Computation of reduced-order models of multivariable systems by balanced truncation

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Previous algorithms to obtain reduced-order models by balanced truncation in a single step either require a very specific way to solve a pair of Lyapunov equations or are suitable only for scalar or symmetric MIMO systems. In this paper, model reduction is revisited and an algorithm to obtain a reduced order model in one step only is proposed. As in the previous algorithms, the key point is to construct two rectangular matrices whose smaller dimensions are equal to the number of Hankel singular values to be kept in the lower model. Unlike the one-step algorithms available in the literature, the algorithm proposed here does not make any restriction to the way the Lyapunov equations necessary to obtain the controllability and observability grammians are solved. Furthermore, since the algorithm only relies on singular value decomposition, it is expected to be robust.

1. Introduction

Balanced realization (Moore 1981) has been proved crucial in model reduction (Glover 1984) and also in the computation of $H_\infty$ optimal controllers in the 1984 approach (Doyle 1984). The idea behind its use in model reduction is to measure the degree of controllability and observability of the system modes and then to discard those modes which are weakly controllable or observable. The computation of reduced-order models by balanced truncation for non-minimal order systems was initially carried out in three steps: (1) computation of a minimal realization for the system; (2) construction of a similarity transformation that relates the state-space realization obtained in step (1) to a balanced realization (Moore 1981, Laub et al. 1987 and references therein); and (3) for a given error bound, balanced truncation is deployed to reduce the system order (Glover 1984). This three-step approach has the drawback that a minimal order realization has to be found whose computation is known to be problematic. To avoid such a difficulty, Tombs and Postlethwaite (1987) proposed an algorithm to compute a lower-order model in a single step. Despite its usefulness, this algorithm has the disadvantage that it requires a particular way (Hammarling 1982) to solve the Lyapunov equations necessary to find the controllability and observability grammians of the system. More recently, Aldhaheri (1991) has proposed another single-step algorithm based on the computation of the eigenvectors associated with the largest eigenvalues (in modules) of the cross-gramian $W_{co}$ (Fernando and Nicholson 1982, 1985). The drawbacks of this approach are that the cross-gramian requires the system to be either scalar or symmetric (in the multivariable case), and the realization obtained is not balanced. In common, Tombs and Postlethwaite (1987) and Aldhaheri (1991) have the fact that the reduced-order model is obtained via pre- and post-multiplication of the state matrix by rectangular matrices.

In this paper, an algorithm is proposed to obtain a reduced-order model for a non-minimal state-space realization, whose key point, as in Tombs and Postlethwaite (1987) and Aldhaheri (1991), is the construction of two rectangular matrices whose smaller dimension corresponds to the number of Hankel singular values to be kept in the lower model. Differently from the previous algorithms, it does not make any restriction on the way the Lyapunov equations, necessary to compute the grammians, are solved, and is suitable.
for both scalar and multivariable systems. Furthermore, since the algorithm relies only on singular value decomposition, it is expected to be robust.

This paper is structured as follows. In Section 2, the problem of finding reduced-order models by balanced truncation is reviewed and, in the sequel, the problem of obtaining directly a balanced realization for the reduced-order model of a given non-minimal order realization is formulated. Some preliminary mathematical results are presented in Section 3. The main result is given in Section 4, where a rectangular matrix and its right-inverse are constructed. The balanced reduced-order system will be obtained by appropriate pre- and post-multiplications by these rectangular matrices. The paper results are summarized in Section 5, where an algorithm is presented. In Section 6, the results are illustrated by means of a numerical example. Finally, conclusions are drawn in Section 7.

2. Problem formulation

Assume that a $p \times m$ stable transfer matrix $G(s)$ has the following state-space realization:

$$G(s) = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right],$$

where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{p \times n}$, $D \in \mathbb{R}^{p \times m}$ and $m$, $n$, $p \in \mathbb{N}^*$ with $(A, B)/(C, A)$ are possibly uncontrollable/unobservable. In addition, let $W_c$ and $W_o$ denote, respectively, the controllability and observability gramians of (1), i.e. $W_c$ and $W_o$ are solutions of the following Lyapunov equations:

$$A W_c + W_c A^T = -BB^T$$

$$A^T W_o + W_o A = -C^T C.$$

It is well known that since $G(s)$ is stable, then $W_c$ and $W_o$ are positive semidefinite. It is also known that although the gramians are not invariant under similarity transformation, their product is invariant in the sense that the eigenvalues of the product of the gramians remain the same no matter what state-space realization is being used. In the control literature, the square roots of the non-zero eigenvalues of $W_c W_o$ are usually referred to as the Hankel singular values of $G(s)$.

