Research Article

Disturbance Observer-Based Adaptive Control of Hypersonic Vehicles with Constrained Actuators

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Received 5 August 2020; Revised 28 August 2020; Accepted 13 September 2020; Published 10 October 2020

1. Introduction

Air-breathing hypersonic vehicles (AHVs) have attracted considerable attention in both military and civil fields. Though AHVs have been developed for more than half a century, various studies have shown that some critical technologies, such as thermal protection, high-fidelity modeling, and robust control, still need to be further investigated. Based on the control-oriented model developed by Parker et al. [1], numerous theoretical studies on the control of AHVs have sprung up [2]. Early research interests focus on state feedback linearization or nonlinear dynamic inverse-based control of AHVs [3]. In recent years, the adaptive control has always been a research hotspot in the field of AHV control, and one can refer to the robust adaptive control in [4, 5], the sliding mode adaptive control in [6], the trajectory restricted adaptive control in [7], the fault-tolerant adaptive control in [8, 9], and the neural adaptive control in [10, 11].

Nevertheless, a practical problem in AHV control is the physically limited actuators [12, 13]. Recently, some outstanding works have come out focusing on the input saturation problem of AHVs [14, 15]. [16] used the disturbance observer-based (DOB) feedback linearization control to overcome constraints on actuators. A combination of backstepping design and nonlinear disturbance observer (NDO) is utilized to handle actuator constraints [17]. By introducing the sliding-mode differentiator techniques and the auxiliary systems, [11] developed the distributed finite time fault-tolerant containment control to handle the input saturation. On the other side, control allocation (CA) also plays a good role in dealing with the actuator constraints [18]. For instance, [19] handled the saturation using a robust model predictive control algorithm. [20] developed a control allocation scheme for fault-tolerant control based on a linear parameter varying (LPV) system. But the employed LPV model was complex and time-demanding for the CA problem.

Motivated by the above investigations, this paper proposes a disturbance observer-based adaptive control AHV’s with constrained actuators. In the velocity loop, the auxiliary system developed in [12] is employed to handle the saturation issue of the scramjet input. In the altitude loop, a control allocation module is exploited for the magnitude/ rate constraints and the dynamics of aerodynamic control surfaces. The main contributions of this paper can be briefly outlined as follows:
9} (1) The saturation problem of the scramjet, as well as the dynamics and magnitude/rate constraints of the control surface, is taken into consideration.

(2) NDO is developed in each step to further improve the disturbance rejection property. The observation errors are suppressed by adaptive control laws.

The rest of this paper is organized as follows: Section 2 formulates the considered problem; Section 3 presents the design process; Sections 4 and 5 give a stability analysis and a simulation study, respectively; Section 6 concludes this study.

2. Problem Statement and Preliminaries

2.1. Control-Oriented AHV Model. In this paper, we consider the longitudinal control-oriented AHV model developed by Parker et al. [1]:

\[
\ddot{V} = \frac{T \cos \alpha - D}{m} - g \sin \gamma + d_V, \quad (1)
\]

\[
\dot{h} = V \sin \gamma, \quad (2)
\]

\[
\dot{\gamma} = \frac{L + T \sin \alpha}{mV} - \frac{g \cos \gamma}{V} + d_\gamma, \quad (3)
\]

\[
\dot{\alpha} = \dot{\gamma} - \dot{\alpha}, \quad (4)
\]

\[
\dot{\eta}_i = -\bar{\xi}_i \omega_i \dot{\eta}_i - \omega_i^2 \eta_i + N_i + \psi_i \dot{\gamma}, \quad i = 1, 2. \quad (5)
\]

In (1)–(5), velocity \( V \), altitude \( h \), flight path angle (FPA) \( \gamma \), angle of attack (AoA) \( \alpha \), and pitch rate (PR) \( Q \) are system states; \( g \) is the acceleration due to gravity; mass \( m \) and moment of inertia \( I_{yy} \) are decided by the design and fuel level of AHVs; the disturbances \( d_V \), \( d_\gamma \), and \( d_\alpha \) are external disturbances with bounded derivatives; \( \eta_i \) is the ith generalized elastic coordinate; \( \omega_i \) and \( \xi_i \) are the natural frequency and the damping ratio for flexible mode \( \eta_i \), respectively; \( \psi_i \) and \( \psi'_i \) are coupling coefficients between rigid-body and flexible dynamics; thrust \( T \), lift \( L \), drag \( D \), pitching moment \( M \), and ith generalized force \( N_i \) have the following curve-fitted approximations:

\[
T = \bar{q}S(C_{T,\alpha} \alpha^3 + C_{T,\alpha'\alpha} \alpha' + C_{T,\alpha''} \alpha'' + C_{T,\theta} \theta + C_{T,\theta\alpha} \theta \alpha + C_{T,\theta''} \theta'' \theta) \Phi,
\]

\[
L = \bar{q}S(C_{L,\alpha} \alpha^3 + C_{L,\alpha'\alpha} \alpha' + C_{L,\alpha''} \alpha'' + C_{L,\alpha''} \alpha'' + C_{L,\theta} \theta + C_{L,\theta\alpha} \theta \alpha + C_{L,\theta''} \theta'' \theta),
\]

\[
D = \bar{q}S(C_{D,\alpha} \alpha^3 + C_{D,\alpha'\alpha} \alpha' + C_{D,\alpha''} \alpha'' + C_{D,\theta} \theta + C_{D,\theta\alpha} \theta \alpha + C_{D,\theta''} \theta'' \theta),
\]

\[
M = z_T T + \bar{q}cS(C_{M,\alpha} \alpha^3 + C_{M,\alpha'\alpha} \alpha' + C_{M,\alpha''} \alpha'' + C_{M,\alpha''} \alpha'' + C_{M,\theta} \theta + C_{M,\theta\alpha} \theta \alpha + C_{M,\theta''} \theta'' \theta),
\]

2.2. Problem Formulation. From a practical viewpoint, FER should comply with the saturation property:

\[
\Phi = \text{Sat}(\Phi_\epsilon) = \begin{cases} 
\Phi_\text{cooling}, & \Phi_\epsilon < \Phi_\text{cooling}, \\
\Phi_\epsilon > \Phi_\text{choking}, & \Phi_\epsilon > \Phi_\epsilon_\text{choking}, \\
\Phi_\epsilon, & \text{otherwise},
\end{cases}
\]

where \( \Phi_\epsilon \) is the FER command; \( \Phi_\text{cooling} \) is the constant lower threshold to keep sufficient heat dissipation through active cooling; the upper bound \( \Phi_\text{choking} \) is required to prevent the scramjet from thermal choking [12].

In engineering, the magnitude and rate constraints of the actuators (canard/elevator) have a great influence on control performances.

\[
\mathcal{M}_L \leq \delta \leq \mathcal{M}_U, \\
\mathcal{R}_L \leq \delta \leq \mathcal{R}_U,
\]

where \( \delta \) denote EDA \( \delta_e \) or CDA \( \delta_\alpha \); \( \mathcal{M}_L < 0 \) and \( \mathcal{M}_U > 0 \) are lower and upper bounds of the magnitude; \( \mathcal{R}_L < 0 \) and \( \mathcal{R}_U > 0 \) are the corresponding rate bounds. It should be pointed out that the upper and lower amplitude limits of a control surface can be asymmetric in reality, that is, the more common case \( \mathcal{M}_L \neq \mathcal{M}_U \) or \( \mathcal{R}_L \neq \mathcal{R}_U \) is also taken into consideration.

At the end of this subsection, several lemmas crucial for the following design are given.

Lemma 1 (see [21]). \(|x| \leq x \tanh (c x/\epsilon) + \epsilon, \) where \( x \in \mathbb{R}, \epsilon \in \mathbb{R}_+, \) and \( c = 0.2785.\)

Lemma 2 (see [22]). If a function \( W \) satisfies \( W \leq -a W + b \) with \( a \in \mathbb{R}_+, b \in \mathbb{R}_+, \) it has

\[
W(t) \leq \frac{b}{a} + \left[ W(0) - \frac{b}{a} \right] e^{-at}.
\]

Lemma 3 (see [23]). Consider a scalar nonlinear system

\[
\dot{x} = H(x) + G(x) u + d,
\]

where \( x \in \mathbb{R} \) and \( u \in \mathbb{R} \) are system state and control input, respectively; \( H(x) \) and \( G(x) \) are known scalar functions with

\[
N_1 = C_{N_1}^2 \alpha^2 + C_{N_1}^0 \alpha + C_{N_1}^0,
\]

\[
N_2 = C_{N_2}^2 \alpha^2 + C_{N_2}^0 \alpha + C_{N_2}^0 \delta_e + C_{N_2}^0 \delta_c.
\]
respect to $x$. $d \in \mathbb{R}$ represents the disturbance and $|d| \leq M_d$, where $M_d$ is an unknown positive constant.

