrix pair in generalized Schur form back transformed by  $Z$ )

- line 2: The number of values of  $M$  and  $N$
- line 3: The values of  $M$  the matrix row dimension
- line 3: The values of  $N$  the matrix column dimension
- line 4: The number of values of the parameters  $NB$  and  $LDA$
- line 5: The values of  $NB$  the blocksize
- line 6: The values of  $LDA$  the leading dimension
- line 7: The minimum time in seconds that a routine will be timed
- line 8:  $NYPES$ , the number of matrix types to be used

If  $0 < NYPES < 5$ , then line 9 specifies  $NYPES$  integer values which are the numbers of the matrix types to be used. The remaining lines specify a path name and the specific computations to be timed. For the SVD, the path names for the four data types are SBD, DBD, CBD, and ZBD. The (optional) characters after the path name indicate the computations to be timed, as in the input file for the nonsymmetric eigenvalue problem.

## 6.6 Timing the Generalized Nonsymmetric Eigenproblem

A separate input file drives the timing codes for the generalized nonsymmetric eigenproblem. The input file specifies

- $N$  the matrix size
- six-tuples of parameter values ( $NB$ ,  $NS$ ,  $MXB$ ,  $MNB$ ,  $MINBK$ ,  $LDA$ ) specifying the block size  $NB$ , the number of shifts  $NS$ , the values of  $MXB$  the minimum blocksize  $MNB$ , the minimum blocksize  $MINBK$  and the leading dimension  $LDA$
- the test matrix types
- the routines or sequences of routines from LAPACK or EISPACK to be timed

The parameters  $NB$ ,  $MNB$ ,  $MINBK$ ,  $NS$ , and  $MXB$  apply only to the QZ iteration routine xHQZ. A goal of this timing code is to determine the values of  $NB$ ,  $NS$ , and  $MXB$  (as well as  $MNB$  and  $MINBK$ ) which maximize the speed of the codes.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main timing program.

Parameter	Description	Value
$MXN$	Maximum value for $N$ , $NB$ , $NS$ , $MXB$ , $MNB$ , or $MINBK$	400
$IDMX$	Maximum value for $LDA$	420
$MXN$	Maximum number of values of $N$	12
$MXPM$	Maximum number of parameter sets ( $NB$ , $NS$ , $MXB$ , $MNB$ , $MINBK$ , $LDA$ )	10

The computations that may be timed for the REAL version are

1. SGBRD (LAPACK reduction to generalized upper Hessenberg form without computing  $U$  or  $V$ .)

13. LINSVD(1) (LINPACKsingular values and min(MN) left singular vectors of a dense matrix using SSVDC; to be compared to LAPSD(1))
14. LINSVD(L) (LINPACKsingular values and Mleft singular vectors of a dense matrix using SSVDC; to be compared to LAPSD(L))
15. LINSVD(R) (LINPACKsingular values and Nright singular vectors of a dense matrix using SSVDC; to be compared to LAPSD(R))
16. LINSVD(B) (LINPACKsingular values, min(MN) left singular vectors and Nright singular vectors of a dense matrix using SSVDC; to be compared to LAPSD(B)).

Five different matrix types are provided for timing the singular value decomposition routines. Matrix types 1-3 are of the form  $UDV$ , where  $U$  and  $V$  are orthogonal or unitary, and  $D$  is diagonal with entries

- evenly spaced entries from 1 down to  $\epsilon$  with random signs (matrix type 1),
- geometrically spaced entries from 1 down to  $\epsilon$  with random signs (matrix type 2), or
- "clustered" entries  $1, \epsilon, \dots, \epsilon$  with random signs (matrix type 3).

Matrix type 4 has in each entry a random number drawn from  $[-1, 1]$ . Matrix type 5 is a nearly bidiagonal matrix, where the upper bidiagonal entries are  $\exp(-2r \log \epsilon)$  and the nonbidiagonal entries are  $r\epsilon$ , where  $r$  is a uniform random number drawn from  $[0, 1]$  (a different  $r$  for each entry).

An annotated example of an input file for timing the REAL singular value decomposition routines is shown below

```

SVD:  Data file for timing Singular Value Decomposition routines
7                                           Number of values of M and N
50  50 100 100 100 200 200                Values of M (row dimension)
50 100  50 100 200 100 200                Values of N (column dimension)
5                                           Number of values of parameters
1   16  32  48  64                        Values of NB (blocksize)
201 201 201 201 201                       Values of LDA (leading dimension)
0.0                                        Minimum time in seconds
4                                           Number of matrix types
1 2 3 4
SBD   T T T T T T T T T T T T T T T T

```

The first line of the input file must contain the characters SVD in columns 1-3. Lines 2-9 are read using list-directed input and specify the following values:

- the test matrix types
- the routines or sequences of routines from LAPACK or LINPACK to be timed.

A goal of this timing code is to determine the values of NB which maximize the speed of the block algorithms.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main timing program

Parameter	Description	Value
MXN	Maximum value for M, N, or NB	400
IDAMX	Maximum value for LDA	420
MXNN	Maximum number of pairs of values (M, N)	12
MXPRM	Maximum number of pairs of values (NB, LDA)	10

The computations that may be timed for the REAL version are

1. SCFBD (LAPACK reduction to bidiagonal form)
2. SDBQR (LAPACK computation of singular values only of a bidiagonal matrix)
3. SDBQR(L) (LAPACK computation of the singular values and left singular vectors of a bidiagonal matrix)
4. SDBQR(R) (LAPACK computation of the singular values and right singular vectors of a bidiagonal matrix)
5. SDBQR(B) (LAPACK computation of the singular values and right and left singular vectors of a bidiagonal matrix)
6. SDBQR(V) (LAPACK computation of the singular values and multiply square matrix of dimension  $\min(M, N)$  by transpose of left singular vectors)
7. LAFSD (LAPACK singular values only of a dense matrix, using SCFBD and SDBQR)
8. LAFSD(L) (LAPACK singular values and  $\min(M, N)$  left singular vectors of a dense matrix, using SCFBD, SCGBR and SDBQR(L))
9. LAFSD(L) (LAPACK singular values and M left singular vectors of a dense matrix, using SCFBD, SCGBR and SDBQR(L))
10. LAFSD(R) (LAPACK singular values and N right singular vectors of a dense matrix, using SCFBD, SCGBR and SDBQR(R))
11. LAFSD(B) (LAPACK singular values,  $\min(M, N)$  left singular vectors, and N right singular vectors of a dense matrix, using SCFBD, SCGBR and SDBQR(B))
12. LINSVD (LINPACK singular values only of a dense matrix using SVD, to be compared to LAFSD)



- evenly spaced entries from 1 down to  $\varepsilon$  with random signs (matrix type 1),
- geometrically spaced entries from 1 down to  $\varepsilon$  with random signs (matrix type 2),
- “clustered” entries  $1, \varepsilon, \dots, \varepsilon$  with random signs (matrix type 3), or
- eigenvalues randomly chosen from the interval  $(\varepsilon, 1)$  (matrix type 4).

An annotated example of an input file for timing the REAL symmetric eigenproblem routines is shown below

```

SEP:  Data file for timing Symmetric Eigenvalue Problem routines
5                                     Number of values of N
50 100 200 300 400                   Values of N (dimension)
5                                     Number of values of parameters
1   16  32  48  64                   Values of NB (blocksize)
401 401 401 401 401                 Values of LDA (leading dimension)
0.0                                   Minimum time in seconds
4                                     Number of matrix types
SST   T T T T T T T T

```

The first line of the input file must contain the characters SEP in columns 1-3. Lines 2-8 are read using list-directed input and specify the following values:

- line 2: The number of values of N
- line 3: The values of N, the matrix dimension
- line 4: The number of values of the parameters NB and LDA
- line 5: The values of NB, the blocksize
- line 6: The values of LDA, the leading dimension
- line 7: The minimum time in seconds that a routine will be timed
- line 8: NIMPES, the number of matrix types to be used

If  $0 < \text{NIMPES} < 4$ , then line 9 specifies NIMPES integer values which are the numbers of the matrix types to be used. The remaining lines specify a path name and the specific computations to be timed. For the symmetric eigenvalue problem the path names for the four data types are SST, DST, CST, and ZST. The (optional) characters after the path name indicate the computations to be timed, as in the input file for the nonsymmetric eigenvalue problem

## 6.5 Timing the Singular Value Decomposition

A separate input file drives the timing codes for the Singular Value Decomposition (SVD). The input file specifies

- pairs of parameter values (M N) specifying the matrix row dimension M and the matrix column dimension N
- pairs of parameter values (NB LDA) specifying the block size NB and the leading dimension LDA

11. **IMQL1** (EISPACK computation of eigenvalues only of a symmetric tridiagonal matrix, to be compared to **SSIEQR(N)**)
12. **IMQL2** (EISPACK computation of eigenvalues and eigenvectors of a symmetric tridiagonal matrix, to be compared to **SSIEQR(V)**)
13. **TQIRAT** (EISPACK computation of eigenvalues only of a symmetric tridiagonal matrix, to be compared to **SSIERF**).
14. **TRIDB** (EISPACK computation of the eigenvalues of  $A$ ) (compare with **SSIEBZ - RANGE=I'**)
15. **HSECT** (EISPACK computation of the eigenvalues of  $A$ ) (compare with **SSIEBZ - RANGE=V**)
16. **TINMT** (EISPACK computation of the eigenvectors of a triangular matrix using inverse iteration) (compare with **SSIEIN**)

For complex matrices the possible computations are

1. **CHEIRD** (LAPACK reduction of a complex Hermitian matrix to real symmetric tridiagonal form)
2. **CSIEQR(N)** (LAPACK computation of eigenvalues only of a symmetric tridiagonal matrix)
3. **CUNGR+(CSIEQR(V)** (LAPACK computation of the eigenvalues and eigenvectors of a symmetric diagonal matrix)
4. **CPIEQR(VECT= N)** (LAPACK computation of the eigenvalues only of a symmetric positive definite tridiagonal matrix)
5. **CUNGR+(CPIEQR(VECT= V)** (LAPACK computation of the eigenvalues and eigenvectors of a symmetric positive definite tridiagonal matrix)
6. **SSIEBZ+(SSIEIN+(CUNMR** (LAPACK computation of the eigenvalues and eigenvectors of a symmetric tridiagonal matrix)
7. **HIRIDI** (EISPACK reduction to symmetric tridiagonal form to be compared to **CHEIRD**)
8. **IMQL1** (EISPACK computation of eigenvalues only of a symmetric tridiagonal matrix, to be compared to **CSIEQR(V)**)
9. **IMQL2-HIRIBK** (EISPACK computation of eigenvalues and eigenvectors of a complex Hermitian matrix given the reduction to real symmetric tridiagonal form to be compared to **CUNGR+(CSIEQR)**).

Four different matrix types are provided for timing the symmetric eigenvalue routines. The matrices used for timing are of the form  $AXDX^{-1}$ , where  $X$  is orthogonal and  $D$  is diagonal with entries

- $N$  the matrix size
- pairs of parameter values ( $NB$  LDA) specifying the block size  $NB$  and the leading dimension LDA
- the test matrix types
- the routines or sequences of routines from LAPACK or EISPACK to be timed.

Goal of this timing code is to determine the values of  $NB$  which maximize the speed of the block algorithms.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main timing program

Parameter	Description	Value
MAXN	Maximum value for $N$ or $NB$	400
LDAMX	Maximum value for LDA	420
MAXN	Maximum number of values of $N$	12
MAXPM	Maximum number of pairs of values ( $NB$ LDA)	10

The computations that may be timed depend on whether the data is real or complex. For the REAL version the possible computations are

1. SSYTRD (LAPACK reduction to symmetric tridiagonal form)
2. SSIEQR(N) (LAPACK computation of eigenvalues only of a symmetric tridiagonal matrix)
3. SSIEQR(V) (LAPACK computation of the eigenvalues and eigenvectors of a symmetric tridiagonal matrix)
4. SSIERF (LAPACK computation of the eigenvalues only of a symmetric tridiagonal matrix using a square-root free algorithm)
5. SPIEQR(COMP= N) (LAPACK computation of the eigenvalues of a symmetric positive definite tridiagonal matrix)
6. SPIEQR(COMP= V) (LAPACK computation of the eigenvalues and eigenvectors of a symmetric positive definite tridiagonal matrix)
7. SSIEBZ(RANGE= I') (LAPACK computation of the eigenvalues in a specified interval for a symmetric tridiagonal matrix)
8. SSIEBZ(RANGE= V) (LAPACK computation of the eigenvalues in a half-open interval for a symmetric tridiagonal matrix)
9. SSIEIN (LAPACK computation of the eigenvectors of a symmetric tridiagonal matrix corresponding to specified eigenvalues using inverse iteration)
10. TRED (EISPACK reduction to symmetric tridiagonal form to be compared to SSYTRD)

```

4                               Number of values of parameters
1  16  32  48                   Values of NB (blocksize)
4   6   8  12                   Values of NS (number of shifts)
40  40  40  40                  Values of MAXB (multishift crossover pt)
301 301 301 301                Values of LDA (leading dimension)
0.0                             Minimum time in seconds
4                               Number of matrix types
1  3  4  6
SHS   T T T T T T T T T T T T

```

The first line of the input file must contain the characters NEP in columns 1-3. Lines 2-10 are read using list-directed input and specify the following values:

- line 2: The number of values of N
- line 3: The values of N, the matrix dimension
- line 4: The number of values of the parameters NB, NS, MAXB, and LDA
- line 5: The values of NB, the blocksize
- line 6: The values of NS, the number of shifts
- line 7: The values of MAXB, the maximum blocksize
- line 8: The values of LDA, the leading dimension
- line 9: The minimum time in seconds that a routine will be timed
- line 10: NYPES, the number of matrix types to be used

If  $0 < NYPES < 8$ , then line 11 specifies NYPES integer values which are the numbers of the matrix types to be used. The remaining lines specify a path name and the specific computations to be timed. For the nonsymmetric eigenvalue problem the path names for the four data types are SHS, DHS, CHS, and ZHS. A line to request all the routines in the REAL path has the form

```
SHS   T T T T T T T T T T T T
```

where the first 3 characters specify the path name, and up to 12 nonblank characters may appear in columns 4-80. If the  $k$ <sup>th</sup> such character is 'T' or 't', the  $k$ <sup>th</sup> routine will be timed. If at least one but fewer than 12 nonblank characters are specified, the remaining routines will not be timed. If columns 4-80 are blank, all the routines will be timed, so the input line

```
SHS
```

is equivalent to the line above.

The output is in the form of a table which shows the absolute times in seconds, floating point operation counts, and megaflop rates for each routine over all relevant input parameters. For the blocked routines, the table has one line for each different value of NB, and for the SHEQR routine, one line for each different combination of NS and MAXB as well.

## 6.4 Timing the Symmetric Eigenproblem

A separate input file drives the timing codes for the symmetric eigenproblem. The input file specifies

4. `SHSEQRV` (LAPACK computation of the Schur form and Schur vectors of a Hessenberg matrix)
5. `SIREM(L)` (LAPACK computation of the left eigenvectors of a matrix in Schur form)
6. `SIREM(R)` (LAPACK computation of the right eigenvectors of a matrix in Schur form)
7. `SHEN(L)` (LAPACK computation of the left eigenvectors of an upper Hessenberg matrix using inverse iteration)
8. `SHEN(R)` (LAPACK computation of the right eigenvectors of an upper Hessenberg matrix using inverse iteration)
9. `CRHES` (EISPACK reduction to upper Hessenberg form to be compared to `SGHRD`)
10. `HQR` (EISPACK computation of eigenvalues only of a Hessenberg matrix, to be compared to `SHSEQRE`)
11. `HQR2` (EISPACK computation of eigenvalues and eigenvectors of a Hessenberg matrix, to be compared to `SHSEQRV` plus `SIREM(R)`)
12. `INMT` (EISPACK computation of the right eigenvectors of an upper Hessenberg matrix using inverse iteration, to be compared to `SHEN(R)`).

Eight different matrix types are provided for timing the nonsymmetric eigenvalue routines. A variety of matrix types is allowed because the number of iterations to compute the eigenvalues, and hence the timing, can depend on the type of matrix whose eigendecomposition is desired. The matrices used for timing are of the form  $XTX^{-1}$  where  $X$  is either orthogonal (for types 1-4) or random with condition number  $1/\sqrt{\epsilon}$  (for types 5-8), where  $\epsilon$  is the machine roundoff error. The matrix  $T$  is upper triangular with random  $O(1)$  entries in the strict upper triangle and has on its diagonal

- evenly spaced entries from 1 down to  $\epsilon$  with random signs (matrix types 1 and 5)
- geometrically spaced entries from 1 down to  $\epsilon$  with random signs (matrix types 2 and 6)
- “clustered” entries  $1, \epsilon, \dots, \epsilon$  with random signs (matrix types 3 and 7), or
- real or complex conjugate paired eigenvalues randomly chosen from the interval  $(\epsilon, 1)$  (matrix types 4 or 8).

An annotated example of an input file for timing the REAL nonsymmetric eigenproblem routines is shown below

```

NEP: Data file for timing Nonsymmetric Eigenvalue Problem routines
4                                     Number of values of N
50 100 200 300                       Values of N (dimension)

```

```

1           Values of INCX
2           Number of values of LDA
512 513     Values of LDA
0.0        Minimum time in seconds
none       Do not time the sample BLAS
SB2
SB3

```

Since the Fortran BLAS do not contain any sub-blocking, the block size NB is not required and its value is replaced by that of INCX the increment between successive elements of a vector in the Level 2 BLAS. Note that we have specified “none” on line 13 to suppress timing of the sample BLAS, which are redundant in this case.