Consider now the eigenvalue decomposition of $W_c W_o$ (Zhou et al. 1996, p. 77):

$$W_c W_o = W \begin{bmatrix} \Sigma^2 & 0 \\ 0 & 0 \end{bmatrix} W^{-1},$$

where

$$\Sigma^2 = \begin{bmatrix} \Sigma_L^2 \\ \Sigma_s^2 \\ 0 \end{bmatrix},$$

and where $\Sigma_L^2 = \text{diag} \{ \sigma^2_1 I_{m_1}, \sigma^2_2 I_{m_2}, \ldots, \sigma^2_p I_{m_p} \}$ and $\Sigma_s^2 = \text{diag} \{ \sigma^2_{r+1} I_{m_{r+1}}, \sigma^2_{r+2} I_{m_{r+2}}, \ldots, \sigma^2_{r+k} I_{m_{r+k}} \}$, with $\sigma_i > 0$, $i = 1, 2, \ldots, k$ and $\sigma_i > \sigma_j$, $i < j$. Note that $\Sigma_L$ and $\Sigma_s$ are formed, respectively, with the largest and smallest Hankel singular values of $G(s)$, i.e. those which are to be kept and discarded in the model reduction. Suppose that we are interested in obtaining a reduced-order model $\tilde{G}(s)$ for $G(s)$ such that the error between $G(s)$ and $\tilde{G}(s)$ is $\leq 2(\sigma_{r+1} + \sigma_{r+2} + \ldots + \sigma_k)$ in $\mathcal{H}_\infty$ sense, namely:

$$e = \| G - \tilde{G} \|_\infty \leq 2(\sigma_{r+1} + \sigma_{r+2} + \ldots + \sigma_k).$$

(4)

Then, the problem of finding a balanced realization directly from the state-space representation (1) for the reduced order model $\tilde{G}(s)$ of $G(s)$ can be stated as follows: find rectangular matrices $T_L \in \mathbb{R}^{m \times n}$ and $T_r \in \mathbb{R}^{k \times m}$, $r = m_1 + m_2 + \ldots + m_r$, $T_L T_r^\dagger = I_r$ ($I_r$ denoting the identity matrix of order $r$), such that

$$\tilde{G}(s) = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] = \left[ \begin{array}{c|c} T_L A T_r^\dagger & T_r B \\ \hline C T_r^\dagger & D \end{array} \right]$$

(5)

with controllability and observability gramians being given by:

$$\Sigma_L = \text{diag} \{ \sigma_1 I_{m_1}, \sigma_2 I_{m_2}, \ldots, \sigma_i I_{m_i} \}.$$
1. These facts allow us to conclude that $T_1^*T_1$ has the following eigenvalue decomposition:

$$T_1^*T_1 = W \begin{bmatrix} I_k & 0 \\ 0 & 0 \end{bmatrix} W^{-1}$$

for some full rank matrix $W$. Therefore:

$$I_n - T_1^*T_1 = I_n - W \begin{bmatrix} I_k & 0 \\ 0 & 0 \end{bmatrix} W^{-1} = W \begin{bmatrix} 0_k & 0 \\ 0 & I_{n-k} \end{bmatrix} W^{-1},$$

which proves the lemma.

The use of Lemma 1 leads to the following result.

**Lemma 2:** Given $T_1$ and $T_1^*$ satisfying the conditions above, there exist two matrices $T_2 \in \mathbb{R}^{(n-k)\times n}$ and $T_2^* \in \mathbb{R}^{n\times(n-k)}$ such that:

$$\begin{bmatrix} T_1 \\ T_2 \end{bmatrix} \begin{bmatrix} T_1^* \\ T_2^* \end{bmatrix} = \begin{bmatrix} T_1T_1^* & T_1T_2^* \\ T_2T_1^* & T_2T_2^* \end{bmatrix} = I. \quad (8)$$

**Proof:** From (7), we may write:

$$I_n - T_1^*T_1 = [W_1 \ W_2] \begin{bmatrix} 0_k & 0 \\ 0 & I_{n-k} \end{bmatrix} \begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix},$$

where $W_1, V_1 \in \mathbb{R}^{n\times k}, W_2, V_2 \in \mathbb{R}^{(n-k)\times(n-k)}$ and $W_1V_1^T + W_2V_2^T = I_n$. Defining $T_2 = V_2^T$ and $T_2^* = W_2$, we have:

