

# Improving the efficiency and accuracy of the MATLAB Control Toolbox using SLICOT-based gateways

Vasile Sima

Research Institute for Informatics  
Bd. Maresal Al. Averescu, Nr. 8–10  
71316 Bucharest 1, Romania  
vsima@u3.ici.ro

Sabine Van Huffel

Department of Electrical Engineering  
Katholieke Universiteit Leuven  
Kardinaal Mercierlaan 94  
B–3001 Leuven–Heverlee, Belgium  
Sabine.VanHuffel@esat.kuleuven.ac.be

Peter Benner

Zentrum für Technomathematik  
Fachbereich 3–Mathematik und Informatik  
Universität Bremen  
D–28334 Bremen, Germany  
benner@math.uni-bremen.de

Andras Varga

German Aerospace Center  
Institute of Robotics and System Dynamics  
DLR Oberpfaffenhofen  
D-82234 Wessling, Germany  
Andreas.Varga@dlr.de

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## Abstract

The paper presents performance results for some components of the new, public-domain version of the SLICOT library (Subroutine Library in Control Theory), in comparison with equivalent computations performed by some MATLAB functions included in the Control Toolbox. SLICOT incorporates the new algorithmic developments in numerical linear algebra, implemented in the state-of-the-art software packages LAPACK and BLAS. The results show that, at comparable or better accuracy, SLICOT routines are several times faster than MATLAB computations.

## 1 Introduction

MATLAB<sup>1</sup> [7] is an excellent tool for developing and testing new algorithmic ideas or new control analysis and synthesis methods. However, the practical experience has shown a sometimes poor performance of MATLAB in a dedicated production-quality *computer-aided control system design environment* (CACSD). To achieve the robustness and efficiency needed for solving possibly ill-conditioned or large-scale real-life control problems, a new public-domain version of the SLICOT library (Subroutine Library in Control Theory) has been recently developed. SLICOT incorporates the new algorithmic de-

velopments in numerical linear algebra, implemented in the state-of-the-art software packages LAPACK [1] and BLAS [4], and can thus exploit the potential of modern high-performance computer architectures. The conversion of the SLICOT library to a freely available software package offered the opportunity to improve the modularity, functionality, reliability, as well as the performance of the codes, by using calls to Level 3 BLAS and LAPACK block algorithms whenever possible, exploiting any special problem structure, etc.

The paper presents performance results (efficiency and accuracy) for some components of the new SLICOT release<sup>2</sup>, in comparison with equivalent computations performed by some MATLAB functions included in the Control Toolbox [6]. The calculations have been performed on a SUN Ultra 2 Creator 2200 workstation under SunOS 5.5, by calling from MATLAB the SLICOT gateways produced by the NAGWare Gateway Generator [8] of the Numerical Algorithms Group (NAG). The results show that, at comparable or better accuracy, SLICOT routines are several times faster than MATLAB computations; moreover, less memory is required by SLICOT routines, because the problem structure is fully exploited. Additional performance results and comparisons can be found in [3, 10]. Reference [3] also includes a brief history and perspective of system and control software.

## 2 Structure preserving algorithms

Among the various reasons for the development of SLICOT library [3], a special emphasis has been put

<sup>1</sup>MATLAB is a registered trademark of The MathWorks, Inc.

<sup>2</sup>see <http://www.win.tue.nl/wgs/slicot.html> and use the link to the FTP site.

on *structure-preserving algorithms*. The main advantage of such algorithms is that the structural properties of a problem are preserved during finite precision computations. This allows the computed result to be interpreted as the exact solution of the original problem with perturbed input data, which may otherwise not be the case. This not only increases the reliability of the returned results, but often also improves their accuracy, as shown below for an important computational problem.