### 6.3 Timing the Nonsymmetric Eigenproblem

A separate input file drives the timing codes for the nonsymmetric eigenproblem. The input file specifies

- N the matrix size
- four-tuples of parameter values (NB, NS, MNB, LDA) specifying the block size NB, the number of shifts NS, the matrix size MNB less than which an unblocked routine is used, and the leading dimension LDA
- the test matrix types
- the routines or sequences of routines from LAPACK or EISPACK to be timed

The parameters NS and MNB apply only to the QR iteration routine xHEQR, and NB is used only by the block algorithms. A goal of this timing code is to determine the values of NB, NS and MNB which maximize the speed of the codes.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main timing program

Parameter	Description	Value
MXN	Maximum value for N, NB, NS, or MNB	400
IDAMX	Maximum value for LDA	420
MXNN	Maximum number of values of N	12
MXPRM	Maximum number of parameter sets (NB, NS, MNB, LDA)	10

The computations that may be timed for the REAL version are

1. SGEHD (LAPACK reduction to upper Hessenberg form)
2. SHSEQRE (LAPACK computation of eigenvalues only of a Hessenberg matrix)
3. SHSEQRS (LAPACK computation of the Schur form of a Hessenberg matrix)

$N$  are specified in ordered pairs  $(M, N)$ . An annotated example of an input file for timing the REAL linear equation routines that operate on dense rectangular matrices is shown below. The input file is read in the same way as the one for dense square matrices.

```
LAPACK timing, REAL rectangular matrices
7                               Number of values of M
100 200 100 200 400 200 400    Values of M (row dimension)
7                               Number of values of N
100 100 200 200 200 400 400    Values of N (column dimension)
2                               Number of values of K
100 400                         Values of K
5                               Number of values of NB
1 16 32 48 64                 Values of NB (blocksize)
0 48 128 128 128              Values of NX (crossover point)
2                               Number of values of LDA
400 401                       Values of LDA (leading dimension)
0.0                            Minimum time in seconds
none
SQR      T T T
SLQ      T T T
SQL      T T T
SRQ      T T T
SQP      T
SBR      T T F
```

## 6.2 Timing the Level 2 and 3 BLAS

Timing of the Level 2 and 3 BLAS routines may be requested from one of the linear equation input files, or by using a special BLAS format provided for compatibility with previous releases of LAPACK. The BLAS input format is the same as the linear equation input format, except that values of NX are not read in. The BLAS input format is requested by specifying 'BLAS' on the first line of the file.

Three input files are provided for timing the BLAS with the matrix shapes encountered in the LAPACK routines. In each of these files, one of the parameters  $M$ ,  $N$ , and  $K$  for the Level 3 BLAS is on the order of the blocksize while the other two are on the order of the matrix size. The first of these input files also times the Level 2 BLAS, and we include the single precision real version of this data file here for reference:

```
BLAS timing, REAL data, K small
5                               Number of values of M
100 200 300 400 500           Values of M
5                               Number of values of N
100 200 300 400 500           Values of N
5                               Number of values of K
2 16 32 48 64                 Values of K
1                               Number of values of INCX
```

```

SRQ    T T F
SQP    T
SHR    T T F F
STD    T T F F
SBR    T F F
SLU    T T T T T T T
SCH    T T T T T T T

```

The first 13 lines of the input file are read using list-directed input and are used to specify the values of  $M$ ,  $N$ ,  $K$ ,  $NB$ ,  $NX$ ,  $LDA$ , and  $TIMIN$  (the minimum time). By default, `xCEM` and `xCEM` are called to sample the BLAS performance on square matrices of order  $N$  but this option can be controlled by entering one of the following on line 14:

```

BAND   Time xCEM (instead of xCEM) using matrices of order M and
       bandwidth K and time xCEM using matrices of order K

```

```

NONE   Do not do the sample timing of xCEM and xCEM

```

The timing paths or routine names which follow may be specified in any order.

When timing the band routines it is more interesting to use one large value of the matrix size and vary the bandwidth. An annotated example of an input file for timing the REAL linear equation routines that operate on banded matrices is shown below

LAPACK timing, REAL band matrices

```

1           Number of values of M
1000        Values of M (row dimension)
0           Number of values of N
200         Values of N (column dimension)
5           Number of values of K
25 50 100 150 200  Values of K (bandwidth)
5           Number of values of NB
1 16 32 48 64    Values of NB (blocksize)
0 48 128 128 128 Values of NX (crossover point)
2           Number of values of LDA
601 602      Values of LDA (leading dimension)
0.0         Minimum time in seconds
BAND        Time sample banded BLAS
SGB
SPB
STB

```

Here  $M$  specifies the matrix size and  $K$  specifies the bandwidth for the test paths `SCB`, `SIB`, and `SIB`. Note that we request timing of the sample BLAS for banded matrices by specifying "BAND" on line 14.

We also provide a separate input file for timing the orthogonal factorization and reduction routines that operate on rectangular matrices. For these routines, the values of  $M$  and



block variants of the LU factorization (left-looking, Gout, and right-looking) for  $N \times N$  matrices, as well as the corresponding unblocked variants on matrices of size  $N \times NB$  and the Linpack routine xGEEA. The xCH timing path requests timing of three block variants of the Cholesky factorization and the corresponding Linpack routine xPCEA. The LU and Cholesky variants are not strictly part of LAPACK and will not be included in the public release.

The timing programs have their own matrix generator that supplies random Toeplitz matrices (constant along a diagonal) for many of the timing paths. Toeplitz matrices are used because they can be generated more quickly than dense matrices, and the call to the matrix generator is inside the timing loop. The LAPACKtest matrix generator is used to generate matrices of known condition for the xQR, xRQ, xLQ, xQL, xQP, xHR, xID, and xBR paths.

The user specifies a minimum time for which each routine should run and the computation is repeated if necessary until this time is used. In order to prevent inflated performance due to a matrix remaining in the cache from one iteration to the next, the paths that use random Toeplitz matrices regenerate the matrix before each call to the LAPACK routine in the timing loop. The time for generating the matrix at each iteration is subtracted from the total time.

An annotated example of an input file for timing the REAL linear equation routines that operate on dense square matrices is shown below. The first line of input is printed as the first line of output and can be used to identify different sets of results.

```
LAPACK timing, REAL square matrices
5                               Number of values of M
100 200 300 400 500           Values of M (row dimension)
5                               Number of values of N
100 200 300 400 500           Values of N (column dimension)
2                               Number of values of K
100 400                        Values of K
5                               Number of values of NB
1 16 32 48 64                 Values of NB (blocksize)
0 48 128 128 128              Values of NX (crossover point)
2                               Number of values of LDA
512 513                        Values of LDA (leading dimension)
0.0                            Minimum time in seconds
SGE      T T T
SPO      T T T
SPP      T T T
SSY      T T T
SSP      T T T
STR      T T
STP      T T
SQR      T T F
SLQ      T T F
SQL      T T F
```

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main timing program

Parameter	Description	Value
NMAX	Maximum value of M, N, K and NB for dense matrices	512
IDAMX	Maximum value of IDA	532
NMNB	Maximum value of M for banded matrices	5000
MAXN	Maximum number of values of M, N, K or NB	12
MNLDA	Maximum number of values of IDA	4

The parameter IDAMX should be at least NMAX. For the xCB path, we must have  $(IDA + K)M \leq 3(IDAMX - 1)(NMAX - 1)$ , where  $IDA \geq 3K + 1$ , which restricts the value of K. These limits allow K to be as big as 200 for M=1000. For the xPB and xTB paths, the condition is  $(2K + 1)M \leq 3(NMAX - 1)(IDAMX - 1)$ .

The input file also specifies a set of LAPACK routine names or LAPACK path names to be timed. The path names are similar to those used for the test program and include the following standard paths:

{S, C, D, Z}	GE	General matrices (LU factorization)
{S, C, D, Z}	GB	General banded matrices
{S, C, D, Z}	PO	Positive definite matrices (Cholesky factorization)
{S, C, D, Z}	PP	Positive definite packed
{S, C, D, Z}	PB	Positive definite banded
{S, C, D, Z}	SY	Symmetric indefinite matrices (Bunch-Kaufman factorization)
{S, C, D, Z}	SP	Symmetric indefinite packed
{C, Z}	HE	Hermitian indefinite matrices (Bunch-Kaufman factorization)
{C, Z}	HP	Hermitian indefinite packed
{S, C, D, Z}	TR	Triangular matrices
{S, C, D, Z}	TP	Triangular packed matrices
{S, C, D, Z}	TB	Triangular band
{S, C, D, Z}	QR	QR decomposition
{S, C, D, Z}	RQ	RQ decomposition
{S, C, D, Z}	LQ	LQ decomposition
{S, C, D, Z}	QL	QL decomposition
{S, C, D, Z}	QP	QR decomposition with column pivoting
{S, C, D, Z}	HR	Reduction to Hessenberg form
{S, C, D, Z}	TD	Reduction to real tridiagonal form
{S, C, D, Z}	BR	Reduction to bidiagonal form
{S, C, D, Z}	LU	Variants of the LU factorization
{S, C, D, Z}	CH	Variants of the Cholesky factorization

For timing the Level 2 and 3 BLAS, two extra paths are provided:

{S, C, D, Z}	B2	Level 2 BLAS
{S, C, D, Z}	B3	Level 3 BLAS

The paths xLU, xCH, xHR, and xTD include timing of the equivalent ITPACK or EISPACK factorizations and reductions for comparison. The xLU path requests timing of three

timing program also times the Level 2 and 3 BLAS, variants of the LU and Cholesky factorizations, and the reductions to bidiagonal, tridiagonal, or Hessenberg form for eigenvalue computations. Results from the linear equation timing program are given in megaflops, and the operation counts are computed from a formula (see Appendix C). Results from the eigensystem timing program are given in execution times, operation counts, and megaflops, where the operation counts are calculated during execution using special versions of the LAPACK routines which have been instrumented to count operations. Each program has its own style of input, and the eigensystem timing program accepts four different sets of parameters, for the nonsymmetric eigenvalue problem, the symmetric eigenvalue problem, the singular value decomposition, and the generalized nonsymmetric eigenvalue problem. The following sections describe the different input formats and timing parameters.

Both timing programs, but the linear equation timing program in particular, are intended to be used to collect data to determine optimal values for the block routines. All of the block factorization, inversion, reduction, and orthogonal transformation routines in LAPACK are included in the linear equation timing program. Currently, the block parameters NB and NX, as well as others, are passed to the block routines by the environment inquiry function ILAENV, which in turn receives these values through a common block set in the timing program. Future implementations of ILAENV may be tuned to a specific machine so that users of LAPACK will not have to set the block size. For a brief introduction to ILAENV and guidelines on setting some of the parameters, see the Preliminary LAPACK User's Guide [1].

The main timing procedure for the REAL linear equation routines is found in LAPACK/TIMING/LIN/stimaa.f in the Unix version and is the first program unit in SLIN.TIME in the non-Unix version. The main timing procedure for the REAL eigenvalue routines is found in LAPACK/TIMING/EIG/stimee.f in the Unix version and is the first program unit in SEIGTIME in the non-Unix version.

## 6.1 The Linear Equation Timing Program

The timing program for the linear equation routines is driven by a data file from which the following parameters may be varied:

- M the matrix row dimension
- N the matrix column dimension
- K the bandwidth for the band routines, or the third dimension for the Level 3 BLAS
- NB the block size for the blocked routines
- NX the crossover point, the point in a block algorithm at which we switch to an unblocked algorithm
- LDA the leading dimension of the dense and banded matrices.

For banded matrices, the values of M are used for the matrix row and column dimensions, and for symmetric or Hermitian matrices that are not banded, the values of N are used for the matrix dimension.

```

SGG:  Data file for testing Nonsymmetric Eigenvalue Problem routines
7          Number of values of N
0 1 2 3 5 10 16  Values of N (dimension)
4          Number of parameter values
1  1  2  2      Values of NB (blocksize)
100 100 2  2    Values of NBMIN (minimum blocksize)
2  4  2  4      Values of NS (no. of shifts)
100 100 2  2    Values of MAXB (multishift crossover pt)
100 100 2  2    Values of NBCOL (minimum col. dimension)
20.0      Threshold value
T          Put T to test the LAPACK routines
T          Put T to test the driver routines
T          Put T to test the error exits
1          Code to interpret the seed
SGG  26

```

The first line of the input file must contain the characters *SGG* in columns 1-3. Lines 2-14 are read using list-directed input and specify the following values:

line 2: The number of values of *N*  
line 3: The values of *N*, the matrix dimension  
line 4: Number of values of the parameters *NB*, *NBMIN*, *NS*, *MAXB*, *NBCOL*  
line 5: The values for the blocksize *NB*  
line 6: The values for the minimum blocksize, *NBMIN*  
line 7: The values for the number of shifts *NS*  
line 8: The values of *MAXB*, the multishift crossover point  
line 9: The values of *NBCOL*, the minimum column dimension for blocks  
line 10: The threshold value for the test ratios  
line 11: *TSCHK* flag to test LAPACK routines  
line 12: *TSIDR* flag to test driver routines  
line 13: *TSERR* flag to test error exits from LAPACK and driver routines  
line 14: An integer code to interpret the random number seed  
=0: Set the seed to a default value before each run  
=1: Initialize the seed to a default value only before the first run  
=2: Like 1, but use the seed values on the next line  
line 15: If line 14 was 2, four integer values for the random number seed

The remaining lines are used to specify the matrix types for one or more sets of tests, as in the nonsymmetric case. The valid 3-character codes are *SGG* (*CGG* in complex, *DGG* in double precision, and *ZGG* in complex\*16).

## 6 More About Timing

There are two distinct timing programs for LAPACK routines in each data type, one for the linear equations routines and one for the eigensystem routines. The linear equation

All norms are  $\|\cdot\|_1$ . The scalings in the test ratios assure that the ratios will be  $O(1)$ , independent of  $\|A\|$  and  $\varepsilon$ , and nearly independent of  $n$ .

When the test programs run, these test ratios will be compared with a user-specified threshold `THRESH`, and for each test ratio that exceeds `THRESH`, a message is printed specifying the test matrix, the ratio that failed, and its value. An example message is

```
Matrix order= 25, type=18, seed=2548,1429,1713,1411, result 8 is 11.33
```

In this example, the test matrix was of order  $n=25$  and of type 18 from Table 11, “seed” is the initial 4-integer seed of the random number generator used to generate  $A$  and  $B$ , and “result” specifies that test ratio  $r_8$  failed to pass the threshold, and its value was 11.33.

The normalization of the eigenvectors will also be checked. If the absolute value of the largest entry in an eigenvector is not within  $\varepsilon \times \text{THRESH}$  of 1, then a message is printed specifying the error. An example message is

```
SCHK51: Right Eigenvectors from STGEVC(JOB=B) incorrectly normalized.
Error/precision=0.103E+05, n= 25, type= 18, seed=2548,1429,1713,1411.
```

## 5.6.5 Tests Performed on the Generalized Nonsymmetric Eigenvalue

The two driver routines have slightly different tests applied to them. For `SGES` the following tests are computed:

$$r_1 = \frac{\|A - QSZ^T\|}{\|A\| n\varepsilon} \quad r_2 = \frac{\|B - QTZ^T\|}{\|B\| n\varepsilon}$$

$$r_3 = \frac{\|I - QQ^T\|}{n\varepsilon} \quad r_4 = \frac{\|I - ZZ^T\|}{n\varepsilon}$$

$$r_5 = \max_j D(j) = \begin{cases} \frac{|\alpha(j) - S(j,j)|}{\max(|\alpha(j)|, |S(j,j)|)} + \frac{|\beta(j) - T(j,j)|}{\max(|\beta(j)|, |T(j,j)|)} & \text{if } \alpha(j) \text{ is real} \\ \frac{|\det(sS - wT)|}{\varepsilon \max(|s|, |w|, \|T\|) \|sS - wT\|} & \text{if } \alpha(j) \text{ is complex,} \end{cases}$$

where  $S$  and  $T$  are the  $2 \times 2$  diagonal blocks of  $S$  and  $T$  corresponding to the  $j$ <sup>th</sup> eigenvalue.

For `SGEV` the following tests are computed:

$$r_6 = \max_{\text{left eigenvalue/-vector pairs } (\beta/\alpha, l)} \frac{|(\beta A - \alpha B)^T l|}{\varepsilon \max(|\beta A|, |\alpha B|)}$$

$$r_7 = \max_{\text{right eigenvalue/-vector pairs } (\beta/\alpha, r)} \frac{|(\beta A - \alpha B) r|}{\varepsilon \max(|\beta A|, |\alpha B|)}$$

## 5.6.6 Input File for Testing the Generalized Nonsymmetric Eigenvalue Routines and Drivers

An annotated example of an input file for testing the generalized nonsymmetric eigenvalue routines is shown below

### 5.6.3 Test Matrices for the Generalized Nonsymmetric Eigenvalue

The same twenty-six different types of test matrix pairs may be generated for the generalized nonsymmetric eigenvalue drivers. Tables 10 and 11 show the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries.

### 5.6.4 Tests Performed on the Generalized Nonsymmetric Eigenvalue

Finding the eigenvalues and eigenvectors of a pair of nonsymmetric matrices  $A$ ,  $B$  is done in the following stages:

1.  $A$  is decomposed as  $UHV^*$  and  $B$  as  $UTV^*$ , where  $U$  and  $V$  are unitary,  $H$  is upper Hessenberg,  $T$  is upper triangular, and  $U^*$  is the conjugate transpose of  $U$ .
2.  $H$  is decomposed as  $QSZ^*$  and  $T$  as  $QPZ^*$ , where  $Q$  and  $Z$  are unitary,  $P$  is upper triangular with non-negative real diagonal entries and  $S$  is in Schur form; this also gives the generalized eigenvalues  $\lambda_i$ , which are expressed as pairs  $(\alpha_i, \beta_i)$ , where  $\lambda_i = \alpha_i / \beta_i$ .
3. The left and right generalized eigenvectors  $l_i$  and  $r_i$  for the pair  $S, P$  are computed, and from them the back-transformed eigenvectors  $\hat{l}_i$  and  $\hat{r}_i$  for the matrix pair  $H, T$ . The eigenvectors are normalized so that their largest element has absolute value 1.<sup>1</sup> (Note that eigenvectors corresponding to singular eigenvalues, i.e., eigenvalues for which  $\alpha = \beta = 0$ , are not well defined, these are not tested in the eigenvector tests described below)

To check these calculations, the following test ratios are computed:

$$\begin{aligned}
 r_1 &= \frac{\|A - UHV^*\|}{n\varepsilon \|A\|} & r_2 &= \frac{\|B - UTV^*\|}{n\varepsilon \|B\|} \\
 r_3 &= \frac{\|I - UU^*\|}{n\varepsilon} & r_4 &= \frac{\|I - VV^*\|}{n\varepsilon} \\
 r_5 &= \frac{\|H - QSZ^*\|}{n\varepsilon \|H\|} & r_6 &= \frac{\|T - QPZ^*\|}{n\varepsilon \|T\|} \\
 r_7 &= \frac{\|I - QQ^*\|}{n\varepsilon} & r_8 &= \frac{\|I - ZZ^*\|}{n\varepsilon} \\
 r_9 &= \max_i \frac{\|(\beta_i S - \alpha_i P)^T l_i\|}{\varepsilon \max(\|\beta_i S\|, \|\alpha_i P\|)} & r_{10} &= \max_i \frac{\|(\beta_i H - \alpha_i T)^T \hat{l}_i\|}{\varepsilon \max(\|\beta_i H\|, \|\alpha_i T\|)} \\
 r_{11} &= \max_i \frac{\|(\beta_i S - \alpha_i P) r_i\|}{\varepsilon \max(\|\beta_i S\|, \|\alpha_i P\|)} & r_{12} &= \max_i \frac{\|(\beta_i H - \alpha_i T) \hat{r}_i\|}{\varepsilon \max(\|\beta_i H\|, \|\alpha_i T\|)}
 \end{aligned}$$

<sup>1</sup>For the purpose of normalization, the ‘‘absolute value’’ of a complex number  $z = x + iy$  is computed as  $|x| + |y|$ .