NDO is constructed as

$$
\begin{align*}
\dot{p} &= -l(x)[l(x) + p + H(x) + G(x)u], \\
\dot{l}(x) &= \frac{\partial l(x)}{\partial x}, \\
\dot{d} &= p + l(x),
\end{align*}
$$

where $\dot{d} \in \mathbb{R}$ is the estimation of $d$; $\lambda(x) = L_d x$ with $L_d > 0$; $\dot{l}(x) = \partial \lambda(x)/\partial x = L_d$.

Defining the estimation error $\dot{d} = d - \dot{d}$, $\tilde{d}$ is governed by

$$
\dot{\tilde{d}} = \dot{d} - \dot{p} - \frac{\partial l(x)}{\partial x} \lambda = -l(x)\tilde{d} + \dot{d}.
$$

The estimation error system (18) is bounded-input-bounded output stability, that is, $\tilde{d}$ is bounded with bounded $\dot{d}$.

### 3. Controller Design

#### 3.1. Velocity Tracking Control Design

This section presents an adaptive tracking controller for velocity subsystem (1) ensuring $\dot{V} \rightarrow V_{ref}$. The velocity dynamics is first expressed in the following form:

$$
\dot{V} = H_V + G_V \Phi + D_V,
$$

where

$$
\begin{align*}
H_V &= \frac{\bar{q}s}{m} \left[ (C_{f,0}^1 \alpha^3 + C_{f,0}^2 \alpha^2 + C_{f,0}^4 \alpha + C_{f,0}^6) \cos \alpha \\
&\quad - (C_{f,0}^2 \sin \alpha + C_{f,0}^3 \eta) \right] - g \sin \gamma, \\
G_V &= (C_{r,0}^3 \alpha^3 + C_{r,0}^4 \alpha^2 + C_{r,0}^6 \alpha + C_{r,0}^8) \cos \alpha, \\
D_V &= -\bar{q}s \left( \frac{\delta^0}{C_{D,0}^1} \delta_e + C_D^1 \delta_e + C_D^2 \delta_e + C_D^3 \delta_e \right) + d_V.
\end{align*}
$$

First, similar to [23], we introduce a modified tracking error to handle the saturated $\Phi$ as

$$
\epsilon_V = V - V_{ref} - \chi_V,
$$

where $\chi_V$ is governed by

$$
\dot{\chi}_V = -k_V \chi_V + G_V(\Phi - \Phi_V),
$$

with $k_V > 0$. Then, $\dot{\epsilon}_V$ is calculated as

$$
\dot{\epsilon}_V = k_V \chi_V + H_V + G_V \Phi + D_V - \dot{V}_{ref}.
$$

Taking into account the bounded $\dot{d}_V$ and the finite deflecting velocity of physical control surfaces, $\dot{D}_V$ is assumed to be bounded. Hence, the NDO for $D_V$ is constructed.

#### 3.2. Altitude Tracking Control Design

In the altitude subsystem, control allocation is employed as an attractive technology to handle both canard and elevator servomechanism dynamics and constraints. For convenience, $\delta_e^0$ and $\delta_e^2$ represent the desired control command to be designed, $\delta_e^{com}$ and $\delta_e^{com}$ denote the control command after the process of control allocation, and $\delta_e$ and $\delta_e$ are the real control inputs applied to AHVs with the dynamics $\delta_e/\delta_e^{com} = 20/(s + 20)$ and $\delta_e/\delta_e^{com} = 20/(s + 20)$ [24]. We also define $[\Delta \delta_e, \Delta \delta_e]^{T} = [\delta_e - \delta_e^0, \delta_e - \delta_e^2]^{T}$ for the following design.

3.2.1. Design for $(h, \gamma)$-Dynamics. Recalling the kinematical equation (2), the realization of altitude reference tracking can be transferred to the FPA tracking problem. In this paper, to achieve $h \rightarrow h_{ref}$, the control system steers the AHVs to track the FPA reference:

$$
\gamma_{ref} = \arcsin \left[ \frac{k_V(h - h_{ref}) + \dot{h}_{ref}}{V} \right],
$$

where $h_{ref}$ is the smooth altitude reference generated by the traditional second-order command filter; $k_h > 0$.

