

NFL-H_∞에 기준한 SMC 의 안정도 증명 : Part 8

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Stability Proof of NFL-H_∞-based SMC : Part 8

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[Abstract] In this paper, the standard Dole, Glover, Khargoneker, and Francis (abbr. : DGKF 1989) H_∞ controller (H_∞C) is extended to the nonlinear feedback linearization-H_∞-based sliding mode controller (NFL-H_∞-based SMC). A stability proof of the closed-loop stability is done by a Lyapunov function candidate using an addition form of the sliding surface vector and the estimation error.

Keywords : nonlinear feedback linearization-H_∞-based sliding mode controller, Lyapunov function, stability proof

1. Introduction

In this paper, the standard Dole, Glover, Khargoneker, and Francis (abbr. : DGKF 1989) H_∞ controller (H_∞C) [1] is extended to the nonlinear feedback linearization-H_∞-based sliding mode controller (NFL-H_∞-based SMC) [2-18]. The proposed NFL-H_∞-based SMC is obtained by the estimated state variable based on H_∞ estimator 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-H_∞-based SMC

The state equations under worst case based on nonlinear feedback linearization (NFL) are [19]

$$z(t) = T(x(t)) \quad (1)$$

$$\dot{z}(t) = Az(t) + B_1 w_{\text{worst}}(t) + B_2 u(t) \quad (2)$$

$$p(t) = C_1 z(t) + D_{11} w_{\text{worst}}(t) + D_{12} u(t) \quad (3)$$

$$y(t) = C_2 z(t) + D_{21} w_{\text{worst}}(t) + D_{22} u(t) \quad (4)$$

The standard H_∞ estimator state equation based on NFL is [1]

$$\dot{\hat{z}}(t) = A\hat{z}(t) + B_2 u(t) + B_1 \hat{w}_{\text{worst}}(t) + Z_{\infty} K_e (y(t) - \hat{y}(t)) \quad (5)$$

$$\text{where } \hat{w}_{\text{worst}}(t) = \gamma^{-2} B_1^T X_{\infty} \hat{z}(t) \quad (6)$$

$$\hat{y}(t) = [C_2 + \gamma^{-2} D_{21} B_1^T X_{\infty}] \hat{z}(t) \quad (7)$$

The controller gain K_c is given by

$$K_c = \tilde{D}_{12} (B_2^T X_{\infty} + D_{12}^T C_1) \quad (8)$$

$$\text{where } \tilde{D}_{12} = (D_{12}^T D_{12})^{-1} \quad (9)$$

The estimator gain K_e is given by

$$K_e = (Y_{\infty} C_2^T + B_1 D_{11}^T) \tilde{D}_{21} \quad (10)$$

$$\text{where } \tilde{D}_{21} = (D_{21} D_{21}^T)^{-1} \quad (11)$$

The term Z_{∞} is given by

$$Z_{\infty} = (I - \gamma^{-2} Y_{\infty} X_{\infty})^{-1} \quad (12)$$

The controller Riccati equation term X_{∞} is

$$X_{\infty} = \text{Ric} \begin{bmatrix} A - B_1 \tilde{D}_{12} D_{11}^T C_1 & \gamma^{-2} B_1 B_1^T - B_1 \tilde{D}_{12} B_1^T \\ -\tilde{C}_1^T \tilde{C}_1 & -(A - B_1 \tilde{D}_{12} D_{11}^T C_1)^T \end{bmatrix} \quad (13)$$

$$\text{where } \tilde{C}_1 = (I - D_{11} \tilde{D}_{12} D_{11}^T) C_1 \quad (14)$$

The estimator Riccati equation term is

$$Y_{\infty} = \text{Ric} \begin{bmatrix} (A - B_1 \tilde{D}_{12} D_{11}^T C_1)^T & \gamma^{-2} C_2 C_2^T - C_2^T \tilde{D}_{21} C_2 \\ -\tilde{B}_1 \tilde{B}_1^T & -(A - B_1 \tilde{D}_{12} D_{11}^T C_1) \end{bmatrix} \quad (15)$$

$$\text{where } \tilde{B}_1 = B_1 (I - D_{21} \tilde{D}_{21} D_{21}^T) \quad (16)$$

