

행렬 부호함수를 이용한 대규모 선형 시불변 계통의 준최적화

Near optimal regulator problem for the large-scale linear time-invariant system via matrix sign function

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1. Introduction

For a large-scale dynamical systems, mathematical complexities and computational difficulties are encountered when we try to apply them in design of optimal regulator. To solve these, near optimality of large-scale system has been investigated by simplifying original system's model into lower dimensional model through aggregation or singular perturbation method. But determination of aggregation matrix or permutation, rescaling must be carried out in prior to apply these method.

In this paper another approach to near optimal regulator problem is introduced by using matrix sign function. Given large-scale system can be decomposed in its fast and slow subsystems and neglecting its fast subsystem, near optimal regulator can be designed with only its slow subsystem.

2. Block-decomposition via matrix sign function

To develop the method for block-decomposition of large-scale system, it is required to review the property of sign function as in the following definition.⁽³⁾

Definition 1.

The scalar sign function of a complex variable λ with $\text{Re}(\lambda) \neq 0$ is defined by

$$\text{sign}(\lambda) = \begin{cases} +1 & \text{when } \text{Re}(\lambda) > 0 \\ -1 & \text{when } \text{Re}(\lambda) < 0 \end{cases}$$

From definition 1 we can extend it to matrix sign function as follows:

Let M be the modal matrix of $A(n \times n)$ and

$$J = M^{-1}AM = \text{block diag} [J_+, J_-] \quad (1)$$

where $J_+ \in \mathbb{C}^{n_1 \times n_1}$ and $J_- \in \mathbb{C}^{n_2 \times n_2}$, with $n_1 + n_2 = n$, are the collection of Jordan blocks associated with the spectrum $\sigma(A) \subset \mathbb{C}^+$ and $\sigma(A) \subset \mathbb{C}^-$, respectively, where \mathbb{C}^+ and \mathbb{C}^- are the open right and left planes of \mathbb{C} , respectively. Then we can write the matrix sign function.

ction as follows:

$$\begin{aligned} \text{sign}(A) &= M [\text{sign}(J_+) \oplus \text{sign}(J_-)] M^{-1} \\ &= M [I_{n_1} \oplus (-I_{n_2})] M^{-1} \quad (2) \end{aligned}$$

And from the definition of the matrix sign function using Riesz projector, we can write:

$$\begin{aligned} \text{sign}(A) &= 2\text{sign}^+(A) - I_n \\ &= I_n - 2\text{sign}^-(A) \\ \text{or } \text{sign}^+(A) &= \frac{1}{2} [I_n + \text{sign}(A)] \\ \text{sign}^-(A) &= \frac{1}{2} [I_n - \text{sign}(A)] \quad (3) \end{aligned}$$

The computation of $\text{sign}(A)$ is possible by the algorithm proposed by Robert and Shieh.

Then the block diagonalization of a system map can be accomplished via the matrix sign function as follows:

Let us rewrite Eq (2) as,

$$\begin{aligned} \text{sign}(A) &= M [I_{n_1} \oplus (-I_{n_2})] W \\ &\triangleq \begin{bmatrix} M_{11} & M_{12} \\ M_{21} & M_{22} \end{bmatrix} [I_{n_1} \oplus (-I_{n_2})] \begin{bmatrix} W_{11} & W_{12} \\ W_{21} & W_{22} \end{bmatrix} \\ \text{where } W &= M^{-1} \triangleq \begin{bmatrix} W_{11} & W_{12} \\ W_{21} & W_{22} \end{bmatrix} \quad (4) \end{aligned}$$

and from Eq (4), Eq (3) becomes

$$\begin{aligned} \text{sign}^+(A) &= M [I_{n_1} \oplus 0_{n_2}] W \\ &= \begin{bmatrix} M_{11} \\ M_{21} \end{bmatrix} \begin{bmatrix} W_{11} & W_{12} \end{bmatrix} \\ &\triangleq M_1 W_1 \quad (5) \end{aligned}$$

and $\text{sign}^-(A) = M [0_{n_1} \oplus I_{n_2}] W$

$$\begin{aligned} &= \begin{bmatrix} M_{12} \\ M_{22} \end{bmatrix} \begin{bmatrix} W_{21} & W_{22} \end{bmatrix} \\ &\triangleq M_2 W_2 \quad (6) \end{aligned}$$