Define a time-related function as $\alpha_f$, which decides the final AoA. The FPA dynamics can be calculated as

$$
\dot{\gamma} = R_f (\alpha - \alpha_f) + H_f + G_f \delta_e^0 + D_f,
$$

where

$$
\begin{align*}
\dot{p}_V &= -l_V[\lambda_V(V) + p_V + H_V + G_V \Phi], \\
\dot{l}_V &= \frac{\partial l_V(V)}{\partial V}, \\
\dot{D}_V &= p_V + \lambda_V(V).
\end{align*}
$$

Defining estimation error as $\dot{D}_V = D_V - \dot{D}_V$, it has

$$
\dot{D}_V = -l_V \dot{D}_V + D_V.
$$

For the fact that $\dot{D}_V$ is bounded, $\dot{D}_V$ is also bounded from Lemma 3, i.e., $|\dot{D}_V| \leq D_V$ with $\Theta_V > 0$.

We design the control law

$$
\Phi_{\epsilon} = \frac{1}{G_v} \left[ -k_* (V - V_{ref}) - H_v + \dot{V}_{ref} - D_V - \Theta_V \tanh \left( \frac{\epsilon_{\epsilon} V}{\epsilon_{\epsilon}} \right) \right],
$$

where $\dot{D}_V$ is estimated by NDO and $\Theta_V$ is the estimate of $\Theta_V$, generated by the adaptive law:

$$
\dot{\Theta}_V = \sigma_V \left[ -k_* \Theta_V - \epsilon_{\epsilon} \tanh \left( \frac{\epsilon_{\epsilon} V}{\epsilon_{\epsilon}} \right) \right],
$$

where $\sigma_V > 0$, $k_* > 0$, and $\epsilon_{\epsilon} > 0$. 

3.2.2. Design for $(h, \gamma)$-Dynamics. 

Recalling the kinematical equation (2), the realization of altitude reference tracking can be transferred to the FPA tracking problem. In this paper, to achieve $h \rightarrow h_{ref}$, the control system steers the AHVs to track the FPA reference:

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\gamma_{ref} = \arcsin \left[ \frac{k_V(h - h_{ref}) + \dot{h}_{ref}}{V} \right],
$$

where $h_{ref}$ is the smooth altitude reference generated by the traditional second-order command filter; $k_h > 0$.

Define a time-related function as $\alpha_f$, which decides the final AoA. The FPA dynamics can be calculated as

$$
\dot{\gamma} = R_f (\alpha - \alpha_f) + H_f + G_f \delta_e^0 + D_f,
$$

where ...

...
where
\[
R_y = \frac{qS}{mV} \left[ C_1^0 + (C_3^0 + C_7^0) \alpha^2 + C_{1,7}^0 \alpha + C_{1,7}^0 \right] \Phi + C_1^0 \alpha f^2 + C_{1,7}^0 \alpha f + C_{1,7}^0, \\
H_y = \frac{qS}{mV} \left[ C_1^0 + (C_3^0 + C_7^0) \alpha^2 + C_{1,7}^0 \alpha f + C_{1,7}^0 \right] \Phi + C_1^0 \alpha f^2 + C_{1,7}^0 \alpha f + C_{1,7}^0, \\
G_y = \frac{qS}{mV} \varepsilon_y, \\
D_y = d_{y,a} + d_{y,b} + d_y, \\
d_{y,a} = \frac{qS}{mV} \left[ (C_3^0 + C_7^0) \alpha^2 + C_{1,7}^0 \alpha \right] \Phi + (C_3^0 + C_7^0) \alpha f + (C_3^0 + C_7^0) \alpha f + C_1^0 \alpha f + C_{1,7}^0, \\
d_{y,b} = \frac{qS}{mV} \left( C_1^0 \delta_\varepsilon + C_{1,7}^0 \Delta \delta_\varepsilon \right). \\
(30)
\]

Further noting the bounded \( d_{y,a}, d_{y,b}, \) and \( d_y \), the boundedness of \( D_y \) is guaranteed. From Lemma 3, \( \tilde{D}_y \) is bounded for the fact that \( \dot{D}_y \) is bounded, i.e., \( |\dot{D}_y| \leq \Theta_y \) with \( \Theta_y > 0 \).