1. $T_1T_1^* = I_k$, by definition;
2. $T_2T_2^* = V_2^TW_2 = I_{n-k};$
3. $(I_n - T_1^*T_1)T_2^* = (I_n - T_1^*T_1)W_2 = W_2 = T_2^*$. Therefore $T_1(I_n - T_1^*T_1)T_2^* = T_1T_2^*$, which implies that $T_1T_1^*T_2^* = 0$ or equivalently $T_1T_2^* = 0$.
4. $T_2(I_n - T_1^*T_1) = V_2^T(I_n - T_1^*T_1) = V_2^T = T_2^*$. Thus, $T_2(I_n - T_1^*T_1)T_1^* = T_2T_1^*$ and proceeding as in (3) we obtain $T_2T_1^* = 0$, which completes the proof.

**Remark 1:** Note from Lemma 2 that given a full row rank rectangular matrix $T_1$ and its right-inverse $T_1^*$, the construction of a square matrix $T_2$, whose first $k$ rows are $T_1$, and its inverse $T_2^{-1}$, whose first $k$ columns are $T_1^*$, is not a matter of adding a bottom matrix $T_2$ whose rows are linearly independent on the rows of $T_1$. This is so because the right-inverse $T_1^*$ is also given and, hence, as stated in the lemma, the rows of $T_2$ must lie in the left null space of $T_1^*$, and the columns of its right-inverse $T_2^*$ must lie in the right null space of $T_1$. This is achieved, as shown in the lemma, by taking, respectively, the eigenvectors and dual-eigenvectors of $I_n - T_1^*T_1$ associated with the unity eigenvalues.

4. **Main results**

In this section, we will initially obtain an expression for two matrices $T_1 \in \mathbb{R}^{n\times n}$ and $T_1^* \in \mathbb{R}^{n\times k}$ ($T_1T_1^* = I_k$), which leads to a minimal realization in balanced form for the non-minimal state-space realization given in (1), i.e.

$$G(s) = \begin{bmatrix} A_h & B_h \\ C_h & D_h \end{bmatrix} = \begin{bmatrix} T_1AT_1^* & T_1B \\ CT_1 & D \end{bmatrix}, \quad (9)$$

where $T_1AT_1^*$ has all the controllable and observable modes of $G(s)$. At this point, it is important to find the relationship between the modes of a given realization and the eigenvalues of $W_c$ and $W_o$, as far as controllability and observability are concerned. This is given by the following results.

**Lemma 3:** Let

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

be a state-space realization of a not-necessarily stable transfer matrix $G(s)$ and assume that there exists a symmetric matrix $P, P = P^*$ solution to the Lyapunov equation

$$AP + PA^* + BB^* = 0,$$

with $P$ non-singular. If we partition $(A, B, C, D)$ compatibly with $P$, i.e.

$$\begin{bmatrix} A_{11} & A_{12} & B_1 \\ A_{21} & A_{22} & B_2 \\ C_{11} & C_{12} & D \end{bmatrix},$$

then

$$\begin{bmatrix} A_{11} & B_1 \\ C_{11} & D \end{bmatrix}$$

is also a realization of $G(s)$. Moreover, $(A_{11}, B_1)$ is controllable if $A_{11}$ is stable.

**Proof:** See Zhou et al. (1996, pp. 72, 73).

**Lemma 4:** Let

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

be a state-space realization of a not-necessarily stabilizable transfer matrix $G(s)$ and assume that there exists a symmetric matrix $Q$,

$$Q = Q^* = \begin{bmatrix} Q_1 & 0 \\ 0 & 0 \end{bmatrix}$$

solution to the Lyapunov equation
\[ QA + A^*Q + C^*C = 0, \]

with \( Q_1 \) non-singular. If we partition \( (A, B, C, D) \) compatibly with \( Q \), i.e.
\[
\begin{bmatrix}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{bmatrix}
\begin{bmatrix}
B_1 \\
B_2
\end{bmatrix}
\begin{bmatrix}
C_1 \\
C_2
\end{bmatrix}
\begin{bmatrix}
D
\end{bmatrix},
\]
then
\[
\begin{bmatrix}
A_{11} & B_1 \\
C_1 & D
\end{bmatrix}
\]
is also a realization of \( G(s) \). Moreover, \( (C_1, A_{11}) \) is observable if \( A_{11} \) is stable.


From Lemmas 3 and 4, it is possible to conclude that the number of non-hidden modes of \( G(s) \) is equal to the number of non-zero eigenvalues of \( W_c W_o \), the product of the controllability and observability gramians associated with the non-minimal realization (1). It is necessary therefore to compute the eigenvalues of \( W_c W_o \), which can be done in a more robust way as follows.