The estimated control input based on NFL is

$$u_e(t) = -K_c \hat{z}(t) \quad (17)$$

The closed loop system can be expressed as

$$\begin{bmatrix} \dot{z}(t) \\ \dot{\hat{z}}(t) \end{bmatrix} = \begin{bmatrix} A & -B_2 K_c \\ Z_{\infty} K_e C_2 & A_1 \end{bmatrix} \begin{bmatrix} z(t) \\ \hat{z}(t) \end{bmatrix} + \begin{bmatrix} B_1 \\ Z_{\infty} K_e D_{21} \end{bmatrix} w_{\text{worst}}(t) \quad (20)$$

$$\begin{bmatrix} p(t) \\ y(t) \end{bmatrix} = \begin{bmatrix} C_1 & -D_{12}K_c \\ C_2 & 0 \end{bmatrix} \begin{bmatrix} z(t) \\ \hat{z}(t) \end{bmatrix} + \begin{bmatrix} 0 \\ D_{21} \end{bmatrix} w_{\text{worst}}(t) \quad (21)$$

where $A_1 := A - B_1K_c + \gamma^{-2}D_{21}B_1^T X_w$
 $-Z_w K_e (C_2 + \gamma^{-2}D_{21}B_1^T X_w)$ (22)

The state equation based on NFL is set to

$$\begin{aligned} \dot{z}(t) &= Az(t) + B_1 w_{\text{worst}}(t) + B_2 u(t) \\ &= (A + B_1(\gamma^{-2}B_1^T X_w))z(t) + B_2 u(t) \end{aligned} \quad (23)$$

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

$$\sigma(\hat{z}(t)) = G_{ss}^T \hat{z}(t) \quad (24)$$

$$\dot{\sigma}(\hat{z}(t)) = G_{ss}^T \dot{\hat{z}}(t) \quad (25)$$

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

The Lyapunov's function candidate is chosen by

$$V(\hat{z}(t)) = \sigma^2(\hat{z}(t)) / 2 \quad (26)$$

$$\begin{aligned} \dot{V}(\hat{z}(t)) &= \sigma(\hat{z}(t))\dot{\sigma}(\hat{z}(t)) = G_{ss}^T \hat{z}(t) G_{ss}^T \dot{\hat{z}}(t) \\ &= G_{ss}^T \hat{z}(t) G_{ss}^T \left[(A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] \right. \\ &\quad \left. + B_1 \gamma^{-2} B_1^T X_w) \hat{z}(t) + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{equal}}(t) + Z_w K_e y(t) \right] \\ &= G_{ss}^T \hat{z}(t) \left\{ G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] \right. \\ &\quad \left. + B_1 \gamma^{-2} B_1^T X_w) \hat{z}(t) + G_{ss}^T B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{equal}}(t) + G_{ss}^T Z_w K_e y(t) \right\} \\ &\leq 0 \end{aligned} \quad (27)$$

From (27), the estimated control input with switching function can be reduced as

$$\begin{aligned} \hat{u}_{\text{H}\infty\text{-SMC}}^+(t) &\geq -(G_{ss}^T B_2)^{-1} \left[G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] \right. \\ &\quad \left. + B_1 \gamma^{-2} B_1^T X_w) \hat{z}(t) + G_{ss}^T Z_w K_e y(t) \right] \\ &\text{for } G_{ss}^T \hat{z}(t) > 0 \end{aligned} \quad (28)$$

$$\begin{aligned} \hat{u}_{\text{H}\infty\text{-SMC}}^-(t) &\leq -(G_{ss}^T B_2)^{-1} \left[G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] \right. \\ &\quad \left. + B_1 \gamma^{-2} B_1^T X_w) \hat{z}(t) + G_{ss}^T Z_w K_e y(t) \right] \\ &\text{for } G_{ss}^T \hat{z}(t) < 0 \end{aligned} \quad (29)$$