The *controllability* and *observability Gramians*  $P_c$ ,  $P_o$ , respectively, of a stable state-space realization  $(A, B, C)$  of a continuous-time linear time-invariant system, where  $A$  is  $n \times n$ ,  $B$  is  $n \times m$ , and  $C$  is  $p \times n$ , are given by the solutions of the stable *Lyapunov equations*

$$AP_c + P_cA^T = -BB^T, \quad (1)$$

$$A^TP_o + P_oA = -C^TC. \quad (2)$$

By Lyapunov stability theory,  $P_c$  and  $P_o$  are positive semidefinite. The nonnegative square roots of the eigenvalues of the product  $P_cP_o$ , called the *Hankel singular values*, play a fundamental role in finding balanced realizations and in model reduction. To guarantee the symmetry and semidefiniteness of the computed Gramians, the special problem structure should be taken into account. Besides this, exploiting the structure usually results in a reduction of necessary computational operations and memory requirements. No MATLAB function specialized for equations like (1) or (2) is available in the standard Control Toolbox. The new SLICOT release includes the routine `SB030D`, which solves for  $X = \text{op}(U)^T \text{op}(U)$  either continuous-time or discrete-time Lyapunov equations,

$$\text{op}(A)^TX + X\text{op}(A) = -\sigma^2 \text{op}(M)^T \text{op}(M), \quad (3)$$

$$\text{op}(A)^TX\text{op}(A) - X = -\sigma^2 \text{op}(M)^T \text{op}(M), \quad (4)$$

where  $\text{op}(K) = K$  or  $K^T$ ,  $A$  is square and stable (in the continuous- or discrete-time sense, respectively),  $M$  is rectangular,  $U$  is upper triangular, and  $\sigma$  is a scale factor, set less than or equal to 1 to avoid overflow in computing  $X$ .<sup>3</sup> The routine uses the same real Schur form of  $A$  for all different computations, and optionally  $A$  can be given in such a form as input.

Results for two sets of experiments will be summarized. The data for the first set have been generated as

$$\begin{aligned} A &= \text{rand}(n,n) - n \cdot \text{eye}(n); \\ B &= \text{rand}(n,m); \quad C = \text{rand}(1,n); \end{aligned}$$

and the actual solutions were unknown.

Table 1 gives comparative results using the gateway for the SLICOT routine `SB030D` and MATLAB function `lyap`, for solving the two Lyapunov equations (1)

<sup>3</sup>Note that the chosen functionality allows to easily implement LAPACK-style condition estimators for such equations, based on `SB030D`.

Table 1: Comparison between `SB030D` and MATLAB results ( $m = p = n/2$ ).

$n$	Time		Relative residuals in $X$	
	<code>SB030D</code>	MATLAB	<code>SB030D</code>	MATLAB
16	0.01	0.06	3.17e-14	2.97e-14
32	0.04	0.24	8.41e-14	7.23e-14
64	0.24	1.30	1.48e-13	1.82e-13
128	1.68	9.44	4.13e-13	4.21e-13

and (2).<sup>4</sup> It is apparent that the speed-up compared with MATLAB is about 5. Besides improved efficiency, the accuracy of the SLICOT routine is sometimes better. More important, the Hankel singular values can be obtained as the singular values of the triangular matrix  $VU$ , where  $U$  and  $V$  are the Cholesky factors computed by `SB030D` for the solutions of the equations (3), for  $\text{op}(K) = K^T$  and  $\text{op}(K) = K$ , respectively. On the contrary, MATLAB calculations need to be based on `sqrt(eig(X*Y))`, where  $X = P_c$  and  $Y = P_o$  are the solutions of the Lyapunov equations (1) and (2), respectively. Similar results have been obtained for the relative residuals in  $Y$ .

For one sample of the largest example, we find that two eigenvalues of the matrix  $XY$  were negative (close to the machine accuracy), and four were complex conjugate. In other words, in extreme cases, MATLAB could give physically meaningless negative or complex Hankel “singular values”. This, of course, does not happen for well-conditioned problems, considered in the next experiment.