\textbf{Lemma 5:} The Hankel singular values of a stable \( G(s) \) are identical to the non-zero singular values of \( W_o^{1/2} W_c^{1/2} \).

\textbf{Proof:} Let \( \sigma(.) \), \( \lambda(.) \) denote singular values and eigenvalues, respectively. Then:
\[
\sigma(W_o^{1/2} W_c^{1/2}) = \lambda^{1/2}(W_o^{1/2} W_c^{1/2} W_c^{1/2} W_o^{1/2}) = \lambda^{1/2}(W_c W_o).
\]

To complete the proof, note that the Hankel singular values are the square root of the non-zero eigenvalues of the product \( W_c W_o \).

\textbf{Remark 2:} At this point it is important to note that since \( W_c \) and \( W_o \) are symmetric positive semidefinite matrices, their square roots can be simply computed by finding the corresponding eigenvalue decomposition (or, equivalently, the singular value decomposition) and squaring down the eigenvalues (singular values).

Besides being a robust way to compute the eigenvalues of \( W_c W_o \), the singular value decomposition of \( W_o^{1/2} W_c^{1/2} \) also plays an important role in the construction of the matrices \( T_1 \) and \( T_1^\dagger \) as shown below.

\textbf{Theorem 1:} Let \( G(s) \) be a stable transfer matrix and assume that (1) is any state–space representation of \( G(s) \) with gramians \( W_c \) and \( W_o \). In addition, suppose that \( W_c W_o \) has the following eigenvalue decomposition (Zhou et al. 1996, p. 77):
\[
W_c W_o = W \begin{bmatrix} \Sigma^2 & 0 \\ 0 & 0 \end{bmatrix} W^{-1},
\]
or, equivalently (Lemma 5), that \( W_o^{1/2} W_c^{1/2} \) has the following singular value decomposition:
\[
W_o^{1/2} W_c^{1/2} = [X_1 \quad X_2] \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Y_1^T \\ Y_2^T \end{bmatrix} = X_1 \Sigma Y_1^T. \tag{11}
\]

Defining
\[
T_1 = \Sigma^{-1/2} X_1^T W_o^{1/2} \quad \text{and} \quad T_1^\dagger = W_c^{1/2} Y_1 \Sigma^{-1/2}, \tag{12}
\]
then the realization (9) is minimal, and has controllability and observability gramians both equal to \( \Sigma \).

\textbf{Proof:} For \( T_1 \) and \( T_1^\dagger \), defined in (12), find two matrices \( T_2 \) and \( T_2^\dagger \) (Lemma 2) and construct a similarity transformation matrix \( T \) and its inverse \( T^{-1} \), as follows:
\[
T = \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} \quad \text{and} \quad T^{-1} = \begin{bmatrix} T_1^\dagger & T_2^\dagger \end{bmatrix} \tag{13}
\]
Thus
\[
G(s) = \begin{bmatrix} \hat{A} & \hat{B} \\ \hat{C} & \hat{D} \end{bmatrix} = \begin{bmatrix} T A T^{-1} & T B \\ C T^{-1} & D \end{bmatrix}
\]
has controllability and observability gramians, \( \hat{W}_c \) and \( \hat{W}_o \), respectively, given by:
\[
\hat{W}_c = \begin{bmatrix} T_1^T & T_2^T \end{bmatrix} W_c [T_1^T \quad T_2^T] = \begin{bmatrix} T_1 W_c T_1^T & T_1 W_c T_2^T \\ T_2 W_c T_1^T & T_2 W_c T_2^T \end{bmatrix}
\]
\[
= \begin{bmatrix} \Sigma & 0 \\ 0 & F \end{bmatrix}
\]
\[
\hat{W}_o = \begin{bmatrix} (T_1^T) & (T_2^T) \end{bmatrix} W_o [T_1^T \quad T_2^T] = \begin{bmatrix} (T_1^T) W_o T_1^T & (T_1^T) W_o T_2^T \\ (T_2^T) W_o T_1^T & (T_2^T) W_o T_2^T \end{bmatrix} = \begin{bmatrix} \Sigma & 0 \\ 0 & L \end{bmatrix},
\]
where \( F = T_2 W_c T_2^T \) and \( L = (T_2^T) W_o T_2^T \). Note that
\[
F L = T_2 W_c T_2^T (T_2^T) W_o T_2^T = T_2 W_c^{1/2} W_o^{1/2} T_2^T (T_2^T) W_o^{1/2} W_c^{1/2} T_2^T = T_2 W_c^{1/2} (W_o^{1/2} T_2^T T_2 W_c^{1/2}) T_2 W_o^{1/2} T_2^T = 0.
\]
since, according to Lemma 2 and equation (12),
\[ W_o^{1/2}T_2^rT_2W_c^{1/2} = W_o^{1/2}(I_n - T_1^rT_1)W_c^{1/2} \]
\[ = W_o^{1/2}W_c^{1/2} - W_o^{1/2}T_1^rT_1W_c^{1/2} \]
\[ = W_o^{1/2}W_c^{1/2} - W_o^{1/2}W_c^{1/2}Y_1 \]
\[ \times \Sigma^{-1/2} \Sigma^{-1/2} X_1^r W_o^{1/2}W_c^{1/2} = 0 \]