The estimated control input with sign function can be reformed as

$$\begin{aligned} \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}(t) &= -(G_{ss}^T B_2)^{-1} \left[G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] \right. \\ &\quad \left. + B_1 \gamma^{-2} B_1^T X_w) \hat{z}(t) + G_{ss}^T Z_w K_e y(t) \right] \text{sign}(\sigma(\hat{z})) \end{aligned} \quad (30)$$

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

$\text{sign}(\sigma(\hat{z}(t))) = 0$ for $\sigma(\hat{z}(t)) = 0$, and $\text{sign}(\sigma(\hat{z}(t))) = -1$

for $\sigma(\hat{z}(t)) < 0$.

The equation (30) can be simplified as follows:

$$\hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}(t) = -[WK_1^{\text{equal}} \hat{z}(t) + WK_2^{\text{equal}} y(t)] \text{sign}(\sigma(\hat{z}(t))) \quad (31)$$

where, $WK_1^{\text{equal}} := (G_{ss}^T B_2)^{-1} G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] + B_1 \gamma^{-2} B_1^T X_w]$ (32)

$$WK_2^{\text{equal}} := (G_{ss}^T B_2)^{-1} G_{ss}^T Z_w K_e \quad (33)$$

Theorem 1: Consider the state equations and the H_∞ estimator based on NFL for the regulation problem under a worst case

$$\dot{z} = Az + B_1 w_{\text{worst}} + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}$$

$$p = C_1 z + D_{11} w_{\text{worst}} + D_{12} \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}$$

$$y = C_2 z + D_{21} w_{\text{worst}} + D_{22} \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}$$

$$\dot{\hat{z}} = A\hat{z} + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}} + B_1 \hat{w}_{\text{worst}} + Z_w K_e (y - \hat{y})$$

The estimated sliding mode control law with sign function based on NFL that keeps the system stable is guaranteed an asymptotically stable for the system (2)

$$\hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}} = -[WK_1^{\text{equal}} \hat{z} + WK_2^{\text{equal}} y] \text{sign}(\sigma(\hat{z}))$$

$$WK_1^{\text{equal}} := (G_{ss}^T B_2)^{-1} G_{ss}^T [A - Z_w K_e [C_2 + \gamma^{-2}D_{21}B_1^T X_w] + B_1 \gamma^{-2} B_1^T X_w]$$

$$WK_2^{\text{equal}} := (G_{ss}^T B_2)^{-1} G_{ss}^T Z_w K_e$$

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

$\text{sign}(\sigma(\hat{z}(t))) = 0$ for $\sigma(\hat{z}(t)) = 0$, and $\text{sign}(\sigma(\hat{z}(t))) = -1$

for $\sigma(\hat{z}(t)) < 0$.

Proof. Let us define the estimation error equation

$$e = z - \hat{z}$$

$$\dot{e} = \dot{z} - \dot{\hat{z}} = Az + B_1 w_{\text{worst}} + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}}$$

$$- [A\hat{z} + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}} + B_1 \hat{w}_{\text{worst}} + Z_w K_e (y - \hat{y})]$$

$$= Az + B_1 w_{\text{worst}} - A\hat{z} - B_1 \hat{w}_{\text{worst}} - Z_w K_e y + Z_w K_e \hat{y}$$

$$= Az + B_1 w_{\text{worst}} - A\hat{z} - B_1 \hat{w}_{\text{worst}}$$

$$- Z_w K_e C_2 z - Z_w K_e D_{21} w_{\text{worst}} - Z_w K_e D_{22} (-K_c \hat{z})$$