In the second experiment, the matrix  $A$  has been defined as above, but  $X$  and  $Y$  have been constructed as positive definite, with given eigenvalues (different for  $X$  and  $Y$ ), generated by `rand(n,1) + 1`. Then,  $B$  and  $C$  have been computed by the MATLAB sequence

$$\begin{aligned} \text{BBT} &= -(A*X + X*A'); \quad \text{BBT} = (\text{BBT} + \text{BBT}')/2; \\ \text{CTC} &= -(A'*Y + Y*A); \quad \text{CTC} = (\text{CTC} + \text{CTC}')/2; \\ B &= \text{chol}(\text{BBT})'; \quad C = \text{chol}(\text{CTC}); \end{aligned}$$

In this way, relative errors of the solutions could be obtained. The relative errors have been of order  $10^{-15}$ , slightly better for `SB030D` than for MATLAB. The timings and relative residuals have been similar with those presented above, and there has been a good agreement between the Hankel singular values.

### 3 Performance Results

Tables 2 and 3 give performance results for SLICOT’s general Lyapunov solver `SB03MD`, which solves both continuous- and discrete-time equations, and MATLAB

<sup>4</sup>The relative residuals or errors have been computed using the Frobenius norm.

Table 2: Comparison between SB03MD and MATLAB results (continuous-time case, relative residuals in  $X$ ).

$n$	Time		Relative residuals in $X$	
	SB03MD	MATLAB	SB03MD	MATLAB
16	0.01	0.05	3.69e-15	3.59e-15
32	0.03	0.12	3.54e-15	8.96e-15
64	0.21	0.70	3.56e-15	1.96e-14
128	1.50	5.38	1.73e-14	2.86e-14

Table 3: Comparison between SB03MD and MATLAB results (discrete-time case, relative errors in  $X$ ).

$n$	Time		Relative errors in $X$	
	SB03MD	MATLAB	SB03MD	MATLAB
16	0.01	0.03	4.01e-13	2.22e-13
32	0.03	0.11	1.63e-13	1.14e-12
64	0.22	0.70	3.48e-12	9.04e-12
128	1.73	5.49	1.51e-11	5.82e-11

functions `lyap` and `dlyap`. The SLICOT routine can solve equations defined by (3) and (4), but with the right-hand side replaced by  $\sigma C$ . Moreover, an estimate of the condition number is returned. Specifically, Table 2 reports results for continuous-time problems with  $\text{op}(A) = A$ , generated using the MATLAB statements

$$A = \text{rand}(n,n); \quad C = \text{rand}(n,n); \quad C = C + C';$$

so the solutions were not known. Therefore, only relative residuals are given in Table 2. The relative errors between the SLICOT and MATLAB computed solutions have been of the order of  $10^{-13}$  or less. Similar results have been obtained for the transposed case, or for discrete-time equations. SLICOT residuals were usually less than MATLAB residuals in our tests. Note also that the speed-up has been even larger when compared with the previous MATLAB versions of `lyap` and `dlyap`.

Table 3 reports results for discrete-time problems with  $\text{op}(A) = A^T$ . The problems were generated using the MATLAB statements

$$\begin{aligned} A &= \text{rand}(n,n); \\ X0 &= \text{rand}(n,n) + n \cdot \text{eye}(n); \quad X0 = X0 + X0'; \\ C &= A \cdot X0 \cdot A' - X0; \quad C = (C + C')/2; \end{aligned}$$

so the relative errors could be computed. The relative errors between the SLICOT and MATLAB computed solutions have been of the order of  $10^{-11}$  or less.

Table 4 gives performance results for the SLICOT Riccati solver SB020D, which solves both continuous- and discrete-time equations, and the MATLAB functions `care` and `dare`, all based on the generalized Schur vector approaches [2]. The SLICOT solver can also compute the

Table 4: Comparison between SB020D and MATLAB results ( $m = n$ , relative residuals in  $X$ ).

$n$	Time		Relative residuals in $X$	
	SB020D	MATLAB	SB020D	MATLAB
16	0.07	0.37	1.83e-14	3.81e-14
32	0.30	0.86	4.99e-14	1.13e-13
64	2.80	4.01	7.21e-12	2.10e-11
128	24.00	79.96	2.44e-11	6.72e-11

Table 5: Comparison between SB02MD and MATLAB results ( $m = n/2$ , relative residuals in  $X$ ).

