# Subspace Controller Reduction based on Experimental Data

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#### Abstract

This paper proposes a controller reduction method based on experimental data. This method is applicable to unstable controllers and provides us some indices by which we can determine the reduced order. Furthermore, we evaluate its effectiveness by experiment on an inverted pendulum system.

#### 1 Introduction

So far, many researchers attacked the controller reduction problem and various nice results have been obtained [1, 2]. All of these results were obtained from the view point of control or system approximation. However, as for the system approximation, it seems to be nice to take advantage of the advanced approaches in system identification. In addition, as for controller redesign, it would be reasonable to utilize the closed loop experimental data. Therefore, in this paper, we proposed a new approach to the controller reduction problem based on the Subspace state-space system identification (4SID) method [3], which is known as a powerful tool for identification. Furthermore, we evaluate its effectiveness through experiments.

## 2 Reduction problem

In this paper, we consider the closed loop system shown in Fig.1, where P is the plant, K is the controller and both  $r_1$  and  $r_2$  are given reference signals. The controller K is represented by time invariant discrete-time state-space equation as

$$K: \begin{cases} x_{[k+1]} = Ax_{[k]} + By_{[k]} \\ u_{[k]} = Cx_{[k]} + Dy_{[k]} \end{cases}, x_{[k]} \in \mathbb{R}^n$$
 (1)

We assume that the above realization is minimal and that its state  $x_{[k]}$   $(k = 1, 2, 3 \cdots)$  is available. Our goal

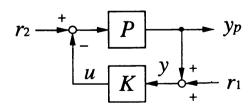


Figure 1: Closed loop system

is to obtain the reduced order controller  $K_r$  given by

$$K_r: \left\{ \begin{array}{l} x_r[k+1] = A_r x_r[k] + B_r y[k] \\ u_r[k] = C_r x_r[k] + D y[k] \end{array} \right., \ x_r[k] \in \mathbf{R}^p \quad (2)$$

which approximately satisfies  $u_r[k] \simeq u[k]$  using the closed loop experimental data.

#### 3 Reduction method

# 3.1 Basic idea

Define the following matrices.

$$\mathcal{Y}^{j} := \left[ \begin{array}{ccc} y_{[1]} & y_{[2]} & \cdots & y_{[N]} \\ \vdots & \vdots & & \vdots \\ y_{[j]} & y_{[j+1]} & \cdots & y_{[N+j-1]} \end{array} \right]$$
(3)

$$X := \begin{bmatrix} x_{[1]} & x_{[2]} & \dots & x_{[N+1]} \end{bmatrix}$$
 (4)

$$\Gamma^{j} := \begin{bmatrix} C \\ \vdots \\ CA^{j-1} \end{bmatrix}, \quad H^{j} := \begin{bmatrix} D & 0 \\ \vdots & \ddots & \\ CA^{j-2}B & \cdots & D \end{bmatrix}$$

$$(5)$$

Similarly,  $\mathcal{U}^j$ ,  $\mathcal{U}^j_r$ ,  $X_r$ ,  $\Gamma^j_r$  and  $H^j_r$  are defined through u[k],  $u_r[k]$ ,  $x_r[k]$ , and (2). Then, from (1) and (2), we have

$$\mathcal{U}^{j} - H^{j}\mathcal{Y}^{j} = \Gamma^{j}X, \ \mathcal{U}^{j}_{r} - H^{j}_{r}\mathcal{Y}^{j} = \Gamma^{j}_{r}X_{r} \tag{6}$$

Now we consider the situation that

$$\Gamma_r^j X_r \simeq \widehat{\Gamma}^j \widehat{X} \simeq \widehat{\Gamma}^j \left[ \frac{\widehat{X}_1}{0} \right]$$
 (7)

$$\widehat{X} = TX = \left[ \frac{\widehat{X}_1}{\widehat{X}_2} \right], \quad \widehat{X}_1 \in \mathbb{R}^{p \times N}$$
 (8)

$$\widehat{\Gamma}^j := \Gamma^j T^{-1} \tag{9}$$

are satisfied. Then, from (6) and (7), the following relations hold.

$$\begin{cases}
 u[k] & \simeq u_r[k] \\
 \widehat{C}\widehat{B}y[k] & \simeq C_rB_ry[k] \\
 \vdots & , \quad k = 1, \dots, N (10) \\
 \widehat{C}\widehat{A}^{j-1}\widehat{B}y[k] & \simeq C_rA_r^{j-1}B_ry[k] \\
 \widehat{A} &= TAT^{-1}, \widehat{B} &= BT^{-1}, \widehat{C} &= TC
\end{cases} (11)$$