$$+ Z_w K_e C_2 \hat{z} + Z_w K_e D_{21} \hat{w}_{\text{worst}} + Z_w K_e D_{22} (-K_c \hat{z})$$

$$= Az - Z_w K_e C_2 z - A\hat{z} + Z_w K_e C_2 \hat{z} + B_1 w_{\text{worst}}$$

$$- Z_w K_e D_{21} w_{\text{worst}} - B_1 \hat{w}_{\text{worst}} + Z_w K_e D_{21} \hat{w}_{\text{worst}}$$

$$= (A - Z_w K_e C_2) z - (A - Z_w K_e C_2) \hat{z} + B_1 w_{\text{worst}} - Z_w K_e D_{21} w_{\text{worst}}$$

$$- B_1 \hat{w}_{\text{worst}} + Z_w K_e D_{21} \hat{w}_{\text{worst}}$$

$$= (A - Z_w K_e C_2) z - (A - Z_w K_e C_2)(e + z) \quad \because \hat{z} := e + z$$

$$+ B_1 \gamma^{-2} B_1^T X_w z - Z_w K_e D_{21} \gamma^{-2} B_1^T X_w z - B_1 \gamma^{-2} B_1^T X_w \hat{z}$$

$$+ Z_w K_e D_{21} \gamma^{-2} B_1^T X_w \hat{z}$$

$$= -(A - Z_w K_e C_2)e - B_1 \gamma^{-2} B_1^T X_w e + Z_w K_e D_{21} \gamma^{-2} B_1^T X_w e$$

$$= -(A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_e (C_2 - \gamma^{-2} D_{21} B_1^T X_w))e$$

Lyapunov's function candidate is chosen by

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

$$\dot{V} = \sigma^T \dot{\sigma} + e^T \dot{e}$$

$$= \sigma^T (G_{ss}^T \dot{\hat{z}}) - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_e (C_2 - \gamma^{-2} D_{21} B_1^T X_w))e$$

$$= \sigma^T G_{ss}^T (A\hat{z} + B_2 \hat{u}_{\text{H}\infty\text{-SMC}}^{\text{sign}} + B_1 \hat{w}_{\text{worst}} + Z_w K_e (y - \hat{y}))$$

$$- e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_e (C_2 - \gamma^{-2} D_{21} B_1^T X_w))e$$

$$\begin{aligned}
&= \sigma^T G_{ss}^T (A\hat{z} + B_2 \hat{u}_{H_{\infty}-SMC}^{H_{\infty}} + B_1 \hat{w}_{\text{error}} + Z_w K_w \gamma - Z_w K_w \hat{y}) \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T G_{ss}^T (A\hat{z} + B_2 (-K_{w-SMC}^{\text{equal}} \hat{z} \text{sign}(\sigma(\hat{z}))) + B_1 \hat{w}_{\text{error}} \\
&\quad + Z_w K_w C_2 z + Z_w K_w D_{21} \hat{w}_{\text{error}} + Z_w K_w D_{22} (-K_{w-SMC}^{\text{equal}} \hat{z} \text{sign}(\sigma(\hat{z}))) \\
&\quad - Z_w K_w C_2 \hat{z} - Z_w K_w D_{21} \hat{w}_{\text{error}} - Z_w K_w D_{22} (-K_{w-SMC}^{\text{equal}} \hat{z} \text{sign}(\sigma(\hat{z}))) \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e
\end{aligned}$$