Let us now partition \( \hat{A}, \hat{B} \) and \( \hat{C} \) compatibly with \( T \) as follows:

\[
G(s) = \begin{bmatrix}
    \hat{A} & \hat{B} \\
    \hat{C} & D
\end{bmatrix} = \begin{bmatrix}
    \hat{A}_{11} & \hat{A}_{12} & B_1 \\
    \hat{A}_{21} & \hat{A}_{22} & \hat{B}_2 \\
    \hat{C}_1 & \hat{C}_2 & \hat{C}_3 & D
\end{bmatrix},
\]

where \( \hat{A}_{11} = T_1^rAT_1^r, \hat{B}_1 = T_1^rB \) and \( \hat{C}_1 = CT_1^r \). To prove that \( A_b, B_b, \text{ and } C_b \) given in (9), are, respectively, equal to \( \hat{A}_{11}, \hat{B}_1 \) and \( \hat{C}_1 \), given above, note that since \( FL = 0 \) then one of the following possibilities must occur:

1. Either \( L \) or \( F \) is identically zero. In this case, the result follows directly by application of Lemmas 3 and 4.
2. The matrices \( L \) and \( F \) are both non-identically zero. This implies that the columns of \( L \) must lie in the right null space of \( F \) or, equivalently, the rows of \( F \) must lie in the left null space of \( L \). This fact implies that \( F \) has the following eigenvalue decomposition:

\[
F = U_F^\top \begin{bmatrix} \Lambda_F & 0 \\ 0 & 0 \end{bmatrix} U_F,
\]

where \( \Lambda_F = \text{diag}\{\lambda_1, \lambda_2, \ldots, \lambda_f\}, \lambda_i \geq \lambda_j > 0, i < j \) and \( f < n - k \).

Consider now the similarity transformation:

\[
T = \begin{bmatrix} I_k & 0 \\ 0 & U_F \end{bmatrix} \quad \text{and} \quad T^{-1} = \begin{bmatrix} I_k & 0 \\ 0 & U_F^\top \end{bmatrix}.
\]

Therefore

\[
G(s) = \begin{bmatrix}
    T\hat{A}T^{-1} & T\hat{B} \\
    \hat{C}T^{-1} & D
\end{bmatrix} = \begin{bmatrix}
    \hat{A}_{11} & \hat{A}_{12}U_F^\top & B_1 \\
    U_F\hat{A}_{21} & U_F\hat{A}_{22}U_F^\top & U_FB_2 \\
    \hat{C}_1 & \hat{C}_2U_F^\top & D
\end{bmatrix},
\]

has gramians

\[
W_c^{(1)} = TW_cT^\top = \begin{bmatrix}
    \Sigma & 0 \\ 0 & U_FU_F^\top
\end{bmatrix} = \begin{bmatrix}
    \Sigma & 0 \\ 0 & 0
\end{bmatrix}
\]

and

\[
W_o^{(1)} = (T^{-1})^\top W_oT^{-1} = \begin{bmatrix}
    \Sigma & 0 \\ 0 & U_FLU_F^\top
\end{bmatrix} = \begin{bmatrix}
    \Sigma & 0 \\ 0 & L_{11} \end{bmatrix}.
\]