Distribution of Eigenvalues	Magnitude of $A, B$				
	$\ A\  \approx 1, \ B\  \approx 1$	$\ A\  \approx \frac{1}{\omega}, \ B\  \approx \omega$	$\ A\  \approx \omega, \ B\  \approx \omega$	$\ A\  \approx \frac{1}{\omega}, \ B\  \approx \frac{1}{\omega}$	$\ A\  \approx \omega, \ B\  \approx \frac{1}{\omega}$
All Ones	16				
(Same as type 15)	17				
Arithmetic	19	22	24	25	23
Geometric	20				
Clustered	18				
Random	21				
Random Entries	26				

Table 11: Dense test matrices for the generalized nonsymmetric eigenvalue problem

$K$ : A  $(k+1) \times (k+1)$  transposed Jordan block which is a diagonal block within a  $(2k+1) \times (2k+1)$  matrix. Thus,  $\begin{pmatrix} K & 0 \\ 0 & I \end{pmatrix}$  has all zero entries except for the last  $k$  diagonal entries and the first  $k$  entries on the first subdiagonal. (Note that the matrices  $\begin{pmatrix} K & 0 \\ 0 & I \end{pmatrix}$  and  $\begin{pmatrix} I & 0 \\ 0 & K \end{pmatrix}$  have odd order; if an even order matrix is needed, a zero row and column are added at the end.)

$D_1$ : A diagonal matrix with the entries  $0, 1, 2, \dots, n-1$  on the diagonal, where  $n$  is the order of the matrix.

$D_2$ : A diagonal matrix with the entries  $0, 0, 1, 2, \dots, n-3, 0$  on the diagonal, where  $n$  is the order of the matrix.

$D_3$ : A diagonal matrix with the entries  $0, n-3, n-4, \dots, 1, 0, 0$  on the diagonal, where  $n$  is the order of the matrix.

Except for matrices with random entries, all the matrix pairs include at least one infinite, one zero, and one singular eigenvalue. For arithmetic, geometric, and clustered eigenvalue distributions, the eigenvalues lie between  $\varepsilon$  (the machine precision) and 1 in absolute value. The eigenvalue distributions have the following meanings:

Arithmetic: Difference between adjacent eigenvalues is a constant.

Geometric: Ratio of adjacent eigenvalues is a constant.

Clustered: One eigenvalue is 1 and the rest are  $\varepsilon$  in absolute value.

Random: Eigenvalues are logarithmically distributed.

Random entries: Matrix entries are uniformly distributed random numbers.

## 5.6.1 The Generalized Nonsymmetric Eigenvalue Drivers

The driver routines for the generalized nonsymmetric eigenvalue problem are

**xGEGS** factors  $A$  and  $B$  into generalized Schur form and computes the generalized eigenvalues

**xGEGV** computes the generalized eigenvalues and the left and right generalized eigenvectors

## 5.6.2 Test Matrices for the Generalized Nonsymmetric Eigenvalue

Twenty-six different types of test matrix pairs may be generated for the generalized nonsymmetric eigenvalue routines. Tables 10 and 11 show the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries.

Matrix A:	Matrix B:									
	0	I			$J^t$	$\begin{pmatrix} I & 0 \\ 0 & K \end{pmatrix}$	$D_1$			$D_3$
		$\times 1$	$\times \omega$	$\times \frac{1}{\omega}$			$\times 1$	$\times \omega$	$\times \frac{1}{\omega}$	
0	1	3								
I	2	4					8			
$I \times \omega$									12	
$I \times \frac{1}{\omega}$								11		
$J^t$					5					
$\begin{pmatrix} K & 0 \\ 0 & I \end{pmatrix}$						6				
$D_1$		7								
$D_1 \times \omega$			14	10						
$D_1 \times \frac{1}{\omega}$			9	13						
$D_2$										15

Table 10: Sparse test matrices for the generalized nonsymmetric eigenvalue problem

The following symbols and abbreviations are used:

**0**: The zero matrix.

**I**: The identity matrix.

**$\omega$** : Generally, the underflow threshold times the order of the matrix divided by the machine precision. In other words, this is a very small number, useful for testing the sensitivity to underflow and division by small numbers. Its reciprocal tests for overflow problems.

**$J^t$** : Transposed Jordan block, i.e., matrix with ones on the first subdiagonal and zeros elsewhere. (Note that the diagonal is zero.)



line 2: The number of values of  $M$  and  $N$   
 line 3: The values of  $M$  the matrix row dimension  
 line 4: The values of  $N$  the matrix column dimension  
 line 5: The number of values of the parameters  $NB$ ,  $NBMN$ ,  $NX$ ,  $NRES$   
 line 6: The values of  $NB$  the blocksize  
 line 7: The values of  $NBMN$  the minimum blocksize  
 line 8: The values of  $NX$  the crossover point  
 line 9: The values of  $NRES$ , the number of right hand sides  
 line 10: The threshold value for the test ratios  
 line 11:  $TSCHK$  the flag to test LAPACK routines  
 line 12:  $TSIDR$  the flag to test driver routines  
 line 13:  $TSERR$  the flag to test error exits from the LAPACK and driver routines  
 line 14: An integer code to interpret the random number seed.  
           =0: Set the seed to a default value before each run  
           =1: Initialize the seed to a default value only before the first run  
           =2: Like 1, but use the seed values on the next line  
 line 15: If line 14 was 2, four integer values for the random number seed

The remaining lines are used to specify the matrix types for one or more sets of tests, as in the nonsymmetric case. The valid 3-character codes are SVD or SBD (CBD in complex, DBD in double precision, and ZBD in complex\*16).

## 5.6 Testing the Generalized Nonsymmetric Eigenvalue Routine

The test routine for the LAPACK generalized nonsymmetric eigenvalue routines has the following parameters which may be varied:

- the order  $N$  of the pair of test matrices  $A$ ,  $B$
- the type of the pair of test matrices  $A$ ,  $B$
- five numerical parameters:
  - the blocksize  $NB$  ;
  - the minimum blocksize  $NBMN$  ;
  - the number of shifts  $NS$  for the multishift QZ method;
  - the minimum blocksize  $MAXB$  ;
  - $MINBK$  the minimum number of rows/columns to be updated by a block of Householder transformations in order for blocking to be used.

The test program thus consists of a triply-nested loop, the outer one over quintuples  $(NB, NBMN, NS, MAXB, MINBK)$ , the next over  $N$  and the inner one over matrix types. On each iteration of the innermost loop, a pair of matrices  $A, B$  is generated and used to test the eigenvalue routines.

$$r_4 = \begin{cases} 0 & \text{if } S \text{ contains } MN \text{ nonnegative values in decreasing order.} \\ \frac{1}{\epsilon} & \text{otherwise} \end{cases}$$

$$r_5 = \frac{\|U - U_p\|}{M\epsilon}, \text{ where } U_p \text{ is a partially computed } U.$$

$$r_6 = \frac{\|VT - VT_p\|}{N\epsilon}, \text{ where } VT_p \text{ is a partially computed } VT.$$

$$r_7 = \frac{\|S - S_p\|}{MNMN\epsilon\|S\|}, \text{ where } S_p \text{ is the vector of singular values from the partial SVD}$$

### 5.5.6 Input File for Testing the Singular Value Decomposition Routines

An annotated example of an input file for testing the singular value decomposition routines and driver routine is shown below

```
SVD: Data file for testing Singular Value Decomposition routines
20                               Number of values of M
0 0 0 0 1 1 1 1 2 2 2 2 3 3 3 3 10 10 16 16  Values of M
0 1 2 3 0 1 2 3 0 1 2 3 0 1 2 3 10 16 10 16  Values of N
5                               Number of parameter values
1 3 3 3 20                      Values of NB (blocksize)
2 2 2 2 2                      Values of NBMIN (minimum blocksize)
1 0 5 9 1                      Values of NX (crossover point)
2 0 2 2 2                      Values of NRHS
20.0                          Threshold value
T                               Put T to test the LAPACK routines
T                               Put T to test the driver routines
T                               Put T to test the error exits
1                               Code to interpret the seed
SVD 16
```

The first line of the input file must contain the characters SVD in columns 1-3. Lines 2-14 are read using list-directed input and specify the following values:

To check these calculations, the following test ratios are computed:

$$\begin{aligned}
r_1 &= \frac{\|A - QBP^*\|}{\tilde{n}\varepsilon \|A\|} & r_2 &= \frac{\|I - Q^*Q\|}{m\varepsilon} \\
r_3 &= \frac{\|I - P^*P\|}{n\varepsilon} & r_4 &= \frac{\|B - U\Sigma V^*\|}{\tilde{n}\varepsilon \|B\|} \\
r_5 &= \frac{\|Y - UZ\|}{\max(\tilde{n}, k)\varepsilon \|Y\|}, & \text{where } Y &= Q^*X \text{ and } Z = U^*Y. \\
r_6 &= \frac{\|I - U^*U\|}{\tilde{n}\varepsilon} & r_7 &= \frac{\|I - VV^*\|}{\tilde{n}\varepsilon} \\
r_8 &= \begin{cases} 0 & \text{if } S1 \text{ contains } \tilde{n} \text{ nonnegative values in decreasing order.} \\ \frac{1}{\varepsilon} & \text{otherwise} \end{cases} \\
r_9 &= \begin{cases} 0 & \text{if eigenvalues of } B \text{ are within } THRESH \text{ of those in } S1. \\ 2 * THRESH & \text{otherwise} \end{cases} \\
r_{10} &= \frac{\|S1 - S2\|}{\varepsilon \|S1\|} & r_{11} &= \frac{\|A - (QU)\Sigma(PV)^*\|}{\tilde{n}\varepsilon \|A\|} \\
r_{12} &= \frac{\|X - (QU)Z\|}{\max(m, k)\varepsilon \|X\|} & r_{13} &= \frac{\|I - (QU)^*(QU)\|}{m\varepsilon} \\
r_{14} &= \frac{\|I - (VP)(VP)^*\|}{n\varepsilon}
\end{aligned}$$

where the subscript 1 indicates that  $U$  and  $V$  were computed at the same time as  $\Sigma$ , and 0 that they were not. (All norms are  $\|\cdot\|_1$ .) The scalings in the test ratios assure that the ratios will be  $O(1)$  (typically less than 10 or 100), independent of  $\|A\|$  and  $\varepsilon$ , and nearly independent of  $m$  or  $n$ .

### 5.5.5 Tests Performed on the Singular Value Decomposition Driver

For the driver routine, the following tests are computed:

$$\begin{aligned}
r_1 &= \frac{\|A - U \text{diag}(S)VT\|}{\|A\| \max(M, N)\varepsilon} \\
r_2 &= \frac{\|I - U^T U\|}{M\varepsilon} \\
r_3 &= \frac{\|I - VT^T V\|}{N\varepsilon}
\end{aligned}$$

### 5.5.3 Test Matrices for the Singular Value Decomposition Driver

Five different types of test matrices may be generated for the singular value decomposition driver. Table 9 shows the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries.

Type	Eigenvalue Distribution				
	Arithmetic	Geometric	Clustered	Random	Other
Zero					1
Identity					2
$UDV$	3, 4 <sup>†</sup> , 5 <sup>‡</sup>				

<sup>†</sup>— matrix entries are multiplied by the underflow threshold  $\epsilon$   
<sup>‡</sup>— matrix entries are multiplied by the overflow threshold  $\epsilon^*$

Table 9: Test matrices for the singular value decomposition driver

### 5.5.4 Tests Performed on the Singular Value Decomposition Routine

Finding the singular values and singular vectors of a dense,  $m \times n$  matrix  $A$  is done in the following stages:

1.  $A$  is decomposed as  $QBP^*$ , where  $Q$  and  $P$  are unitary and  $B$  is real bidiagonal.
2.  $B$  is decomposed as  $U\Sigma V$ , where  $U$  and  $V$  are real orthogonal and  $\Sigma$  is a positive real diagonal matrix of singular values. This is done three times to compute
  - (a)  $B = U\Sigma_1 V^*$ , where  $\Sigma_1$  is the diagonal matrix of singular values and the columns of the matrices  $U$  and  $V$  are the left and right singular vectors, respectively, of  $B$ .
  - (b) Same as above, but the singular values are stored in  $\Sigma_2$  and the singular vectors are not computed.
  - (c)  $A = (UQ)S(VP)^*$ , the SVD of the original matrix  $A$ .

For each pair of matrix dimensions  $(m, n)$  and each selected matrix type, an  $m$  by  $n$  matrix  $A$  and an  $m$  by  $\text{NRFB}$  matrix  $X$  are generated. The problem dimensions are as follows

$A$	$m \times n$
$Q$	$m \times \tilde{n}$ (but $m \times m$ if $\text{NRFB} > 0$ )
$P$	$\tilde{n} \times n$
$B$	$\tilde{n} \times \tilde{n}$
$U, V$	$\tilde{n} \times \tilde{n}$
$S1, S2$	diagonal, order $\sim n$
$X$	$m \times \text{NRFB}$

where  $\tilde{n} = \min(m, n)$ .

The test program thus consists of a triply-nested loop, the outer one over NB pairs (MN), and the inner one over matrix types. On each iteration of the innermost loop, a matrix A is generated and used to test the SVD routines.

### 5.5.1 The Singular Value Decomposition Driver

The driver routine for the singular value decomposition is

**xGESVD** singular value decomposition of A

### 5.5.2 Test Matrices for the Singular Value Decomposition Routine

Sixteen different types of test matrices may be generated for the singular value decomposition routines. Table 8 shows the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrix types other than the random bidiagonal matrices have  $O(1)$  entries.

Type	Singular Value Distribution			
	Arithmetic	Geometric	Clustered	Other
Zero				1
Identity				2
Diagonal	3, 6 <sup>†</sup> , 7 <sup>‡</sup>	4	5	
UDV	8, 11 <sup>†</sup> , 12 <sup>‡</sup>	9	10	
Random entries				13, 14 <sup>†</sup> , 15 <sup>‡</sup>
Random bidiagonal				16

<sup>†</sup>—matrix entries are  $O(\sqrt{\text{overflow}})$   
<sup>‡</sup>—matrix entries are  $O(\sqrt{\text{underflow}})$

Table 8: Test matrices for the singular value decomposition

Matrix types identified as “Zero”, “Diagonal”, and “Random entries” should be self-explanatory. The other matrix types have the following meanings:

**Identity:**  $\min(M, N) \times \min(M, N)$  identity matrix with zero rows or columns added to the bottom or right to make it  $M \times N$

**UDV:** Real  $M \times N$  diagonal matrix  $D$  with  $O(1)$  entries multiplied by unitary (or real orthogonal) matrices on the left and right

**Random bidiagonal:** Upper bidiagonal matrix whose entries are randomly chosen from a logarithmic distribution on  $[\varepsilon^{-2}, \varepsilon^{-2}]$

The QR algorithm used in **xDQR** should compute all singular values, even small ones, to good relative accuracy, even of matrices with entries varying over many orders of magnitude, and the random bidiagonal matrix is intended to test this. Thus, unlike the other matrix types, the random bidiagonal matrix is neither  $O(1)$ , nor an  $O(1)$  matrix scaled to some other magnitude.

The singular value distributions are analogous to the eigenvalue distributions in the nonsymmetric eigenvalue problem (see Section 6.2.1).

0 1 2 3 5 10 16	Values of N (dimension)
5	Number of values of NB, NBMIN, and NX
1 3 3 3 20	Values of NB (blocksize)
2 2 2 2 2	Values of NBMIN (minimum blocksize)
1 0 5 9 1	Values of NX (crossover point)
20.0	Threshold value
T	Put T to test the LAPACK routines
T	Put T to test the driver routines
T	Put T to test the error exits
1	Code to interpret the seed
SEP 15	

The first line of the input file must contain the characters SEP in columns 1-3. Lines 2-12 are read using list-directed input and specify the following values:

line 2:	The number of values of N
line 3:	The values of N, the matrix dimension
line 4:	The number of values of the parameters NB, NBMIN, NX
line 5:	The values of NB, the blocksize
line 6:	The values of NBMIN, the minimum blocksize
line 7:	The values of NX, the crossover point
line 8:	The threshold value for the test ratios
line 9:	TESTCK flag to test LAPACK routines
line 10:	TESTDZ flag to test driver routines
line 11:	TESTERR flag to test error exits from LAPACK and driver routines
line 12:	An integer code to interpret the random number seed
	=0: Set the seed to a default value before each run
	=1: Initialize the seed to a default value only before the first run
	=2: Like 1, but use the seed values on the next line
line 13:	If line 12 was 2, four integer values for the random number seed

The remaining lines are used to specify the matrix types for one or more sets of tests, as in the nonsymmetric case. The valid 3-character codes are SEP or SST (CST in complex, DST in double precision, and ZST in complex\*16).

## 5.5 Testing the Singular Value Decomposition Routines

The test routine for the LAPACK singular value decomposition (SVD) routines has the following parameters which may be varied:

- the number of rows M and columns N of the test matrix A
- the type of the test matrix A
- the blocksize NB

When  $S$  is also diagonally dominant by a factor  $\gamma < 1$ ,

$$r_{13} = \max_i \frac{\|D4(i) - WR(i)\|}{\|D4(i)\| \omega},$$

$$\text{where } \omega = 2(2n - 1)\varepsilon \frac{1 + 8 * \gamma^2}{(1 - \gamma)^4}$$

$$r_{14} = \frac{\|WA1 - D3\|}{\varepsilon \|D3\|}$$

$$r_{15} = \frac{\max_i (\min_j (\|WA2(i) - WA3(j)\|)) + \max_j (\min_i (\|WA3(i) - WA2(j)\|))}{\varepsilon \|D3\|}$$

$$r_{16} = \frac{\|S - YWA1Y^*\|}{n\varepsilon \|S\|}$$

$$r_{17} = \frac{\|I - YY^*\|}{n\varepsilon}$$

where the subscript 1 indicates that the eigenvalues and eigenvectors were computed at the same time, and 0 that they were computed in separate steps. (All norms are  $\|\cdot\|_1$ .) The scalings in the test ratios assure that the ratios will be  $O(1)$  (typically less than 10 or 100), independent of  $\|A\|$  and  $\varepsilon$ , and nearly independent of  $n$ .