So we can regard (2) as a reduced order approximation of (1). From the above observation, if we can find a similarity transformation T satisfying (7), we obtain the reduced controller  $K_r$  by

$$\widehat{A} =: \left[ \begin{array}{cc} A_r & * \\ * & * \end{array} \right], \ \widehat{B} =: \left[ \begin{array}{cc} B_r \\ * \end{array} \right], \ \widehat{C} =: \left[ \begin{array}{cc} C_r & * \end{array} \right]$$

$$\tag{12}$$

## 3.2 Algorithm

First, by using the experimental data of  $x_{[k]}$ , determine  $X \in \mathbb{R}^{n \times N}$  in (4). We assume rank X = n.

[Step 1] By using A, C, X and a large number j, set  $\Xi$  as

$$\Xi := \Gamma X \in \mathbf{R}^{jm \times N} \tag{13}$$

[Step 2] Using  $L_1$  calculated by

$$L_1 := \operatorname{diag} \left\{ \|\xi_1\|^{-1} \cdots \|\xi_{jm}\|^{-1} \right\}$$
 (14)

$$\Xi = \left[ \begin{array}{ccc} \xi_1^T & \cdots & \xi_{jm}^T \end{array} \right]^T \tag{15}$$

determine  $\Xi_L$  by

$$\Xi_L := L_1 \Xi \tag{16}$$

[Step 3] Calculate the singular value decomposition of  $\Xi_L$ 

$$\Xi_{L} = \begin{bmatrix} E \mid * \end{bmatrix} \begin{bmatrix} S \mid 0 \\ \hline 0 \mid 0 \end{bmatrix} \begin{bmatrix} V^{T} \\ * \end{bmatrix}$$

$$= ESV^{T}$$

$$S := \operatorname{diag} \{ \sigma_{1} \quad \sigma_{2} \quad \cdots \quad \sigma_{n} \}$$
(17)

In (17), if we consider

$$\Xi_L = L_1 \widehat{\Gamma}^j \widehat{X}, \quad \widehat{\Gamma}^j := L_1^{-1} ES, \quad \widehat{X} := V^T$$
 (18)

then by using T as

$$X = T\widehat{X}, \ T := XV \tag{19}$$

and if

$$\sigma_1 \ge \dots \ge \sigma_p \gg \sigma_{p+1} \ge \dots \ge \sigma_n$$
 (20)

is satisfied, it is easy to see that we obtain

$$\Xi_L \simeq L_1 \widehat{\Gamma}^j \left[ \frac{\widehat{X}_1}{0} \right] \tag{21}$$

So the similarity transformation T given by (19) satisfies (7).

[Step 4] From T in (19), we can obtain the reduced controller  $K_r$  by (12).

# 4 Experiment

# 4.1 Plant description

In this section, we consider the inverted pendulum system shown in Fig.2. In this system, the angle of the pendulum  $\theta$  and the arm  $\phi$  are controlled by the input torque  $\tau$ , and it is a single input and 2 output system.

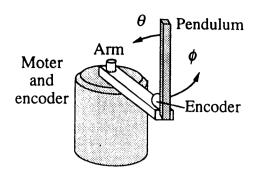


Figure 2: Inverted pendulum system

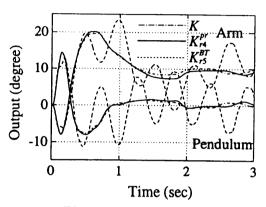


Figure 3: Step responses

#### 4.2 Controller reduction

We design a controller based on the  $H_{\infty}$  loop shaping control[4] which is of the order 12. We transformed K(s) to the discrete time controller K[z] and obtained the 3~7th order reduced controller  $K_{ri}^{pr}$ , (i=3,4,..7) by the proposed method (using j=30). The 4~7th order controller could stabilize P. Next, we obtained the 3~7th order reduced controller  $K_{ri}^{BT}$ , (i=3,4,..7) by BT method using the normalized coprime factorization. Then, the 5~7th order controller could stabilize P. Fig.3 shows step responses of closed loop systems using  $K, K_{rd}^{pr}$  and  $K_{r5}^{BT}$ .

## References

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