Partitioning the realization (15) compatibly with \( W_c^{(1)} \), gives:

\[
G(s) = \begin{bmatrix}
    \hat{A}_{11} & \hat{A}_{12} & \hat{A}_{13} & \hat{B}_1 \\
    \hat{A}_{21} & \hat{A}_{22} & \hat{A}_{23} & \hat{B}_2 \\
    \hat{C}_1 & \hat{C}_2 & \hat{C}_3 & D
\end{bmatrix},
\]

and applying Lemma 3 to the realization above, we obtain:

\[
G(s) = \begin{bmatrix}
    \hat{A}_{11} & \hat{A}_{12} & \hat{B}_1 \\
    \hat{A}_{21} & \hat{A}_{22} & \hat{B}_2 \\
    \hat{C}_1 & \hat{C}_2 & D
\end{bmatrix},
\]

whose gramians are:

\[
W_c^{(2)} = \begin{bmatrix} \Sigma & 0 \\ 0 & \Lambda_F \end{bmatrix}
\]

and

\[
W_o^{(2)} = \begin{bmatrix} \Sigma & 0 \\ 0 & L_{11} \end{bmatrix}.
\]

Note that

\[
\Lambda_FL_{11} = 0
\]

since

\[
W_c^{(1)}W_o^{(1)} = \begin{bmatrix} \Sigma^2 & 0 \\ 0 & \Lambda_FL_{11} \end{bmatrix}.
\]

and

\[
W_c^{(1)}W_o^{(1)} = TW_cW_oT^{-1} = \begin{bmatrix} \Sigma^2 & 0 \\ 0 & 0 \end{bmatrix}
\]

In addition, note that \( \Lambda_F \) is, by definition, full rank and hence, \( L_{11} \) must be identically zero.

Therefore:

\[
W_o^{(2)} = \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix}.
\]

Finally, applying Lemma 4 to (16), gives equation (9), with \( \hat{A}_{11} = A_b, \hat{B}_1 = B_b \) and \( \hat{C}_1 = C_b \).

Once a minimal realization for \( G(s) \) in a balanced form has been obtained, model reduction by balanced truncation can be employed directly to realization (16). This can be done in accordance with the following theorem (Glover 1984).
Theorem 2: Consider the balanced realization of $G(s)$ given in (16) and assume that its controllability and observability gramians are $\Sigma = \text{diag} \{ \Sigma_L, \Sigma_s \}$, where $\Sigma_L = \text{diag} \{ \sigma_1 I_{m_1}, \sigma_2 I_{m_2}, \ldots, \sigma_r I_{m_r} \}$ and $\Sigma_s = \text{diag} \{ \sigma_{r+1} I_{m_{r+1}}, \sigma_{r+2} I_{m_{r+2}}, \ldots, \sigma_k I_{m_k} \}$ with $\sigma_i > \sigma_j$, $i = 1, 2, \ldots, k$, $i < j$ and $m_i$ being the multiplicity of $\sigma_i$.

Partition (16) in accordance with $\Sigma$, namely:

$$ G(s) = \begin{bmatrix} A_{b_{11}} & A_{b_{12}} & B_{b_{1}} \\ A_{b_{21}} & A_{b_{22}} & B_{b_{2}} \\ C_{b_{1}} & C_{b_{2}} & D_{b} \end{bmatrix}. $$

Then the truncated system

$$ \tilde{G}(s) = \begin{bmatrix} A_{b_{11}} & B_{b_{1}} \\ C_{b_{1}} & D_{b} \end{bmatrix} $$

is balanced and stable. Moreover

$$ ||G - \tilde{G}||_\infty \leq 2(\sigma_{r+1} + \sigma_{r+2} + \cdots + \sigma_k). $$


The direct application of Theorem 2 to equation (9) leads to the following result.

Corollary 1: Let us partition the matrices $T_1$ and $T_1^\dagger$ in accordance with the gramian $\Sigma$, given in Theorem 2, as follows:

$$ T_1 = \begin{bmatrix} T_{11} \\ T_{1s} \end{bmatrix} \quad \text{and} \quad T_1^\dagger = \begin{bmatrix} T_{11}^\dagger \\ T_{s1}^\dagger \end{bmatrix}. $$

Then

$$ \tilde{G}(s) = \begin{bmatrix} T_{11} A T_{11}^\dagger + T_{1s} B \\ C T_{1s} \end{bmatrix} $$

is such that $||G - \tilde{G}||_\infty \leq 2(\sigma_{r+1} + \sigma_{r+2} + \cdots + \sigma_k)$.