As in the nonsymmetric case, the test ratios for each test matrix are compared to a user-specified threshold `THRESH`, and a message is printed for each test that exceeds this threshold.

#### 5. 4. 5 Tests Performed on the Symmetric Eigenvalue Drivers

For each driver routine, the following tests will be performed:

$$r_1 = \frac{\|A - ZDZ^*\|}{n\varepsilon \|A\|}$$

$$r_2 = \frac{\|I - ZZ^*\|}{n\varepsilon}$$

$$r_3 = \frac{\|D1 - D2\|}{\varepsilon \|D1\|}$$

where  $Z$  is the matrix of eigenvectors returned when the eigenvector option is given and  $D1$  and  $D2$  are the eigenvalues returned with and without the eigenvector option.

#### 5. 4. 6 Input File for Testing the Symmetric Eigenvalue Routines and Drivers

An annotated example of an input file for testing the symmetric eigenvalue routines and drivers is shown below

```
SEP: Data file for testing Symmetric Eigenvalue Problem routines
7                               Number of values of N
```

4.  $S$  is decomposed as  $Z_4 D_4 Z_4^*$ , for a symmetric positive definite tridiagonal matrix.  $D_5$  is the matrix of eigenvalues computed when  $Z$  is not computed.
5. Selected eigenvalues ( $WA1$ ,  $WA2$ , and  $WA3$ ) are computed and denote eigenvalues computed to high absolute accuracy, with different range options.  $WR$  will denote eigenvalues computed to high relative accuracy.
6. Given the eigenvalues, the eigenvectors of  $S$  are computed in  $Y$ .

To check these calculations, the following test ratios are computed:

$$r_1 = \frac{\|A - VSV^*\|}{n\varepsilon \|A\|}$$

computed by `SSYTRD(UPLO = 'U')`

$$r_2 = \frac{\|I - UV^*\|}{n\varepsilon}$$

test of `SORGTR(UPLO = 'U')`

$$r_3 = \frac{\|A - VSV^*\|}{n\varepsilon \|A\|}$$

computed by `SSYTRD(UPLO = 'L')`

$$r_4 = \frac{\|I - UV^*\|}{n\varepsilon}$$

test of `SORGTR(UPLO = 'L')`

$$r_5 = \frac{\|S - ZD_1Z^*\|}{n\varepsilon \|S\|}$$

$$r_6 = \frac{\|I - ZZ^*\|}{n\varepsilon}$$

$$r_7 = \frac{\|D_1 - D_2\|}{\varepsilon \|D_1\|}$$

$$r_8 = \frac{\|D_1 - D_3\|}{\varepsilon \|D_1\|}$$

$$r_9 = \begin{cases} 0 & \text{if eigenvalues of } S \text{ are within } THRESH \text{ of those in } D_1. \\ 2 * THRESH & \text{otherwise} \end{cases}$$

For  $S$  positive definite,

$$r_{10} = \frac{\|S - Z_4 D_4 Z_4^*\|}{n\varepsilon \|S\|}$$

$$r_{11} = \frac{\|I - Z_4 Z_4^*\|}{n\varepsilon}$$

$$r_{12} = \frac{\|D_4 - D_5\|}{100\varepsilon \|D_4\|}$$



Type	Eigenvalue Distribution			
	Arithmetic	Geometric	Clustered	Other
Zero				1
Identity				2
Diagonal	3	4, 6 <sup>†</sup> , 7 <sup>‡</sup>	5	
$UDU^{-1}$	8, 11 <sup>†</sup> , 12 <sup>‡</sup> , 16*, 19*, 20 <sup>•</sup>	9, 17*	10, 18*	
Symmetric w/Random entries				13, 14 <sup>†</sup> , 15 <sup>‡</sup>
Dag. Dominant		21		

† – matrix entries are  $O(\sqrt{\text{overflow}})$

‡ – matrix entries are  $O(\sqrt{\text{underflow}})$

\* – diagonal entries are positive

★ – matrix entries are  $O(\sqrt{\text{overflow}})$  and diagonal entries are positive

• – matrix entries are  $O(\sqrt{\text{underflow}})$  and diagonal entries are positive

Table 6: Test matrices for the symmetric eigenvalue problem

Type	Eigenvalue Distribution			
	Arithmetic	Geometric	Clustered	Other
Zero				1
Identity				2
Diagonal	3	4, 6 <sup>†</sup> , 7 <sup>‡</sup>	5	
$UDU^{-1}$	8, 11 <sup>†</sup> , 12 <sup>‡</sup>	9	10	
Symmetric w/Random entries				13, 14 <sup>†</sup> , 15 <sup>‡</sup>
Band				16, 17 <sup>†</sup> , 18 <sup>‡</sup>

† – matrix entries are  $O(\sqrt{\text{overflow}})$

‡ – matrix entries are  $O(\sqrt{\text{underflow}})$

Table 7: Test matrices for the symmetric eigenvalue drivers

#### 5.4.4 Tests Performed on the Symmetric Eigenvalue Routines

Finding the eigenvalues and eigenvectors of a symmetric matrix  $A$  is done in the following stages:

1.  $A$  is decomposed as  $USU^*$ , where  $U$  is unitary,  $S$  is real symmetric tridiagonal, and  $U^*$  is the conjugate transpose of  $U$ .  $U$  is represented as a product of Householder transformations, whose vectors are stored in the first  $n-1$  columns of  $V$ , and whose scale factors are in  $TAU$ .
2.  $S$  is decomposed as  $ZDZ^*$ , where  $Z$  is real orthogonal and  $D1$  is a real diagonal matrix of eigenvalues.  $D2$  is the matrix of eigenvalues computed when  $Z$  is not computed.
3. The “PK method is used to compute  $D3$ , the matrix of eigenvalues, using a square-root-free method which does not compute  $Z$ .

- the order  $N$  of the test matrix  $A$
- the type of the test matrix  $A$
- the blocksize  $NB$

The testing program thus consists of a triply-nested loop, the outer one over  $NB$ , the next over  $N$  and the inner one over matrix types. On each iteration of the innermost loop, a matrix  $A$  is generated and used to test the eigenvalue routines.

#### 5.4.1 The Symmetric Eigenvalue Drivers

The driver routines for the symmetric eigenvalue problem are

**xSTEVE** eigenvalue/eigenvector driver for symmetric tridiagonal matrix,

**xSTE VX** selected eigenvalue/eigenvectors for symmetric tridiagonal matrix,

**xSYEV** eigenvalue/eigenvector driver for symmetric matrix,

**xSYEVX** selected eigenvalue/eigenvectors for symmetric matrix,

**xSPEV** eigenvalue/eigenvector driver for symmetric matrix in packed storage,

**xSPEVX** selected eigenvalue/eigenvectors for symmetric matrix in packed storage,

**xSBEV** eigenvalue/eigenvector driver for symmetric band matrix,

**xSBEVX** selected eigenvalue/eigenvectors for symmetric band matrix.

#### 5.4.2 Test Matrices for the Symmetric Eigenvalue Routines

Twenty-one different types of test matrices may be generated for the symmetric eigenvalue routines. Table 6 shows the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries. The expression  $UDU^{-1}$  means a real diagonal matrix  $D$  with  $O(1)$  entries conjugated by a unitary (or real orthogonal) matrix  $U$ . The eigenvalue distributions have the same meanings as in the nonsymmetric case (see Section 5.2.1).

#### 5.4.3 Test Matrices for the Symmetric Eigenvalue Drivers

Eighteen different types of test matrices may be generated for the symmetric eigenvalue drivers. The first 15 test matrices are the same as the types of matrices used to test the symmetric eigenvalue computational routines, and are given in Table 6. Table 7 shows the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries. The expression  $UDU^{-1}$  means a real diagonal matrix  $D$  with  $O(1)$  entries conjugated by a unitary (or real orthogonal) matrix  $U$ . The eigenvalue distributions have the same meanings as in the nonsymmetric case (see Section 5.2.1).

real part of the eigenvalue, the imaginary part of the eigenvalue, the reciprocal condition number of the eigenvalue, and the reciprocal condition number of the vector eigenvector. The end of data is indicated by dimension  $N=0$ . Even if no data is to be tested, there must be at least one line containing  $N=0$ .

The input data for testing `xGESX` also consists of two parts. The first part is identical to that for `xGES` (using `SSXi` instead of `SES` and `CSXi` instead of `CES`). The second consists of precomputed data for testing the eigenvalue/vector condition estimation routines. Each matrix is stored on  $3 \times N$  lines, where  $N$  is its dimension ( $3 \times N^2$  lines for complex data). The first line contains the dimension  $N$  and the dimension  $M$  of an invariant subspace (for complex data, a third integer `ISRT` indicating how the data is sorted is also provided). The second line contains  $M$  integers, identifying the eigenvalues in the invariant subspace (by their position in a list of eigenvalues ordered by increasing real part (or imaginary part, depending on `ISRT` for complex data)). The next  $N$  lines contains the matrix ( $N^2$  lines for complex data). The last line contains the reciprocal condition number for the average of the selected eigenvalues, and the reciprocal condition number for the corresponding right invariant subspace. The end of data is indicated by a line containing  $N=0$  and  $M=0$ . Even if no data is to be tested, there must be at least one line containing  $N=0$  and  $M=0$ .

### 5.3 Testing the Nonsymmetric Eigenvalue Condition Estimation Routines

The main routines tested are `xTREC`, `xTRSL`, `xTRNA` and `xTRSE`. `xTREC` reorders eigenvalues on the diagonal of a matrix in Schur form. `xTRSL` solves the Sylvester equation  $AX + XB = C$  for  $X$  given  $A$ ,  $B$  and  $C$ , `xTRNA` computes condition numbers for individual eigenvalues and right eigenvectors, and `xTRSE` computes condition numbers for the average of a cluster of eigenvalues, as well as their corresponding right invariant subspace. Several auxiliary routines `xLAQU`, `xLAFXC`, `xLAIN`, `xLAQTR` and `xLASZ` are also tested; these are only used with real (`x=S` or `x=D`) data.

No parameters can be varied; the data files contain precomputed test problems along with their precomputed solutions. The reason for this approach is threefold. First, there is no simple residual test ratio which can test correctness of a condition estimator. Second, no comparable code in another library exists to compare solutions. Third, the condition numbers we compute can themselves be quite ill-conditioned, so that we need the precomputed solution to verify that the computed result is within acceptable bounds.

The test program `xeigtsts` reads in the data from the data file `sec.in` (for the `REAL` code). If there are no errors, a single message saying that all the routines pass the tests will be printed. If any routine fails its tests, an error message is printed with the name of the failed routine along with the number of failures, the number of the example with the worst failure, and the test ratio of the worst failure.

For more details on eigencondition estimation, see LAPACK Wiki page Note 13 [4].

### 5.4 Testing the Symmetric Eigenvalue Routines

The test routine for the LAPACK symmetric eigenvalue routines has the following parameters which may be varied:

## 5.2. Input File for Testing the Nonsymmetric Eigenvalue Drivers

There is a single input file to test all four drivers. The input data for each path (testing xCEEV, xCEES, xCEEX and xCESX) is preceded by a single line identifying the path (SEV, SES, SVX and SSX respectively, when x=S, and CEV, CES, CX and CSX respectively, when x=C). We discuss each set of input data in turn.

An annotated example of input data for testing SCEEV is shown below (testing CEEV is identical except CEV replaces SEV):

```

SEV          Data file for the Real Nonsymmetric Eigenvalue Driver
6           Number of matrix dimensions
0 1 2 3 5 10 Matrix dimensions
3 3 1 4 1   Parameters NB, NBMIN, NX, NS, NBCOL
15.0       Threshold for test ratios
T          Put T to test the error exits
2          Read another line with random number generator seed
2518 3899 995 397 Seed for random number generator
SEV 21     Use all matrix types

```

The first line must contain the characters SEV in columns 1-3. The remaining lines are read using list-directed input and specify the following values:

- line 2: The number of values of matrix dimension N
- line 3: The values of N the matrix dimension
- line 4: The values of the parameters NB, NBMIN, NX, NS and NBCOL
- line 5: The threshold value THRESH for the test ratios
- line 6: T to test the error exits
- line 7: An integer code to interpret the random number seed
  - =0: Set the seed to a default value before each run
  - =1: Initialize the seed to a default value only before the first run
  - =2: Like 1, but use the seed values on the next line
- line 8: If line 7 was 2, four integer values for the random number seed
- line 9: Contains 'SEV' in columns 1-3, followed by the number of matrix types (an integer from 0 to 21)
- line 9: (and following) if the number of matrix types is at least one and less than 21, a list of integers between 1 and 21 indicating which matrix types are to be tested.

The input data for testing xCEES has the same format as for xCEEV, except SES replaces SEV when testing SCEES, and CES replaces CEV when testing CEES.

The input data for testing xCEEX consists of two parts. The first part is identical to that for xCEEV (using SVXi instead of SEV and CXi instead of CEV). The second consists of precomputed data for testing the eigenvalue/vector condition estimation routines. Each matrix is stored on  $1+2*N$  lines, where N is its dimension ( $1+N*N*2$  lines for complex data). The first line contains the dimension, a single integer (for complex data, a second integer ISRT indicating how the data is sorted is also provided). The next N lines contain the matrix, one row per line ( $N*2$  lines for complex data, one i temper row). The last N lines correspond to each eigenvalue. Each of these last N lines contains 4 real values: the

5		Number of values of NB, NS, and MAXB
1	3 3 3 20	Values of NB (blocksize)
2	2 2 2 2	Values of NBMIN (minimum blocksize)
1	0 5 9 1	Values of NX (crossover point)
2	4 2 4 6	Values of NS (no. of shifts)
20	20 6 10 10	Values of MAXB (min. blocksize)
20.0		Threshold value
T		Put T to test the error exits
1		Code to interpret the seed
NEP	21	

The first line of the input file must contain the characters NEP in columns 1-3. Lines 2-11 are read using list-directed input and specify the following values:

line 2: The number of values of N  
line 3: The values of N, the matrix dimension  
line 4: The number of values of the parameters NB, NBMIN, NX, NS, and MAXB  
line 5: The values of NB, the blocksize  
line 6: The values of NBMIN, the minimum blocksize  
line 7: The values of NX, the crossover point  
line 8: The values of NS, the number of shifts  
line 9: The values of MAXB, the minimum blocksize  
line 10: The threshold value for the test ratios  
line 11: An integer code to interpret the random number seed  
=0: Set the seed to a default value before each run  
=1: Initialize the seed to a default value only before the first run  
=2: Like 1, but use the seed values on the next line  
line 12: If line 9 was 2, four integer values for the random number seed

The remaining lines occur in sets of 1 or 2 and allow the user to specify the matrix types. Each line contains a 3-character identification in columns 1-3, which must be either NEP or SHS (CHS in complex, DHS in double precision, and ZHS in complex\*16), and the number of matrix types must be the first nonblank item in columns 4-80. If the number of matrix types is at least 1 but is less than the maximum number of possible types, a second line will be read to get the numbers of the matrix types to be used. For example,

```
NEP 21
```

requests all of the matrix types for the nonsymmetric eigenvalue problem, while

```
NEP 4
 9 10 11 12
```

requests only matrices of type 9, 10, 11, and 12.

These test ratios are compared to the input parameter THRESH. If a ratio exceeds THRESH, a message is printed specifying the test matrix, the ratio that failed and its value, just like the tests performed on the nonsymmetric eigenvalue problem computational routines.

In addition to the above tests, xGEMV is tested by computing the test ratios  $r_8$  through  $r_{11}$ .  $r_8$  tests whether the output quantities SCALE, ILO, IH, and ANRM are identical independent of which other output quantities are computed.  $r_9$  tests whether the output quantity RCOND is independent of the other outputs.  $r_{10}$  and  $r_{11}$  are only applied to the matrices in the precomputed examples:

$$r_{10} = \max \frac{|RCONDV - RCDVIN|}{\text{cond}(RCONDV)} \quad r_{11} = \max \frac{|RCONDE - RCDEIN|}{\text{cond}(RCONDE)}$$

RCOND(RONE) is the array of output reciprocal condition numbers of eigenvectors (eigenvalues), RDMN(RDEIN) is the array of precomputed reciprocal condition numbers, and  $\text{cond}(RCOND)$  ( $\text{cond}(RONE)$ ) is the condition number of RCOND (RONE).

xGES takes the input matrix  $A$  and computes its Schur decomposition  $A = VS \cdot T \cdot VS'$  where  $VS$  is orthogonal and  $T$  is (quasi) upper triangular, optionally sorts the eigenvalues on the diagonal of  $T$ , and computes a vector of eigenvalues  $W$ . The following test ratios are computed without sorting eigenvalues in  $T$ , and compared to THRESH

$$\begin{aligned} r_1 &= (T \text{ in Schur form}) & r_2 &= \frac{\|A - VS \cdot T \cdot VS'\|}{n \epsilon \|A\|} \\ r_3 &= \frac{\|I - VS \cdot VS'\|}{n \epsilon} & r_4 &= (W \text{ agrees with diagonal of } T) \\ r_5 &= (T(\text{partial}) = T(\text{full})) & r_6 &= (W(\text{partial}) = W(\text{full})) \end{aligned}$$

$r_7$  through  $r_{12}$  are the same test ratios but with sorting the eigenvalues.  $r_{13}$  indicates whether the sorting was done successfully.

In addition to the above tests, xGESX is tested via ratios  $r_{14}$  through  $r_{17}$ .  $r_{14}$  ( $r_{15}$ ) tests if RONE(ROND) is the same no matter what other quantities are computed.  $r_{16}$  and  $r_{17}$  are only applied to the matrices in the precomputed examples:

$$r_{16} = \max \frac{|RCONDE - RCDEIN|}{\text{cond}(RCONDE)} \quad r_{17} = \max \frac{|RCONDV - RCDVIN|}{\text{cond}(RCONDV)}$$

ROND(RONE) is the output reciprocal condition number of the selected invariant subspace (eigenvalue cluster), RDMN(RDEIN) is the precomputed reciprocal condition number, and  $\text{cond}(ROND)$  ( $\text{cond}(RONE)$ ) is the condition number of ROND (RONE).