We design the control law to regulate the \( \dot{e}_y \) system as
\[
\delta_y^0 = \frac{1}{C_y} \left[ -k_y e_y - R_y - H_y \right] + \dot{y}_{\text{ref}} - \Theta_y \tan \left( \frac{e_y}{\varepsilon_y} \right), \\
(35)
\]
where \( \Theta_y \) is governed by the adaptive law
\[
\dot{\Theta}_y = \sigma_y \left[ -k_y \hat{\Theta}_y - e_y \tan \left( \frac{e_y}{\varepsilon_y} \right) \right], \\
(36)
\]
with \( \sigma_y > 0, \varepsilon_y > 0, \) and \( k_y > 0 \).

3.2.2. Design for \((\alpha, Q)\)-Dynamics. For the altitude subsystem, the AoA tracking error dynamics have the relationship
\[
\dot{e}_\alpha = \dot{\alpha} - \dot{\alpha}_a = \dot{\alpha} - \ddot{\alpha}_a + \dot{e}_\alpha = Q - \dot{y}_{\text{ref}} - \dot{\alpha}_f. \\
(37)
\]
Define \( e_Q = Q - Q_d \) with \( Q_d = -k_a e_a + \dot{y}_{\text{ref}} + \dot{\alpha}_f, \) and \( \dot{e}_a \) can be rewritten as
\[
\dot{e}_a = -k_a e_a + e_Q, \\
\dot{Q}_d = k_a^2 e_a + k_a e_a + \dot{y}_{\text{ref}} + \ddot{\alpha}_f. \\
(38)
\]

Then, the PR dynamics is formulated as
\[
\dot{Q} = H_Q + G_Q \delta_\varepsilon + D_Q, \\
(39)
\]
where
\[
H_Q = \frac{z_y qS}{I_{yy}} \left[ (C_3^0 + C_7^0) \alpha^2 + C_{1,7}^0 \alpha + C_{1,7}^0 \right] \Phi + (C_3^0 \alpha^2 + C_7^0 \alpha^2 + C_{1,7}^0 \alpha + C_{1,7}^0), \\
G_Q = \frac{qS}{I_{yy}} C_{M,w}, \\
D_Q = d_Q + d_{Q,b} + \sum_{i=1}^{2} \frac{\psi_j \tilde{y}_j}{I_{yy}}, \\
d_{Q,b} = \frac{qS}{I_{yy}} \left( C_{M,w} \delta_\varepsilon + C_{M,w} \Delta \delta_\varepsilon \right). \\
(40)
\]

Following a similar analysis to previous steps, the boundedness of \( D_Q \) is also guaranteed. Thus, the PR tracking error dynamics is formulated as
\[
\dot{e}_Q = H_Q + G_Q \delta_\varepsilon + D_Q - k_a^2 e_a - \dot{y}_{\text{ref}} - \ddot{\alpha}_f. \\
(41)
\]
Then, NDO for $D_Q$ is constructed as
\[
\begin{aligned}
\dot{D}_Q &= -I_Q \dot{D}_Q + D_Q, \\
\dot{D}_Q &= -I_Q \dot{D}_Q + D_Q, \\
\end{aligned}
\]
(43)

Define estimation error as $\hat{D}_Q = D_Q - D_Q$, it has
\[
\begin{aligned}
\dot{\hat{D}}_Q &= -I_Q \hat{D}_Q + D_Q. \\
\end{aligned}
\]
(44)

The estimation error $\hat{D}_Q$ is bounded according to the corresponding analysis in [25, 26]. Considering the bounded $\hat{D}_Q$, $\hat{D}_Q$, and $\sum_{i=1}^{2} (\psi_i \mu_i)$, the estimation error $\hat{D}_Q$ is bounded. Similarly, $\hat{D}_Q$ is bounded for the fact that $\hat{D}_Q$ is bounded, i.e., $|\hat{D}_Q| \leq \Theta Q$ with $\Theta Q > 0$.

We design the control law to regulate the $\hat{E}_Q$ system as
\[
\delta_e = \frac{1}{c_Q} \left[ -k_o \delta_e + k_o \sigma + \delta_n - H + \bar{\gamma}_n + \bar{\delta} \right],
\]
(45)

with $\sigma Q > 0$, $\epsilon Q > 0$, and $k_o > 0$.