Let $w_{\text{error}} = \gamma^{-2} B_1^T X_w z(t)$, and $\hat{w}_{\text{error}} = \gamma^{-2} B_1^T X_w \hat{z}(t)$

$$\begin{aligned}
\dot{V} &= \sigma^T G_{ss}^T (A\hat{z} + B_2 (-K_{w-SMC}^{\text{equal}} \hat{z} \text{sign}(\sigma(\hat{z}))) + B_1 \gamma^{-2} B_1^T X_w \hat{z} \\
&\quad + Z_w K_w C_2 z + Z_w K_w D_{21} \gamma^{-2} B_1^T X_w z - Z_w K_w C_2 \hat{z} \\
&\quad - Z_w K_w D_{21} \gamma^{-2} B_1^T X_w \hat{z}) \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T G_{ss}^T [A - B_2 K_{w-SMC}^{\text{equal}} \text{sign}(\sigma(\hat{z})) + B_1 \gamma^{-2} B_1^T X_w] \hat{z} \\
&\quad + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) (z - \hat{z}) \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T G_{ss}^T [A - B_2 K_{w-SMC}^{\text{equal}} \text{sign}(\sigma(\hat{z})) + B_1 \gamma^{-2} B_1^T X_w] \hat{z} \\
&\quad + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) e \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e
\end{aligned}$$

Let $K_{w-SMC}^{\text{equal}} := (G_{ss}^T B_2)^{-1} [G_{ss}^T (A + B_1 \gamma^{-2} B_1^T X_w)]$,

$$\begin{aligned}
\dot{V} &= \sigma^T G_{ss}^T [A - B_2 (G_{ss}^T B_2)^{-1} [G_{ss}^T (A + B_1 \gamma^{-2} B_1^T X_w)]] \text{sign}(\sigma(\hat{z})) \\
&\quad + B_1 \gamma^{-2} B_1^T X_w \hat{z} + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) e \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T [G_{ss}^T A - (G_{ss}^T B_2 (G_{ss}^T B_2)^{-1} G_{ss}^T A \\
&\quad + G_{ss}^T B_2 (G_{ss}^T B_2)^{-1} G_{ss}^T B_1 \gamma^{-2} B_1^T X_w) \text{sign}(\sigma(\hat{z})) \\
&\quad + G_{ss}^T B_1 \gamma^{-2} B_1^T X_w] \hat{z} + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) e \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T [G_{ss}^T A - G_{ss}^T A \text{sign}(\sigma(\hat{z})) - G_{ss}^T B_1 \gamma^{-2} B_1^T X_w \text{sign}(\sigma(\hat{z})) \\
&\quad + G_{ss}^T B_1 \gamma^{-2} B_1^T X_w] \hat{z} + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) e \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e \\
&= \sigma^T G_{ss}^T A [1 - \text{sign}(\sigma(\hat{z}))] \hat{z} + \sigma^T G_{ss}^T B_1 \gamma^{-2} B_1^T X_w \\
&\quad (1 - \text{sign}(\sigma(\hat{z}))) \hat{z} + \sigma^T G_{ss}^T (Z_w K_w (C_2 + D_{21} \gamma^{-2} B_1^T X_w)) e \\
&\quad - e^T (A - \gamma^{-2} B_1 B_1^T X_w - Z_w K_w (C_2 - \gamma^{-2} D_{21} B_1^T X_w)) e
\end{aligned}$$

The estimation error is $e \rightarrow 0$ as $t \rightarrow 0$.

$$\dot{V} = \sigma^T G_{ss}^T A [1 - \text{sign}(\sigma(\hat{z}))] \hat{z} + G_{ss}^T B_1 \gamma^{-2} B_1^T X_w$$

$$(1 - \text{sign}(\sigma(\hat{z}))) \hat{z} \leq 0$$

subject to

$$\text{if } \sigma > 0, \quad \dot{V} = 0$$

$$\text{if } \sigma = 0, \quad \dot{V} = 0$$

$$\text{if } \sigma < 0, \quad \dot{V} \leq -2kG_{ss}^T A\hat{z} - 2kG_{ss}^T B_1 \gamma^{-2} B_1^T X_w \hat{z} < 0$$

k is positive constant.

The above condition is satisfied on negative definite, and is *asymptotically stable*. This completes the proof of this theorem. \square

Conclusion

A stability proof of a nonlinear feedback linearization- H_{∞} -based sliding mode controller (NFL- H_{∞} -based SMC) has been done.

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