$n$	Time		Relative residuals in $X$	
	SB02MD	MATLAB	SB02MD	MATLAB
16	0.03	0.22	3.03e-14	8.90e-14
32	0.14	1.48	1.67e-13	7.31e-13
64	1.02	3.89	3.10e-13	3.85e-13
128	7.78	35.86	4.66e-11	9.97e-11

anti-stabilizing solutions. The problems solved were generated using the MATLAB statements

$$\begin{aligned} A &= \text{rand}(n,n); \quad B = \text{rand}(n,m); \\ G &= B \cdot \text{inv}(R) \cdot B'; \quad G = (G + G')/2; \end{aligned}$$

(with  $m = n$ ), and the weighting matrices  $R$  and  $Q$  have been computed to have given random (non-negative) eigenvalues, using a gateway for the test routine DLATMS from LAPACK. Table 4 reports the timings and relative residuals for computing stabilizing solutions for continuous-time problems. The reciprocal condition number of the linear system which is solved for obtaining the stabilizing Riccati solution  $X$  was of the order  $10^{-2}$  or larger. The relative errors between the SLICOT and MATLAB computed solutions have been of the order of  $10^{-12}$  or less.

Table 5 gives performance results for the SLICOT Riccati solver SB02MD, which solves both continuous- and discrete-time equations, using the Schur vectors approach [5]. For discrete-time problems, SB02MD includes an option to use the inverse of the symplectic  $2n \times 2n$  matrix [9], which is always faster than the standard approach. The continuous-time case, with  $m = n/2$ , is illustrated in Table 5. The reciprocal condition number of the linear system which is solved for obtaining  $X$  was of the order  $10^{-3}$  or larger. The relative errors between the SLICOT and MATLAB computed solutions have been of the order of  $10^{-11}$  or less. Discrete-time problems can be solved even faster by SB02MD, but the accuracy depends on the conditioning of the matrix  $A$ .

Finally, some timing results and relative residuals for several other SLICOT gateways, in comparison with

Table 6: Comparison of some SLICOT gateways and MATLAB functions (timings).

Routine/function	$n$	$m$	Time	
			SLICOT	MATLAB
AB01ND/ctrbf	128	16	0.32	1.44
AB01ND/ctrbf	256	1	2.78	362.09
SB01MD/place	128	1	0.55	104.00
SB04MD/lyap2	128	64	0.88	3.85

Table 7: Comparison of some SLICOT gateways and MATLAB functions (accuracy).

Routine/function	$n$	$m$	Relative residuals	
			SLICOT	MATLAB
AB01ND/ctrbf	128	16	6.46e-16	1.57e-15
AB01ND/ctrbf	256	1	1.39e-15	5.98e-15
SB01MD/place	128	1	1.32e-14	4.10e-14
SB04MD/lyap2	128	64	4.86e-13	7.38e-13

MATLAB calculations, are reported in Tables 6 and 7. It is apparent that the MATLAB function `ctrbf` is rather slow, especially for single-input systems. Additional results are given in Table 8.

## 4 Conclusions

Performance results for some basic components of the SLICOT library, in comparison with similar computations performed by some MATLAB functions included in the Control Toolbox have been presented. Using the NAGWare Gateway Generator, it is possible to embed SLICOT routines into MATLAB. The MATLAB functions produced in this way often outperform the available functions from the MATLAB toolboxes or even built-in functions. The library is in the public-domain and its development is continuing under the framework of the

Table 8: Comparison between AB01ND and MATLAB results ( $m = 1$ ).

$n$	Time		Relative residuals in $X$	
	AB01ND	MATLAB	AB01ND	MATLAB
16	0.00	0.03	4.94e-16	1.04e-15
32	0.00	0.17	7.01e-16	9.30e-16
64	0.04	1.29	5.83e-16	1.23e-15
128	0.32	19.12	7.41e-16	3.58e-15

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