3.2.3. Control Allocation Design for CDA-EDA. In this subsection, a control allocation algorithm on CDA and EDA is integrated to the proposed control considering the hard magnitude and rate constraints of canard and elevator deflections.

Considering the components $M$ in (10), we briefly define
\[
\begin{aligned}
\delta^{\text{com}}_M &= C_H \delta^H + C_T \delta^T, \\
\delta^{\text{com}}_M &= C_H \delta^H + C_T \delta^T,
\end{aligned}
\]
(46)

where $\delta^{\text{com}}_M$ denotes the commands generated by control allocation; $\delta^{\text{com}}_M$ presents the desired values.

Meanwhile, assume the future control sequence with $z$ prediction intervals at the current interval $k$ as
\[
\Xi_k = \begin{bmatrix}
\delta^{\text{com}}_c(k|k) & \delta^{\text{com}}_c(k+1|k) & \cdots & \delta^{\text{com}}_c(k+z-1|k) \\
\delta^{\text{com}}_e(k|k) & \delta^{\text{com}}_e(k+1|k) & \cdots & \delta^{\text{com}}_e(k+z-1|k)
\end{bmatrix}^T.
\]
(47)

As the flight control system is a sampled-data system, the magnitude and rate can establish the following simple first-order approximation:
\[
\dot{\delta} = \frac{\delta(t) - \delta(t - t_s)}{t_s},
\]
(48)

where $t_s$ is the control interval and $\delta(t - t_s)$ is the CDA/EDA command generated in the last control interval; the rate constraint can be absorbed into the magnitude constraint:
\[
\delta(t - t_s) + \mathcal{R}_1 t_s \leq \delta(t) \leq \delta(t - t_s) + \mathcal{R}_U t_s.
\]
(49)

And new constraints can be defined as
\[
\begin{aligned}
\mathcal{M}_{\text{max}} &= \max \{ \mathcal{M}_1, \delta(t - t_s) + \mathcal{R}_U t_s \}, \\
\mathcal{M}_{\text{min}} &= \min \{ \mathcal{M}_1, \delta(t - t_s) + \mathcal{R}_1 t_s \}.
\end{aligned}
\]
(50)

Therefore, the optimization problem of the above control allocation can be expressed in the following form:
\[
\begin{aligned}
\min_{\Xi_k} & J(\Xi_k) \\
\text{subject to} & \quad \mathcal{M}_{\text{min}} \leq \delta \leq \mathcal{M}_{\text{max}}
\end{aligned}
\]
(51)

with
\[
J(\Xi_k) = J_{\Delta_1}(\Xi_k) + J_{\Delta_2}(\Xi_k),
\]
(52)

\[
J_{\Delta_1}(\Xi_k) = \sum_{j=1}^{z} \left( w_j^\delta \left[ \delta^{\text{com}}_c(k+j|k) - \delta^{\text{des}}_c(k+j|k) \right] \right)^2,
\]
(53)

\[
J_{\Delta_2}(\Xi_k) = \sum_{j=1}^{z} \left( w_j^\epsilon \left[ \delta^{\text{com}}_e(k+j|k) - \delta^{\text{des}}_e(k+j-1|k) \right] \right)^2
\]
\[
+ \sum_{j=1}^{z} \left( w_j^\delta \left[ \delta^{\text{com}}_c(k+j|k) - \delta^{\text{des}}_c(k+j-1|k) \right] \right)^2,
\]
(54)

where $w_j^\delta$ is the tuning weight matrices for the $j$th step.

Remark 1. The cost function in (52) is set for two main practical purposes. One goal reflected in $J_{\Delta_1}(\Xi_k)$ is to pursue approximating precision, and the other goal reflected in $J_{\Delta_2}(\Xi_k)$ is to maximize energy savings.

4. Stability Analysis

In this section, stability and performance properties of the closed-loop systems are summarised.
Consider the Lyapunov function $\mathcal{L} = \mathcal{L}_V + \mathcal{L}_{h,y} + \mathcal{L}_{a,Q}$, where

$$\mathcal{L}_V = \frac{1}{2} \left( \dot{e}_V^2 + \sigma_V \Theta_V^2 \right),$$

$$\mathcal{L}_{h,y} = \frac{1}{2} \left( \dot{e}_y^2 + \sigma_y \Theta_y^2 \right),$$

$$\mathcal{L}_{a,Q} = \frac{1}{2} \left( \dot{e}_a^2 + \dot{e}_Q^2 + \epsilon_Q \Theta_Q^2 \right).$$

