

POLYNOMIAL COMPUTATION OF HANKEL SINGULAR VALUES

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Abstract

The paper presents a revised and improved version of a polynomial algorithm published by N. J. Young in 1990 for the computation of the singular values and vectors of the Hankel operator defined by a linear time-invariant system with rational transfer matrix.

Introduction

We describe an algorithm for the computation of the Hankel singular values and vectors by polynomial methods. It is a much streamlined version of that of Young [1]. The computation of Hankel singular values arises in determining lower-order approximations of linear multi-variable systems [2], and in \mathcal{H}_∞ optimization [3]. We consider continuous-time systems rather than discrete-time systems as Young [1] does. The results can easily be adapted to the discrete-time case.

Let G be the $n \times m$ strictly proper transfer matrix of a strictly stable continuous-time system, with left and right coprime polynomial matrix fraction representations

$$G = D^{-1}N = \bar{N}\bar{D}^{-1}, \quad (1)$$

with D row reduced and \bar{D} column reduced. The Hankel operator G_H defined by G maps any finite-energy input u such that $u(t) = 0$ for $t \geq 0$ to the finite-energy signal y_+ that is obtained from the corresponding output y as

$$y_+(t) = \begin{cases} 0 & \text{for } t < 0, \\ y(t) & \text{for } t \geq 0. \end{cases} \quad (2)$$

The Hankel singular values of G are the singular values of the Hankel operator G_H .

Good references for the polynomial methods used in this paper are the books by Kučera [4], Kailath [5], and Callier and Desoer [6].

Representation of the Hankel Operator in Expedient Bases

Following Young [1], we successively consider the *kernel* (or nullspace,) the *cokernel* and the *range* of the Hankel operator G_H , and find suitable bases for the latter two spaces.

The kernel of the Hankel operator consists of all inputs in the domain of G_H that are mapped to the zero signal. The cokernel of G_H consists of all signals u in the domain of G_H that are *orthogonal* to the kernel. It is not difficult to argue that the cokernel of G_H is a *finite-dimensional* signal space, whose basis may conveniently be chosen as the signals whose (two-sided) Laplace transform is given by

$$\bar{D}^\sim(s)^{-1}e_i s^k, \quad (3)$$

$k = 0, 1, \dots, c_i - 1, i = 1, 2, \dots, m$, where c_i is the degree of the i th column of \bar{D} , and e_i the i th m -dimensional unit vector. The tilde denotes the adjoint. Juxtapose these basis vectors to form the rational matrix

$$\bar{D}^\sim(s)^{-1}[e_1 \ e_1 s \ \dots \ e_1 s^{c_1-1} \ e_2 \ \dots \ e_2 s^{c_2-1} \\ e_3 \ \dots \ e_m s^{c_m-1}] = \bar{D}^\sim(s)^{-1}V(s), \quad (4)$$

with V a polynomial matrix. Then our basis for the cokernel of G_H consists of those signals whose Laplace transforms are the columns of

$\mathcal{D} = (\bar{D}^\sim)^{-1}V$. Young [1] calls this an *expedient basis*. The Laplace transform of any element u in the cokernel of G_H may be represented as $\hat{u} = \mathcal{D}v$, with v a suitable constant finite-dimensional vector.

Given any such input u in the cokernel of G_H , the corresponding output y has Laplace transform

$$\hat{y} = G\hat{u} = D^{-1}N(\bar{D}^\sim)^{-1}Vv. \quad (5)$$

We determine the corresponding truncated output y_+ by partial fraction expansion. Define the right coprime polynomial matrices \tilde{V} and \tilde{D} by the left-to-right conversion

$$(\bar{D}^\sim)^{-1}V = \tilde{V}\tilde{D}^{-1}, \quad (6)$$

and perform the partial fraction expansion

$$D^{-1}N\tilde{V}\tilde{D}^{-1} = D^{-1}A + B\tilde{D}^{-1}. \quad (7)$$

The polynomial coefficient matrices A and B may be obtained by solving the bilateral linear polynomial matrix equation

$$N\tilde{V} = A\tilde{D} + DB. \quad (8)$$

It follows that

$$\hat{y} = D^{-1}Av + B\tilde{D}^{-1}v. \quad (9)$$

Since D^{-1} has all its poles in the left-half plane and \tilde{D}^{-1} has all its poles in the right-half plane, we immediately conclude that the Laplace transform \hat{y}_+ of y_+ is

$$\hat{y}_+ = D^{-1}Av. \quad (10)$$

Hence, a convenient basis for the *range* of the Hankel operator consists of the signals whose Laplace transforms are the columns of the rational matrix $\mathcal{R} = D^{-1}A$. Moreover, in terms of the bases we selected for the cokernel and range, the Hankel operator simply is represented by the *unit matrix*.

Our approach differs from that of Young [1] in the choice of the basis for the range of the Hankel operator.