## 5.2. Input File for Testing the Nonsymmetric Eigenvalue Routines

An annotated example of an input file for testing the nonsymmetric eigenvalue routines is shown below

```
NEP: Data file for testing Nonsymmetric Eigenvalue Problem routines
7                               Number of values of N
0 1 2 3 5 10 16                Values of N (dimension)
```

To check these calculations, the following test ratios are computed:

$$\begin{aligned}
r_1 &= \frac{\|A - UHU^*\|}{n\varepsilon \|A\|} & r_2 &= \frac{\|I - UU^*\|}{n\varepsilon} \\
r_3 &= \frac{\|H - ZTZ^*\|}{n\varepsilon \|H\|} & r_4 &= \frac{\|I - ZZ^*\|}{n\varepsilon} \\
r_5 &= \frac{\|A - (UZ)T(UZ)^*\|}{n\varepsilon \|A\|} & r_6 &= \frac{\|I - (UZ)(UZ)^*\|}{n\varepsilon} \\
r_7 &= \frac{\|T_1 - T_0\|}{\varepsilon \|T\|} & r_8 &= \frac{\|\Lambda_1 - \Lambda_0\|}{\varepsilon \|\Lambda\|} \\
r_9 &= \frac{\|TR - R\Lambda\|}{\varepsilon \|T\| \|R\|} & r_{10} &= \frac{\|LT - \Lambda L\|}{\varepsilon \|T\| \|L\|} \\
r_{11} &= \frac{\|HX - X\Lambda\|}{n\varepsilon \|H\| \|X\|} & r_{12} &= \frac{\|YH - \Lambda Y\|}{n\varepsilon \|H\| \|Y\|}
\end{aligned}$$

where the subscript 1 indicates that the eigenvalues and eigenvectors were computed at the same time, and 0 that they were computed in separate steps. (All norms are  $\|\cdot\|_1$ .) The scalings in the test ratios assure that the ratios will be  $O(1)$ , independent of  $\|A\|$  and  $\varepsilon$ , and nearly independent of  $n$ .

When the test programs run, these test ratios will be compared with a user-specified threshold `THRESH`, and for each test ratio that exceeds `THRESH`, a message is printed specifying the test matrix, the ratio that failed, and its value. A sample message is

```
Matrix order= 25, type=11, seed=2548,1429,1713,1411, result 8 is 11.33
```

In this example, the test matrix was of order  $n=25$  and of type 11 from Table 5, “seed” is the initial 4-integer seed of the random number generator used to generate  $A$ , and “result” specifies that test ratio  $r_8$  failed to pass the threshold, and its value was 11.33.

## 5.2. Tests Performed on the Nonsymmetric Eigenvalue Drivers

The four drivers have slightly different tests applied to them.

`xGEV` takes the input matrix  $A$  and computes a matrix of its right eigenvectors  $VR$ , a matrix of its left eigenvectors  $VL$ , and a (block) diagonal matrix  $W$  of eigenvalues. If  $W$  is real it may have 2 by 2 diagonal blocks corresponding to complex conjugate eigenvalues.

The test ratios computed are:

$$\begin{aligned}
r_1 &= \frac{\|A \cdot VR - VR \cdot W\|}{n\varepsilon \|A\|} & r_2 &= \frac{\|A' \cdot VL - VL \cdot W\|}{n\varepsilon \|A\|} \\
r_3 &= \frac{\|VR_i\|_{-1}}{\varepsilon} & r_4 &= \frac{\|VL_i\|_{-1}}{\varepsilon} \\
r_5 &= (W(full) = W(partial)) & r_6 &= (VR(full) = VR(partial)) \\
r_7 &= (VL(full) = VL(partial))
\end{aligned}$$

$r_5$ ,  $r_6$  and  $r_7$  check whether  $W$  or  $VR$  or  $VL$  is computed identically independent of whether other quantities are computed or not.  $r_3$  and  $r_4$  also check that the component of  $VR$  or  $VL$  of largest absolute value is real.

(Jordan Block)<sup>T</sup>: Matrix with ones on the diagonal and the first subdiagonal, and zeros elsewhere

$UTU^{-1}$ : Schur-form matrix  $T$  with  $O(1)$  entries conjugated by a unitary (or real orthogonal) matrix  $U$

$XTX^{-1}$ : Schur-form matrix  $T$  with  $O(1)$  entries conjugated by an ill-conditioned matrix  $X$

For eigenvalue distributions other than “Other”, the eigenvalues lie between  $\varepsilon$  (the machine precision) and 1 in absolute value. The eigenvalue distributions have the following meanings:

Arithmetic: Difference between adjacent eigenvalues is a constant

Geometric: Ratio of adjacent eigenvalues is a constant

Clustered: One eigenvalue is 1 and the rest are  $\varepsilon$  in absolute value

Random: Eigenvalues are logarithmically distributed

## 5.2.3 Test Matrices for the Nonsymmetric Eigenvalue Drivers

All four drivers are tested with up to 21 types of random matrices. These are nearly the same as the types of matrices used to test the nonsymmetric eigenvalue computational routines, and are given in Table 3. The only differences are that matrix types 7 and 17 are scaled by a number close to the underflow threshold (rather than its square root), types 8 and 18 are scaled by a number close to the overflow threshold, and types 20 and 21 have certain rows and columns zeroed out. The reason for these changes is to activate the automatic scaling features in the driver, and to test the balancing routine.

In addition, the condition estimation features of the expert drivers `xCHEX` and `xCHSX` are tested by the same precomputed sets of test problems used to test their constituent pieces `xTRNA` and `xTRSN`.

## 5.2.4 Tests Performed on the Nonsymmetric Eigenvalue Routines

Finding the eigenvalues and eigenvectors of a nonsymmetric matrix  $A$  is done in the following stages:

1.  $A$  is decomposed as  $UHU^*$ , where  $U$  is unitary,  $H$  is upper Hessenberg, and  $U^*$  is the conjugate transpose of  $U$ .
2.  $H$  is decomposed as  $ZTZ^*$ , where  $Z$  is unitary and  $T$  is in Schur form; this also gives the eigenvalues  $\lambda_i$ , which may be considered to form a diagonal matrix  $\Lambda$ .
3. The left and right eigenvector matrices  $L$  and  $R$  of the Schur matrix  $T$  are computed.
4. Inverse iteration is used to obtain the left and right eigenvector matrices  $Y$  and  $X$  of the matrix  $H$ .



## 5.2. The Nonsymmetric Eigenvalue Drivers

The driver routines for the nonsymmetric eigenvalue problem are

**xGEEV** eigenvalue/eigenvector driver,

**xGEEVX** expert version of **xGEEV** (includes condition estimation),

**xGEES** Schur form driver, and

**xGEESX** expert version of **xGEES** (includes condition estimation).

For these subroutines, some tests are done by generating random matrices of a dimension and type chosen by the user, and computing error bounds similar to those used for the nonsymmetric eigenvalue computational routines. Other tests use a file of precomputed matrices and condition numbers, identical to that used for testing the nonsymmetric eigenvalue/vector condition estimation routines.

The parameters that can be varied in the random matrix tests are:

- the order  $N$  of the matrix  $A$
- the type of test matrix  $A$
- five numerical parameters: **NB** (the block size), **NBMN** (minimum block size), **NX** (minimum dimension for blocking), **NS** (number of shifts in **xHSEQR**), and **NBCL** (minimum column dimension for blocking).

## 5.2.2 Test Matrices for the Nonsymmetric Eigenvalue Routines

Twenty-one different types of test matrices may be generated for the nonsymmetric eigenvalue routines. Table 5 shows the types available, along with the numbers used to refer to the matrix types. Except as noted, all matrices have  $O(1)$  entries.

Type	Eigenvalue Distribution				
	Arithmetic	Geometric	Clustered	Random	Other
Zero					1
Identity					2
(Jordan Block) <sup>T</sup>					3
Diagonal	4, 7 <sup>†</sup> , 8 <sup>‡</sup>	5	6		
$UTU^{-1}$	9	10	11	12	
$XTX^{-1}$	13	14	15	16, 17 <sup>†</sup> , 18 <sup>‡</sup>	
Random entries					19, 20 <sup>†</sup> , 21 <sup>‡</sup>

<sup>†</sup>— matrix entries are  $O(\sqrt{\text{overflow}})$

<sup>‡</sup>— matrix entries are  $O(\sqrt{\text{underflow}})$

Table 5: Test matrices for the nonsymmetric eigenvalue problem

Matrix types identified as “Zero”, “Identity”, “Diagonal”, and “Random entries” should be self-explanatory. The other matrix types have the following meanings:

requests only matrices of type 4, 5, and 6.

When the tests are run, each test ratio that is greater than or equal to the threshold value causes a line of information to be printed to the output file. The first such line is preceded by a header that lists the matrix types used and the tests performed for the current path. An example line for a test from the SGE path that did not pass when the threshold was set to 1.0 is

```
M =    4, N =    4, NB =    1, type  2, test 13, ratio =  1.14270
```

To get this information for every test, set the threshold to zero. After all the unsuccessful tests have been listed, a summary line is printed of the form

```
SGE:    11 out of  1960 tests failed to pass the threshold
```

If all the tests pass the threshold, only one line is printed for each path:

```
All tests for SGE passed the threshold ( 1960 tests run)
```

## 5.2 Testing the Nonsymmetric Eigenvalue Routines

The test routine for the LAPACK nonsymmetric eigenvalue routines has the following parameters which may be varied:

- the order  $N$  of the test matrix  $A$
- the type of the test matrix  $A$
- three numerical parameters: the blocksize  $NB$ , the number of shifts  $NS$  for the multishift QR method, and the (sub)matrix size  $MNB$  below or equal to which an unblocked, EISPACK style method will be used

The test program thus consists of a triply-nested loop, the outer one over triples  $(NB, NS, MNB)$ , the next over  $N$  and the inner one over matrix types. On each iteration of the innermost loop, a matrix  $A$  is generated and used to test the eigenvalue routines.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main test program

Parameter	Description	Value
$NMAX$	Maximum value for $N$ , $NB$ , $NS$ , and $MNB$	132
$MAXN$	Maximum number of values of the parameters	20

For the nonsymmetric eigenvalue input file,  $MAXN$  is both the maximum number of values of  $N$  and the maximum number of 3-tuples  $(NB, NS, MNB)$ . Similar restrictions exist for the other input files for the eigenvalue test program

```

1 3 3 3 20      Values of NB (the blocksize)
1 0 5 9 1      Values of NX (crossover point)
20.0           Threshold value of test ratio
T             Put T to test the LAPACK routines
T             Put T to test the driver routines
T             Put T to test the error exits
SGE  11
SGB   8
SGT  12
SPO   9
SPP   9
SPB   8
SPT  12
SSY  10
SSP  10
STR  18
STP  18
STB  17
SQR   8
SRQ   8
SLQ   8
SQL   8
SQP   6
STZ   3
SLS   6
SEQ
SLU  11
SCH   9

```

The first 14 lines of the input file are read using list-directed input and are used to specify the values of `M`, `N`, `NB`, and `THRESH` (the threshold value). Lines 12-14 specify if the LAPACK routines, the driver routines, or the error exits are to be tested. The remaining lines occur in sets of 1 or 2 and allow the user to specify the matrix types. Each line contains a 3-character path name in columns 1-3, followed by the number of test matrix types. If the number of matrix types is omitted, as in the above example for `SEQ`, or if a character is encountered before an integer, all the possible matrix types are tested. If the number of matrix types is at least 1 but is less than the maximum number of possible types, a second line will be read to get the numbers of the matrix types to be used. For example, the input line

```
SGE  8
```

requests all of the matrix types for path `SGE`, while

```
SGE  3
  4 5 6
```

- If  $s$  are the true singular values of  $A$ , and  $\tilde{s}$  are the singular values of  $T$ , we compute

$$\|s - \tilde{s}\| / (\|s\| \epsilon)$$

for SGEISX and

$$\|s - \sigma\| / (\|s\| \epsilon)$$

for SGEISS.

- Compute the ratio

$$\|AX - B\| / (\max(m, n) \|A\| \|X\| \epsilon)$$

- If  $m > r$ , form  $R = AX - B$ , and check whether  $R$  is orthogonal to the column space of  $A$  by computing

$$\|R^H A\| / (\max(m, n, nrhs) \|A\| \|B\| \epsilon)$$

- If  $n > r$ , check if  $X$  is in the row space of  $A$  by forming the LQ factorization of  $D = [A^H, X]^H$ . Letting  $E = D(m+1 : m+nrhs, m+1 : m+nrhs)$ , we return

$$\max |d_{ij}| / (\max(m, n, nrhs) \epsilon)$$

### 5.1.5 Tests for the Equilibration Routines

The equilibration routines are xGHEQU, xCHEQU, xFHEQU, xPHEQU and xSHEQU. These routines perform diagonal scaling on various kinds of matrices to reduce their condition number prior to linear equation solving. All of them attempt to somehow equalize the norms of the rows and/or columns of the input matrix by diagonal scaling. This is tested by generating a few matrices for which the answer is known exactly, and comparing the output with the correct answer. There are no testing parameters for the user to set.

Equilibration is also an option to the driver routines for the test paths xGE, xGB, xFO, xFP, and xBB, so it is tested in context there.

### 5.1.6 Input File for Testing the Linear Equation Routines

From the test program's input file, one can control the size of the test matrices, the block size and crossover point for the blocked routines, the paths to be tested, and the matrix types used in testing. We have set the options in the input files to run through all of the test paths. An annotated example of an input file for the REAL test program is shown below

```
Data file for testing REAL LAPACK linear equation routines
7           Number of values of M
0 1 2 3 5 10 16 100   Values of M (row dimension)
7           Number of values of N
0 1 2 3 5 10 16 100   Values of N (column dimension)
1           Number of values of NRHS
2           Values of NRHS (number of right hand sides)
5           Number of values of NB
```

- Apply the orthogonal matrix  $Z$  to  $T$  from the right using SLAIZM and compute the ratio

$$\|R - TZ\|/(m\|R\|\epsilon)$$

- Form  $Z^T Z$  using SLAIZM and compute the ratio

$$\|I - Z^T Z\|/(m\epsilon)$$

## 5.1. Tests for the Least Squares Driver Routines

In the SLS path, driver routines are tested for computing solutions to over- and under-determined, possibly rank-deficient systems of linear equations  $AX = B$  ( $A$  is  $m \times n$ ). For each test matrix type, we generate three matrices: One which is scaled near underflow, a matrix with moderate norm and one which is scaled near overflow.

The SCELS driver computes the least-squares solutions (when  $m \geq n$ ) and the minimum norm solution (when  $m < n$ ) for an  $m \times n$  matrix  $A$  of full rank. To test SCELS, we generate a diagonally dominant matrix  $A$ , and for  $C = A$  and  $C = A^H$ , we

- generate a consistent right-hand side  $B$  such that  $X$  is in the range space of  $C$ , compute a matrix  $X$  using SCELS, and compute the ratio

$$\|AX - B\|/(\max(m, n)\|A\|\|X\|\epsilon)$$

- If  $C$  has more rows than columns (i.e. we are solving a least-squares problem), form  $R = AX - B$ , and check whether  $R$  is orthogonal to the column space of  $A$  by computing

$$\|R^H C\|/(\max(m, n, nrhs)\|A\|\|B\|\epsilon)$$

- If  $C$  has more columns than rows (i.e. we are solving an overdetermined system), check whether the solution  $X$  is in the row space of  $C$  by scaling both  $X$  and  $C$  to have norm one, and forming the QR factorization of  $D = [A, X]$  if  $C = A^H$ , and the LQ factorization of  $D = [A^H, X]^H$  if  $C = A$ . Letting  $E = D(n+1:n+nrhs, n+1:n+nrhs)$  in the first case, and  $E = D(m+1:m+nrhs, m+1:m+nrhs)$  in the latter, we compute

$$\max |d_{ij}|/(\max(m, n, nrhs)\epsilon)$$

The SCELSX and SCLESS drivers solve a possibly rank-deficient system  $AX = B$  using a complete orthogonal factorization (SCELSX) or singular value decomposition (SCLESS), respectively. We generate matrices  $A$  that have rank  $r = \min(m, n)$  or rank  $r = 3 \min(m, n)/4$  and are scaled to be near underflow, of moderate norm or near overflow. We also generate the null matrix (which has rank  $r = 0$ ). Given such a matrix, we then generate a right-hand side  $B$  which is in the range space of  $A$ .

In the process of determining  $X$ , SCELSX computes a complete orthogonal factorization  $AP = QTZ$ , whereas SCLESS computes the singular value decomposition  $A = U \text{diag}(\sigma) V^T$ .

- Compute the least-squares solution to a system of equations  $Ax = b$  using `SGQRS`, and compute the ratio

$$7. \|b - Ax\| / (\|A\| \|x\| \epsilon)$$

In the SQP test path, we test the QR factorization with column pivoting (`SGQPF`), which decomposes a matrix  $A$  into a product of a permutation matrix  $P$ , an orthogonal matrix  $Q$ , and an upper triangular matrix  $R$  such that  $AP = QR$ . We generate three types of matrices  $A$  with singular values  $s$  as follows:

- all singular values are zero,
- all singular values are 1, except for  $\sigma_{\min(m,n)} = 1/\epsilon$ , and
- the singular values are  $1, r, r^2, \dots, r^{\min(m,n)-1} = 1/\epsilon$ .

The following tests are performed:

- Compute the QR factorization with column pivoting using `SGQPF`, compute the singular values  $\tilde{s}$  of  $R$  using `SCH2` and `SBSQR`, and compute the ratio

$$\|\tilde{s} - s\| / (m \|s\| \epsilon)$$

- Generate the orthogonal matrix  $Q$  from the Householder vectors using `SRMQR` and compute the ratio

$$\|AP - QR\| / (m \|A\| \epsilon)$$

- Test the orthogonality of the computed matrix  $Q$  by computing the ratio

$$\|I - Q^H Q\| / (m \epsilon)$$

In the STZ path, we test the trapezoidal reduction (`STZRF`), which decomposes an  $m \times n$  ( $m < n$ ) upper trapezoidal matrix  $R$  (i.e.  $r_{ij} = 0$  if  $i > j$ ) into a product of a strictly upper triangular matrix  $T$  (i.e.  $t_{ij} = 0$  if  $i > j$  or  $j > m$ ) and an orthogonal matrix  $Z$  such that  $R = TZ$ . We generate matrices with the following three singular value distributions:

- all singular values are zero,
- all singular values are 1, except for  $\sigma_{\min(m,n)} = 1/\epsilon$ , and
- the singular values are  $1, r, r^2, \dots, r^{\min(m,n)-1} = 1/\epsilon$ .