(55)

Utilizing (23), (27) together with Lemma 1 and applying inequalities: $2\Theta_V \Theta_V \leq \Theta_V^2$, $\mathcal{L}_V$ satisfies

$$\dot{\mathcal{L}}_V = e_V \dot{e}_V + \frac{1}{\sigma_V} \dot{\Theta}_V \dot{\Theta}_V \leq -k_v e_V^2 - \frac{k'_V}{2} \dot{\Theta}_V^2 + \left( \frac{k'_V}{2} \Theta_V + \epsilon_V \right) \Theta_V.$$  

(56)

Similarly, $\mathcal{L}_{h,y}$ and $\mathcal{L}_{a,Q}$ can be calculated from (32), (35), (36) and (41), (44), (45), respectively.

$$\dot{\mathcal{L}}_{h,y} = e_y \dot{e}_y + \frac{1}{\sigma_y} \dot{\Theta}_y \dot{\Theta}_y \leq -k_y e_y^2 - \frac{k'_y}{2} \dot{\Theta}_y^2 + \left( \frac{k'_y}{2} \Theta_y + \epsilon_y \right) \Theta_y,$$

(57)

$$\dot{\mathcal{L}}_{a,Q} = e_a \dot{e}_a + k'_Q \dot{\Theta}_Q \dot{\Theta}_Q \leq -k_a e_a^2 - k'_Q \dot{\Theta}_Q^2 + \left( \frac{k'_Q}{2} \Theta_Q + \epsilon_Q \right) \Theta_Q.$$  

(58)

By adding inequalities (56)–(58), we obtain

$$\dot{\mathcal{L}} \leq -k_v e_V^2 - (k_y + R_y) e_y^2 - k_a e_a^2 - (k_Q - k_a) e_Q^2 - \frac{k'_V}{2} \dot{\Theta}_V^2 - \frac{k'_y}{2} \dot{\Theta}_y^2 - \frac{k'_Q}{2} \dot{\Theta}_Q^2 + \left( \frac{k'_V}{2} \Theta_V + \epsilon_V \right) \Theta_V + \left( \frac{k'_y}{2} \Theta_y + \epsilon_y \right) \Theta_y + \left( \frac{k'_Q}{2} \Theta_Q + \epsilon_Q \right) \Theta_Q \leq -a_x \mathcal{L} + b_x,$$

(59)

where $k_Q > k_a$, $k_y > -R_y$, and $a_x, b_x \in \mathbb{R}_+$ are constants expressed as

$$a_x = \min \left\{ 2k_v, 2(k_y + R_y), 2k_a, 2(k_Q - k_a), k'_V \sigma_V, k'_y \sigma_y, k'_Q \sigma_Q \right\},$$

$$b_x = \left( \frac{k'_V}{2} \Theta_V + \epsilon_V \right) \Theta_V + \left( \frac{k'_y}{2} \Theta_y + \epsilon_y \right) \Theta_y + \left( \frac{k'_Q}{2} \Theta_Q + \epsilon_Q \right) \Theta_Q.$$  

(60)

Obviously, $\dot{\mathcal{L}} < 0$ provided that $\mathcal{L} > b_x/a_x$. Therefore, all signals involved in the Lyapunov function and utilized in the design are uniformly ultimately bounded.

Applying Lemma 2 to (59), it is concluded that the boundedness of the related variables and signals in the proposed control can be ensured. Meanwhile, the tracking error ex with $x = V, h$ retains the uniformly ultimate boundedness property:

$$\max \{ |e_V(t)|, |e_h(t)| \} \leq \sqrt{2 \frac{b_x}{a_x}} \left[ \mathcal{L}(0) - \frac{b_x}{a_x} e^{-a_x t} \right],$$  

(61)

with

$$\lim_{t \to \infty} \max \{ |e_V(t)|, |e_h(t)| \} \leq \sqrt{2 \frac{b_x}{a_x}}.$$  

(62)

### 5. Simulation Results

This section implements a simulation study to verify the proposed controller. In the simulating model, the nominal parameter values of the AHV model are selected according to [1]. For comparison, a slightly modified version of the adaptive fault-tolerant control presented in [9] is applied to the same scenario (denoted as “compared control” for convenience).

