

# NFL-O 에 기준한 SMMFC 의 안정도 증명 : Part 7

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## Stability Proof of NFL-O-based SMMFC : Part 7

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**[Abstract]** This paper presents a stability proof for the nonlinear feedback linearization-observer-based sliding mode model following controller (NFL-O-based SMMFC). The closed-loop stability is proved by a Lyapunov function candidate using an addition form of the sliding surface vector and the estimation error.

**Keywords :** nonlinear feedback linearization-observer-based sliding mode model following controller, Lyapunov function, stability proof

### 1. Introduction

In this paper, the nonlinear feedback linearization-observer-based sliding mode model following controller (NFL-O-based SMMFC) for unmeasurable plant state variables to solve the problem associated with the full state feedback is developed [1-17,19]. The proposed NFL-O-based SMMFC is obtained by the estimated state variable based on observer state in designing a sliding surface gain to ensure the stability by Lyapunov's method. The closed-loop stability is proved by a Lyapunov function candidate using an addition form of the sliding surface vector and the estimation error.

### 2. NFL-O-based SMMFC

The NFL-based reference model state equation is

$$z_m(t) = T(x_m(t)) \quad (1)$$

$$\dot{z}_m(t) = A_m z_m(t) + B_m u_m(t) \quad (2)$$

where  $x_m \in R^n$  is the state vector for model,  $z_m \in R^n$  is the transformed state vector for model,  $u_m \in R^p$  is the control input for model,  $A_m$  is the  $n \times n$  system matrix for model, and  $B_m$  is the

$n \times p$  control vector for model.

The control input for a reference model is

$$u_m(t) = -K_m z_m(t) \quad (3)$$

$$K_m = R_m^{-1} B_m^T P_m \quad (4)$$

$$P_m A_m + A_m^T P_m - P_m B_m R_m^{-1} B_m^T P_m + Q_m = 0 \quad (5)$$

where  $K_m$  is a  $p \times n$  optimal feedback gain for model, and  $P_m$  is the algebraic matrix Riccati equation.

The closed loop feedback system is

$$\dot{z}_m(t) = (A_m - B_m K_m) z_m(t) \quad (6)$$

$$A_{im} := A_m - B_m K_m \quad (7)$$

The NFL-based state equation including CLF is reformed as

$$\dot{z}_m(t) = A_{im} z_m(t) \quad (8)$$

The control input for a controlled plant is

$$u_p(t) = -K_p z_p(t) \quad (9)$$

$$K_p = R_p^{-1} B_p^T P_p \quad (10)$$

$$P_p A_p + A_p^T P_p - P_p B_p R_p^{-1} B_p^T P_p + Q_p = 0 \quad (11)$$

The NFL-based state equation for the controlled plant and the output equation are

$$z_p(t) = T(x_p(t)) \quad (12)$$

$$\dot{z}_p(t) = A_p z_p(t) + B_p u_p(t) \quad (13)$$

$$y_p(t) = C_p z_p(t) \quad (14)$$

The NFL-based observer equation is [18]

$$\begin{aligned} \dot{\hat{z}}_p(t) &= A_p \hat{z}_p(t) + B_p u_p(t) + L_p (y_p(t) - C_p \hat{z}_p(t)) \\ &= (A_p - L_p C_p) \hat{z}_p(t) + B_p u_p(t) + L_p y_p(t) \end{aligned} \quad (15)$$

$$L_p = P_p C_p^T R_p^{-1} \quad (16)$$

$$A_p P_p + P_p A_p^T - P_p C_p^T R_p^{-1} C_p P_p + Q_p = 0 \quad (17)$$

The NFL-based state equation for the controlled plant including CLF is expressed as

$$\dot{z}_p(t) = (A_p - B_p K_p) z_p(t) \quad (18)$$

$$\text{Let } A_{ip} := A_p - B_p K_p \quad (19)$$

$$\dot{z}_p(t) = A_{kp}z_p(t) + B_p u_{op}(t) \quad (20)$$

The error and the differential error vectors are

$$e(t) = z_m(t) - \hat{z}_p(t) \quad (21)$$

$$\dot{e}(t) = \dot{z}_m(t) - \dot{\hat{z}}_p(t) \quad (22)$$

$$\begin{aligned} \dot{e}(t) = \dot{z}_m(t) - \dot{\hat{z}}_p(t) = & \left[ A_{km}z_m(t) \right] - \left[ (A_{kp} - L_p C_p) \hat{z}_p(t) \right. \\ & \left. + B_p \hat{u}_{O-SMFC}(t) + L_p y_p(t) \right] \end{aligned} \quad (23)$$