To obtain an upper trapezoidal matrix with the specified singular value distribution, we generate a dense matrix using `SLANS` and reduce it to upper triangular form using `SGQRF`.

The following tests are performed:

- Compute the trapezoidal reduction `STZRF`, compute the singular values  $\tilde{s}$  of  $T$  using `SCH2` and `SBSQR`, and compute the ratio

$$\|\tilde{s} - s\| / (m \|s\| \epsilon)$$

## 5.1.3 Tests for the Orthogonal Factorization Routines

The orthogonal factorization routines are contained in the test paths xQR, xRQ, xIQ, xQL, xQP, and xTZ. The first four of these test the QR, RQ, IQ, and QL factorizations without pivoting. The subroutines to generate or multiply by the orthogonal matrix from the factorization are also tested in these paths. There is not a separate test path for the orthogonal transformation routines, since the important thing when generating an orthogonal matrix is not whether or not it is, in fact, orthogonal, but whether or not it is the orthogonal matrix we wanted. The xQP test path is used for QR with pivoting, and xTZ tests the reduction of a trapezoidal matrix by an RQ factorization.

The test paths xQR, xRQ, xIQ, and xQL all use the same set of test matrices and compute similar test ratios, so we will only describe the xQR path. Also, we will refer to the subroutines by their single precision real names, SGEQRF, SGEQRS, SRCQR, and SCRMQR. In the complex case, the orthogonal matrices are unitary, so the names beginning with SQR are changed to CUN. Each of the orthogonal factorizations can operate on  $m \times n$  matrices, where  $m > n$ ,  $m = n$ , or  $m < n$ .

Eight test matrices are used for SQR and the other orthogonal factorization test paths. All are generated with a predetermined condition number (by default,  $\kappa = 2$ ).

- |                          |   |
|--------------------------|---|
| 1. Diagonal              | 5. Random $\kappa = \sqrt{0.1/\varepsilon}$ |
| 2. Upper triangular      | 6. Random $\kappa = 0.1/\varepsilon$        |
| 3. Lower triangular      | 7. Scaled near underflow                    |
| 4. Random $\kappa = 2$ . | 8. Scaled near overflow                     |

The tests for the SQR path are as follows:

- Compute the QR factorization using SGEQRF, generate the orthogonal matrix  $Q$  from the Householder vectors using SRCQR, and compute the ratio
  1.  $\|A - QR\|/(m\|A\|\varepsilon)$
- Test the orthogonality of the computed matrix  $Q$  by computing the ratio
  2.  $\|I - Q^H Q\|/(m\varepsilon)$
- Generate a random matrix  $C$  and multiply it by  $Q$  or  $Q^H$  using SCRMQR with UHO='L', and compare the result to the product of  $C$  and  $Q$  (or  $Q^H$ ) using the explicit matrix  $Q$  generated by SRCQR. The different options for SCRMQR are tested by computing the 4 ratios
  3.  $\|QC - QC\|/(m\|C\|\varepsilon)$
  4.  $\|CQ - CQ\|/(m\|C\|\varepsilon)$
  5.  $\|Q^H C - Q^H C\|/(m\|C\|\varepsilon)$
  6.  $\|CQ^H - CQ^H\|/(m\|C\|\varepsilon)$

where the first product is computed using SCRMQR and the second using the explicit matrix  $Q$ .

Test matrix type	TR	TP	TB
Diagonal	1		
Random $\kappa = 2$	2		1
Random $\kappa = \sqrt{0.1/\varepsilon}$	3		2
Random $\kappa = 0.1/\varepsilon$	4		3
Scaled near underflow	5		4
Scaled near overflow	6		5
Identity	7		6
Unit triangular, $\kappa = 2$	8		7
Unit triangular, $\kappa = \sqrt{0.1/\varepsilon}$	9		8
Unit triangular, $\kappa = 0.1/\varepsilon$	10		9
Matrix elements are Q(1), large right hand side	11		10
First diagonal causes overflow, offdiagonal column norms $< 1$	12		11
First diagonal causes overflow, offdiagonal column norms $> 1$	13		12
Growth factor underflows, solution does not overflow	14		13
Small diagonal causes gradual overflow	15		14
One zero diagonal element	16		15
Large offdiagonals cause overflow when adding a column	17		16
Unit triangular with large right hand side	18		17

Table 3: Test matrices for triangular linear systems

Test ratio	TR	TP	TB
$\ I - AA^{-1}\ /(n\ A\  \ A^{-1}\ \varepsilon)$	1		
$\ b - Ax\ /(\ A\  \ x\ \varepsilon)$	2		1
$\ x - x^*\ /(\ x^*\ \kappa\varepsilon)$	3		2
$\ x - x^*\ /(\ x^*\ \kappa\varepsilon)$ , refined	4		3
(backward error)/ $\varepsilon$	5		4
$\ x - x^*\ /(\ x^*\ (\text{error bound}))$	6		5
RCOND * $\kappa$	7		6
$\ sb - Ax\ /\ A\  \ x\ \varepsilon)$	8		7

Table 4: Tests performed for triangular linear systems



is returned. Since the same value of ANORM is used in both cases, this test measures the accuracy of the estimate computed for  $A^{-1}$ .

The solve and iterative refinement steps are also tested with  $A$  replaced by  $A^T$  or  $A^H$  where applicable. The test ratios computed for the general and symmetric test paths are listed in Table 2. Here we use  $\|LU - A\|$  to describe the difference in the recomputed matrix, even though it is actually  $\|LL^T - A\|$  or some other form for other paths.

Test ratio	GE, PO, IP, SY, SP		GB, GF, PB, PT	
	routines	drivers	routines	drivers
$\ LU - A\ /(n\ A\ \varepsilon)$	1	1	1	1
$\ I - AA^{-1}\ /(n\ A\ \ A^{-1}\ \varepsilon)$	2			
$\ b - Ax\ /(\ A\ \ x\ \varepsilon)$	3	2	2	2
$\ x - x^*\ /(\ x^*\ \kappa\varepsilon)$	4		3	
$\ x - x^*\ /(\ x^*\ \kappa\varepsilon)$ , refined	5	3	4	3
(backward error)/ $\varepsilon$	6	4	5	4
$\ x - x^*\ /(\ x^*\ (\text{error bound}))$	7	5	6	5
RCOND * $\kappa$	8	6	7	6

Table 2: Tests performed for general and symmetric linear systems

### 5.1. Tests for Triangular Matrices

The triangular test paths, xTR, xTP, and xTB include a number of pathological test matrices for testing the auxiliary routines xLAIRS, xLAIPS, and xLABS, which are robust triangular solves used in condition estimation. The triangular test matrices are summarized in Table 3. To generate unit triangular matrices of predetermined condition number, we choose a special unit triangular matrix and use plane rotations to fill in the zeros without destroying the ones on the diagonal. For the xTB path, all combinations of the values 0, 1,  $n - 1$ ,  $(3n - 1)/4$ , and  $(n - 1)/4$  are used for the number of offdiagonal s  $\kappa$ , so the diagonal type is not necessary.

Types 11-18 for the xTR and xTP paths, and types 10-17 for xTB are used only to test the scaling options in xLAIRS, xLAIPS, and xLABS. These subroutines solve a scaled triangular system  $Ax = sb$  or  $A^T x = sb$ , where  $s$  is allowed to underflow to 0 in order to prevent overflow in  $x$ . A growth factor is computed using the norms of the columns of  $A$ , and if the solution can not overflow the Level 2 HES routine is called. Types 11 and 18 test the scaling of  $b$  when  $b$  is initially large, types 12-13 and 15-16 test scaling when the diagonal of  $A$  is small or zero, and type 17 tests the scaling if overflow occurs when adding multiples of the columns to the right hand side. In type 14, no scaling is done, but the growth factor is too large to call the equivalent HES routine.

The tests performed for the triangular routines are similar to those for the general and symmetric routines, including tests of the inverse, solve, iterative refinement, and condition estimation routines. One additional test ratio is computed for the robust triangular solves:

$$\|sb - Ax\|/(\|A\|\|x\|\varepsilon)$$

Table 4 shows the test ratios computed for the triangular test paths.

Test matrix type	GE	GB	GF	PQ	PP	PB	PT	SY, SP, HE, HP
Diagonal	1		1	1			1	1
Upper triangular	2							
Lower triangular	3							
Random $\kappa=2$	4	1	2	2	1	2		2
Random $\kappa = \sqrt{0.1/\epsilon}$	8	5	3	6	5	3		7
Random $\kappa=0.1/\epsilon$	9	6	4	7	6	4		8
First column zero	5	2	8	3	2	8		3
Last column zero	6	3	9	4	3	9		4
Middle column zero				5	4	10		5
Last $n/2$ columns zero	7	4	10					6
Scaled near underflow	10	7	5, 11	8	7	5, 11		9
Scaled near overflow	11	8	6, 12	9	8	6, 12		10
Random unspecified $\kappa$			7			7		
Block diagonal								11 <sup>†</sup>

<sup>†</sup>— complex symmetric test paths only

Table 1: Test matrices for general and symmetric linear systems

diagonal; and for the paths xSY, xSP, xHE, or xHP, replace  $LU$  by  $LDL^T$  or  $UDU^T$ , where  $D$  is diagonal with 1-by-1 and 2-by-2 diagonal blocks.

- Invert the matrix  $A$  using xxxIR, and compute the ratio

$$\|I - AA^{-1}\| / (n\|A\| \|A^{-1}\| \epsilon)$$

For triangular and banded matrices, inversion routines are not available because the inverse would be dense.

- Solve the system  $Ax = b$  using xxxRS, and compute the ratios

$$\begin{aligned} & \|b - Ax\| / (\|A\| \|x\| \epsilon) \\ & \|x - x^*\| / (\|x^*\| \kappa \epsilon) \end{aligned}$$

where  $x^*$  is the exact solution and  $\kappa$  is the condition number of  $A$ .

- Use iterative refinement (xxxRS) to improve the solution, and compute the ratios

$$\begin{aligned} & \|x - x^*\| / (\|x^*\| \kappa \epsilon) \\ & (\text{backward error}) / \epsilon \\ & \|x - x^*\| / (\|x^*\| (\text{error bound})) \end{aligned}$$

- Compute the reciprocal condition number RCOND using xxxCN and compare to the value RCONDC which was computed as  $1/\text{ANORM}^* \text{ANORM}$  after forming the inverse. The larger of the ratios

$$\text{RCOND} / \text{RCONDC} \quad \text{and} \quad \text{RCONDC} / \text{RCOND}$$

{S, C, D, Z} PP	Positive definite packed
{S, C, D, Z} PB	Positive definite band
{S, C, D, Z} PT	Positive definite tridiagonal
{C, Z} HE	Hermitian indefinite matrices
{C, Z} HP	Hermitian indefinite packed
{S, C, D, Z} SY	Symmetric indefinite matrices
{S, C, D, Z} SP	Symmetric indefinite packed
{S, C, D, Z} TR	Triangular matrices
{S, C, D, Z} TP	Triangular packed
{S, C, D, Z} TB	Triangular band
{S, C, D, Z} QR	QR decomposition
{S, C, D, Z} RQ	RQ decomposition
{S, C, D, Z} LQ	LQ decomposition
{S, C, D, Z} QL	QL decomposition
{S, C, D, Z} QP	QR decomposition with column pivoting
{S, C, D, Z} TZ	Trapezoidal matrix (RQ factorization)
{S, C, D, Z} LS	Least Squares driver routines
{S, C, D, Z} EQ	Equilibration routines
{S, C, D, Z} LU	LU variants
{S, C, D, Z} CH	Cholesky variants

The xQR, xRQ, xLQ, and xQL test paths also test the routines for generating or multiplying by an orthogonal or unitary matrix expressed as a sequence of elementary Householder transformations.

### 5.1. Tests for General and Symmetric Matrices

For each LAPACK test path specified in the input file, the test program generates test matrices, calls the LAPACK routines in that path, and computes a number of test ratios to verify that each operation has performed correctly. The test matrices used in the test paths for general and symmetric matrices are shown in Table 1. Both the computational routines and the driver routines are tested with the same set of matrix types. In this context,  $\varepsilon$  is the machine epsilon and  $\kappa$  is the condition number of the matrix  $A$ . Matrix types with one or more columns set to zero (or rows and columns, if the matrix is symmetric) are used to test the error return codes. For band matrices, all combinations of the values 0, 1,  $n-1$ ,  $(3n-1)/4$ , and  $(n-1)/4$  are used for  $\mathbf{KL}$  and  $\mathbf{KU}$  in the  $\mathbf{GB}$  path, and for  $\mathbf{KD}$  in the  $\mathbf{PB}$  path. For the tridiagonal test paths xGT and xPT, types 1-6 use matrices of predetermined condition number, while types 7-12 use random tridiagonal matrices.

For the LAPACK test paths shown in Table 1, each test matrix is subjected to the following tests:

- Factor the matrix using xxxTRF, and compute the ratio

$$\|LU - A\|/(n\|A\|\varepsilon)$$

This form is for the paths xGE, xGB, and xGF. For the paths xFQ, xFP, or xFB, replace  $LU$  by  $LL^T$  or  $U^TU$ ; for xPT, replace  $LU$  by  $LDL^T$  or  $U^TDU$ , where Dis

## 5 More About Testing

There are two distinct test programs for LAPACK routines in each data type, one for the linear equation routines and one for the eigensystem routines. Each program has its own style of input, and the eigensystem test program accepts 13 different sets of input, although four of these may be concatenated into one data set, for a total of 10 input files. The following sections describe the different input formats and testing styles.

The main test procedure for the REAL linear equation routines is in LAPACK/TESTING/LIN/schkaa.f in the Unix version and is the first program unit in SUNISIF in the non-Unix version. The main test procedure for the REAL eigenvalue routines is in LAPACK/TESTING/EIG/schkee.f in the Unix version and is the first program unit in SEIGSIF in the non-Unix version.

### 5.1 The Linear Equation Test Program

The test program for the linear equation routines is driven by a data file from which the following parameters may be varied:

- $M$  the matrix row dimension
- $N$  the matrix column dimension
- $NRHS$ , the number of right hand sides
- $NB$  the blocksize for the blocked routines
- $NX$  the crossover point, the point in a block algorithm at which we switch to an unblocked algorithm

For symmetric or Hermitian matrices, the values of  $N$  are used for the matrix dimension.

The number and size of the input values are limited by certain program maxima which are defined in PARAMETER statements in the main test programs. For the linear equation test program these are:

Parameter	Description	Value
MAX	Maximum value of $M$ or $N$ for rectangular matrices	132
MAXN	Maximum number of values of $M$ , $N$ , $NB$ or $NX$	12
MAXRHS	Maximum value of $NRHS$	10

The input file also specifies a set of LAPACK path names and the test matrix types to be used in testing the routines in each path. Path names are 3 characters long; the first character indicates the data type, and the next two characters identify a matrix type or problem type. The test paths for the linear equation test program are as follows:

- {S, C, D, Z} GE General matrices (LU factorization)
- {S, C, D, Z} GB General band matrices
- {S, C, D, Z} GT General tri-diagonal
- {S, C, D, Z} PO Positive definite matrices (Cholesky factorization)

#### 4.9 Send the Results to Tennessee

Congratulations! You have now finished installing and testing LAPACK. Your participation is greatly appreciated. If possible, results and comments should be sent by electronic mail to

sost@s.utk.edu

Otherwise, results may be submitted either by sending the authors a hard copy of the output files or by returning the distribution tape with the output files stored on it.

We encourage you to make the LAPACK library available to your users and provide us with feedback from their experiences. You should make it clear that this software is still under development, and parts of it may be changed before the project is completed. The changes may affect the calling sequences of some routines, so the public release of LAPACK is not guaranteed to be compatible with this version.

If you would like to do more, please contact us so that we may coordinate your efforts with the development of the final test release of LAPACK. One option is to look at ways to improve the performance of LAPACK on your machine. If you do not have optimized BLAS, tuning the BLAS would likely have a dramatic effect on performance. Other suggestions on fine-tuning specific algorithms are also welcome. For example, one of our test sites noticed that the row interchanges in the LU factorization routine `SGEIRF` were degrading performance on the IBM3090 because of the non-unit stride in `SSWP` [2]. In response we added the auxiliary routine `SLASWP` to interchange a block of rows, so that users of the IBM3090 could easily replace this routine with one in which the row interchanges are applied to one column at a time.

and `LOADOPTS` to refer to the loader and desired load options for your machine. Then type `make` followed by the data types desired, as in the examples of Section 3.5. The library of instrumented code is created in `LAPACK/TIMING/EIG/eigsrc.a`.

- b) To make the eigensystem timing programs, go to `LAPACK/TIMING/EIG` and edit the `makefile`. Define `FORTRAN` and `OPTS` to refer to the compiler and desired compiler options for your machine, and define `LOADER` and `LOADOPTS` to refer to the loader and desired load options for your machine. If you are not using the Fortran BLAS, define `BLAS` to point to your system's BLAS library, instead of `../..blas.a`.
- c) Type `make` followed by the data types desired, as in the examples of Section 3.5. The executable files are called `xeigtims`, `xeigtimc`, `xeigtimd`, and `xeigtimz` and are created in `LAPACK/TIMING`.
- d) Go to `LAPACK/TIMING` and make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value, or to restrict the number of tests if you are using a computer with performance in between that of a workstation and that of a supercomputer. Instead of decreasing the matrix dimensions to reduce the time, it would be better to reduce the number of matrix types to be timed, since the performance varies more with the matrix size than with the type. For example, for the nonsymmetric eigenvalue routines, you could use only one matrix of type 4 instead of four matrices of types 1, 3, 4, and 6. See Section 6 for further details.
- e) Run the program for each data type you are using. For the `REAL` version, the commands for the small data sets are

```
xeigtims < sneptim.in > sneptim.out
xeigtims < sseptim.in > sseptim.out
xeigtims < ssvdtim.in > ssvdtim.out
xeigtims < sgeptim.in > sgeptim.out
```

or the commands for the large data sets are

```
xeigtims < SNEPTIM.in > SNEPTIM.out
xeigtims < SSEPTIM.in > SSEPTIM.out
xeigtims < SSVDTIM.in > SSVDTIM.out
xeigtims < SGEPTIM.in > SGEPTIM.out
```

Similar commands should be used for the other data types.