Let matrices S_1 and S_2 be defined as

$$S_1 \triangleq \text{ind} [\text{sign}^+(A)] = [s_1, s_2, \dots, s_{n_1}]$$

$$S_2 \triangleq \text{ind} [\text{sign}^-(A)] = [s_1, s_2, \dots, s_{n_2}]$$

where $\text{ind}(\cdot)$ indicates independent columns of $\text{sign}^+(A)$ and $\text{sign}^-(A)$, respectively.

These independent column vectors are selected from the $n(n_1, i=1, 2)$ column vectors of $\text{sign}^+(A)$ and $\text{sign}^-(A)$, respectively

These matrices S_1 and S_2 can be written as follows:

$$S_1 = \text{sign}^+(A) E_1, \quad E_1 \triangleq [e_n^{k1}, e_n^{k2}, \dots, e_n^{kn1}] \in \mathcal{C}^{n \times n_1}$$

$$S_2 = \text{sign}^-(A) E_2, \quad E_2 \triangleq [e_n^{k1}, e_n^{k2}, \dots, e_n^{kn2}] \in \mathcal{C}^{n \times n_2}$$

where E_1 and E_2 are elementary matrices.

From Eq (5) and Eq (6) we obtain.

$$S_1 = M_1 W_1 E_1 \triangleq M_1 H_1, \quad \text{where } H_1 \triangleq W_1 E_1 \quad (7)$$

$$S_2 = M_2 W_2 E_2 \triangleq M_2 H_2, \quad \text{where } H_2 \triangleq W_2 E_2 \quad (8)$$

According to Sylvester's inequality we obtain $\text{rank}(H_1) = n_1$ and $\text{rank}(H_2) = n_2$ or H_1 and H_2 are nonsingular.

If we define right block modal matrix M_R as

$$M_R \triangleq [S_1 \ S_2] = M H \quad (9)$$

where $H = [H_1 \oplus H_2]$. And from Eq. (1)

$$\begin{aligned} A M_R &= A M H = M J M^{-1} M H \\ &= M H \left(H^{-1} \begin{bmatrix} J_+ & 0_{n_1, n_2} \\ 0_{n_2, n_1} & J_- \end{bmatrix} H \right) \triangleq M_R A_R \quad (10) \end{aligned}$$

Hence we can write

$$A_R = M_R^{-1} A M_R$$

We can apply the previous result to block decomposition of a system in the following theorem.

Theorem 1

Let a linear time invariant system be given by

$$\dot{X} = AX + Bu, \quad X \in R^{n \times 1}, u \in R^{m \times 1} \quad (11-a)$$

$$y = CX, \quad y \in R^{p \times 1} \quad (11-b)$$

and

$$X = M_F X_R \overset{\Delta}{=} M_F [X_S^T \ X_F^T]^T, \quad X_S \in R^{n_1 \times 1}, \quad X_F \in R^{n_2 \times 1} \quad (12)$$