Proof: Note that the matrices $A_{b_{11}}$, $B_{b_{1}}$, and $C_{b_{1}}$ of (17) are obtained from $A_b$, $B_b$, and $C_b$ as follows:

$$ A_{b_{11}} = \begin{bmatrix} I_r & 0_{r \times (k-r)} \end{bmatrix} A_b \begin{bmatrix} I_r \\ 0_{(k-r) \times r} \end{bmatrix}, $$

$$ B_{b_{1}} = \begin{bmatrix} I_r & 0_{r \times (k-r)} \end{bmatrix} B_b \quad \text{and} \quad C_{b_{1}} = C_b \begin{bmatrix} I_r \\ 0_{(k-r) \times r} \end{bmatrix}, $$

where $r = \sum_{i=1}^{r} m_i$ and $k = \sum_{i=1}^{k} m_i$. Hence, substituting $A_b = T_1 A T_1^\dagger$, $B_b = T_1 B$ and $C_b = C T_{1s}^\dagger$ in equation above, and noting that

$$ T_L = \begin{bmatrix} I_r & 0_{r \times (k-r)} \end{bmatrix} T_1 \quad \text{and} \quad T_L^\dagger = \begin{bmatrix} I_r \\ 0_{(k-r) \times r} \end{bmatrix} T_1^\dagger $$

gives the result.

Remark 3: Note that, from the definitions of $T_1$ and $T_1^\dagger$, given in (12), and of $T_L$ and $T_L^\dagger$, given in (19), we may write:

$$ T_1 = \begin{bmatrix} T_{1L} \\ T_{1s} \end{bmatrix} = \begin{bmatrix} \Sigma_L^{-1/2} & 0 \\ 0 & \Sigma_s^{-1/2} \end{bmatrix} \begin{bmatrix} X_L^T \\ X_s^T \end{bmatrix} W_o^{1/2}, $$

$$ T_1^\dagger = W_c^{1/2} \begin{bmatrix} Y_L \\ Y_s \end{bmatrix} \begin{bmatrix} \Sigma_L^{-1/2} & 0 \\ 0 & \Sigma_s^{-1/2} \end{bmatrix}, $$

and therefore

$$ T_L = \Sigma_L^{-1/2} X_L^T W_o^{1/2} \quad \text{and} \quad T_L^\dagger = W_c^{1/2} Y_L \Sigma_s^{-1/2}. \quad (20) $$

5. The algorithm

The results obtained in the previous section may be summarized in the following algorithm.

Algorithm 1: For a $p \times m$ stable, rational and proper transfer matrix $G(s)$ with a non-minimal state-space representation given by

$$ G(s) = \begin{bmatrix} A & B \\ C & D \end{bmatrix}, $$

a reduced-order model $\tilde{G}(s)$, in balanced form, can be obtained as follows:

Step 1. Compute the observability and controllability gramians, $W_c$ and $W_o$ respectively, by solving the following Lyapunov equations:

$$ A W_c + W_c A^T = -B B^T $$

$$ A^T W_o + W_o A = -C^T C. $$

Step 2. Compute the singular value decompositions of $W_c$ and $W_o$,

$$ W_c = U_c \Lambda_c U_c^T \quad \text{and} \quad W_o = U_o \Lambda_o U_o^T, $$

respectively, and find

$$ W_c^{1/2} = U_c \Lambda_c^{1/2} U_c^T \quad \text{and} \quad W_o^{1/2} = U_o \Lambda_o^{1/2} U_o^T. $$

Step 3. Compute the singular value decomposition of the product $W_o^{1/2} W_c^{1/2}$ and partition it as follows:

$$ W_o^{1/2} W_c^{1/2} = \begin{bmatrix} X_L & X_s & X_2 \end{bmatrix} \begin{bmatrix} \Sigma_L & 0 \\ 0 & \Sigma_s \end{bmatrix} \begin{bmatrix} Y_L^T \\ Y_s^T \\ 0 \end{bmatrix}, $$

where $\Sigma_L = \text{diag} \{ \sigma_1 I_{m_1}, \sigma_2 I_{m_2}, \ldots, \sigma_r I_{m_r} \}$ and $\Sigma_s = \text{diag} \{ \sigma_{r+1} I_{m_{r+1}}, \sigma_{r+2} I_{m_{r+2}}, \ldots, \sigma_k I_{m_k} \}$ are formed with the Hankel singular values to be kept and discarded, respectively.
Step 4. Compute

\[ T_L = \Sigma_L^{-1/2} X_L^T W_o^{1/2} \quad \text{and} \quad T_L^\dagger = W_c^{1/2} Y_L \Sigma_L^{-1/2}. \]

Step 5. Obtain the reduced order model \( \tilde{G}(s) \):

\[ \tilde{G}(s) = \begin{bmatrix} T_L A T_L^\dagger & T_L B \\ C T_L & D \end{bmatrix}. \]