- f) Send the output files to the authors as directed in Section 4.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

- a) Go to LAPACK/TIMING and make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value. If you modified the values of `NorNB` in Section 4.8.1, set `MN` and `K` accordingly. The large parameters among `MN` and `K` should be the same as the matrix sizes used in timing the linear equation routines, and the small parameter should be the same as the block sizes used in timing the linear equation routines. If necessary, the large data set can be simplified by using only one value of `LDA`.
- b) Run the program for each data type you are using. For the `REAL` version, the commands for the small data sets are

```
xtims < sblas.in1 > sblas.out1
xtims < sblas.in2 > sblas.out2
xtims < sblas.in3 > sblas.out3
```

or the commands for the large data sets are

```
xtims < SBLAS.in1 > SBLAS.out1
xtims < SBLAS.in2 > SBLAS.out2
xtims < SBLAS.in3 > SBLAS.out3
```

Similar commands should be used for the other data types.

- c) Send the output files to the authors as directed in Section 4.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the `BLAS` library or libraries that you used.

#### 4.8.3 Timing the Eigensystem Routines

The eigensystem timing programs found in `LAPACK/TIMING/EIG` and the input files are in `LAPACK/TIMING`. Four input files are provided in each data type for timing the eigensystem routines, one for the nonsymmetric eigenvalue problem, one for the symmetric eigenvalue problem, one for the singular value decomposition, and one for the generalized nonsymmetric eigenvalue problem. For the `REAL` version, the small data sets are called `sneptim.in`, `sseptim.in`, `ssvdtim.in`, and `sgeptim.in`, respectively, and the large data sets are called `SNEPTIM.in`, `SSEPTIM.in`, `SSVDTIM.in`, and `SGEPTIM.in`. Each of the four input files reads a different set of parameters, and the format of the input is indicated by a 3-character code on the first line.

The timing program for eigenvalue/singular value routines accumulates the operation count as the routines are executing using special instrumented versions of the `LAPACK` routines. The first step in compiling the timing programs is therefore to make a library of the instrumented routines.

- a) To make a library of the instrumented `LAPACK` routines, first go to `LAPACK/TIMING/EIG/EIGSRC` and edit the `makefile`. Define `FORTRAN` and `OPTS` to refer to the compiler and desired compiler options for your machine, and define `LOADER`

desired load options for your machine. If you are not using the Fortran BLAS, define BLAS to point to your system's BLAS library, instead of ../../blas.a.

- b) Type `make` followed by the data types desired, as in the examples of Section 3.5. The executable files are called `xtims`, `xtimc`, `xtimd`, and `xtimz` and are created in LAPACK/TIMING.
- c) Go to LAPACK/TIMING and make any necessary modifications to the input files. You may need to set the minimum time a subroutine will be timed to a positive value, or to restrict the size of the tests if you are using a computer with performance in between that of a workstation and that of a supercomputer. The computational requirements can be cut in half by using only one value of `IDA`. If it is necessary to also reduce the matrix sizes or the values of the blocksize, corresponding changes should be made to the BLAS input files (see Section 4.8.2).
- d) Run the program for each data type you are using. For the REAL version, the commands for the small data sets are

```
xtims < stime.in > stime.out
xtims < sband.in > sband.out
xtims < stime2.in > stime2.out
```

or the commands for the large data sets are

```
xtims < STIME.in > STIME.out
xtims < SBAND.in > SBAND.out
xtims < STIME2.in > STIME2.out
```

Similar commands should be used for the other data types.

- e) Send the output files to the authors as directed in Section 4.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

#### 4.8.2 Timing the BLAS

The linear equation timing program is also used to time the BLAS. Three input files are provided in each data type for timing the Level 2 and 3 BLAS. These input files time the BLAS using the matrix shapes encountered in the LAPACK routines, and we will use the results to analyze the performance of the LAPACK routines. For the REAL version, the small data files are `sblas.in1`, `sblas.in2`, and `sblas.in3` and the large data files are `SBLAS.in1`, `SBLAS.in2`, and `SBLAS.in3`. There are three sets of inputs because there are three parameters in the Level 3 BLAS,  $M$ ,  $N$ , and  $K$ , and in most applications one of these parameters is small (on the order of the blocksize) while the other two are large (on the order of the matrix size). In `sblas.in1`,  $M$  and  $N$  are large but  $K$  is small, while in `sblas.in2` the small parameter is  $M$  and in `sblas.in3` the small parameter is  $N$ . The Level 2 BLAS are timed only in the first data set, where  $K$  is also used as the bandwidth for the banded routines.



## 4.8 Run the LAPACK Timing Programs

There are two distinct timing programs for LAPACK routines in each data type, one for the linear equation routines and one for the eigensystem routines. The timing program for the linear equation routines is also used to time the BLAS. We encourage you to conduct these timing experiments in REAL and COMPLEX or in DOUBLE PRECISION and COMPLEX\*16; it is not necessary to send timing results in all four data types.

Two sets of input files are provided, a small set and a large set. The small data sets are appropriate for a standard workstation or other non-vector machine. The large data sets are appropriate for supercomputers, vector computers, and high-performance workstations. We are mainly interested in results from the large data sets, and it is not necessary to run both the large and small sets. The values of  $N$  in the large data sets are about five times larger than those in the small data set, and the large data sets use additional values for parameters such as the block size  $NB$  and the leading array dimension  $LDA$ . Small data sets are indicated by lower case names, such as `stime.in`, and large data sets are indicated by upper case names, such as `STIME.in`. Except as noted, the leading 's' (or 'S') in the input file name must be replaced by 'd', 'c', or 'z' ('D', 'C', or 'Z') for the other data types.

We encourage you to obtain timing results with the large data sets, as this allows us to compare different machines. If this would take too much time, suggestions for paring back the large data sets are given in the instructions below. We also encourage you to experiment with these timing programs and send us any interesting results, such as results for larger problems or for a wider range of block sizes. The main programs are dimensioned for the large data sets, so the parameters in the main program may have to be reduced in order to run the small data sets on a small machine, or increased to run experiments with larger problems.

The minimum time each subroutine will be timed is set to 0.0 in the large data files and to 0.05 in the small data files, and on many machines this value should be increased. If the timing interval is not long enough, the time for the subroutine after subtracting the overhead may be very small or zero, resulting in megaflop rates that are very large or zero. (To avoid division by zero, the megaflop rate is set to zero if the time is less than or equal to zero.) The minimum time that should be used depends on the machine and the resolution of the clock.

For more information on the timing programs and how to modify the input files, see Section 6.

### 4.8.1 Timing the Linear Equations Routines

The linear equation timing programs are found in LAPACK/TIMING/LIN and the input files are in LAPACK/TIMING. Three input files are provided in each data type for timing the linear equation routines, one for square matrices, one for band matrices, and one for rectangular matrices. The small data sets for the REAL version are `stime.in`, `sband.in`, and `stime2.in`, respectively, and the large data sets are `STIME.in`, `SBAND.in`, and `STIME2.in`.

- a) To make the linear equation timing programs, go to LAPACK/TIMING/LIN and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine, and define LOADER and LOADOPTS to refer to the loader and

- d) Send the output files to the authors as directed in Section 4.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

#### 4.7. Testing the Eigensystem Routines

- a) Go to LAPACK/TESTING/EIG and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine, and define LOADER and LOADOPTS to refer to the loader and desired load options for your machine. If you are not using the Fortran BLAS, define BLAS to point to your system's BLAS library, instead of ../.. /blas.a.
- b) Type make followed by the data types desired, as in the examples of Section 3.5. The executable files are called xeigtsts, xeigtstc, xeigtstd, and xeigtstz and are created in LAPACK/TESTING.
- c) Go to LAPACK/TESTING and run the tests for each data type. The tests for the eigensystem routines use ten separate input files, for testing the generalized nonsymmetric eigenvalue problem routines, the nonsymmetric eigenvalue problem drivers, the nonsymmetric eigenvalue problem routines, the symmetric eigenvalue problem routines, and the singular value decomposition routines. The tests for the REAL version are as follows:

```
xeigtsts < nep.in > snep.out
xeigtsts < sep.in > ssep.out
xeigtsts < svd.in > ssvd.out
xeigtsts < sec.in > sec.out
xeigtsts < sed.in > sed.out
xeigtsts < sgg.in > sgg.out
xeigtsts < ssg.in > ssg.out
xeigtsts < ssb.in > ssb.out
xeigtsts < sbal.in > sbal.out
xeigtsts < sbak.in > sbak.out
```

The tests using xeigtstc, xeigtstd, and xeigtstz also use the input files nep.in, sep.in, and svd.in, but the leading 's' in the other input file names must be changed to 'c', 'd', or 'z'. We have shown the output of these ten tests going to ten different output files, but we would prefer to receive one file containing the results of all the tests.

- d) Send the output files to the authors as directed in Section 4.9. Please tell us the type of machine on which the tests were run, the compiler options that were used, and details of the BLAS library or libraries that you used.

#### 4.5 Create the LAPACK Library

- a) Go to the directory LAPACK/SRC and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine.
- b) Type `make` followed by the data types desired, as in the examples of Section 3.5. The `make` command can be run more than once to add another data type to the library if necessary.

The LAPACK library is created in LAPACK/lapack.a.

#### 4.6 Create the Test Matrix Generator Library

- a) Go to the directory LAPACK/TESTING/MATGEN and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine.
- b) Type `make` followed by the data types desired, as in the examples of Section 3.5. The `make` command can be run more than once to add another data type to the library if necessary.

The test matrix generator library is created in LAPACK/tmglib.a.

#### 4.7 Run the LAPACK Test Programs

There are two distinct test programs for LAPACK routines in each data type, one for the linear equation routines and one for the eigensystem routines. In each data type, there is one input file for testing the linear equation routines and ten input files for testing the eigenvalue routines. The input files reside in LAPACK/TESTING. For more information on the test programs and how to modify the input files, see Section 5.

##### 4.7.1 Testing the Linear Equations Routines

- a) Go to LAPACK/TESTING/LIN and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine, and define LOADER and LOADOPTS to refer to the loader and desired load options for your machine. If you are not using the Fortran BLAS, define BLAS to point to your system's BLAS library, instead of `../..blas.a`.
- b) Type `make` followed by the data types desired, as in the examples of Section 3.5. The executable files are called `xchks`, `xchkc`, `xchkd`, or `xchkz` and are created in LAPACK/TESTING.
- c) Go to LAPACK/TESTING and run the tests for each data type. For the REAL version, the command is

```
xchks < stest.in > stest.out
```

The tests using `xchkd`, `xchkc`, and `xchkz` are similar with the leading 's' in the input and output file names replaced by 'd', 'c', or 'z'.

- b) Type `make` followed by the data types desired, as in the examples of Section 3.5. The `make` command can be run more than once to add another data type to the library if necessary.

The BLAS library is created in `LAPACK/blas.a` and not in the current directory.

#### 4.4 Run the BLAS Test Programs

Test programs for the Level 2 and 3 BLAS are in the directory `LAPACK/BLAS/TESTING`. A test program for the Level 1 BLAS is not included, in part because only a subset of the original set of Level 1 BLAS is actually used in LAPACK and the old test program was designed to test the full set of Level 1 BLAS. The original Level 1 BLAS test program is available from netlib as `T05` algorithm 539.

- a) To make the Level 2 BLAS test program, go to `LAPACK/BLAS/TESTING` and edit the makefile called `makeblat2`. Define `FORTTRAN` and `OPTS` to refer to the compiler and desired compiler options for your machine, and define `LOADER` and `LOADOPTS` to refer to the loader and desired load options for your machine. If you are not using the Fortran BLAS, define `BLAS` to point to your system's BLAS library, instead of `../../../../blas.a`.
- b) Type `make -f makeblat2` followed by the data types desired, as in the examples of Section 3.5. The executable files are called `xblat2s`, `xblat2d`, `xblat2c`, and `xblat2z` and are created in `LAPACK/BLAS`.
- c) Go to `LAPACK/BLAS` and run the Level 2 BLAS tests. For the REAL version, the command is

```
xblat2s < sblat2.in
```

Similar commands should be used for the other test programs, with the leading 's' in the input file name replaced by 'd', 'c', or 'z'. The name of the output file is indicated on the first line of the input file and is currently defined to be `SBLAT2.SUMM` for the REAL version, with similar names for the other data types.

- d) To compile and run the Level 3 BLAS test program, repeat steps a-c using the makefile `makeblat3`. For step c, the executable program in the REAL version is `xblat3s`, the input file is `sblat3.in`, and output is to the file `SBLAT3.SUMM`, with similar names for the other data types.

If the tests using the supplied data files were completed successfully, consider whether the tests were sufficiently thorough. For example, on a machine with vector registers, at least one value of  $N$  greater than the length of the vector registers should be used; otherwise, important parts of the compiled code may not be exercised by the tests. If the tests were not successful, either because the program did not finish or the test ratios did not pass the threshold, you will probably have to find and correct the problem before continuing. If you have been testing a system-specific BLAS library, try using the Fortran BLAS for the routines that did not pass the tests. For more details on the BLAS test program, see [8 and [6]. ]

'U': Underflow threshold

Some people may be familiar with RUMCH(DIMCH), a primitive routine for setting machine parameters in which the user must comment out the appropriate assignment statements for the target machine. If a version of RUMCH is on hand, the assignments in SLAMCH can be made to refer to RUMCH using the correspondence

SLAMCH('U') = RUMCH(1)

SLAMCH('O') = RUMCH(2)

SLAMCH('E') = RUMCH(3)

SLAMCH('B') = RUMCH(5)

The safe minimum returned by SLAMCH('S') is initially set to the underflow value, but if  $1/(\text{overflow}) \geq (\text{underflow})$  it is recomputed as  $(1/(\text{overflow})) * (1 + \epsilon)$ , where  $\epsilon$  is the machine precision.

#### 4.2. Installing SECOND and DSECOND

Both the timing routines and the test routines call SECOND(DSECOND), a real function with no arguments that returns the time in seconds from some fixed starting time. Our version of this routine returns only "user time", and not "user time + system time". The version of SECOND in `second.f` calls EIIME, a Fortran library routine available on some computer systems. If EIIME is not available or a better local timing function exists, you will have to provide the correct interface to SECOND and DSECOND on your machine.

The test program in `secondtst.f` performs a million operations using 5000 iterations of the SAXPY operation  $y := y + ax$  on a vector of length 100. The total time and megaflops for this test is reported, then the operation is repeated including a call to SECOND on each of the 5000 iterations to determine the overhead due to calling SECOND. Run the test program by typing `testsecond` (or `testdsecond`). There is no single right answer, but the times in seconds should be positive and the megaflop ratios should be appropriate for your machine. If you modify SECOND or DSECOND, copy `second.f` and/or `dsecond.f` to LAPACK/SRC/ for inclusion in the LAPACK library.

#### 4.3 Create the BLAS Library

Ideally, a highly optimized version of the BLAS library already exists on your machine. In this case you can go directly to Section 4.4 to make the BLAS test program. You may already have a library containing some of the BLAS, but not all (Level 1 and 2, but not Level 3, for example). If so, you should use your local version of the BLAS wherever possible.

- a) Go to LAPACK/BLAS/SRC and edit the makefile. Define FORTRAN and OPTS to refer to the compiler and desired compiler options for your machine. If you already have some of the BLAS, comment out the lines defining the BLAS you have.

ASCII character set  
Tests completed

If any modifications were required to ISAME, copy `lsame.f` to both `LAPACK/BLAS/SRC/` and `LAPACK/SRC/`. The function ISAME is needed by both the BLAS and LAPACK so it is safer to have it in both libraries as long as this does not cause trouble in the link phase when both libraries are used.

#### 4.2. Installing SLAMCH and DLAMCH

SLAMCH and DLAMCH are real functions with a single character parameter that indicates the machine parameter to be returned. The test program `slamchtst.f` simply prints out the different values computed by SLAMCH so you need to know something about what the values should be. For example, the output of the test program for SLAMCH on a Sun SPARCstation is

```
Epsilon           =      5.96046E-08
Safe minimum      =      1.17549E-38
Base              =      2.00000
Precision         =      1.19209E-07
Number of digits in mantissa = 24.0000
Rounding mode     =      1.00000
Minimum exponent  =     -125.000
Underflow threshold =      1.17549E-38
Largest exponent  =      128.000
Overflow threshold =      3.40282E+38
Reciprocal of safe minimum =      8.50706E+37
```

On a Gay machine, the safe minimum underflows its output representation and the overflow threshold overflows its output representation, so the safe minimum is printed as 0.00000 and overflow is printed as R. This is normal. If you would prefer to print a representable number, you can modify the test program to print `SEMIN*100.` and `RMAX/100.` for the safe minimum and overflow thresholds.

Run the test program by typing `testslamch`. If any modifications were made to SLAMCH, copy `slamch.f` to `LAPACK/SRC/`. Do the same for DLAMCH and the test program `testdlamch`. If both tests were successful, go to Section 4.2.3.

If SLAMCH (or DLAMCH) returns an invalid value, you will have to create your own version of this function. The following options are used in LAPACK and must be set:

- 'B': Base of the machine
- 'E': Epsilon (relative machine precision)
- 'O': Overflow threshold
- 'P': Precision = Epsilon \* Base
- 'S': Safe minimum (often same as underflow threshold)

requirements will be less if you do not use all four data types. The total space requirements including the object files is approximately 70 MB for all four data types.

If you received a tar file of LAPACK via the file transfer program `ftp`, enter the following command to untar the file:

```
tar xvf file (where file is the name of the tar file)
```

Since single precision (`REAL+COMPLEX`) and double precision (`DOUBLEPRECISION+COMPLEX*16`) are separated into two separate tar files, the space requirements for each tar file will be half what they are for the tape.

## 4.2 Test and Install the Machine-Dependent Routines.

There are five machine-dependent functions in the test and timing package, at least three of which must be installed. They are

<code>LSAME</code>	<code>LOGICAL</code>	Test if two characters are the same regardless of case
<code>SLAMCH</code>	<code>REAL</code>	Determine machine-dependent parameters
<code>DLAMCH</code>	<code>DOUBLEPRECISION</code>	Determine machine-dependent parameters
<code>SECON</code>	<code>REAL</code>	Return time in seconds from a fixed starting time
<code>DSECON</code>	<code>DOUBLEPRECISION</code>	Return time in seconds from a fixed starting time

If you are working only in single precision, you do not need to install `DLAMCH` and `DSECON` and if you are working only in double precision, you do not need to install `SLAMCH` and `SECON`.