Then the block decomposed system of Eq. (11-a,b) using Eq.(12) becomes

$$\dot{X}_R = A_R X_R + B_R u \quad (13-a)$$

$$y = C_R X_R \quad (13-b)$$

where

$$A_R = M_F^{-1} A M_F = \text{block diag}[A_{R1}, A_{R2}] \quad (14-a)$$

$$B_R = M_F^{-1} B = [B_{R1}^T, B_{R2}^T]^T \quad (14-b)$$

$$C_R = C M_F = [C_{R1}, C_{R2}] \quad (14-c)$$

Proof) From Eq. (1)

$$WM = \begin{bmatrix} W_1 \\ W_2 \end{bmatrix} \begin{bmatrix} M_1 & M_2 \end{bmatrix} = \begin{bmatrix} W_1 M_1 & W_1 M_2 \\ W_2 M_1 & W_2 M_2 \end{bmatrix} = \begin{bmatrix} I_n & 0_{n \times n_2} \\ 0_{n_2 \times n_1} & I_{n_2} \end{bmatrix}$$

and

$$M_F^{-1} M_F = \begin{bmatrix} S_1^+ \\ S_2^+ \end{bmatrix} \begin{bmatrix} S_1 & S_2 \end{bmatrix} = \begin{bmatrix} S_1^+ S_1 & S_1^+ S_2 \\ S_2^+ S_1 & S_2^+ S_2 \end{bmatrix}$$

where '+' denotes pseudo inverse and

$$S_1^+ S_1 = I_{n_1}, \quad S_2^+ S_2 = I_{n_2}, \quad S_1^+ S_2 = 0_{n_1 \times n_2}, \quad S_2^+ S_1 = 0_{n_2 \times n_1} \quad (15)$$

Therefore Eq. (14-a) is

$$M_F^{-1} A M_F = \begin{bmatrix} S_1^+ \\ S_2^+ \end{bmatrix} A \begin{bmatrix} S_1 & S_2 \end{bmatrix} = \begin{bmatrix} S_1^+ A S_1 & S_1^+ A S_2 \\ S_2^+ A S_1 & S_2^+ A S_2 \end{bmatrix} \quad (16-a)$$

Define

$$A_{R1} \overset{\Delta}{=} S_1^+ A S_1, \quad S_1 A_{R1} = A S_1 \quad (16-b)$$

$$A_{R2} \overset{\Delta}{=} S_2^+ A S_2, \quad S_2 A_{R2} = A S_2 \quad (16-c)$$

From Eq.(16-b,c), off-diagonal block of Eq.

(16-a) can be obtained as follows

$$S_1^+ A S_2 = S_1^+ S_2 A_{R2} = 0_{n_1 \times n_2}, \quad S_2^+ A S_1 = S_2^+ S_1 A_{R1} = 0_{n_2 \times n_1}$$

Eq.(14-b,c) can be proved in the same way.

(Q.E.D.)

From theorem 1 and reference (5) we can obtain the reduced order model of a system with its dominant(slow) modes as follows

$$\dot{X}_S = A_{R1} X_S + B_{R1} u_R \quad (17-a)$$

$$y = C_{R1} X_S + C_{R2} (-A_{R2})^{-1} B_{R2} u_R \quad (17-b) \approx y_R$$

3. Near optimal regulator problem

To obtain the optimal control law for Eq.(1), we set the performance index J and J_R as follows:

$$J = \frac{1}{2} \int_0^{\infty} (y^T y + u^T R u) dt \quad (18)$$

$$\approx \frac{1}{2} \int_0^{\infty} (y_R^T y_R + u_R^T R u_R) dt \overset{\Delta}{=} J_R \quad (19)$$

Now the problem is to find a near optimal control law u_R^* to minimize the Eq.(19).

From Eq.(17-b) we can get

$$J_R = \frac{1}{2} \int_0^{\infty} (X_S^T C_{R1}^T C_{R1} X_S + 2u_R^T D^T C_{R1} X_S + u_R^T R u_R) dt \quad (20)$$

where $D = -C_{R2} A_{R2}^{-1} B_{R2}$, $R_0 = D^T D + R$.