$$z_m(t) = e(t) + \hat{z}_p(t) \quad (24)$$

By substituting (24) into (23), we have

$$\begin{aligned} \dot{e}(t) = & A_{km}z_m(t) - (A_{kp} - L_p C_p) \hat{z}_p(t) - B_p \hat{u}_{O-SMFC}(t) - L_p y_p(t) \\ = & A_{km}(e(t) + \hat{z}_p(t)) - (A_{kp} - L_p C_p) \hat{z}_p(t) \\ & - B_p \hat{u}_{O-SMFC}(t) - L_p y_p(t) \\ = & A_{km}e(t) + (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \\ & - B_p \hat{u}_{O-SMFC}(t) - L_p y_p(t) \end{aligned} \quad (25)$$

The sliding surface vector and the differential sliding surface vector can be expressed as

$$\sigma(e(t)) = G_{ss}^T e(t) = G_{ss}^T z_m(t) - G_{ss}^T \hat{z}_p(t) \Rightarrow 0 \quad (26)$$

$$\begin{aligned} \dot{\sigma}(e(t)) = & G_{ss}^T \dot{e}(t) = G_{ss}^T A_{km}e(t) + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \\ & - G_{ss}^T B_p \hat{u}_{O-SMFC}(t) - G_{ss}^T L_p y_p(t) \Rightarrow 0 \end{aligned} \quad (27)$$

where  $G_{ss}^T$  is the sliding surface gain [ ].

The Lyapunov's function candidate is chosen as

$$V(e(t)) = \sigma^T(e(t)) / 2 \quad (28)$$

The time derivative of  $V(e(t))$  is given by

$$\dot{V}(e(t)) = \sigma(e(t)) \dot{\sigma}(e(t)) \quad (29)$$

$$= G_{ss}^T e(t) G_{ss}^T \dot{e}(t) \quad (30)$$

$$\begin{aligned} = & G_{ss}^T e(t) G_{ss}^T \left[ A_{km}e(t) + (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \right. \\ & \left. - B_p \hat{u}_{O-SMFC}(t) - L_p y_p(t) \right] \\ = & G_{ss}^T e(t) \left[ G_{ss}^T A_{km}e(t) + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \right. \\ & \left. - G_{ss}^T B_p \hat{u}_{O-SMFC}(t) - G_{ss}^T L_p y_p(t) \right] \leq 0 \end{aligned} \quad (31)$$

From equation (31), the estimated control inputs with switching function are represented by

$$\begin{aligned} \hat{u}_{O-SMFC}(t) \geq & \left( G_{ss}^T B_p \right)^{-1} \left[ G_{ss}^T A_{km}e(t) + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \right. \\ & \left. - G_{ss}^T L_p y_p(t) \right] \quad \text{for } G_{ss}^T e(t) > 0 \end{aligned} \quad (32)$$

$$\begin{aligned} \hat{u}_{O-SMFC}(t) \leq & \left( G_{ss}^T B_p \right)^{-1} \left[ G_{ss}^T A_{km}e(t) + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p(t) \right. \\ & \left. - G_{ss}^T L_p y_p(t) \right] \quad \text{for } G_{ss}^T e(t) < 0 \end{aligned} \quad (33)$$

The estimated control input vector with sign function for the controlled plant is reformed as

$$\begin{aligned} \hat{u}_{O-SMFC}^{sign}(t) = & \left[ E_{O-SMFC}^{equal} e(t) + P_{O-SMFC}^{equal} \hat{z}_p(t) \right. \\ & \left. + O_{O-SMFC}^{equal} y_p(t) \right] \text{sign}(\sigma(e(t))) \end{aligned} \quad (34)$$

subject to  $\text{sign}(\sigma(e(t))) = 1$  for  $\sigma(e(t)) > 0$

$$\text{sign}(\sigma(e(t))) = 0 \quad \text{for } \sigma(e(t)) = 0$$

$$\text{sign}(\sigma(e(t))) = -1 \quad \text{for } \sigma(e(t)) < 0$$

$$\text{where } E_{O-SMFC}^{equal} := \left( G_{ss}^T B_p \right)^{-1} G_{ss}^T A_{km} \quad (35)$$

$$P_{O-SMFC}^{equal} := \left( G_{ss}^T B_p \right)^{-1} G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \quad (36)$$

$$O_{O-SMFC}^{equal} := -\left( G_{ss}^T B_p \right)^{-1} G_{ss}^T L_p \quad (37)$$