6. Example

With the view to illustrating the algorithm proposed in this paper, let us consider the MIMO system

\[ G(s) = \frac{1}{(s + 1)^2(s + 1)} \begin{bmatrix} s + 1 & (s + 1)(2s + 1) & s(s + 1) \\ s + 2 & (s + 2)(s^2 + 5s + 3) & s(s + 2) \end{bmatrix}, \]

which has originally appeared in (Zhou et al. 1996, p. 82). From the Smith–McMillan form of \( G(s) \), we can conclude that its poles are \(-2\) and \(-1\) (multiplicity 3) and its unique zero is \(-2\). Therefore, any minimal order realization for \( G(s) \) must have four states. An immediate state–space representation for \( G(s) \) is as follows:

\[
G(s) = \begin{bmatrix}
-3 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0.5 \\
-1 & 0 & 0 & 0 & 0 & 0 & 0.25 & 0.25 & 0 \\
0 & 0 & -2 & 2 & 0 & 0 & 0 & 1.5 & 0.5 \\
0 & 0 & -0.5 & 0 & 0 & 0 & 0.25 & 0.5 & 0 \\
0 & 0 & 0 & -4 & 4 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & -1.25 & 0 & 1 & 0.5 & 0.25 \\
0 & 0 & 0 & 0 & -0.5 & 0 & 0 & 0.25 & 0.25 & 0 \\
2.0000 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 2 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

which is clearly non-minimal.

The next step towards obtaining a reduced model for \( G(s) \) is to perform the singular value decomposition of \( W_o^{1/2} W_c^{1/2} = \Sigma Y^T \). In doing so, we find out that \( G(s) \) has the following Hankel singular values:

\[ \mathcal{H} = \{1.34599705068499, 0.561442935324430, 0.229553347221876, 0.121751694629952, 0.994480795128365 \times 10^{-8}, 0.377508751606880 \times 10^{-8}, 0.137796554002854 \times 10^{-8} \}. \]

Note that if the last three Hankel singular values are discarded, then the error of approximation \( \| G - \tilde{G} \|_\infty \) will be \( \leq 3.0196 \times 10^{-8} \). Therefore, forming the matrices \( T_L (T_L^\dagger) \) with the first four columns (rows) of \( X (Y^T) \) we obtain the following reduced order model \( \tilde{G}(s) \) of \( G(s) \):

\[
\tilde{G}(s) = \begin{bmatrix}
-1.19912 & 1.17669 & -0.20410 & 0.08115 & 0.31978 & 1.71569 & 0.42683 \\
-0.22153 & -0.61453 & -0.23408 & 0.09276 & 0.66409 & -0.22411 & -0.44588 \\
-1.20966 & 0.99303 & -2.28659 & 0.50952 & -0.09831 & 0.85714 & 0.55265 \\
-0.23454 & -0.34857 & -0.25357 & -0.89976 & 0.10835 & 0.25235 & -0.37904 \\
0.77478 & -0.19715 & 0.82738 & -0.32752 & 0 & 0 & 0 \\
1.58965 & -0.78316 & 0.06424 & 0.16676 & 0 & 1 & 0 \\
0.31744 & -0.19452 & -0.60092 & -0.28986 & 0 & 0 & 0
\end{bmatrix}
\]
for which $W_c = W_o = \text{diag} \{1.34600, 0.56144, 0.22955, 0.12175\}$, and therefore the realization above is balanced. Note that the actual $H_\infty$ error between $G(s)$ and $\tilde{G}(s)$ is approximately $1.5 \times 10^{-15}$.

**Remark 4:** A realization, with less states then that given above, could be obtained by following the steps of algorithm (1). Indeed, suppose that we are interested in keeping the first three Hankel singular values. In this case, proceeding according to algorithm (1), we obtain a three-state balanced realization for which the actual $H_\infty$ error between $G(s)$ and $\tilde{G}(s)$ and the error bound (18) are approximately equal to 0.2435.

## 7. Conclusions

The problem of model reduction by balanced truncation has been revisited and a simple algorithm presented. The reduced-order model is obtained by pre- and post-multiplication of the non-minimal order state-space realization by rectangular matrices. Moreover, there is no restriction on the way the Lyapunov equations, required to calculate the gramians, are computed. Other features of the algorithm are the realization obtained for the reduced model is in a balanced form, and the algorithm is expected to be robust, since it relies solely on singular value decompositions.

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**References**


Doyle, J. C., 1984, Lecture Notes in Advances in Multivariable Control. (Minneapolis: ONR/Honeywell Workshop).