These five subroutines are provided on the tape in `LAPACK/INSTALL`, along with five test programs and a `makefile`. To compile the five test programs, go to `LAPACK/INSTALL` and edit the `makefile`. Define `FORTRAN` and `OPTS` to refer to the compiler and desired compiler options for your machine. Then type `make` to create test programs called `testlsame`, `testslamch`, `testdlamch`, `testsecond`, and `testdsecnd`. The expected results of each test program are described below.

### 4.2. Installing LSAME

`LSAME` is a logical function with two character parameters, `A` and `B`. It returns `.TRUE` if `A` and `B` are the same regardless of case, or `.FALSE` if they are different. For example, the expression

```
LSAME( UPL0, 'U' )
```

is equivalent to

```
( UPL0.EQ.'U' ).OR.( UPL0.EQ.'u' )
```

The supplied version works correctly on all systems that use the ASCII code for internal representations of characters. For systems that use the EBCDIC code, one constant must be changed. For CDC systems with 6-12 bit representation, alternative code is provided in the comments. The test program in `lsametst.f` tests all combinations of the same character in upper and lower case for `A` and `B` and two cases where `A` and `B` are different characters.

Run the test program by typing `testlsame`. If `LSAME` works correctly, the only message you should see is

```

xeigtsts < ssb.in > ssb.out
repeat for c, d, and z (except for nep.in, sep.in, and svd.in, replace the leading s
in the input file by c, d, or z, respectively)

```

#### 8. Run the LAPACK Timing Program

```

cd LAPACK/TIMING/LIN
make
cd LAPACK/TIMING
xtims < stime.in > stime.out
xtims < sband.in > sband.out
xtims < stime2.in > stime2.out
repeat for c, d, and z
xtims < sblas.in1 > sblas.out1
xtims < sblas.in2 > sblas.out2
xtims < sblas.in3 > sblas.out3
repeat for c, d, and z
cd LAPACK/TIMING/EIG/EIGSRC
make
cd LAPACK/TIMING/EIG
make
cd LAPACK/TIMING
xeigtims < sgeptim.in > sgeptim.out
xeigtims < sneptim.in > sneptim.out
xeigtims < sseptim.in > sseptim.out
xeigtims < ssvdtim.in > ssvdtim.out
repeat for c, d, and z

```

### 4.1 Read the Tape or Untar the File

If you received a tar tape of LAPACK type one of the following commands to unload the tape (the device name may be different at your site):

```

tar xvf /dev/rst0 (cartridge tape), or
tar xvf /dev/rmt8 (9-track tape)

```

This will create a top-level directory called LAPACK. You will need about 28 megabytes to read in the complete tape. On a Sun SPARCstation, the libraries used 14 MB and the LAPACK executable files used 20 MB. In addition, the object files used 18 MB, but the object files can be deleted after creating the libraries and executable files. Your actual space



```
xblat3d < dblat3.in
xblat3c < cblat3.in
xblat3z < zblat3.in
```

5. Geate the LAPACK library

```
cp LAPACK/INSTALL/lsame.f LAPACK/SRC/
cp LAPACK/INSTALL/slamch.f LAPACK/SRC/
cp LAPACK/INSTALL/dlamch.f LAPACK/SRC/
cp LAPACK/INSTALL/second.f LAPACK/SRC/
cp LAPACK/INSTALL/dsecnd.f LAPACK/SRC/
cd LAPACK/SRC
make
```

6. Geate the Library of Test Matrix Generators

```
cd LAPACK/TESTING/MATGEN
make
```

7. Run the LAPACK Test Programs

```
cd LAPACK/TESTING/LIN
make
cd LAPACK/TESTING
xchks < stest.in > stest.out
xchkd < dtest.in > dtest.out
xchkc < ctest.in > ctest.out
xchkz < ztest.in > ztest.out
cd LAPACK/TESTING/EIG
make
cd LAPACK/TESTING
xeigtsts < nep.in > snep.out
xeigtsts < sep.in > ssep.out
xeigtsts < svd.in > ssvd.out
xeigtsts < sec.in > sec.out
xeigtsts < sed.in > sed.out
xeigtsts < ssg.in > ssg.out
xeigtsts < sgg.in > sgg.out
xeigtsts < sbal.in > sbal.out
xeigtsts < sbak.in > sbak.out
```

## Quick Reference Guide for Installation of LAPACK

If you insist on not reading the instructions, here is an abbreviated set of directions for installing and testing LAPACK

To install, test, and time LAPACK

1. Read the tape or untar the file.

```
tar xvf /dev/rst0 (cartridge tape), or
tar xvf /dev/rmt8 (9-track tape), or
tar xvf file (from a file)
```

2. Test and Install the Machine-Dependent Routines

```
cd LAPACK/INSTALL
make
testlsame
testslamch
testdlamch
testsecond
testdsecnd
```

3. Create the BLAS Library, *if necessary*

*(NOTE For best performance, it is recommended you use the manufacturer's BLAS)*

```
cp LAPACK/INSTALL/lsame.f LAPACK/BLAS/SRC/
cd LAPACK/BLAS/SRC
make
```

4. Run the Level 2 and 3 BLAS Test Programs

```
cd LAPACK/BLAS/TESTING
make -f makeblat2
cd LAPACK/BLAS
xblat2s < sblat2.in
xblat2d < dblat2.in
xblat2c < cblat2.in
xblat2z < zblat2.in
cd LAPACK/BLAS/TESTING
make -f makeblat3
cd LAPACK/BLAS
xblat3s < sblat3.in
```

without any options creates a library of all four data types. The `make` command can be run more than once to add another data type to the library if necessary. Because of the quantity of software in LAPACK, compiling all four data types into one library may not be advisable; see Appendix D for alternate suggestions.

Similarly, the makefiles for the test routines create separate test programs for each data type. These programs can be created one at a time:

```
make single
make double
. . .
```

or all at once:

```
make single double complex complex16
```

where the last command is equivalent to typing `make` by itself. In the case of the BLAS test programs, where the makefile has a name other than `makefile`, the `-f` option must be added to specify the file name, as in the following example:

```
make -f makeblat2 single
```

The makefiles used to create libraries call `ranlib` after each `ar` command. Some computers (for example, CRAY computers running UNICOS) do not require `ranlib` to be run after creating a library. On these systems, references to `ranlib` should be commented out or removed from the makefiles in LAPACK/SRC, LAPACK/BLAS/SRC, LAPACK/TESTING/MATGEN, and LAPACK/TIMING/EIG/EIGSRC.

## 4 Installing LAPACK on a Unix System

Installing, testing, and timing the Unix version of LAPACK involves the following steps:

1. Read the tape or untar the file.
2. Test and install the machine-dependent routines.
3. Create the BLAS library, if necessary.
4. Run the Level 2 and 3 BLAS test programs.
5. Create the LAPACK library.
6. Create the library of test matrix generators.
7. Run the LAPACK test programs.
8. Run the LAPACK timing programs.
9. Send the results from steps 7 and 8 to the authors at the University of Tennessee.

use it (but be sure to run the BLAS test programs). If an optimized library of the BLAS is not available, Fortran source code for the Level 1, 2, and 3 BLAS is provided on the tape. Users should not expect too much from the Fortran BLAS; these versions were written to define the basic operations and do not employ the standard tricks for optimizing Fortran code.

The formal definitions of the Level 1, 2, and 3 BLAS are in [9], [7], and [5]. Copies of the BLAS Quick Reference card are available from the authors.

### 3.3 LAPACK Test Routines

This release contains two distinct test programs for LAPACK routines in each data type. One test program tests the routines for solving linear equations and linear least squares problems, and the other tests routines for the matrix eigenvalue problem. The routines for generating test matrices are used by both test programs and are compiled into a library for use by both test programs.

### 3.4 LAPACK Timing Routines

This release also contains two distinct timing programs for the LAPACK routines in each data type. The linear equation timing program gathers performance data in `mgaflops` on the factor, solve, and inverse routines for solving linear systems, the routines to generate or apply an orthogonal matrix given as a sequence of elementary transformations, and the reductions to bidiagonal, tridiagonal, or Hessenberg form for eigenvalue computations. The operation counts used in computing the `mgaflop` rates are computed from a formula; see Appendix C. The eigenvalue timing program is used with the eigensystem routines and returns the execution time, number of floating point operations, and `mgaflop` rate for each of the requested subroutines. In this program the number of operations is computed while the code is executing using special instrumented versions of the LAPACK subroutines.

### 3.5 make files

The libraries and test programs are created using the `makefile` in each directory. Target names are supplied for each of the four data types and are called `single`, `double`, `complex`, and `complex16`. To create a library from one of the files called `makefile`, you simply type `make` followed by the data types desired. Here are some examples:

```
make single
make double complex16
make single double complex complex16
```

Alternatively,

```
make
```

## 3 Overview of Tape Contents

Most routines in LAPACK occur in four versions: REAL, DOUBLE PRECISION, COMPLEX, and COMPLEX\*16. The first three versions (REAL, DOUBLE PRECISION, and COMPLEX) are written in standard Fortran 77 and are completely portable; the COMPLEX\*16 version is provided for those compilers which allow this data type. For convenience, we often refer to routines by their single precision names; the leading 'S' can be replaced by a 'D' for double precision, a 'C' for complex, or a 'Z' for complex\*16. For LAPACK use and testing you must decide which version(s) of the package you intend to install at your site (for example, REAL and COMPLEX on a Cray computer or DOUBLE PRECISION and COMPLEX\*16 on an IBM computer).

### 3.1 LAPACK Routines

There are three classes of LAPACK routines:

- **driver** routines solve a complete problem such as solving a system of linear equations or computing the eigenvalues of a real symmetric matrix. Users are encouraged to use a driver routine if there is one that meets their requirements. The driver routines are listed in Appendix A.
- **computational** routines, also called simply LAPACK routines, perform a distinct computational task, such as computing the  $LU$  decomposition of an  $m \times n$  matrix or finding the eigenvalues and eigenvectors of a symmetric tridiagonal matrix using the  $QR$  algorithm. The LAPACK routines are listed in Appendix A; see also LAPACK Working Note #5 [3].
- **auxiliary** routines are all the other subroutines called by the driver routines and computational routines. Among them are subroutines to perform subtasks of block algorithms, in particular, the unblocked versions of the block algorithms; extensions to the BLAS, such as matrix-vector operations involving complex symmetric matrices; the special routines ISAME and XERBLA which first appeared with the BLAS; and a number of routines to perform common low level computations, such as computing a matrix norm, generating an elementary Householder transformation, and applying a sequence of plane rotations. Many of the auxiliary routines may be of use to numerical analysts or software developers, so we have documented the Fortran source for these routines with the same level of detail used for the LAPACK routines and driver routines. The auxiliary routines are listed in Appendix B.

### 3.2 Level 1, 2, and 3 BLAS

The BLAS are a set of Basic Linear Algebra Subprograms that perform vector-vector, matrix-vector, and matrix-matrix operations. LAPACK is designed around the Level 1, 2, and 3 BLAS, and nearly all of the parallelism in the LAPACK routines is contained in the BLAS. Therefore, the key to getting good performance from LAPACK lies in having an efficient version of the BLAS optimized for your particular machine. If you have access to a library containing optimized versions of some or all of the BLAS, you should certainly

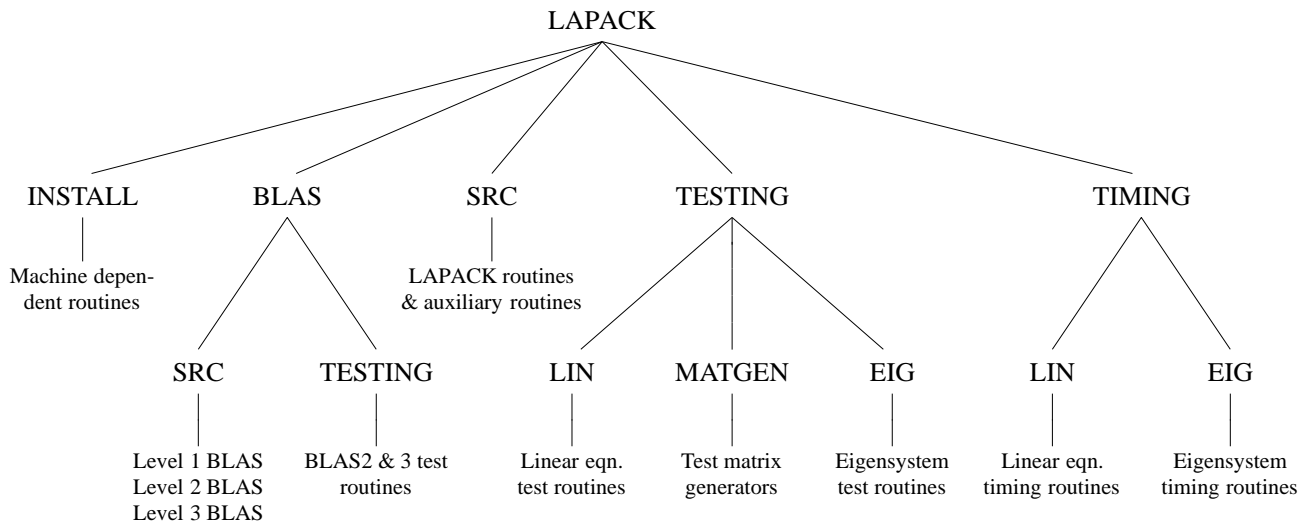


Figure 1: Unix organization of LAPACK

There have also been a number of revisions to correct bugs, improve efficiency, simplify calling sequences, and improve the appearance of output. You should destroy any previous release versions of LAPACK

## 2 File Format

The software for LAPACK is distributed in the form of a tape or tar file which contains the Fortran source for LAPACK, the Basic Linear Algebra Subprograms (the Level 1, 2, and 3 BLAS) needed by LAPACK, the testing programs, and the timing programs. This section describes the organization of the software for users who have received a Unix tar tape or a tar file via the file-transfer program ftp. Users who have an ASCII or EBCDIC tape should go to appendix F, although the overview in section 3 applies to both the Unix and non-Unix versions.

The software on a tar tape or in a tar file is organized in a number of directories as shown in Figure 1. Each of the lowest level directories in the tree structure contains a makefile to create a library or a set of executable programs for testing and timing. Libraries are created in the LAPACK directory and executable files are created in one of the directories BLAS, TESTING or TIMING. Input files for the test and timing programs are also found in these three directories so that testing may be carried out in the directories LAPACK/BLAS, LAPACK/TESTING and LAPACK/TIMING

This guide combines the instructions for the Unix and non-Unix versions of the LAPACK test package (the non-Unix version is in Appendix F).

Section 2 describes how the files are organized on the tape, and Section 3 gives a general overview of the parts of the test package. Step-by-step instructions appear in Section 4 for the Unix version and in the appendix for the non-Unix version.

For users desiring additional information, Sections 5 and 6 give details of the test and timing programs and their input files. Appendices A and B briefly describe the LAPACK routines and auxiliary routines provided in this release. Appendix C lists the operation counts we have computed for the BLAS and for some of the LAPACK routines. Appendix D entitled "Caveats", is a compendium of the known problems from our own experiences, with suggestions on how to overcome them. Appendix E contains the execution times of the different test and timing runs on two sample machines. Appendix F contains the instructions to install LAPACK on a non-Unix system.

Release 3 of LAPACK includes updates of all of the software from Release 2, with the following additions:

- Driver routines for solving systems of linear equations, for solving least squares problems, for computing some or all eigenvalues and/or eigenvectors of a matrix, and for computing the SVD; test code for all the driver routines is also included with this package.
- New functionality for QR, which now includes the QR, RQ, LQ, and QL factorizations and also QR with pivoting; the matrix to be factored may be  $m$  by  $n$  with no restrictions on the relative sizes of  $m$  and  $n$ .
- New functionality for the reduction to specialized forms for eigenvalue computations, including a block algorithm for reduction to bidiagonal form, provision for upper or lower triangular storage of a symmetric matrix in the reduction to tridiagonal form and provision for packed storage of a symmetric matrix.
- A complete set of subroutines to generate an orthogonal matrix from a sequence of elementary transformations or to multiply a matrix  $C$  by an orthogonal matrix given as a sequence of elementary transformations, using all the possible storage schemes for the orthogonal matrix  $Q$  from the orthogonal factorization and reduction routines.
- Additional techniques for computing eigenvalues, including bisection and inverse iteration for the symmetric eigenvalue problem, a block multi-shift QZ algorithm for the generalized nonsymmetric eigenvalue problem, and a subroutine to find the eigenvalues and eigenvectors of a symmetric positive definite tridiagonal matrix by performing a Cholesky factorization followed by a high-accuracy method for finding the eigenvalues of the bidiagonal factor.
- Software for the generalized symmetric eigenvalue problem and the generalized nonsymmetric eigenvalue problem.
- Subroutines for solving triangular and tridiagonal linear systems, and for computing and applying the scaling factors to equilibrate a matrix.

# LAPACK Working Note 35

## Implementation Guide for LAPACK \*

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August 9, 1991

### Abstract

This working note describes how to install, test, and time the third and final test release of LAPACK, a linear algebra package for high-performance computers. Separate instructions are provided for the Unix and non-Unix versions of the test package. Further details are also given on the design of the test and timing programs.

## 1 Introduction

LAPACK is planned to be a linear algebra library for high-performance computers. The library will include Fortran 77 subroutines for the analysis and solution of systems of simultaneous linear algebraic equations, linear least-squares problems, and matrix eigenvalue problems. Our approach to achieving high efficiency is based on the use of a standard set of Basic Linear Algebra Subprograms (the BLAS), which can be optimized for each computing environment. By confining most of the computational work to the BLAS, the subroutines should be transportable and efficient across a wide range of computers.

This working note describes how to install, test, and time the third and final test release of LAPACK. This release is being made available only to our test sites and is intended only for testing, and not for general distribution. After we receive the results from our test sites and make any necessary corrections, we will make the LAPACK routines available to the public. We do not expect any major changes to the software in this release before the public release, but this software should still be regarded as a preliminary version.

The instructions for installing, testing, and timing are designed for a person whose responsibility is the maintenance of a mathematical software library. We assume the installer has experience in compiling and running Fortran programs and in creating object libraries. The installation process involves reading the tape, creating a set of libraries, and compiling and running the test and timing programs.

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\*This work was supported by NSF Grant No. ASC-8715728.

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