**Theorem 1:** Consider the state equations of the reference model and the controlled plant based on NFL for the regulation problem

$$\dot{z}_m = A_{km}z_m \quad \text{and} \quad y_m = C_m z_m$$

$$\dot{z}_p = A_{kp}z_p + B_p \hat{u}_{O-SMFC}^{sign} \quad \text{and} \quad y_p = C_p z_p$$

Consider the observer state equation based on NFL

$$\dot{\hat{z}}_p = A_p \hat{z}_p + B_p \hat{u}_{O-SMFC}^{sign} + L_p (y_p - C_p \hat{z}_p)$$

Consider  $G_{ss}^T B_p \left( G_{ss}^T B_p \right)^{-1} = I$ ,  $y_p = C_p z_p$ ,  $e = z_m - \hat{z}_p$ , and  $z_p = e_p + \hat{z}_p$ . Suppose that  $(A_p, C_p)$  is detectable and  $(A_p - L_p C_p)$  is Hurwitz. The estimated sliding mode model following control law with sign function based on NFL is guaranteed an asymptotically stable for the system (13)

$$\hat{u}_{O-SMFC}^{sign} = \left[ E_{O-SMFC}^{equal} e + P_{O-SMFC}^{equal} \hat{z}_p + O_{O-SMFC}^{equal} y_p \right] \text{sign}(\sigma(e))$$

$$E_{O-SMFC}^{equal} := \left( G_{ss}^T B_p \right)^{-1} G_{ss}^T A_{km}$$

$$P_{O-SMFC}^{equal} := \left( G_{ss}^T B_p \right)^{-1} G_{ss}^T (A_{km} + L_p C_p - A_{kp})$$

$$O_{O-SMFC}^{equal} := -\left( G_{ss}^T B_p \right)^{-1} G_{ss}^T L_p$$

subject to  $\text{sign}(\sigma(e)) = 1$  for  $\sigma(e) > 0$

$$\text{sign}(\sigma(e)) = 0 \quad \text{for } \sigma(e) = 0$$

$$\text{sign}(\sigma(e)) = -1 \quad \text{for } \sigma(e) < 0$$

**Proof.** Let us define the error equation

$$e_p = z_p - \hat{z}_p$$

The differential error equation is represented by

$$\begin{aligned} \dot{e}_p = & \dot{z}_p - \dot{\hat{z}}_p \\ = & A_{kp}z_p + B_p \hat{u}_{O-SMFC}^{sign} - A_{kp} \hat{z}_p \\ & - B_p \hat{u}_{O-SMFC}^{sign} - L_p C_p z_p + L_p C_p \hat{z}_p \\ = & A_{kp}z_p - A_{kp} \hat{z}_p - L_p C_p z_p + L_p C_p \hat{z}_p \\ = & (A_{kp} - L_p C_p) e_p \end{aligned}$$

A Lyapunov's function candidate is chosen by

$$V = \frac{1}{2} \sigma^T(e) \sigma(e) + \frac{1}{2} e_p^T e_p$$

The derivative is obtained by

$$\begin{aligned} \dot{V} = & \sigma^T(e) \dot{\sigma}(e) + e_p^T \dot{e}_p \\ = & \sigma^T(e) \left( G_{ss}^T A_{km}e + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \right. \\ & \left. - G_{ss}^T B_p \hat{u}_{O-SMFC}^{sign} - G_{ss}^T L_p y_p \right) + e_p^T (A_{kp} - L_p C_p) e_p \\ = & \sigma^T(e) \left( G_{ss}^T A_{km}e + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \right. \\ & \left. - \left( G_{ss}^T B_p E_{O-SMFC}^{equal} e + G_{ss}^T B_p P_{O-SMFC}^{equal} \hat{z}_p \right) \right. \end{aligned}$$

$$+G_{ss}^T B_p O_{O-SMMFC}^{equal} y_p) \text{sign}(\sigma(e)) - G_{ss}^T L_p y_p) \\ + e_p^T (A_{kp} - L_p C_p) e_p$$

Let  $E_{O-SMMFC}^{equal} := (G_{ss}^T B_p)^{-1} G_{ss}^T A_{km}$   
 $P_{O-SMMFC}^{equal} := (G_{ss}^T B_p)^{-1} G_{ss}^T (A_{km} + L_p C_p - A_{kp})$   
 $O_{O-SMMFC}^{equal} := -(G_{ss}^T B_p)^{-1} G_{ss}^T L_p$