The simulated AHV is initially cruising at $V(0) = 7846$ ft/s, $h(0) = 85000$ ft, $\gamma(0) = 0$ rad, $\alpha(0) = 2.6$ deg, $\eta_1(0) = 0.92$ ft/s, $\dot{\alpha}(0) = 0.61$ ft/s and plans to implement a maneuver in the next 600 s. $\Phi_{\text{cooling}} = 0.01$ and $\Phi_{\text{choking}}$ is selected according to [12]; $M_L = -20$ deg, $M_A = 20$ deg, $\delta_L = -60$ deg/s, and $\delta_R = 60$ deg/s; the simulating control interval is set as $t_f = 10$ ms.

Velocity and altitude references are generated by passing step command signals:

$$V_c(t) = \begin{cases} 7,846, & t = 0s, \\ 9,846, & t \in (0s, 300s], \\ 9,000, & t \in (300s, 600s], \\ 85,000, & t = 0s, \\ 110,000, & t \in (0s, 300s], \\ 95,000, & t \in (300s, 600s], \end{cases}$$

(63)

through a second-order filter [4] with the damping ratio of 0.9 and the natural frequency of 0.03 rad/s.

The control parameters are set as follows. In the velocity loop, $k_v = 5$, $k_y = 10$, $\sigma_V = 0.1$, $\epsilon_V = 1$, and $t_f = 5$. In the altitude loop, $k_y = 20$, $k'_y = 1$, $\sigma_y = 0.5$, $\epsilon_y = 5$, $t_f = 10$; $k_a = 10$, $k_Q = 40$, $k'_Q = 0.3$, $\sigma_Q = 0.8$, $\epsilon_Q = 10$, and $t_Q = 20$.

The main simulating results are given in Figures 1–9. Figures 1 and 2 show the output tracking results during the ascending maneuver and the descending maneuver, from which we see that the proposed control performs better,
Figure 1: Velocity tracking results in the comparison simulation: (a) velocity tracking; (b) velocity tracking error.

Figure 2: Altitude tracking results in the comparison simulation: (a) altitude tracking; (b) altitude tracking error.
though both controls achieve the desired conditions. Time histories of flight attitudes including FPA, AoA, and PR are given in Figure 3. It is seen from Figure 1 that the desired velocity reference cannot be tracked because of the FER saturation property, also see Figure 5. In this situation, the employed auxiliary system generates a modified reference, which is less aggressive and can be well tracked within the ability of scramjet. Figure 4 shows the flexible states in the comparison simulation. Figures 5–7 indicate the control input signals of FER, CDA, and EDA. It is seen from Figures 6 and 7 that both the magnitude constraints and rate constraints of CDA/EDA are well constrained within prescribed bound under the proposed control, which performs better than the compared control. Figure 8 shows the results

Figure 3: Flight attitudes in the comparison simulation: (a) flight path angle; (b) angle of attack; (c) pitch rate.

Figure 4: Flexible states in the comparison simulation.
Figure 5: FER and state of the auxiliary system $\chi_V$: (a) FER; (b) state of the auxiliary system.

Figure 6: CDA and CDA rate in the comparison simulation: (a) canard deflection angle; (b) canard deflection angle rate.
Figure 7: EDA and EDA rate in the comparison simulation: (a) elevator deflection angle; (b) elevator deflection angle rate.

Figure 8: Results of lumped disturbance estimation.

Figure 9: Estimation errors $\hat{\Theta}_V$, $\hat{\Theta}_\gamma$, and $\hat{\Theta}_Q$. 
of lumped disturbance estimation from NDOs. \( \hat{\Theta}_v, \hat{\Theta}_y, \) and \( \hat{\Theta}_Q \) estimated by the adaptive laws are shown in Figure 9.

6. Conclusions

In this paper, a disturbance observer-based adaptive control for a hypersonic vehicle has been addressed to obtain the capability of handling external disturbances and actuator constraints. Fast disturbance estimate has been addressed via NDO and adaptive control compensation in each subsystem. An auxiliary system has been employed subject to the scramjet input saturation. Meanwhile, there is an effective mechanism of dealing with the magnitude/rate constraints and dynamics of aerodynamic control surfaces by introducing a control allocation module. Simulations have been provided to verify the effectiveness of the proposed control.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

Conflicts of Interest

The authors declare that there is no conflict of interest regarding the publication of this paper.

Acknowledgments

This work was supported by the National Natural Science Foundation of China (grant number 61304108).

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