Computation of the Hankel Singular Values and Vectors

Let Γ_d be the Gramian of the basis selected for the cokernel (consisting of the columns of \mathcal{D}), and Γ_r the Gramian of the basis for the range (consisting of the columns of \mathcal{R} .) The computation of these Gramians is detailed later. Again following Young [1], we factor the Gramians as

$$\Gamma_d = T_d^T T_d, \quad \Gamma_r = T_r^T T_r. \quad (11)$$

Then the singular values of the Hankel operator are the singular values of the matrix $T_r T_d^{-1}$.

Summary (Computation of Hankel singular values.) The singular values of the Hankel operator G_H defined by the system with the strictly stable and strictly proper $n \times m$ transfer matrix $G = D^{-1}N = \bar{N}\bar{D}^{-1}$, with D row reduced and \bar{D} column reduced, may be found as follows.

1. Form the polynomial matrix V by juxtaposing the columns $e_i s^k$, $k = 0, 1, \dots, c_i - 1$, $i = 1, 2, \dots, m$, with e_i the i th m -dimensional unit vector and c_i the degree of the i th column of \bar{D} . Compute the right coprime polynomial matrices \tilde{V} and \tilde{D} , with \tilde{D} column reduced, by the left-to-right conversion

$$(\bar{D}^\sim)^{-1}V = \tilde{V}\tilde{D}^{-1}. \quad (12)$$

2. Solve the bilateral linear polynomial equation

$$N\tilde{V} = A\tilde{D} + DB, \quad (13)$$

for polynomial matrices A and B such that $D^{-1}A$ is strictly proper.

3. Compute the Gramian Γ_d of the signals whose Laplace transforms are the columns of $\mathcal{D} = \tilde{V}\tilde{D}^{-1}$. Also compute the Gramian Γ_r of the signals whose Laplace transforms are the columns of $\mathcal{R} = D^{-1}A$.

4. Factor the Gramians as

$$\Gamma_d = T_d^T T_d, \quad \Gamma_r = T_r^T T_r, \quad (14)$$

and compute the singular value decomposition

$$T_r T_d^{-1} = U \Sigma W^H. \quad (15)$$

5. The singular values of the Hankel operator are the diagonal entries of Σ . Its right and left singular vectors are the columns of

$$\mathcal{D}T_d^{-1}W \quad \text{and} \quad \mathcal{R}T_c^{-1}U, \quad (16)$$

respectively. •

Computation of the Gramians

It remains to calculate the Gramians. Here we again offer an alternative to Young's proposal [1]. Consider computing the Gramian Γ of a finite number of time signals $x_i, i = 1, 2, \dots, p$, whose Laplace transforms $\hat{x}_i, i = 1, 2, \dots, p$, are the columns of the strictly stable and strictly proper rational matrix

$$X = QR^{-1}, \quad (17)$$

with Q and R right coprime such that R is column reduced. The (i, k) element Γ_{ik} of Γ is the inner product

$$\begin{aligned} \Gamma_{ik} &= \int_{-\infty}^{\infty} x_i^H(t)x_k(t) dt \\ &= \int_{-\infty}^{\infty} \hat{x}_i^H(j2\pi f)\hat{x}_k(j2\pi f) df. \end{aligned} \quad (18)$$

It follows that

$$\begin{aligned} \Gamma &= \int_{-\infty}^{\infty} X^{\sim}(j2\pi f)X(j2\pi f) df \\ &= \int_{-\infty}^{\infty} (R^{\sim})^{-1}Q^{\sim}QR^{-1} df, \end{aligned} \quad (19)$$

where for simplicity we suppress the argument $j2\pi f$. Perform the partial fraction expansion

$$(R^{\sim})^{-1}Q^{\sim}QR^{-1} = (R^{\sim})^{-1}C^{\sim} + CR^{-1}, \quad (20)$$

where the polynomial coefficient matrix C is a solution of the symmetric bilateral polynomial matrix equation

$$Q^{\sim}Q = R^{\sim}C + C^{\sim}R \quad (21)$$

such that CR^{-1} is strictly proper. It follows that

$$\Gamma = \int_{-\infty}^{\infty} [(R^{\sim})^{-1}C^{\sim} + CR^{-1}] df. \quad (22)$$

Since by assumption R is strictly Hurwitz, CR^{-1} is the (two-sided) Laplace transform of a matrix function γ such that $\gamma(t) = 0$ for $t < 0$, and

$$\Gamma = \lim_{t \downarrow 0} \gamma(t). \quad (23)$$

By the initial value theorem,

$$\lim_{t \downarrow 0} \gamma(t) = \lim_{|s| \rightarrow \infty} sC(s)R^{-1}(s). \quad (24)$$

Define R_l as the leading coefficient matrix of the column reduced matrix R , and C_l as the associated leading coefficient matrix¹ of C . Then,

$$\Gamma = C_l R_l^{-1}. \quad (25)$$

If R is strictly anti-Hurwitz the Gramian of the functions whose Laplace transforms are the columns of QR^{-1} may be found as the Gramian of the functions whose Laplace transforms are the columns of X defined by $X(s) = Q(-s)R^{-1}(-s)$.