Then we can find the near optimal control

$$u_R^* = -R_0^{-1} (D^T C_{R1} + B_{R1}^T K_S) X_S \quad (21)$$

where K_S is a solution of Riccati equation

$$0_{n_1 \times n_1} = K_S (A_{R1} - B_{R1} R_0^{-1} D^T C_{R1}) + (A_{R1} - B_{R1} R_0^{-1} D^T C_{R1})^T K_S + C_{R1}^T (I_{n_1} - D R_0^{-1} D^T) C_{R1} - K_S B_{R1} R_0^{-1} B_{R1}^T K_S$$

From Eq.(12) X_F and X_S was defined by mapping M_F^{-1} as follows

$$\begin{bmatrix} X_S \\ X_F \end{bmatrix} = M_F^{-1} X = \begin{bmatrix} S_1^+ \\ \dots \\ S_2^+ \end{bmatrix} X$$

Hence X_s can be written

$$X_s = S_1^+ X \quad (22)$$

As a result we can obtain near optimal regulator control u_R^* in Eq.(21) in terms of X as follows

$$u_R^* = -R_0^{-1} (D^T C_{R1} + B_{R1}^T K_S) S_1^+ X$$

4. Numerical Example

Consider a fifth-order linear time-invariant system.

$$\dot{X} = \begin{bmatrix} -2. & -1. & -0.5 & -0.1 & -0.05 \\ 0.5 & -1. & 0.2 & -4. & 0.8 \\ 1. & -0.1 & -0.25 & 0. & -1. \\ 0. & 1. & -0.1 & -0.1 & 0. \\ 1. & -1. & -1. & 0.5 & -5. \end{bmatrix} X + \begin{bmatrix} 1. \\ 0. \\ 1. \\ 1. \\ 1. \end{bmatrix} u$$

$$y = \begin{bmatrix} 1. & 0. & 0. & 0.05 & -0.01 \\ 0. & 1. & 0.01 & 0. & 0. \end{bmatrix} X$$

Let the performance index be $J = \frac{1}{2} \int_0^T (y^T y + u^T u) dt$ and $J_R = \frac{1}{2} \int_0^T (y_R^T y_R + u_R^T u_R) dt$ for optimal and near optimal regulator, respectively.

As a result of simulations, optimal and near optimal controls are obtained as follows

$$u^* = -0.1479x_1 + 0.1181x_2 - 0.0195x_3 - 0.8565$$

$$x_4 + 0.04659x_5$$

$$u_R^* = -0.15386x_1 + 0.1245x_2 - 0.0181x_3 +$$

$$0.02427x_5$$

The response curves of y_{opt} , y_R , and y_{spt} are compared in Fig.1 with their performances. Where y_{spt} denotes the response of singular perturbation method.

5. Conclusions

In this paper near optimal regulator for

large scale system has been designed by using lower order model via matrix sign function. As illustrated in the numerical example, proposed near optimal regulator problem has the better response than the one by singular perturbation method and other near optimal problem. Proposed algorithm can give solution to the difficulties in aggregation and singular perturbation and is useful for the near optimal regulator problem of multivariable systems with widely spaced eigenvalues.

References

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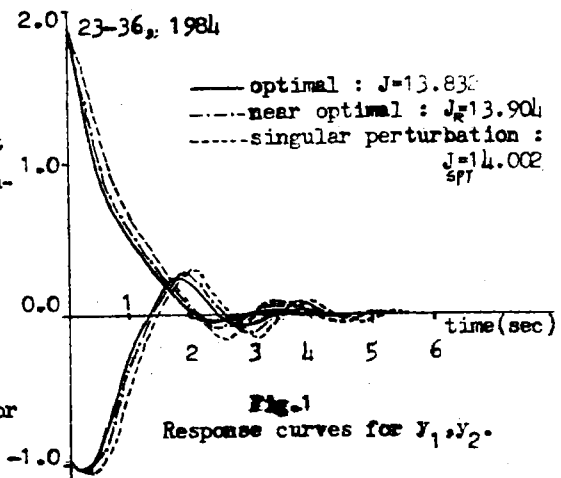


Fig.1
Response curves for y_1, y_2 .