Therefore,

$$\dot{V} = \sigma^T(e) \left( G_{ss}^T A_{km} e + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \right. \\ \left. - G_{ss}^T B_p \left( (G_{ss}^T B_p)^{-1} G_{ss}^T A_{km} \right) e \text{sign}(\sigma(e)) \right. \\ \left. - G_{ss}^T B_p \left( (G_{ss}^T B_p)^{-1} G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \right) \hat{z}_p \text{sign}(\sigma(e)) \right. \\ \left. + G_{ss}^T L_p \left( (G_{ss}^T B_p)^{-1} G_{ss}^T L_p \right) y_p \text{sign}(\sigma(e)) - G_{ss}^T L_p y_p \right) \\ + e_p^T (A_{kp} - L_p C_p) e_p$$

Consider  $G_{ss}^T B_p (G_{ss}^T B_p)^{-1} = I$ ,  $y_p = C_p z_p$ ,  $e = z_m - \hat{z}_p$ , and  $z_p = e_p + \hat{z}_p$

$$\dot{V} = \sigma^T(e) \left( G_{ss}^T A_{km} e + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \right. \\ \left. - G_{ss}^T A_{km} e \text{sign}(\sigma(e)) - G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \text{sign}(\sigma(e)) \right. \\ \left. + G_{ss}^T L_p y_p \text{sign}(\sigma(e)) - G_{ss}^T L_p y_p \right) + e_p^T (A_{kp} - L_p C_p) e_p \\ = \sigma^T(e) \left( G_{ss}^T A_{km} e + G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \right. \\ \left. - G_{ss}^T A_{km} e \text{sign}(\sigma(e)) - G_{ss}^T (A_{km} + L_p C_p - A_{kp}) \hat{z}_p \text{sign}(\sigma(e)) \right. \\ \left. + G_{ss}^T L_p C_p z_p \text{sign}(\sigma(e)) - G_{ss}^T L_p C_p z_p \right) + e_p^T (A_{kp} - L_p C_p) e_p \\ = \sigma^T(e) G_{ss}^T (A_{km} + L_p C_p - A_{kp}) (1 - \text{sign}(\sigma(e))) \hat{z}_p \\ + \sigma^T(e) G_{ss}^T L_p C_p (e_p + \hat{z}_p) \text{sign}(\sigma(e)) \\ - \sigma^T(e) G_{ss}^T L_p C_p (e_p + \hat{z}_p) + \sigma^T(e) G_{ss}^T A_{km} e \\ - \sigma^T(e) G_{ss}^T A_{km} e \text{sign}(\sigma(e)) + e_p^T (A_{kp} - L_p C_p) e_p \\ = \sigma^T(e) G_{ss}^T (A_{km} + L_p C_p - A_{kp}) (1 - \text{sign}(\sigma(e))) \hat{z}_p \\ - \sigma^T(e) G_{ss}^T L_p C_p (1 - \text{sign}(\sigma(e))) \hat{z}_p \\ + \sigma^T(e) G_{ss}^T L_p C_p e_p \text{sign}(\sigma(e)) - \sigma^T(e) G_{ss}^T L_p C_p e_p \\ + \sigma^T(e) G_{ss}^T A_{km} e - \sigma^T(e) G_{ss}^T A_{km} e \text{sign}(\sigma(e)) \\ + e_p^T (A_{kp} - L_p C_p) e_p \\ = \sigma^T(e) G_{ss}^T (A_{km} - A_{kp}) (1 - \text{sign}(\sigma(e))) \hat{z}_p \\ + \sigma^T(e) G_{ss}^T L_p C_p e_p \text{sign}(\sigma(e)) - \sigma^T(e) G_{ss}^T L_p C_p e_p \\ + \sigma^T(e) G_{ss}^T A_{km} e - \sigma^T(e) G_{ss}^T A_{km} e \text{sign}(\sigma(e)) \\ + e_p^T (A_{kp} - L_p C_p) e_p$$

If  $(A_{kp} - L_p C_p)$  is stable, the error is  $e_p \rightarrow 0$ , and  $e \rightarrow 0$  as  $t \rightarrow 0$ .

$$\dot{V} = \sigma^T(e) G_{ss}^T (A_{km} - A_{kp}) (1 - \text{sign}(\sigma(e))) \hat{z}_p \leq 0$$

subject to if  $\sigma(e) > 0$ ,  $\dot{V} = 0$   
if  $\sigma(e) = 0$ ,  $\dot{V} = 0$   
if  $\sigma(e) < 0$ ,  $\dot{V} = -2k G_{ss}^T (A_{km} - A_{kp}) \hat{z}_p < 0$

,  $k$  is positive constant.

i.e.,  $\dot{V} < 0$  and so the system is asymptotically stable.

This completes the proof of this theorem.  $\square$

### 3. Conclusion

A stability proof of a nonlinear feedback linearization-observer-based sliding mode model following controller (NFL-O-based SMMFC) has been done.

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