Summary (Computation of the Gramian.)

Consider the signals whose Laplace transforms are the columns of the strictly stable and strictly proper rational matrix QR^{-1} with the polynomial matrix Q and R right coprime such that R is column reduced. The Gramian Γ of these signals may be computed as follows.

1. Solve the symmetric bilateral linear polynomial matrix equation

$$Q^{\sim}Q = R^{\sim}C + C^{\sim}R \quad (26)$$

for the square polynomial matrix C such that CR^{-1} is strictly proper.

2. Let R_l be the leading coefficient matrix of the column reduced polynomial matrix R , and C_l the associated leading coefficient matrix of C . Then the Gramian Γ is

$$\Gamma = C_l R_l^{-1}. \quad (27)$$

¹Let the column degrees of R be given by c_1, c_2, \dots, c_p . Then the k th column of C_l is the k th column of the coefficient matrix of C corresponding to s^{c_k-1} .

Implementation

Implementation of the algorithm under MATLAB first of all requires a way of representing polynomial matrices, and basic routines for elementary polynomial matrix manipulations such as addition and multiplication. More advanced routines are needed for left-to-right and right-to-left conversion of polynomial matrix fractions, transformation to row- and column-reduced form, and the solution of symmetric and asymmetric bilateral linear polynomial matrix equations. Several of these routines were already available [7] and others have been developed (in particular, for the solution of bilateral equations.)

Left-to-right and right-to-left conversion may be done by Kailath's algorithms ([5], Section 6.7.) Callier [8] describes how to transform to row- and column-reduced form. A provisional and rather inefficient way of solving bilateral equations is to expand them in terms of linear coefficient equations.

Example

We consider Example 6.1 of Glover [2], where

$$\begin{aligned}
 G(s) &= \begin{bmatrix} \frac{.45 + 2s}{.09 + 1.25s + s^2} & 0 \\ 0 & \frac{1}{.5 + s} \end{bmatrix} \\
 &= \underbrace{\begin{bmatrix} .09 + 1.25s + s^2 & 0 \\ 0 & .5 + s \end{bmatrix}}_{D(s)}^{-1} \underbrace{\begin{bmatrix} .45 + 2s & 0 \\ 0 & 1 \end{bmatrix}}_{N(s)} \\
 &= \underbrace{\begin{bmatrix} 0 & .45 + 2s \\ 1 & 0 \end{bmatrix}}_{\bar{N}(s)} \underbrace{\begin{bmatrix} 0 & .09 + 1.25s + s^2 \\ .5 + s & 0 \end{bmatrix}}_{\bar{D}(s)}^{-1}.
 \end{aligned} \tag{28}$$

Since the column degrees of \bar{D} are $c_1 = 1$ and $c_2 = 2$ we have

$$V(s) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & s \end{bmatrix}. \tag{29}$$

By the left-to-right conversion (12) we obtain

$$\tilde{V}(s) = \begin{bmatrix} 0 & 0 & 1 \\ -1 & 0 & 0 \end{bmatrix}, \tag{30}$$

$$\tilde{D}(s) = \begin{bmatrix} -.5 + s & 0 & 0 \\ 0 & s & .09 \\ 0 & -1 & -1.25 + s \end{bmatrix}. \tag{31}$$

Solution of the bilateral polynomial matrix equation (13) results in

$$A(s) = \begin{bmatrix} 0 & 1.7 + 2s & -.18 - .8s \\ 1 & 0 & 0 \end{bmatrix}, \tag{32}$$

$$B(s) = \begin{bmatrix} 0 & -2 & .8 \\ -1 & 0 & 0 \end{bmatrix}. \tag{33}$$

The two Gramians may be found to be

$$\Gamma_c = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4.44444 & 0 \\ 0 & 0 & 0.4 \end{bmatrix}, \tag{34}$$

$$\Gamma_r = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 14.4444 & -2 \\ 0 & -2 & 0.4 \end{bmatrix}, \tag{35}$$

resulting in

$$T_r T_c^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1.8028 & -.8321 \\ 0 & 0 & .5547 \end{bmatrix}. \tag{36}$$

The singular values of this matrix are the correct numbers 2, 1, and 0.5.

Concluding Remarks

The revised algorithm for the polynomial computation of the singular values of the Hankel operator is considerably more polished than Young's original algorithm [1].

Whether it can compete with state space algorithms remains to be seen. Tentative numerical experiments indicate that for high-order systems scaling of the polynomial matrices N and D (so that the constant and leading coefficient matrices are of the same order of magnitude) is mandatory, and that solution of bilateral linear polynomial matrix equations by coefficient expansion is highly inefficient.

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